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Annotate every variant in VCF with information from (subset): dbSNP, UCSC, ClinVar, OMIM, KEGG, Pfam, Ensembl, ESP, 1000G, CADD, Polyphen, SIFT, ENCODE, HPRD, COSMIC, GERP, FliCons, VISTA.

Variants, annotations, phenotypes & genotypes together in a database.

Prioritize genetic variants in various disease contexts based on genome annotations, sample genotypes, and sample relationships.

Powerful ad hoc queries

```sql
$ gemini query -q query
"SELECT chrom, start, end, ref, alt, gene, impact, aaf
FROM variants
WHERE in_dbSNP = 0
and aaf < 0.01
and cadd_scaled_score >= 30
and is_loss_of_function = 1
and my_disease_regions = 1
and gt_types.mom == HET
and gt_types.dad == HET
and gt_types.child == HOM_ALLELE"
```

Family studies

Tumor/Normal

Cohort studies

w/ disease

w/o disease
GEMINI (GEnome MINing) is a flexible framework for exploring genetic variation in the context of the wealth of genome annotations available for the human genome. By placing genetic variants, sample phenotypes and genotypes, as well as genome annotations into an integrated database framework, GEMINI provides a simple, flexible, and powerful system for exploring genetic variation for disease and population genetics.

Using the GEMINI framework begins by loading a VCF file (and an optional PED file) into a database. Each variant is automatically annotated by comparing it to several genome annotations from sources such as ENCODE tracks, UCSC tracks, OMIM, dbSNP, KEGG, and HPRD. All of this information is stored in portable SQLite database that allows one to explore and interpret both coding and non-coding variation using “off-the-shelf” tools or an enhanced SQL engine.

Please also see the original manuscript.

Note:

1. GEMINI solely supports human genetic variation mapped to build 37 (aka hg19) of the human genome.
2. GEMINI is very strict about adherence to VCF format 4.1.
3. For best performance, load and query GEMINI databases on the fastest hard drive to which you have access.
In addition to the documentation, please review the following tutorials if you are new to GEMINI. We recommend that you follow these tutorials in order, as they introduce concepts that build upon one another.

- Introduction to GEMINI, basic variant querying and data exploration. html pdf
- Identifying de novo mutations underlying Mendelian disease html pdf
- Identifying autosomal recessive variants underlying Mendelian disease html pdf
- Identifying autosomal dominant variants underlying Mendelian disease html pdf
- Other GEMINI tools html pdf
CHAPTER 3

Latest news

New Installation

In version 0.18, we have introduced a new installation procedure based on conda that should make the installation more reliable. For users with an existing installation with any trouble using `gemini update --devel`, we suggest to do a fresh install using a command like:

```bash
wget https://github.com/arq5x/gemini/raw/master/gemini/scripts/gemini_install.py
python gemini_install.py $tools $data
PATH=$tools/bin:$data/anaconda/bin:$PATH
```

where `$tools` and `$data` are paths writable on your system.

With an existing `$tool` and `$data` directory from a previous install, you can use the installer to re-install the Python code with the new version, but leave the existing data in place. To do this, first remove the old anaconda directory:

```bash
rm -rf $data/anaconda
```

then run the installation commands above.

Changes to Inheritance Tools

As of version 0.16.0, the built-in Mendelian inheritance tools are more stringent by default (they can be relaxed with the `--lenient`) option. By default, samples with unknown phenotype will not affect the tools, and strict requirements are placed on family structure. See the the `docs` for more info. In addition, the inheritance tools now support multi-generational pedigrees.
New GEMINI Workflow

As version 0.12.2 of GEMINI it is required that your input VCF file undergo additional preprocessing such that multi-allelic variants are decomposed and normalized using the vt toolset from the Abecasis lab. Note that we have also decomposed and normalized all of the VCF-based annotation files (e.g., ExAC, dbSNP, ClinVar, etc.) so that variants and alleles are properly annotated and we minimize false negative and false positive annotations. For a great discussion of why this is necessary, please read this blog post from Eric Minikel in Daniel MacArthur’s lab.

Essentially, VCF preprocessing for GEMINI now boils down to the following steps.

0. If working with GATK VCFs, you need to correct the AD INFO tag definition to play nicely with vt.

1. Decompose the original VCF such that variants with multiple alleles are expanded into distinct variant records; one record for each REF/ALT combination.

2. Normalize the decomposed VCF so that variants are left aligned and represented using the most parsimonious alleles.

3. Annotate with VEP or snpEff.

4. bgzip and tabix.

A workflow for the above steps is given below.

```
# setup
VCF=/path/to/my.vcf
NORMVCF=/path/to/my.norm.vcf.gz
REF=/path/to/human.b37.fasta
SNPEFFJAR=/path/to/snpEff.jar

# decompose, normalize and annotate VCF with snpEff.
# NOTE: can also swap snpEff with VEP
zless $VCF
  | sed 's/ID=AD,Number=./ID=AD,Number=R/'
  | vt decompose -s -
  | vt normalize -r $REF -
  | java -Xmx4G -jar $SNPEFFJAR GRCh37.75
  | bgzip -c > $NORMVCF
  | tabix -p vcf $NORMVCF

# load the pre-processed VCF into GEMINI
gemini load --cores 3 -t snpEff -v $NORMVCF $db

# query away
gemini query -q "select chrom, start, end, ref, alt, (gts).(*) from variants"
  --gt-filter "gt_types.mom == HET and "
  "gt_types.dad == HET and "
  "gt_types.kid == HOM_ALT"
  $db
```
If you use GEMINI in your research, please cite the following manuscript:

Paila U, Chapman BA, Kirchner R, Quinlan AR (2013)
GEMINI: Integrative Exploration of Genetic Variation and Genome Annotations.
CHAPTER 5

Table of contents

Installation

Automated installation

GEMINI contains an automated installation script which installs GEMINI along with required Python dependencies, third party software and data files.

$ wget https://raw.github.com/arq5x/gemini/master/gemini/scripts/gemini_install.py

While other tools (e.g., curl) can be used to download this file, wget is required by the installation script. Please install wget if it is not already available:

Once you have downloaded the above installation script, you can proceed as follows:

$ python gemini_install.py /usr/local /usr/local/share/gemini
$ export PATH=$PATH:/usr/local/gemini/bin

# it would be wise to add the above line to your `.bashrc` or `.bash_profile`

This installs the GEMINI executable as `/usr/local/bin/gemini`, other required third party dependencies in `/usr/local/bin`, and associated data files in `/usr/local/share/gemini`. Please note that this is merely an example: one can place the GEMINI executables and annotation files in any directories one wishes.

Note: The automated installtion script typically requires ~30 minutes, primarily owing to the time required to download the GEMINI genome annotation files. Also, please note that the annotation files requires ~15Gb of storage, so please ensure that the data directory (`/usr/local/share/gemini` in the example above) has sufficient space.

Tip: Some tips and tricks for installation issues:

1. Some older versions of wget have certificate problems with GitHub files. If you run into this problem, you can alternatively download the install script using `wget --no-check-certificate` or curl -O.
2. The installation script is idempotent and you can re-run it multiple times without any issues. If you experience internet connectivity or other transient errors during installation, a re-run can often solve the problem (fingers crossed).

3. If you’re installing behind a proxy you’ll need to set proxy information in a condarc file and also set all_proxy, http_proxy and https_proxy in your .bashrc file. See this mailing list discussion for more information.

4. The installer tries to ignore site-wide environmental variables pointing to other Python installations, but if you have issues with pulling in Python libraries from other locations, you clear these by unsetting/setting these environmental variables before running: unset PYTHONPATH && unset PYTHONHOME && export PYTHONNOUSERSITE=1

Dependencies

The installer requires:

- Python 2.7.x
- git
- wget
- a working C / C++ compiler such as gcc
- zlib (including headers)

These dependencies already exist on most UNIX/LINUX/OS X machines. However, on minimalist machines such as fresh Amazon Cloud instances, one may have to install these dependencies before running the automated installer. In the case of Amazon instances, the following command should take care of all of the above dependencies:

```
$ sudo yum -y install python27 git wget gcc gcc-c++ zlib-devel
```

Installing without root access.

As many users do not have root or sudo access, the automated installer also has options to install in “non-root” environments:

```
$ python gemini_install.py ~/gemini ~/gemini --nosudo
```

Updating your PATH to find the GEMINI executable

At this point, you will have a self-contained installation of GEMINI, including both the software and its associated genome annotations. However, if you have done a custom install in a “non-root” environment, you will first need to update your PATH environment variable to include the path to the bin directory that you just created by running the automated installer.

For example, if, as above, you placed you custom install in ~/gemini, you would need to update your PATH as follows. It would be wise to also add this to your .bashrc or .bash_profile:

```
$ export PATH=$PATH:~/gemini/bin
```

Note that this change will only last for the life of your current terminal session. To make this more permanent, update your .bash_profile so that this change is made each time you login.

If successful, you should be able to run the following command from anywhere on your system:
Running unit tests

gemini only installs the software itself, not the source repository with the tests. To run the tests, do:

```
$ git clone https://github.com/arq5x/gemini
$ git checkout v0.18.3 # or the current version that you installed
$ cd gemini
$ bash master-test.sh
```

Updating the GEMINI executables and annotations

Once installed with the automated installer, it is easy to upgrade the GEMINI programs and associated data files to the latest released version with:

```
$ gemini update
```

There are also flags to update to the latest development version of the code or to only update data files without updating the source:

```
$ gemini update --devel
$ gemini update --dataonly
```

To download optional large files associated with specific analyses in GEMINI, like GERP scores per base pair and CADD scores, pass the `--extra` flag:

```
$ gemini update --dataonly --extra cadd_score
$ gemini update --dataonly --extra gerp_bp
```

Software dependencies

GEMINI depends upon several widely-used genomics command line software as well as multiple Python packages. We recognize that the dependency stack is quite deep and are working on ways to minimize dependencies in the interest of the most streamlined installation process possible. Nonetheless, the following are core dependencies:

1. Python 2.7.x
2. grabix
3. tabix (only by annotate tool)
4. bedtools (only by windower tool)
5. pybedtools (only by windower tool)

Manual installation

Once the above dependencies have been installed, one can begin installing GEMINI itself. To install you should download the latest source code from GitHub, either by going to:
and clicking on “Downloads”, or by cloning the git repository with:

```bash
$ git clone https://github.com/arq5x/gemini.git
```

Once you have the source code, run:

```bash
$ cd gemini
$ sudo python setup.py install
```

to install it. If you don’t have permission to install it in the default directory, you can simply build the source in-place and use the package from the git repository:

```bash
$ python setup.py build_ext --inplace
```

### Installing annotation files

One of the more appealing features in GEMINI is that it automatically annotates variants in a VCF file with several genome annotations. However, you must first install these data files on your system. It’s easy enough — you just need to run the following script and tell it in which full path you’d like to install the necessary data files. The recommended path is `/usr/local/share`, but you can install the data files wherever you want.

```bash
$ python gemini/install-data.py /usr/local/share/
```

**Note:** Annotation files like GERP at base pair resolution and CADD scores are not part of this default installation owing to their large file size. They may however be installed as additional data files using the `gemini update --dataonly` option, with the flag `--extra` for `gerp_bp` and `cadd_score`.

### Using previously installed annotation files

If you have installed GEMINI with the annotation files on a server and you can NFS mount the annotation files, you can tell a local install of GEMINI where those annotation files are by making the file `~/.gemini/gemini-config.yaml`:

```yaml
annotation_dir: /path/to/nfs_mounted/gemini/data
versions:
  GRCh37-gms-mappability.vcf.gz: 2
  hgl9.rmsk.bed.gz: 2
```

### Running the testing suite

GEMINI comes with a full test suite to make sure that everything has installed correctly on your system. We **strongly** encourage you to run these tests.

```bash
$ bash master-test.sh
```
Quick start

gemini is designed to allow researchers to explore genetic variation contained in a VCF file. The basic workflow for working with gemini is outlined below.

Importing VCF files into gemini.

---

**Note:** we now recommend splitting variants with multiple alternates and left-aligning, and trimming all variants before loading into gemini.

See Step 1. split, left-align, and trim variants for a detailed explanation.

Before we can use GEMINI to explore genetic variation, we must first load our VCF file into the GEMINI database framework. We expect you to have first annotated the functional consequence of each variant in your VCF using either VEP or snpEff (Note that v3.0+ of snpEff is required to track the amino acid length of each impacted transcript). Logically, the loading step is done with the gemini load command. Below are two examples based on a VCF file that we creatively name my.vcf. The first example assumes that the VCF has been pre-annotated with VEP and the second assumes snpEff.

```
# VEP-annotated VCF
$ gemini load -v my.vcf -t VEP my.db

# snpEff-annotated VCF
$ gemini load -v my.vcf -t snpEff my.db
```

Assuming you have a valid VCF file produced by standard variation discovery programs (e.g., GATK, FreeBayes, etc.), one loads the VCF into the gemini framework with the load submodule:

```
$ gemini load -v my.vcf my.db
```

In this step, gemini reads and loads the my.vcf file into a SQLite database named my.db, whose structure is described here. While loading the database, gemini computes many additional population genetics statistics that support downstream analyses. It also stores the genotypes for each sample at each variant in an efficient data structure that minimizes the database size.

Loading is by far the slowest aspect of GEMINI. Using multiple CPUs can greatly speed up this process.

```
$ gemini load -v my.vcf --cores 8 my.db
```

Querying the gemini database.

If you are familiar with SQL, gemini allows you to directly query the database in search of interesting variants via the -q option. For example, here is a query to identify all novel, loss-of-function variants in your database:

```
$ gemini query -q "select * from variants where is_lof = 1 and in_dbsnp = 0" my.db
```

Or, we can ask for all variants that substantially deviate from Hardy-Weinberg equilibrium:

```
$ gemini query -q "select * from variants where hwe < 0.01" my.db
```
Annotation with snpEff or VEP

*GEMINI* depends upon external tools to predict the functional consequence of variants in a VCF file. We currently support annotations produced by either *SnpEff* or *VEP*.

**Note:** Versions tested: VEP versions 73 through 75 and core SnpEff versions 3.0 through 3.6. *GEMINI* supports *ENSEMBL* annotations hence users are expected to download genome databases for these tools as represented in the examples below.

**Note:** Version support would be subsequently updated here, as we test along and add or edit changes available with the latest version of these tools.

Recommended instructions for annotating existing VCF files with these tools are summarized here.

**Stepwise installation and usage of VEP**

Download the Variant Effect Predictor “standalone perl script” from *Ensembl*. You can choose a specific version of VEP to download here.

Example:

```
Download version 74
```

Untar the tarball into the current directory

```
$ tar -zxvf variant_effect_predictor.tar.gz
```

This will create the variant_effect_predictor directory. Now do the following for install:

```
$ cd variant_effect_predictor
$ perl INSTALL.pl [options]
```

By default this would install the API’s, bioperl-1.2.3 and the cache files (in the $HOME/.vep directory).

**Homebrew or Anaconda VEP installation**

If you are a *Homebrew, Linuxbrew* or *Anaconda* user, there is an automated recipe to install the main VEP script and plugins in the CloudBioLinux homebrew repository:

```
$ brew tap chapmanb/cbl
$ brew update
$ brew install vep
```

For Anaconda/Miniconda, just make sure you are pointing to the *bioconda* channel:

```
$ conda install variant-effect-predictor -c bioconda
```
Manual installation of VEP

For those (e.g. mac users) who have a problem installing through this install script, try a manual installation of the API’s, BioPerl-1.2.3 and set all pre-requisites for running VEP (DBI and DBD::mysql modules required). The appropriate pre-build caches should be downloaded for Human to the $HOME/.vep directory and then untar.

You may follow instructions at http://www.ensembl.org/info/docs/api/api_installation.html which provides alternate options for the API installation and additional tips for windows/mac users. It also has information for setting up your environment to run VEP.

Example download of the cache files

```bash
```

You may change the release date in this example to get the appropriate cache files for your version of VEP that you have installed.

Example

```bash
```

Cache requires the gzip and zcat utilities. VEP uses zcat to decompress cached files. For systems where zcat may not be installed or may not work, the following option needs to be added along with the --cache option:

```
--compress "gunzip -c"
```

Running VEP

You may now run VEP as:

```bash
$ perl variant_effect_predictor.pl [OPTIONS]
```

We recommend running VEP with the following options as currently we support VEP fields specified as below:

```bash
$ perl variant_effect_predictor.pl --cache \ 
   --sift b \ 
   --polyphen b \ 
   --symbol \ 
   --numbers \ 
   --biotype \ 
   --total_length \ 
   --o output \ 
   --vcf \ 
   --fields Consequence,Codons,Amino_acids,Gene,SYMBOL,Feature,EXON,PolyPhen,SIFT, 
   ...Protein_position,BIOTYPE
```

A documentation for the above specified options may be found at http://www.ensembl.org/info/docs/tools/vep/script/vep_options.html

As of GEMINI version 0.8.0, you can also run VEP with additional fields, which will be automatically added to the variants table as columns. As an example, run VEP on your VCF with the dbNSFP and LOFTEE plugins to annotate potential high impact variations:

```bash
$ variant_effect_predictor.pl --sift b --polyphen b --symbol --numbers --biotype \ 
   --total_length --canonical --ccds \ 
   --fields Consequence,Codons,Amino_acids,Gene,SYMBOL,Feature,EXON,PolyPhen,SIFT, 
   ...Protein_position,BIOTYPE,CANONICAL,CCDS,RadialSVM_score,RadialSVM_pred,LR_score,LR_ 
   ...pred,CADD_raw,CADD_phred,Reliability_index,LoF,LoF_filter,LoF_flags \ 
```

5.3. Annotation with snpEff or VEP
Feeding this into GEMINI produces a variants table with columns for each of the additional VEP metrics. The annotation loader names each column by prefixing `vep_` to the origin VEP name, so select on `vep_radialsvm_score` or `vep_lof_filter` in the final database.

### Stepwise installation and usage of SnpEff

**Note:** Basic Requirements: Java v1.7 or later; at least 4GB of memory

Download the supported versions of SnpEff from [http://snpeff.sourceforge.net/download.html](http://snpeff.sourceforge.net/download.html)

Example:

```
$ wget http://sourceforge.net/projects/snpeff/files/snpeff_v3_6_core.zip
```

**Note:** SnpEff should be installed preferably in `snpeff` directory in your home directory. Else, you must update the `data_dir` parameter in your `snpeff.config` file. For e.g. if the installation of snpEff has been done in `~/src` instead of `~/` then change the `data_dir` parameter in `snpeff.config` to `data_dir = ~/src/snpeff/data/`

Unzip the downloaded package.

```
$ unzip snpeff_v3_6_core.zip
```

Change to the `snpeff` directory and download the genome database.

```
$ cd snpeff_v3_6_core
$ java -jar snpeff.jar download GRCh37.69
```

Unzip the downloaded genome database. This will create and place the genome in the ‘data’ directory

```
$ unzip snpeff_v3_6_GRCh37.69.zip
```

To annotate a vcf using snpEff, use the default options as below:

**Note:** Memory options for the run may be specified as `-Xmx4G` (4GB)

```
$ java -Xmx4G -jar snpeff.jar -i vcf -o vcf GRCh37.69 example.vcf > example_snpeff.vcf
```

If running from a directory different from the installation directory, the complete path needs to be specified as, e.g.:

```
$ java -Xmx4G -jar path/to/snpeff/snpeff.jar -c path/to/snpeff/snpeff.config GRCh37.69 path/to/example.vcf > example_snpeff.vcf
```

**Note:** When using the latest versions of snpEff (e.g. 4.1) annotate your VCF with the additional parameters `-classic` and `-formatEff`. This would ensure proper loading of the gene info columns in the variants table.
Columns populated by snpEff/VEP tools

The following variant consequence columns in the variant/variant_impacts table, are populated with these annotations, which are otherwise set to null.

- anno_id
- gene
- transcript
- exon
- is_exonic
- is_lof
- is_coding
- codon_change
- aa_change
- aa_length
- biotype
- impact
- impact_so
- impact_severity
- polyphen_pred
- polyphen_score
- sift_pred
- sift_score

Standardizing impact definitions for GEMINI

GEMINI uses slightly modified impact terms (for ease) to describe the functional consequence of a given variant as provided by snpEff/VEP.

The table below shows the alternate GEMINI terms used for snpEff/VEP.

<table>
<thead>
<tr>
<th>GEMINI terms</th>
<th>snpEff terms</th>
<th>VEP terms (uses SO by default)</th>
</tr>
</thead>
<tbody>
<tr>
<td>splice_acceptor</td>
<td>SPLICE_SITE_ACCEPTOR</td>
<td>splice_acceptor_variant</td>
</tr>
<tr>
<td>splice_donor</td>
<td>SPLICE_SITE_DONOR</td>
<td>splice_donor_variant</td>
</tr>
<tr>
<td>stop_gain</td>
<td>STOP_GAINED</td>
<td>stop_gained</td>
</tr>
<tr>
<td>stop_loss</td>
<td>STOP_LOST</td>
<td>stop_lost</td>
</tr>
<tr>
<td>frame_shift</td>
<td>FRAME_SHIFT</td>
<td>frameshift_variant</td>
</tr>
<tr>
<td>start_loss</td>
<td>START_LOST</td>
<td>null</td>
</tr>
<tr>
<td>exon_deleted</td>
<td>EXON_DELETED</td>
<td>null</td>
</tr>
<tr>
<td>non_synonymous_start</td>
<td>NON_SYNONYMOUS_START</td>
<td>null</td>
</tr>
<tr>
<td>transcript_codon_change</td>
<td>null</td>
<td>initiator_codon_variant</td>
</tr>
<tr>
<td>chrom_large_del</td>
<td>CHROMOSOME_LARGE_DELETION</td>
<td>null</td>
</tr>
<tr>
<td>rare_amino_acid</td>
<td>RARE_AMINO_ACID</td>
<td>null</td>
</tr>
<tr>
<td>non_syn_coding</td>
<td>NON_SYNONYMOUS_CODING</td>
<td>missense_variant</td>
</tr>
</tbody>
</table>

5.3. Annotation with snpEff or VEP
<table>
<thead>
<tr>
<th>GEMINI terms</th>
<th>snpEff terms</th>
<th>VEP terms (uses SO by default)</th>
</tr>
</thead>
<tbody>
<tr>
<td>inframe_codon_gain</td>
<td>CODON_INSERTION</td>
<td>inframe_insertion</td>
</tr>
<tr>
<td>inframe_codon_loss</td>
<td>CODON_DELETION</td>
<td>inframe_deletion</td>
</tr>
<tr>
<td>inframe_codon_change</td>
<td>CODON_CHANGE</td>
<td>null</td>
</tr>
<tr>
<td>codon_change_del</td>
<td>CODON_CHANGE_PLUS_CODON_DELETION</td>
<td>null</td>
</tr>
<tr>
<td>codon_change_ins</td>
<td>CODON_CHANGE_PLUS_CODON_INSERTION</td>
<td>null</td>
</tr>
<tr>
<td>UTR_5_del</td>
<td>UTR_5_DELETED</td>
<td>null</td>
</tr>
<tr>
<td>UTR_3_del</td>
<td>UTR_3_DELETED</td>
<td>null</td>
</tr>
<tr>
<td>splice_region</td>
<td>SPLICE_SITE_REGION</td>
<td>splice_region_variant</td>
</tr>
<tr>
<td>mature_miRNA</td>
<td>null</td>
<td>mature_miRNA_variant</td>
</tr>
<tr>
<td>regulatory_region</td>
<td>null</td>
<td>regulatory_region_variant</td>
</tr>
<tr>
<td>TF_binding_site</td>
<td>null</td>
<td>TF_binding_site_variant</td>
</tr>
<tr>
<td>regulatory_region_ablation</td>
<td>null</td>
<td>regulatory_region_ablation</td>
</tr>
<tr>
<td>regulatory_region_amplification</td>
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</table>

Note: “null” refers to the absence of the corresponding term in the alternate database

**SO impact definitions in GEMINI**

The below table shows the Sequence Ontology (SO) term mappings for the GEMINI impacts, which is otherwise contained in the impact_so column of the variants/variant_impacts table of the GEMINI database. The last column shows the severity terms defined in GEMINI for these impacts.
### GEMINI terms (column: impact)

<table>
<thead>
<tr>
<th>GEMINI terms (column: impact)</th>
<th>Sequence Ontology terms (column: impact_so)</th>
<th>Impact severity</th>
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<tbody>
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Continued on next page
Table 5.2 – continued from previous page

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</tr>
<tr>
<td>feature truncation</td>
<td>feature_truncation</td>
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</table>

Preprocessing and Loading a VCF file into GEMINI

Step 1. split, left-align, and trim variants

Variants with multiple alternate alleles will not be handled correctly by gemini (or by the tools used to annotate the variants). As projects get more samples it is likely that a non-negligible percentage of site will have multiple alternate alleles.

In addition, variants that are not left-aligned and trimmed can be incorrectly (or not) annotated.

To reduce the number of false negatives, we strongly recommend that gemini users split, left-align, and trim their variants. The tools we recommend for this are either vt:

```
vt decompose -s $VCF | vt normalize -r $REFERENCE > $NEW_VCF
```

gemini uses the allele depths from the AD tag. In order for vt to decompose correctly, users will have to change the #INFO field for AD in the header from Number=. to Number=R.

Then the $NEW_VCF can be annotated with snpEff or VEP.

Step 2. Annotate with snpEff or VEP

Note: Annotate your VCF with SnpEff/VEP, prior to loading it into GEMINI, otherwise the gene/transcript features would be set to None.

GEMINI supports gene/transcript level annotations (we do not use pre-computed values here) from snpEff and VEP and hence we suggest that you first annotate your VCF with either of these tools, prior to loading it into GEMINI. The related database columns would be populated, which would otherwise be set to None if an unannotated VCF file is loaded into GEMINI.

Note: Choose the annotator as per your requirement! Some gene/transcript annotations are available with only one tool (e.g. Polyphen/Sift with VEP). As such these values would be set to None, if an alternate annotator is used during the load step.

Instructions for installing and running these tools can be found in the following section:

Annotation with snpEff or VEP

The basics

Before we can use GEMINI to explore genetic variation, we must first load our VCF file into the GEMINI database framework. We expect you to have first annotated the functional consequence of each variant in your VCF using either VEP or snpEff (Note that v3.0+ of snpEff is required to track the amino acid length of each impacted transcript). Logically, the loading step is done with the gemini load command. Below are two examples based on a VCF file
that we creatively name my.vcf. The first example assumes that the VCF has been pre-annotated with VEP and the second assumes snpEff.

```
# VEP-annotated VCF
$ gemini load -v my.vcf -t VEP my.db

# snpEff-annotated VCF
$ gemini load -v my.vcf -t snpEff my.db
```

As each variant is loaded into the GEMINI database framework, it is being compared against several annotation files that come installed with the software. We have developed an annotation framework that leverages tabix, bedtools, and pybedtools to make things easy and fairly performant. The idea is that, by augmenting VCF files with many informative annotations, and converting the information into a sqlite database framework, GEMINI provides a flexible database-driven API for data exploration, visualization, population genomics and medical genomics. We feel that this ability to integrate variation with the growing wealth of genome annotations is the most compelling aspect of GEMINI. Combining this with the ability to explore data with SQL using a database design that can scale to 1000s of individuals (genotypes too!) makes for a nice, standardized data exploration system.

Many variant callers set filter flags in the VCF file to flag possible problem variants. By default GEMINI will leave these variants in the database during loading but they can be filtered out during the loading step by passing the --passonly flag to load.

You can create a smaller, faster database if you dont need the genotype likelihoods format each sample by passing the --skip-pls flag.

### Using multiple CPUs for loading

Now, the loading step is very computationally intensive and thus can be very slow with just a single core. However, if you have more CPUs in your arsenal, you can specify more cores. This provides a roughly linear increase in speed as a function of the number of cores. On our local machine, we are able to load a VCF file derived from the exomes of 60 samples in about 10 minutes. With a single core, it takes a few hours.

```
$ gemini load -v my.vcf -t snpEff --cores 20 my.db
```

### Using LSF, SGE, SLURM and Torque schedulers

One can load VCF files into GEMINI in parallel using many cores on LSF, SGE, SLURM or Torque clusters. One must simply specify the type of job scheduler your cluster uses and the queue name to which your jobs should be submitted.

For example, let’s assume you use LSF and a queue named preempt_everyone. Here is all you need to do:

```
$ gemini load -v my.vcf \
  -t snpEff \
  --cores 50 \n  --queue preempt_everyone \n  --scheduler lsf \n  my.db
```
Describing samples with a PED file

GEMINI also accepts PED files in order to establish the familial relationships and phenotypic information of the samples in the VCF file.

```
$ gemini load -v my.vcf -p my.ped -t snpEff my.db
```

The PED file format is documented here: [PED](http://example.com). An example PED file looks like this:

```
1 M10475 -9 -9 1 1
1 M10478 M10475 M10500 2 2
1 M10500 -9 -9 2 2
1 M128215 M10475 M10500 1 1
```

The columns are `family_id`, `name`, `paternal_id`, `maternal_id`, `sex` and `phenotype`. For GEMINI, you can use either tabs or spaces, but not both.

You can also provide a PED file with a heading starting with `#`, and include extra fields, like this:

```
#family_id name paternal_id maternal_id sex phenotype hair_color
1 M10475 -9 -9 1 1 brown
1 M10478 M10475 M10500 2 2 brown
1 M10500 -9 -9 2 2 black
1 M128215 M10475 M10500 1 1 blue
```

This will add the extra columns to the `samples` table and allow for you to use those extra columns during queries.

Missing values for `family_id`, `paternal_id` and `maternal_id` can be specified by any of `0`, `−9` or `None` and GEMINI will translate them to `0` in the database.

Load GERP base pair conservation scores

GERP scores at base pair resolution are loaded by default (Note: This requires a prior install of the data file by running `gemini update --dataonly --extra gerp_bp`). However, if not required, one may optionally skip the load process (to save on the loading time) with the `--skip-gerp-bp` option.

```
$ gemini load -v my.vcf --skip-gerp-bp -t snpEff my.db
```

Load CADD scores for deleterious variants

CADD scores ([http://cadd.gs.washington.edu/](http://cadd.gs.washington.edu/)) are loaded by default in GEMINI (Note: This requires a prior install of the data file by running `gemini update --dataonly --extra cadd_score`). However, one may optionally skip the load process using the `--skip-cadd` option.

```
$ gemini load -v my.vcf --skip-cadd my.db
```
Updating the samples table in a database

If, after loading a database, you find more information about your samples or want to add a column to the samples table to query on, you can reload the samples table with a new PED file with `gemini amend --sample`. This is also useful if you forgot to load a PED file when initially loading your database. This file must have the standard first six columns of a PED file, but after that other columns can be added. The top of the PED file also must have a header starting with # which names all of the columns if there are more than the standard six PED file columns:

```
gemini amend --sample your_new_ped_file your.db
```

Loading VCFs without genotypes.

To do.

Querying the GEMINI database

The real power in the GEMINI framework lies in the fact that all of your genetic variants have been stored in a convenient database in the context of a wealth of genome annotations that facilitate variant interpretation. The expressive power of SQL allows one to pose intricate questions of one’s variation data.

**Note:** If you are unfamiliar with SQL, sqlzoo has a decent online tutorial describing the basics. Really all you need to learn is the SELECT statement, and the examples below will give you a flavor of how to compose base SQL queries against the GEMINI framework.

Basic queries

GEMINI has a specific tool for querying a gemini database that has been loaded using the `gemini load` command. That’s right, the tool is called `gemini query`. Below are a few basic queries that give you a sense of how to interact with the gemini database using the `query` tool.

1. Extract all transitions with a call rate > 95%

```
gemini query -q "select * from variants \n    where sub_type = 'ts' \n    and call_rate >= 0.95" my.db
```

2. Extract all loss-of-function variants with an alternate allele frequency < 1%:

```
gemini query -q "select * from variants \n    where is_lof = 1 \n    and aaf >= 0.01" my.db
```

3. Extract the nucleotide diversity for each variant:

```
gemini query -q "select chrom, start, end, pi from variants" my.db
```

4. Combine GEMINI with bedtools to compute nucleotide diversity estimates across 100kb windows:

```
gemini query -q "select chrom, start, end, pi from variants \n    order by chrom, start, end" my.db | \n    bedtools map -a hgl9.windows.bed -b -c 4 -o mean
```
Selecting sample genotypes

The above examples illustrate *ad hoc* queries that do not request or filter upon the genotypes of individual samples. Since GEMINI stores the genotype information for each variant in compressed arrays that are stored as BLOBs in the database, standard SQL queries cannot directly access individual genotypes. However, we have enhanced the SQL syntax to support such queries with C “struct-like” access. For example, to retrieve the alleles for a given sample’s (in this case, sample 1094PC0009), one would add `gts.1094PC0009` to the select statement.

Here is an example of selecting the genotype alleles for four different samples (note the examples below use the `test.snpEff.vcf.db` file that is created in the ./test directory when you run the `bash master-test.sh` command as described above):

```
$ gemini query -q "select chrom, start, end, ref, alt, gene, 
gts.1094PC0005, 
gts.1094PC0009, 
gts.1094PC0012, 
gts.1094PC0013 
from variants" test.snpEff.vcf.db
```

```
chr1 30547 30548 T G FAM138A ./. ./. ./. ./.
chr1 30859 30860 G C FAM138A G/G G/G G/G G/G
chr1 30866 30869 CCT C FAM138A CCT/CCT CCT/CCT CCT/C CCT/CCT
chr1 30894 30895 T C FAM138A T/C T/T T/T T/T
chr1 30922 30923 G T FAM138A ./. ./. ./. ./.
chr1 69269 69270 A G OR4F5 ./. ./. G/G G/G
chr1 69427 69428 T G OR4F5 T/T T/T T/T T/T
chr1 69510 69511 A G OR4F5 ./. ./. A/G A/G
chr1 69760 69761 A T OR4F5 A/A A/T A/A A/A
chr1 69870 69871 G A OR4F5 ./. G/G G/G G/G
```

You can also add a header so that you can keep track of who’s who:

```
$ gemini query -q "select chrom, start, end, ref, alt, gene, 
gts.1094PC0005, 
gts.1094PC0009, 
gts.1094PC0012, 
gts.1094PC0013 
from variants" --header test.snpEff.vcf.db
```

```
chrom start end ref alt gene gts.1094PC0005 gts.1094PC0009
chr1 30547 30548 T G FAM138A ./. ./. ./. ./.
chr1 30859 30860 G C FAM138A G/G G/G G/G G/G
chr1 30866 30869 CCT C FAM138A CCT/CCT CCT/CCT CCT/C CCT/CCT
chr1 30894 30895 T C FAM138A T/C T/T T/T T/T
chr1 30922 30923 G T FAM138A ./. ./. ./. ./.
chr1 69269 69270 A G OR4F5 ./. ./. G/G G/G
chr1 69427 69428 T G OR4F5 T/T T/T T/T T/T
chr1 69510 69511 A G OR4F5 ./. ./. A/G A/G
chr1 69760 69761 A T OR4F5 A/A A/T A/A A/A
chr1 69870 69871 G A OR4F5 ./. G/G G/G G/G
```

Let’s now get the genotype and the depth of aligned sequence observed for a sample so that we can assess the confidence in the genotype:

```
$ gemini query -q "select chrom, start, end, ref, alt, gene, 
gts.1094PC0005, 
```
Selecting sample genotypes based on “wildcards”.

The above examples demonstrate how one can select individual sample genotype information by explicitly listing each column and sample that one wishes to see. Obviously, this can become tedious when a project involves hundreds or thousands of samples — if you wanted to see genotype information for the 345 of 1145 affected samples in your study, you would have to type each and every column/sample name out. Brutal.

The “*” wildcard

To get around this, one can bulk-select sample genotype information using “wildcards”. The column and the wildcard must each be surrounded with parentheses and separated by a period. The “*” is a shortcut (wildcard) meaning “all samples”.

Note: The syntax for SELECTing genotype columns using a wildcard is \( (COLUMN).(*) \).

For example, a shortcut to reporting the genotype for all samples (in this case 4) in the study, one could do the following:

```
$ gemini query --header -q "select chrom, start, end, ref, alt, gene, (gts).(*) \ from variants" extended_ped.db
```

Wildcards based on sample attributes

To report the genotypes for solely those samples that are affected (phenotype == 2) with the phenotype in question, one could do the following:

```
```

5.5. Querying the GEMINI database
$ gemini query --header -q "select chrom, start, end, ref, alt, gene, (gts). 
→(phenotype==2) \ 
→from variants" extended_ped.db

One can add multiple wildcard criteria as well:

$ gemini query --header -q "select chrom, start, end, ref, alt, gene, (gts). 
→(phenotype==1 and hair_color=='blue') \ 
→from variants" extended_ped.db

---gt-filter Filtering on genotypes

Now, we often want to focus only on variants where a given sample has a specific genotype (e.g., looking for homozygous variants in family trios). Unfortunately, we cannot directly do this in the SQL query, but the gemini query tool has an option called --gt-filter that allows one to specify filters to apply to the returned rows. The rules followed in the --gt-filter option follow Python syntax.

**Tip:** As you will see from the examples below, appropriate use of the --gt-filter option will allow you to compose queries that return variants meeting inheritance patterns that are relevant to the disease model of interest in your study.

As an example, let’s only return rows where sample 1094PC0012 is heterozygous. In order to do this, we apply a filter to the gt_types columns for this individual:

$ gemini query -q "select chrom, start, end, ref, alt, gene,
gts.1094PC0005, \
gts.1094PC0009, \
gts.1094PC0012, \
gts.1094PC0013 \
from variants" \
--gt-filter "gt_types.1094PC0012 == HET" \
--header \
test.snpEff.vcf.db

chrom start end ref alt gene gts.1094PC0005 gts.1094PC0009 gts.1094PC0012 gts.1094PC0013

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Now let’s be a bit less restrictive and return variants where either sample 1094PC0012 is heterozygous or sample 1094PC0005 is homozygous for the reference allele:

```
$ gemini query -q "select chrom, start, end, ref, alt, gene,
    gts.1094PC0005, \
    gts.1094PC0009, \
    gts.1094PC0012, \
    gts.1094PC0013 
from variants" \
--gt-filter "gt_types.1094PC0012 == HET or \n    gt_types.1094PC0005 == HOM_REF" \
--header \
test.snpEff.vcf.db
```

Eh, I changed my mind, let’s restrict the above to those variants where sample 1094PC0012 must also be heterozygous:

```
$ gemini query -q "select chrom, start, end, ref, alt, gene,
    gts.1094PC0005, \
    gts.1094PC0009, \
    gts.1094PC0012, \
    gts.1094PC0013 
from variants" \
--gt-filter "(gt_types.1094PC0012 == HET or \n    gt_types.1094PC0005 == HOM_REF) \
    and \
    (gt_types.1094PC0013 == HET)"
--header \
test.snpEff.vcf.db
```

---

**--gt-filter** Wildcard filtering on genotype columns.

Many times, we want to be able to apply the same rule to multiple samples without having to enter the rule over and over again for each sample. For example, let’s imagining there are 100 samples in your study and you only want to report variants where every sample has an observed alignment depth of at least 20 reads. Traditionally, one would have enter each of the 100 samples from the command line as follows:

```
$ gemini query -q "select chrom, start, end, ref, alt, gene from variants" \
    --gt-filter *gt_depths.sample1 >= 20 and \
    gt_depths.sample2 >= 20 and \
    gt_depths.sample3 >= 20 and \
    ...
```

---

## 5.5. Querying the GEMINI database

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The basics.

Obviously, this is deeply painful. Thankfully, there an option allowing the use of “wildcards” to prevent this.

Note: The syntax of the wildcard --gt-filters is (COLUMN).(SAMPLE_WILDCARD). (SAMPLE_WILDCARD_RULE). (RULE_ENFORCEMENT).

For example, using wildcards, the above could be converted to:

```
$ gemini query -q "select chrom, start, end, ref, alt, gene from variants" \
    --gt-filter "(gt_depths).(*).(`=20).(`all)" \ 
    extended_ped.db
```

Obviously, this makes things much simpler.

The all operator

One can also apply wildcards that select samples based on the values in specific columns in the samples table. For example, let’s imagine we wanted to require that variants are returned only in cases where ALL the affected individuals in the study (i.e., the phenotype column in the samples table is equal to 2) have non-reference genotypes. We could do the following:

```
$ gemini query --header \
    -q "SELECT chrom, start, end, ref, alt, gene, (gts).(phenotype==2) \ 
        FROM variants" \
    --gt-filter "(gt_types).(phenotype==2).(!=HOM_REF).(all)" \ 
    extended_ped.db
```

Or perhaps we wanted to be more restrictive. We could also enforce that the affected individuals also had at least 20 aligned reads at such variant sites:

```
$ gemini query --header \
    -q "SELECT chrom, start, end, ref, alt, gene, (gts).(phenotype==2), (gt_depths).(phenotype==2) \ 
        FROM variants" \
    --gt-filter "(gt_types).(phenotype==2).(!=HOM_REF).(all) \ 
        and (gt_depths).(phenotype==2).(`>=20).(`all)" \ 
    extended_ped.db
```

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The any operator

The examples provided thus far have demonstrated how to enforce that ALL of the samples meeting specific criteria meet the same genotype column restrictions. However, clearly there are cases where one would want to be less restrictive. In order to accommodate such queries, there are three other enforcement operators allowed: any, none, and count.

For example, perhaps we want to relax the above query a bit and only require that at least one (i.e., any) of the affected samples has depth > 20:

```
$ gemini query --header \\n- q "SELECT chrom, start, end, ref, alt, gene, (gts).(phenotype==2), (gt_depths).(phenotype==2) \n FROM variants" \\n--gt-filter "(gt_types).(phenotype==2).(!=HOM_REF).(all) \n and \n (gt_depths).(phenotype==2).(>=20).(any)" \\nextended_ped.db
```

The none operator

Or we enforce that none of the affected samples have depth less than 10:

```
$ gemini query --header \\n- q "SELECT chrom, start, end, ref, alt, gene, (gts).(phenotype==2), (gt_depths).(phenotype==2) \n FROM variants" \\n--gt-filter "(gt_types).(phenotype==2).(!=HOM_REF).(all) \n and \n (gt_depths).(phenotype==2).(<10).(none)" \\nextended_ped.db
```

The count operator

Finally, we could enforce that at most 2 of all the samples in the study have depths < 10:

```
$ gemini query --header \\n- q "SELECT chrom, start, end, ref, alt, gene, (gts).(phenotype==2), (gt_depths).(phenotype==2) \n FROM variants" \\n--gt-filter "(gt_types).(phenotype==2).(!=HOM_REF).(all) \n and \n (gt_depths).(phenotype==2).(<10).(count <= 2)" \\nextended_ped.db
```

Note: The count operator allows the following comparisons: >, >=, <, <=, ==, <>, and !=.
**Example scenario.**

One usage of the wildcard functionality is screening for variants that are present in affected individuals yet absent from unaffected individuals (this is obviously an idealized scenario).

```
$ gemini query --header \
- q "SELECT chrom, start, end, ref, alt, gene, (gts).(*) \n  FROM variants" \
- gt-filter "((gt_types).(phenotype==1).(==HOM_REF).(all) \n  and \n  (gt_depths).(phenotype==2).(==HOM_REF).(none)) \n  or gt.NA12878 == HET" \
extended_ped.db
```

**More complexity.**

One can also combine wildcard filters with “basic” genotype filters to create more complex logic.

```
$ gemini query --header \
- q "SELECT chrom, start, end, ref, alt, gene, (gts).(*) \n  FROM variants" \
- gt-filter "((gt_types).(phenotype==1).(==HOM_REF).(all) \n  and \n  (gt_depths).(phenotype==2).(==HOM_REF).(none)) \n  or gt.NA12878 == HET" \
extended_ped.db
```

**Other details.**

The system is fairly flexible in that it allows one to wildcard-select samples based on custom columns that have been added to the `samples` table based upon a custom PED file. For example, let’s imaging our custom PED file had an extra column defining the hair color of each sample. We could use that to restrict interesting variants to those where samples with blue hair were heterozygous:

```
$ gemini query -q "SELECT chrom, start, end, ref, alt, gene FROM variants" \
  -gt-filter "((gt_types).(hair_color='blue')==(HET).(all)" \
extended_ped.db
```

Or possibly, you want to stratify based on sub-population:

```
$ gemini query -q "SELECT chrom, start, end, ref, alt, gene FROM variants" \
  -gt-filter "((gt_types).(population='CEU')==(HET).(all) \n  and \n  (gt_types).(population='YRI')==(HOM_ALT).(any)" \
extended_ped.db
```

One can also base the wildcard on multiple criteria (in this case, brown hair and affected):

```
$ gemini query -q "SELECT chrom, start, end, ref, alt, gene FROM variants" \
  -gt-filter "((gt_types).(hair_color=='brown' and phenotype==2)!(==HET).(all)" \
extended_ped.db
```

There is a special case when we want to select samples based on their genotypes. For example, to show only sites where HET samples have a genotype quality above 20.
In this case, the 2nd value must be one of =HET, =HOM_REF or =HOM_ALT

Lastly, wildcards can, of course, be combined with non-wildcard criteria:

```
$ gemini query -q "SELECT chrom, start, end, gene FROM variants \n   --gt-filter "((gt_types).(ha"r_color=="brown" and phenotype==2).(!="HET).\n   and \n   gt_types.M128215 == HOM_REF" \n   extended_ped.db
```

Hopefully this gives you a sense of what you can do with the “wildcard” genotype filter functionality.

---show-samples Finding out which samples have a variant

While exploring your data you might hit on a set of interesting variants and want to know which of your samples have that variant in them. You can display the samples containing a variant with the --show-sample-variants flag:

```
$ gemini query --header --show-samples -q "select chrom, start, end, ref, alt \n   from variants where is_lof=1 limit 5" test.query.db
```

```
chrom start start ref alt variant_samples het_samples hom_alt_samples
chr1 874815 874816 C CT 1478PC0006B,1478PC0007B,1478PC0010,1478PC0013B,1478PC0022B,1478PC0023B,1478PC0025,1719PC0007,1719PC0009,1719PC0010,1719PC0022 1478PC0006B,1478PC0007B,1478PC0010,1478PC0013B,1478PC0022B,1478PC0023B,1478PC0025,1719PC0007,1719PC0009,1719PC0010,1719PC0013B,1719PC0022B,1719PC0023B,1478PC0007,1719PC0009,1719PC0010 1478PC0006B,1478PC0007B,1478PC0010,1478PC0013B,1478PC0022B,1478PC0023B,1478PC0025,1719PC0007,1719PC0009,1719PC0010
chr1 1140811 1140813 TC G 1478PC0011 1478PC0011
chr1 1219381 1219382 C G 1719PC0012 1719PC0012
chr1 1221487 1221490 CAA C 1478PC0004 1478PC0004
```

variant_samples is a list of all of the samples with a variant, het_samples is the subset of those heterozygous for the variant and hom_alt_samples is the subset homozygous for the variant.

---show-samples --format sampledetail Provide a flattened view of samples

If you’d like to be able to export variants plus associated sample metadata into a downstream tool like R or pandas for additional exploration, adding the --format sampledetail command flattens all found samples and attached metadata information:

```
$ gemini query --header --format sampledetail --show-samples \n   -q "select chrom, start, ref from variants where is_lof=1 limit 1" test.query.db
```

```
chrom start ref family_id name paternal_id maternal_id sex phenotype
chr1 30547 T 0 1478PC0006B 1478PC0007B 1478PC0010 -9 -9
chr1 30547 T 0 1719PC0007 1719PC0009 1719PC0010 1478PC0004 1478PC0004
```

The denormalized results contain a row for each sample associated with a variant, along with information from the sample table. This provides a way to join sample information with a variant query.

5.5. Querying the GEMINI database
--show-families Finding out which families have a variant

This works exactly like --show-samples except lists all of the families with a variant instead of the individual samples.

--region Restrict a query to a specified region

If you are only interested in a specific region, you can restrict queries to that region using the --region tool.

```bash
$ gemini query --region chr1:30859-30900 -q "select chrom, start, end, ref, alt \ from variants" test1.snpeff.db
chr1 30859 30860 G C
```

--sample-filter Restrict a query to specified samples

The --sample-filter option allows you to select samples that a variant must be in by doing a SQL query on the samples table. For example if you wanted to show the set of variants that appear in all samples with a phenotype status of 2, you could do that query with:

```bash
$ gemini query --sample-filter "phenotype=2" -q "select gts, gt_types from variants" test1.snpeff.db
T/T,T/T,T/C,T/T,T/T,T/T,C/C 0,0,1,0,0,0,0,3 1_kid,3_kid 1_kid
T/T,T/T,T/C,T/T,T/T,T/T,C/C 0,0,1,0,0,1,0,1 1_kid,2_kid,3_kid
T/T,T/T,T/T,T/T,T/T,T/T,T/C 0,0,0,0,0,0,0,1 3_kid 3_kid
```

By default --sample-filter will show the variant if at least one sample contains the variant. You can change this behavior by using the --in option along with --sample-filter. --in all will return a variant if all samples matching the query have the variant. in none will return a variant if the variant does not appear in any of the matching samples. --in only will return a variant if the variant is only in the matching samples and not in any of the non-matching samples. --in only all will show all of the variant which are in all of the matching samples and not in any of the non-matching samples.

The --family-wise flag applies the --sample-filter and --in behavior on a family-wise basis. For example to show all variants that are only in samples with a phenotype status of 2 in at least one family:

```bash
$ gemini query --family-wise --in only all --sample-filter "phenotype=2" -q "select gts, gt_types from variants" test1.snpeff.db
T/T,T/T,T/C,T/T,T/T,T/T,C/C 0,0,1,0,0,0,0,3 1_kid,3_kid 1_kid
T/T,T/T,T/C,T/T,T/T,T/T,C/C 0,0,1,0,0,1,0,1 1_kid,2_kid,3_kid
T/T,T/T,T/T,T/T,T/T,T/T,T/C 0,0,0,0,0,0,0,1 3_kid 3_kid
```

You can also specify that a variant passes this filter in multiple families with the --min-kindreds option. So if you want to do the same query above, but restrict it such that at least three families have to pass the filter:

```bash
$ gemini query --family-wise --in only all --sample-filter "phenotype=2" -q "select gts, gt_types from variants" test1.snpeff.db
T/T,T/T,T/C,T/T,T/T,T/T,C/C 0,0,1,0,0,0,0,3 1_kid,3_kid 1_kid
T/T,T/T,T/C,T/T,T/T,T/T,C/C 0,0,1,0,0,1,0,1 1_kid,2_kid,3_kid
T/T,T/T,T/T,T/T,T/T,T/T,T/C 0,0,0,0,0,0,0,1 3_kid 3_kid
```

If the PED file you loaded has extra fields in it, those will also work with the --sample-filter option. For example if you had a hair_color extended field, you could query on that as well as phenotype:
---sample-delim Changing the sample list delimiter

One can modify the default comma delimiter used by the --show-samples option through the use of the --sample-delim option. For example, to use a semi-colon instead of a comma, one would do the following:

```bash
$ gemini query --header --show-samples --sample-delim ";" -q "select chrom, start, end, ref, alt from variants where is_lof=1 limit 5" test.query.db
```

---format Reporting query output in an alternate format.

The results of GEMINI queries can automatically be formatted for use with other programs using the --format command. Supported alternative formats are JSON and TPED (Transposed PED) format.

Reporting query output in JSON format may enable HTML/Javascript apps to query GEMINI and retrieve the output in a format that is amenable to web development protocols.

Here is a basic query:

```bash
$ gemini query -q "select chrom, start, end from variants" my.db | head
```

To report in JSON format, use the --format json option. For example:

```bash
$ gemini query --format json -q "select chrom, start, end from variants" my.db | head
```
If you would to use tools such as PLINK that use the PED format, you can dump out a set of variants matching any query in TPED (Transposed PED) format by adding the `--tped` flag to your query:

```
$ gemini query --format tped -q "select * from variants where chrom=10" test4.snpeff.db
10 rs10794716 0 1142207 C/C C/C C/C C/C
10 rs142685947 0 48003991 T/T C/T C/T C/C
10 rs2842123 0 52004314 ./. ./C C/C
16 rs4935178 0 52497528 ./C C/C ./C
10 rs73373169 0 126678091 G/G G/G G/A
10 rs2265637 0 135210790 T/C C/C T/T
10 rs6537611 0 135336655 ./A/A ./A/A
10 rs3747881 0 135369531 T/T T/C T/C T/T
```

You can pass `--header` to get a header to see which samples have which variant. To use the TPED format you also need to generate a corresponing TFAM file from your data as well, which you can get from the GEMINI dump tool:

```
$ gemini dump --tfam test4.snpeff.db > obs
None M10475 None None None None
None M10478 None None None None
None M10500 None None None None
None M128215 None None None None
```

**--carrier-summary-by-phenotype** Summarize carrier status

For prioritizing variants sometimes it is useful to have summary counts of the carrier status for all samples with a variant stratified across a phenotype. `--carrier-summary-by-phenotype` takes a column in the samples table that you want to summarize the carrier status of and adds a set of counts of carrier/non-carrier status for each phenotype in the given column. For example, to get a summary of how a set of variants segregate with affected status:

```
$ gemini query --show-samples --carrier-summary-by-phenotype affected --header -q "select * from variants" extended_ped_test.db
chrom start ref alt gt_types variant_samples het_samples hom_affected_samples unaffected_samples noncarrier_samples
chr10 1142207 T C 3,3,3,3 M10475,M10478,M10500,M128215 2 2 0 0 0
chr10 48003991 C T 3,1,1,0 M10475,M10478,M10500 2 0 2 0
chr10 52004314 T C 2,2,3,3 M10500,M128215 2 0 2 0
chr10 52497528 G C 2,3,3,2 M10478,M10500 0 2 2 0
chr16 72057434 C T 1,0,0,0 M10475 M10478 1 0
chr10 126678091 G A 0,0,0,1 M128215 1 0
chr10 135210790 T C 0,3,3,0 M10478,M10500 0 2 2 0
chr10 135336655 G A 2,3,2,3 M10478,M128215 0 2 2 0
chr10 135369531 T C 0,1,1,0 M10478,M10500 0 2 2 0
```
Or if you have another phenotypic feature you are interested in summarizing, like hair color:

```sql
$ gemini query --show-samples --carrier-summary-by-phenotype hair_color --header -q "select chrom, start, ref, alt, gt_types from variants" extended_ped.db
```

```
chrom start ref alt gt_types variant_samples het_samples hom_ noncarrier purple_carrier blue_carrier brown_ carrier brown_ noncarrier purple_noncarrier unknown
chr10 1142207 T C 3,3,3 M10475,M10478,M10500,M128215 1 2 1 0 0 0 0
chr10 48003991 C T 3,1,1,0 M10475,M10478,M10500 M10478,M10500 1 1 1 0 0 0 0
chr10 52004314 T C 2,2,3,3 M10500,M128215 1 0 1 0 0 0 2
chr10 52497528 G C 2,3,3,2 M10478,M10500 M10478,M10500 1 1 1 0 0 0 2
chr16 72057434 C T 1,0,0,0 M10475 M10475 0 1 0 1 0
chr10 126678091 G A 0,0,0,1 M128215 M128215 1 0 0 0 2
chr10 135210790 T C 0,3,3,0 M10478,M10500 M10478,M10500 0 1 0 0 0
chr10 135336655 G A 2,3,2,3 M10478,M128215 M10478, 0 1 0 0 0 2
chr10 135369531 T C 0,1,1,0 M10478,M10500 M10478,M10500 0 1 0 0 0
```

### Querying the gene tables

The gene tables viz. `gene_detailed` table and the `gene_summary` table have been built on version 73 of the ensembl genes. The column specifications are available at [The GEMINI database schema](#). These tables contain gene specific information e.g. gene synonyms, RVIS percentile scores(Petrovski et.al 2013), strand specifications, cancer gene census etc. While the former is more detailed, the later lacks transcript wise information and summarizes some aspects of the former. For e.g. while the gene_detailed table lists all transcripts of a gene with their start and end co-ordinates, the gene_summary table reports only the minimum start and maximum end co-ordinates of the gene transcripts. The `chrom`, `gene` and the `transcript` columns of the gene tables may be used to join on the variants and the variant_impacts tables.

### Using the `gene_detailed` & `gene_summary` tables

#### Query the `gene_detailed` table with a join on variants table:

E.g. Get additional transcript info for the most severe impact transcript e.g. transcript status, transcript start,end and the protein length

```sql
$ gemini query --header -q "select v.variant_id, v.gene, v.impact, g.transcript_status, g.transcript, g.transcript_start, g.transcript_end, g.protein_length, FROM variants v, gene_detailed g WHERE v.chrom = g.chrom AND v.gene = g.gene AND
```

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v.transcript = g.transcript AND \n v.impact_severity='HIGH'" test.query.db

<table>
<thead>
<tr>
<th>variant_id</th>
<th>gene</th>
<th>impact</th>
<th>transcript_status</th>
<th>transcript</th>
<th>protein_length</th>
</tr>
</thead>
<tbody>
<tr>
<td>46</td>
<td>SAMD11</td>
<td>frame_shift</td>
<td>KNOWN</td>
<td>ENST00000342066</td>
<td>861118 879955 681</td>
</tr>
<tr>
<td>578</td>
<td>TNFRSF18</td>
<td>frame_shift</td>
<td>PUTATIVE</td>
<td>ENST00000486728</td>
<td>1139224</td>
</tr>
<tr>
<td>733</td>
<td>SCNN1D</td>
<td>stop_gain</td>
<td>NOVEL</td>
<td>ENST00000470022</td>
<td>1217305 1221548 138</td>
</tr>
</tbody>
</table>

### Query the gene_detailed table with a join on the variant_impacts table:

E.g. Get the transcript status for all transcripts of the SCNN1D gene where impact severity is not ‘LOW’.

```bash
$ gemini query --header -q "select v.gene, g.transcript_status, g.transcript, v.impact from variant_impacts v, gene_detailed g \n WHERE v.transcript = g.transcript AND \n v.gene = g.gene AND \n v.gene = 'SCNN1D' \n v.impact_severity!='LOW'" test.query.db
```

<table>
<thead>
<tr>
<th>gene</th>
<th>transcript_status</th>
<th>transcript</th>
<th>impact</th>
</tr>
</thead>
<tbody>
<tr>
<td>SCNN1D</td>
<td>NOVEL</td>
<td>ENST00000470022</td>
<td>non_syn_coding</td>
</tr>
<tr>
<td>SCNN1D</td>
<td>NOVEL</td>
<td>ENST00000470022</td>
<td>frame_shift</td>
</tr>
<tr>
<td>SCNN1D</td>
<td>KNOWN</td>
<td>ENST00000325425</td>
<td>frame_shift</td>
</tr>
<tr>
<td>SCNN1D</td>
<td>KNOWN</td>
<td>ENST00000379116</td>
<td>non_syn_coding</td>
</tr>
<tr>
<td>SCNN1D</td>
<td>KNOWN</td>
<td>ENST00000338555</td>
<td>non_syn_coding</td>
</tr>
<tr>
<td>SCNN1D</td>
<td>KNOWN</td>
<td>ENST00000400928</td>
<td>non_syn_coding</td>
</tr>
</tbody>
</table>

### Query the gene_summary table with a join on the variants table:

E.g. Get the synonym/alternate names, RVIS percentile scores and the min-max start-end of transcripts for genes that have a severely affected transcript of a ‘HIGH’ order.

```bash
$ gemini query --header -q "%select v.chrom, v.gene, g.transcript_min_start, g.transcript_max_end, g.synonym, g.rvis_pct, v.impact from variants v, gene_summary g \n WHERE v.chrom = g.chrom AND \n v.gene = g.gene AND \n v.impact_severity='HIGH'" test.query.db
```

<table>
<thead>
<tr>
<th>chrom</th>
<th>gene</th>
<th>transcript_min_start</th>
<th>transcript_max_end</th>
<th>synonym</th>
<th>rvis_pct</th>
<th>impact</th>
</tr>
</thead>
<tbody>
<tr>
<td>chr1</td>
<td>SAMD11</td>
<td>860260 879955</td>
<td>MGC45873</td>
<td>None</td>
<td>frame_shift</td>
<td></td>
</tr>
<tr>
<td>chr1</td>
<td>TNFRSF18</td>
<td>1138888 1142071</td>
<td>AITR,CD357,GITR</td>
<td>None</td>
<td>frame_shift</td>
<td></td>
</tr>
<tr>
<td>chr1</td>
<td>SCNN1D</td>
<td>1215816 1227409</td>
<td>ENaCdelta,dNaCh</td>
<td>96.77990092</td>
<td>stop_gain</td>
<td></td>
</tr>
</tbody>
</table>

### Query the gene_summary table with a join on the variant_impacts table:

E.g. Get all variants of a gene, the affected transcripts and impacts, where a mammalian phenotype ID is available for the mouse phenotype.
Restrict analysis to transcripts with a valid CCDS_ID

Since the current available transcript sets are more than one (e.g. RefSeq, ENSEMBL and UCSC) we support information (e.g pathways tool) for the ENSEMBL transcripts but provide a mapping of these transcripts to the consensus set agreed upon by all the above three mentioned groups viz. transcripts having a valid CCDS_ID. Here we show, how we can return variants and their impacts for only these restricted set of transcripts using the gene_detailed table.

What if I don’t see my gene in the database?

Most genes are popular by their common names while the representation of gene names in the GEMINI database is mostly HGNC. For e.g ARTEMIS would be DCLRE1C in the GEMINI database. As such one may miss out on variants if looking for specific genes by their common names. While, joining the main tables with the gene tables for synonym information would be useful (as shown in the previous examples), the gene tables may also serve as a quick look up for alternate names of a gene, which could then be looked up in the database.
Built-in analysis tools

As of version 0.14, the tools have a standardized output that is different from previous versions. Requested `--columns` will come first followed by a standard set of columns:

- `variant_id` - unique id from the database
- `family_id` - family id for this row
- `family_members` - which family members were tested
- `family_genotypes` - genotypes of this family
- `samples` - samples contributing to this row appearing in the results
- `family_count` - number of families with this effect

Other tools such as `mendel_errors` additional columns at the end.

When queries are limited to variants in genes, the output will be in `gene, chrom` order as opposed to the usual `gene, position` order.

**Note:** As of version 0.16.0, the inheritance tools (autosomal_dominant, autosomal_recessive, comp_het, mendel_errors, de_novo) are now more strict by default.

A `--lenient` flag allows, e.g. allows some samples to be of unknown phenotype or to not have both parents of known phenotype.

The `--allow-unaffected` flag will result in reporting variants where unaffected samples share the same variants. The default will only report variants that are unique to affected samples.

The design decisions for this change are described here: https://github.com/arq5x/gemini/issues/388 a visual representation is here: https://github.com/arq5x/gemini/blob/master/inheritance.ipynb

**Note:** Candidate variants reported by the built-in inheritance model tools will appear in order by chromosome, then alphabetically by gene. In other words, they will not be in strict positional order for each chromosome. This is in an effort to group all candidate variants by gene since the gene is typically the atomic unit of interest.
Additionally, all current built-in tools (autosomal_dominant, comp_hets, and autosomal_recessive) only analyze autosomal (non sex chromosome) variants. Analysis of X- and Y-linked phenotypes must be done manually with `gt-filter`.

**common_args: common arguments**

The inheritance tools share a common set of arguments. We will describe them here and refer to them in the corresponding sections:

---

--columns

This flag is followed by a comma-delimited list of columns the user is requestin in the output.

--min-kindreds 1

This is the number of families required to have a variant in the same gene in order for it to be reported. For example, we may only be interested in candidates where at least 4 families have a variant in that gene.

--families

By default, candidate variants are reported for all families in the database. One can restrict the analysis to variants in specific families with the `--families` option. Families should be provided as a comma-separated list.

--filter

By default, each tool will report all variants regardless of their putative functional impact. In order to apply additional constraints on the variants returned, one can use the `--filter` option. Using SQL syntax, conditions applied with the `--filter` options become WHERE clauses in the query issued to the GEMINI database.

--d [0] (depth)

Filter variants that do not have at least this depth for all members in a family. Default is 0.

--min-gq [0]

Filter variants that do not have at least this genotype quality for each sample in a family. Default is 0. Higher values are more stringent.

--allow-unaffected

By default, candidates that also appear in unaffected samples are not reported if this flag is specified, such variants will be reported.

--lenient

Loosen the restrictions on family structure. This will allow, for example, finding compound_hets in unaffected samples.
In order to eliminate less confident genotypes, it is possible to enforce a maximum PL value for each sample. On this scale, lower values indicate more confidence that the called genotype is correct. 10 is a reasonable value. This is applied per-family such that all members of a family must meet this level in order to be reported in the final results.

**comp_hets**: Identifying potential compound heterozygotes

Many autosomal recessive disorders are caused by compound heterozygotes. Unlike canonical recessive sites where the same recessive allele is inherited from both parents at the same site in the gene, compound heterozygotes occur when the individual’s phenotype is caused by two heterozygous recessive alleles at different sites in a particular gene.

We are looking for two (typically loss-of-function (LoF)) heterozygous variants impacting the same gene at different loci. The complicating factor is that this is recessive and as such, we must also require that the consequential alleles at each heterozygous site were inherited on different chromosomes (one from each parent). Where possible, comp_hets will phase by transmission. Once this has been done, the comp_hets tool will provide a report of candidate compound heterozygotes for each sample/gene.

Non-exonic/non-coding analyses: comp_hets excludes intronic/non-coding variants for which impact_severity == ‘LOW’ AND is_exonic == FALSE. Therefore, comp_hets will not retrieve most pairs of variants that are downstream or upstream of a gene or are intronic unless otherwise annotated with medium or high impact_severity.

**Note:** As of version 0.16.0 the comp_het tool will perform family-based phasing by default in order to provide better candidates even in the absence of unphased genotypes. Any candidate that could be one element of a comp_het will also be phaseable as long as the parents and their genotypes are known.

As of version 0.16.1, the –ignore-phasing option is removed and there is no –lenient option.

In 0.16.2, a –pattern-only flag was added to find compound hets by inheritance pattern without regard to affection status. A priority code was also added where variants with priority 1 are much more informative. See docs below for further information.

**Genotype Requirements**

- All affected individuals must be heterozygous at both sites.
- No unaffected can be homozygous alternate at either site.
- Neither parent of an affected sample can be homozygous reference at both sites.
- If any unphased-unaffected is het at both sites, the site will be give lower priority
- No phased-unaffected can be heterozygous at both sites.
  1. –allow-unaffected keeps sites where a phased unaffected shares the het-pair
  2. unphased, unaffected that share the het pair are counted and reported for each candidate pair.
- Remove candidates where an affected from the same family does NOT share the same het pair.
- Sites are automatically phased by transmission when parents are present in order to remove false positive candidates.
  1. If data from one or both parents are unavailable and the child’s data was not phased prior to loading into GEMINI, all comp_het variant pairs will automatically be given at most priority == 2. If there’s only a single parent and both the parent and the affected are HET at both sites, the candidate will have priority 3.
2. `--max_priority x` can be used to set the maximum allowed priority level at which candidate pairs are included in the output.

We prioritize with these rules:

<table>
<thead>
<tr>
<th>mom</th>
<th>dad</th>
<th>kid</th>
<th>phaseable</th>
<th>priority</th>
<th>notes</th>
</tr>
</thead>
<tbody>
<tr>
<td>R-H</td>
<td>H-R</td>
<td>H-H</td>
<td>both</td>
<td>1</td>
<td>both sites phaseable and alts on opposite chroms</td>
</tr>
<tr>
<td>n/a</td>
<td>n/a</td>
<td>H-H</td>
<td>NO</td>
<td>2</td>
<td>singleton (unphaseable) HETs have priority 2.</td>
</tr>
<tr>
<td>R-H</td>
<td>H-H</td>
<td>H-H</td>
<td>one</td>
<td>3</td>
<td>should be a rare occurrence</td>
</tr>
<tr>
<td>H-H</td>
<td>H-H</td>
<td>H-H</td>
<td>NO</td>
<td>3</td>
<td>should be a rare occurrence</td>
</tr>
<tr>
<td>H-H</td>
<td>UNK</td>
<td>H-H</td>
<td>NO</td>
<td>3</td>
<td>missing parent and all hets.</td>
</tr>
<tr>
<td>A-R</td>
<td>H-H</td>
<td>H-H</td>
<td>both</td>
<td>NA</td>
<td>exclude hom-alts from un-affecteds</td>
</tr>
<tr>
<td>R-R</td>
<td>H-H</td>
<td>H-H</td>
<td>both</td>
<td>NA</td>
<td>phaseable, but alts are on the same chroms.</td>
</tr>
</tbody>
</table>

**Note:** candidates of priority == 3 are very unlikely (< 1%) to be causal for a rare Mendelian condition (see: [http://www.ncbi.nlm.nih.gov/pmc/articles/PMC3734130/](http://www.ncbi.nlm.nih.gov/pmc/articles/PMC3734130/)); we report them for completeness, but strongly recommend using priority 1 and 2 only. Priority 2 is useful when there are multiple families, some of which consist of only a single sequenced, affected sample.

**Pattern Only**

To find compound heterozygotes by inheritance pattern only, without regard to affections, the following rules are used (with `--pattern-only`):

- Kid must be HET at both sites.
- Kid must have alts on different chromosomes.
- Neither parent can be HOM_ALT at either site.
- If either parent is phased at both sites and matches the kid, it’s excluded.
- If either parent is HET at both sites, priority is reduced.
- When the above criteria are met, and both parents and kid are phased or parents are HET at different sites, the priority is 1.
- If both parents are not phased, the priority is 2.
- For every parent that’s a het at both sites, the priority is incremented by 1.
- The priority in a family is the minimum found among all kids.

**Note:** Each pair of consecutive lines in the output represent the two variants for a compound heterozygote in a given sample. The third column, `comp_het_id`, tracks the distinct compound heterozygote variant pairs.

Example usage with a subset of columns:

```sh
$ gemini comp_hets my.db --columns "chrom, start, end" test.comp_het_default.2.db
```

<table>
<thead>
<tr>
<th>chrom</th>
<th>start</th>
<th>end</th>
<th>gene</th>
<th>alt</th>
<th>variant_id</th>
<th>family_id</th>
<th>family_members</th>
<th>family_genotypes</th>
<th>samples</th>
<th>family_count</th>
<th>comp_het_id</th>
</tr>
</thead>
<tbody>
<tr>
<td>chr1</td>
<td>17362</td>
<td>17366</td>
<td>WASH7P</td>
<td>T</td>
<td>1</td>
<td>dad_3(dad;unaffected),mom_3(mom;unaffected),child_3(child;affected)</td>
<td>TTCT</td>
<td>T,TTCT</td>
<td>TTCT,TTCT</td>
<td>T child_3 2 1</td>
<td></td>
</tr>
<tr>
<td>chr1</td>
<td>17729</td>
<td>17730</td>
<td>WASH7P</td>
<td>A</td>
<td>2</td>
<td>dad_3(dad;unaffected),mom_3(mom;unaffected),child_3(child;affected)</td>
<td>C</td>
<td>A,C</td>
<td>A,A</td>
<td>C child_3 2 1</td>
<td></td>
</tr>
</tbody>
</table>

**5.7. Built-in analysis tools**
This indicates that samples child_3 and child_4 have a candidate compound heterozygotes in WASH.

the following command would further restrict candidate genes to those genes with a compound heterozygote in at least two families:

```bash
$ gemini comp_hets -d 50
   --columns "chrom, start, end, ref, alt"
   --filter "impact_severity = 'HIGH'"
   --allow-unaffected
   --min-kindreds 2
   my.db
```

Now, this does not require that the family members are necessarily restricted to solely those that are affected. To impose this restriction, we remove the `--allow-unaffected` flag

```bash
$ gemini comp_hets -d 50
   --columns "chrom, start, end, ref, alt"
   --filter "impact_severity = 'HIGH'"
   --min-kindreds 2
   my.db
```

We may also specify the families of interest:

```bash
$ gemini comp_hets --families 1 my.db
$ gemini comp_hets --families 1,7 my.db
```

gene-where

The default selection of genes is by the clause: "is_exonic = 1 or impact_severity != 'LOW'" This can be specified to limit to a different subset, e.g. "gene != "

mendelian_error: Identify non-mendelian transmission.

Note: This tool requires that you identify familial relationships via a PED file when loading your VCF into gemini via:

```
gemini load -v my.vcf -p my.ped my.db
```

We can query for mendelian errors in trios including:

- loss of heterozygosity
- implausible de-novo mutations
- de-novo mutations
- uniparental disomy
Genotype Requirements

• (LOH) kind and one parent are opposite homozygotes; other parent is HET
• (uniparental disomy) parents are opposite homozygotes; kid is homozygote;
• (plausible de novo) kid is het. parents are same homozygotes
• (implausible de novo) kid is homozygoes. parents are same homozygotes and opposite to kid.

If allow --only-affected is used, then the tools will only consider samples that have parents and are affected. The default is to consider any sample with parents.

This tool will report the probability of a mendelian error in the final column that is derived from the genotype likelihoods if they are available.

Example:

```bash
$ gemini mendel_errors --columns "chrom,start,end" test.mendel.db --gt-pl-max 1
```

```
+chrom+ start | end  | variant_id | family_id | family_members | family_genotypes | samples | family_count | violation | violation_prob
|chr1|100670|10671|CEPH1463|NA12889(dad;unknown),NA12890(mom;unknown)|G/G,G/G,G/C|NA12877|1|plausible de novo|0.962
|chr1|28493|28494|CEPH1463|NA12889(dad;unknown),NA12890(mom;unknown)|T/C,T/T,C/C|NA12877|1|loss of heterozygosity|0.660
|chr1|28627|28628|CEPH1463|NA12889(dad;unknown),NA12890(mom;unknown)|C/C,C/C,C/T|NA12877|1|plausible de novo|0.989
|chr1|267558|267560|CEPH1463|NA12889(dad;unknown),NA12890(mom;unknown)|C/C,C/C,CT/C|NA12877|1|plausible de novo|0.928
|chr1|537969|537970|CEPH1463|NA12889(dad;unknown),NA12890(mom;unknown)|G/G,G/G,G/T|NA12877|1|plausible de novo|0.940
|chr1|749688|749689|CEPH1463|NA12889(dad;unknown),NA12890(mom;unknown)|T/T,T/T,G/G|NA12877|1|implausible de novo|0.959
|chr1|788944|788945|CEPH1463|NA12889(dad;unknown),NA12890(mom;unknown)|C/C,G/G,G/G|NA12877|1|uniparental disomy|0.914
|chr1|1004248|1004249|CEPH1463|NA12889(dad;unknown),NA12890(mom;unknown)|G/G,G/G,G/C|NA12877|1|plausible de novo|1.000
```

Where, here, we have required the called genotype to have at most a PL of 1 (lower is more confident). Note that the “violation” column indicates the type of mendelian error and the final column can be used for further filtering, with higher numbers indicating a greater probability of mendelian error. We have found > 0.99 to be a reasonable cutoff.

Arguments are similar to the other tools:

```
positional arguments:
db The name of the database to be queried.
```

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optional arguments:
-h, --help show this help message and exit
--columns STRING A list of columns that you would like returned. Def. = "*"
--filter STRING Restrictions to apply to variants (SQL syntax)
--min-kindreds MIN_KINDREDS The min. number of kindreds that must have a candidate variant in a gene.
--families FAMILIES Restrict analysis to a specific set of 1 or more (comma) separated) families
-d MIN_SAMPLE_DEPTH The minimum aligned sequence depth required for each sample in a family (default = 0)
--gt-pl-max GT_PHRED_LL The maximum phred-scaled genotype likelihood (PL) allowed for each sample in a family.
--allow-unaffected consider candidates that also appear in unaffected samples.

de_novo: Identifying potential de novo mutations.

Note: 1. This tool requires that you identify familial relationships via a PED file when loading your VCF into gemini via:

```
gemini load -v my.vcf -p my.ped my.db
```

Genotype Requirements

- all affecteds must be het
- [affected] all unaffected must be homref or homalt
- at least 1 affected kid must have unaffected parents
- [strict] if an affected has affected parents, it’s not de_novo
- [strict] all affected kids must have unaffected (or no) parents
- [strict] warning if none of the affected samples have parents.

The last 3 items, prefixed with [strict] can be turned off with –lenient

If --allow-unaffected is specified, then the item prefixed [affected] is not required.

Example PED file format for GEMINI

<table>
<thead>
<tr>
<th>Family_ID</th>
<th>Individual_ID</th>
<th>Paternal_ID</th>
<th>Maternal_ID</th>
<th>Sex</th>
<th>Phenotype</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>S173</td>
<td>S238</td>
<td>S239</td>
<td>1</td>
<td>caucasian</td>
</tr>
<tr>
<td>1</td>
<td>S238</td>
<td>-9</td>
<td>-9</td>
<td>1</td>
<td>caucasian</td>
</tr>
<tr>
<td>1</td>
<td>S239</td>
<td>-9</td>
<td>-9</td>
<td>2</td>
<td>caucasian</td>
</tr>
<tr>
<td>2</td>
<td>S193</td>
<td>S230</td>
<td>S231</td>
<td>1</td>
<td>caucasian</td>
</tr>
<tr>
<td>2</td>
<td>S230</td>
<td>-9</td>
<td>-9</td>
<td>1</td>
<td>caucasian</td>
</tr>
<tr>
<td>2</td>
<td>S231</td>
<td>-9</td>
<td>-9</td>
<td>2</td>
<td>caucasian</td>
</tr>
<tr>
<td>3</td>
<td>S242</td>
<td>S243</td>
<td>S244</td>
<td>1</td>
<td>caucasian</td>
</tr>
<tr>
<td>3</td>
<td>S243</td>
<td>-9</td>
<td>-9</td>
<td>1</td>
<td>caucasian</td>
</tr>
<tr>
<td>3</td>
<td>S244</td>
<td>-9</td>
<td>-9</td>
<td>2</td>
<td>caucasian</td>
</tr>
<tr>
<td>4</td>
<td>S253</td>
<td>S254</td>
<td>S255</td>
<td>1</td>
<td>caucasianNEuropean</td>
</tr>
</tbody>
</table>
Assuming you have defined the familial relationships between samples when loading your VCF into GEMINI, one can leverage a built-in tool for identifying de novo (a.k.a spontaneous) mutations that arise in offspring.

**example**

```
$ gemini de_novo --columns "chrom,start,end" test.de_novo.db

chrom  start  end  variant_id  family_id  family_members  family_genotypes
chr10  1142207 1142208 1 1 1_dad(dad;unaffected),1_mom(mom;unaffected),1_kid(child;affected) T/T,T/T,T/C
chr10  48003991 48003992 2 2 2_dad(dad;unaffected),2_kid(child;affected) C/C,C/C,C/T
chr10  48004991 48004992 3 3 3_dad(dad;unaffected),3_kid(child;affected) C/C,C/C,C/T
chr10  135336655 135336656 4 4 1_dad(dad;unaffected),1_mom(mom;unaffected),1_kid(child;affected) G/G,G/G,G/A
chr10  135336655 135336656 4 4 2_dad(dad;unaffected),2_kid(child;affected) G/G,G/G,G/A
chr10  135369531 135369532 5 5 1_dad(dad;unaffected),1_kid(child;affected) T/T,T/T,T/C
chr10  135369531 135369532 5 5 2_dad(dad;unaffected),2_kid(child;affected) T/T,T/T,T/C
chr10  135369531 135369532 5 5 3_dad(dad;unaffected),3_kid(child;affected) T/T,T/T,T/C
```

**Note:** The output will always start with the requested columns followed by the 5 columns enumerated at the start of this document.

```
$ gemini de_novo -d 50 --columns "chrom,start,end" test.de_novo.db

chrom  start  end  variant_id  family_id  family_members  family_genotypes
chr10  135369531 135369532 5 5 3_dad(dad;unaffected),3_kid(child;affected) T/T,T/T,T/C
```

**example**

If we wanted to restrict candidate variants to solely those with a HIGH predicted functional consequence, we could use the following:

```
$ gemini de_novo --columns "chrom, start, end, ref, alt"
--filter "impact_severity = 'HIGH'"
--members family_genotypes
--samples family_count
--num_dets 55

chrom  start  end  ref  alt  variant_id  family_id  family_members  family_genotypes
chr10  1142207 1142208 T  C  1 1 1_dad(dad;unaffected),1_mom(mom;unaffected),1_kid(child;affected) T/T,T/T,T/C
```

5.7. Built-in analysis tools
example

The following command would further restrict candidate genes to those genes with a de novo variant in at least two families:

```bash
$ gemini de_novo \
   --columns "chrom, start, end, ref, alt" \
   --filter "impact_severity = 'HIGH'" \
   --min-kindreds 2 \
   test.de_novo.db
```

example

By default, candidate de novo variants are reported for families in the database. One can restrict the analysis to variants in specific families with the `--families` option. Families should be provided as a comma-separated list.

```bash
$ gemini de_novo --families 1 my.db
$ gemini de_novo --families 1,7 my.db
```

**autosomal_recessive:** Find variants meeting an autosomal recessive model.

**Warning:** By default, this tool requires that you identify familial relationships via a PED file when loading your VCF into GEMINI. For example:

```bash
gemini load -v my.vcf -p my.ped my.db
```

However, in the absence of established parent/child relationships in the PED file, GEMINI will issue a WARNING, yet will attempt to identify autosomal recessive candidates for all samples marked as “affected”.

**Genotype Requirements**

- all affecteds must be hom_alt
- [affected] no unaffected can be hom_alt (can be unknown)
- [strict] if parents exist they must be unaffected and het for all affected kids
- [strict] if there are no affecteds that have a parent, a warning is issued.

If `--lenient` is specified, the 2 points prefixed with “[strict]” are not required.

If `--allow-unaffected` is specified, the point prefix with “[affected]” is not required.

**Default behavior**

Assuming you have defined the familial relationships between samples when loading your VCF into GEMINI, one can leverage a built-in tool for identifying variants that meet an autosomal recessive inheritance pattern. The reported variants will be restricted to those variants having the potential to impact the function of affecting protein coding transcripts.

For the following examples, let’s assume we have a PED file for 3 different families as follows (the kids are affected in each family, but the parents are not):
gemini Documentation, Release 0.20.1

$ cat families.ped
1 1_dad 0 0 -1 1
1 1_mom 0 0 -1 1
1 1_kid 1_dad 1_mom -1 2
2 2_dad 0 0 -1 1
2 2_mom 0 0 -1 1
2 2_kid 2_dad 2_mom -1 2
3 3_dad 0 0 -1 1
3 3_mom 0 0 -1 1
3 3_kid 3_dad 3_mom -1 2

$ gemini autosomal_recessive test.auto_rec.db --columns "chrom,start,end,gene"

Note: The output will always start with the requested columns and end with the 5 extra columns enumerated at the start of this document.

To restrict the report to genes with variants (doesn’t have to be the _same_ variant) observed in at least two kindreds, use the following:

$ gemini autosomal_recessive --columns "gene, chrom, start, end, ref, alt, impact, impact_severity" --min-kindreds 2

to report only those with a HIGH predicted functional consequence, we could use the following:

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To limit to confidently called genotypes:

```bash
$ gemini autosomal_dominant \
--columns "gene, chrom, start, end, ref, alt, impact, impact_severity" \
--filter "impact_severity = 'HIGH'" \
--min-kindreds 1 \ 
--gt-pl-max 10 \ 
my.db
```

**autosomal_dominant**: Find variants meeting an autosomal dominant model.

**Warning**: 0. version 0.16.0 changes the behavior of this tool to be more strict. To regain more lenient behavior, specify `--lenient` and `--allow-unaffected`.

By default, this tool requires that you identify familial relationships via a PED file when loading your VCF into GEMINI. For example:

```bash
gemini load -v my.vcf -p my.ped my.db
```

**Genotype Requirements**

- All affecteds must be het
- [affected] No unaffected can be het or homalt (can be unknown)
- `de_novo` mutations are not auto_dom (at least not in the first generation)
- At least 1 affected must have 1 affected parent (or have no parents).
- If no affected has a parent, a warning is issued.
- [strict] All affecteds must have parents with known phenotype.
- [strict] All affected kids must have at least 1 affected parent

If `--lenient` is specified, the items prefixed with “[strict]” are not required.

If `--allow-unaffected` is specified, the item prefix with “[affected]” is not required.

Note that for autosomal dominant `--lenient` allows singleton affecteds to be used to meet the `--min-kindreds` requirement if they are HET.

If there is incomplete penetrance in the kindred (unaffected obligate carriers), these individuals currently must be coded as having unknown phenotype or as being affected.
default behavior

For the following examples, let's assume we have a PED file for 3 different families as follows (the kids are affected in each family, but the parents are not):

```bash
$ cat families.ped
1 1_dad 0 0 -1 1
1 1_mom 0 0 -1 1
1 1_kid 1_dad 1_mom -1 2
2 2_dad 0 0 -1 1
2 2_mom 0 0 -1 2
2 2_kid 2_dad 2_mom -1 2
3 3_dad 0 0 -1 2
3 3_mom 0 0 -1 -9
3 3_kid 3_dad 3_mom -1 2
```

```
$ gemini autosomal_dominant test.auto_dom.db --columns "chrom,start,end,gene"
chr start end gene variant_id family_id family_members
--- ------- ------ --- -------------- ---- ----------------- 
chr10 48003991 48003992 ASAH2C 3 3 2_dad(dad;unaffected),2_mom(mom;affected),2_kid(child;affected) C/C,C/T,C/T 2_mom,2_kid
chr10 48004991 48004992 ASAH2C 4 4 2_dad(dad;unaffected),2_mom(mom;affected),2_kid(child;affected) C/C,C/T,C/T 2_mom,2_kid
chr10 48003991 48003992 ASAH2C 3 3 3_dad(dad;affected),3_mom(mom;unknown),3_kid(child;affected) C/T,C/C,C/T 3_dad,3_kid
chr10 48004991 48004992 ASAH2C 4 4 3_dad(dad;affected),3_mom(mom;unknown),3_kid(child;affected) C/T,C/C,C/T 3_dad,3_kid
chr10 135336655 135336656 SPRN 5 5 3_dad(dad;affected),3_mom(mom;unknown),3_kid(child;affected) G/A,G/G,G/A 3_dad,3_kid
chr10 1142207 1142208 WDR37 1 1 2_dad(dad;unaffected),2_mom(mom;affected),2_kid(child;affected) T/T,T/C,T/C 2_mom,2_kid
chr10 1142207 1142208 WDR37 1 1 3_dad(dad;affected),3_mom(mom;unknown),3_kid(child;affected) T/C,T/T,T/C 3_dad,3_kid
```

```
$ gemini autosomal_dominant test.auto_dom.db --columns "gene, chrom, start, end, ref, alt, impact, impact_severity"

<table>
<thead>
<tr>
<th>gene</th>
<th>chrom</th>
<th>start</th>
<th>end</th>
<th>ref</th>
<th>alt</th>
<th>impact</th>
<th>impact_severity</th>
<th>variant_id</th>
</tr>
</thead>
<tbody>
<tr>
<td>ASAH2C</td>
<td>chr10</td>
<td>48003991</td>
<td>48003992</td>
<td>C</td>
<td>T</td>
<td>non_syn_coding</td>
<td>MED</td>
<td>3 3 2_dad(dad;unaffected),2_mom(mom;affected),2_kid(child;affected) C/C,C/T,C/T 2_mom,2_kid</td>
</tr>
<tr>
<td>ASAH2C</td>
<td>chr10</td>
<td>48004991</td>
<td>48004992</td>
<td>C</td>
<td>T</td>
<td>non_syn_coding</td>
<td>MED</td>
<td>4 4 2_dad(dad;unaffected),2_mom(mom;affected),2_kid(child;affected) C/C,C/T,C/T 2_mom,2_kid</td>
</tr>
<tr>
<td>ASAH2C</td>
<td>chr10</td>
<td>48003991</td>
<td>48003992</td>
<td>C</td>
<td>T</td>
<td>non_syn_coding</td>
<td>MED</td>
<td>3 3 3_dad(dad;affected),3_mom(mom;unknown),3_kid(child;affected) C/T,C/C,C/T 3_dad,3_kid</td>
</tr>
<tr>
<td>ASAH2C</td>
<td>chr10</td>
<td>48004991</td>
<td>48004992</td>
<td>C</td>
<td>T</td>
<td>non_syn_coding</td>
<td>MED</td>
<td>4 4 3_dad(dad;affected),3_mom(mom;unknown),3_kid(child;affected) C/T,C/C,C/T 3_dad,3_kid</td>
</tr>
<tr>
<td>WDR37</td>
<td>chr10</td>
<td>1142207</td>
<td>1142208</td>
<td>T</td>
<td>C</td>
<td>stop_loss</td>
<td>HIGH</td>
<td>1 1 2_dad(dad;unaffected),2_mom(mom;affected),2_kid(child;affected) T/T,T/C,T/C 2_mom,2_kid</td>
</tr>
</tbody>
</table>
```

5.7. Built-in analysis tools
x_linked_recessive: x-linked recessive inheritance

Note that as of version 0.19.0, we do not account for the pseudo autosomal regions. The ‘X’ chromosome can be specifying using –X (defaults to ‘chrX’ and ‘X’)

Genotype Requirements

- Affected females must be HOM_ALT
- Unaffected females are HET or HOM_REF
- Affected males are not HOM_REF
- Unaffected males are HOM_REF

x_linked_dominant: x-linked dominant inheritance

Note that as of version 0.19.0, we do not account for the pseudo autosomal regions. The ‘X’ chromosome can be specifying using –X (defaults to ‘chrX’ and ‘X’)

Genotype Requirements

- Affected males are HET or HOM_ALT
- Affected females must be HET
- Unaffecteds must be HOM_REF
- girls of affected dad must be affected
- boys of affected dad must be unaffected
- mothers of affected males must be het (and affected) [added in 0.19.1]
- at least 1 parent of affected females must be het (and affected). [added in 0.19.1]

x_linked_de_novo: x-linked de novo

Note that as of version 0.19.0, we do not account for the pseudo autosomal regions. The ‘X’ chromosome can be specifying using –X (defaults to ‘chrX’ and ‘X’)

Genotype Requirements

- affected female child must be het
- affected male child must be hom_alt (or het)
- parents should be unaffected and hom_ref
gene_wise: Custom genotype filtering by gene.

The gemini query tool allows querying by variant and the inheritance tools described above enable querying by gene for fixed inheritance patterns. The gene_wise tool allows querying by gene with custom genotype filters to bridge the gap between these tools.

With this tool, multiple --gt-filter s can be specified. Each filter can be any valid filter; often, it will make sense to have 1 filter for each family. For example, given this pedigree:

![Pedigree Diagram]

Where only the orange samples are sequenced, we could devise a query:

```sh
$ gemini gene_wise $db \
    --min-filters 3 \
    --gt-filter "gt_types.fam1_kid == HET and gt_types.fam1_mom == HOM_REF and gt_types.fam1_dad == HOM_REF" \
    --gt-filter "gt_types.fam2_kid == HET" \
    --gt-filter "gt_types.fam3_kid == HET" \
    --columns "chrom,start,end,gene,impact,impact_severity" \
    --filter "max_aaf_all < 0.005"
```

The --min-filters option means that we want all 3 of those filters to be met in a gene in order for variants in that gene to be reported. We can envision a scenario where we have 6 families (and 6 filters) and we want to report genes where 4 of them meet the filters. In that case, the query would have 6 --gt-filter s and --min-filters of 3.

This differs from using gemini query with a single --gt-filter that combines each of those terms with an and because this allows each filter to be met in a different variant but in the same gene while the gemini query tool applies all elements of the single filter to each variant.

The output from the above query is:

<table>
<thead>
<tr>
<th>chrom</th>
<th>start</th>
<th>end</th>
<th>gene</th>
<th>impact</th>
<th>impact_severity</th>
<th>variant_filters</th>
<th>n_gene_variants</th>
<th>gene_filters</th>
</tr>
</thead>
<tbody>
<tr>
<td>chr5</td>
<td>60839982</td>
<td>60839983</td>
<td>ZSWIM6</td>
<td>non_syn_coding</td>
<td>MED</td>
<td>1,2,3</td>
<td>1,2,3</td>
<td></td>
</tr>
<tr>
<td>chr6</td>
<td>32548031</td>
<td>32548032</td>
<td>HLA-DRB1</td>
<td>non_syn_coding</td>
<td>MED</td>
<td>1,2,3</td>
<td>1</td>
<td></td>
</tr>
<tr>
<td>chr6</td>
<td>32552059</td>
<td>32552060</td>
<td>HLA-DRB1</td>
<td>frame_shift</td>
<td>HIGH</td>
<td>2,3</td>
<td>2</td>
<td></td>
</tr>
<tr>
<td>chr6</td>
<td>32552131</td>
<td>32552132</td>
<td>HLA-DRB1</td>
<td>inframe_codon_gain</td>
<td>MED</td>
<td>3,4</td>
<td>4</td>
<td></td>
</tr>
<tr>
<td>chr6</td>
<td>32552136</td>
<td>32552137</td>
<td>HLA-DRB1</td>
<td>non_syn_coding</td>
<td>MED</td>
<td>3,4</td>
<td>4</td>
<td></td>
</tr>
</tbody>
</table>

5.7. Built-in analysis tools 53
Note that the first gene has the same variant for all 3 families, so we could have found this with the gemini query tool. However, for the HLA gene, each of the 3 filters passed in different variant so this would be missed by the query tool which only looks at a single variant at a time.

As with the other tools, this tool orders by chromosome and gene and it applies WHERE (is_exonic = 1 AND impact_severity !='LOW') to the query.

- The \texttt{variant\_filters} column shows which filters were passed by the variant.
- The \texttt{n\_gene\_variants} column shows how many variants in the gene are being reported.
- The \texttt{gene\_filter} column shows which filters in the gene passed by any variant.

Multiple \texttt{--gt-filter-required} filters can also be specified. Each filter added to this argument is required to pass for each variant and it does not contribute to the \texttt{--min-filters} argument. This can be used with, or instead of \texttt{--gt-filter}.

\begin{verbatim}
gemini gene_wise  \
--columns "gene, chrom, start, end, ref, alt, impact, impact_severity"  \
--gt-filter-required "((gt_depths).(*).(>10).(all))"  \
--gt-filter "gt_types.fam1_kid == HET and gt_types.fam1_mom == HOM_REF and gt_  
˓→types.fam1_dad == HOM_REF"  \
--gt-filter "gt_types.fam2_kid == HET"  \
--gt-filter "gt_types.fam3_kid == HET"  \
--min-filters 2  \
test.db
\end{verbatim}

will required that all samples meet the minimum depth filter and then keep the subset of those that meet 2 out of the 3 \texttt{--gt-filters}.

\texttt{--where}

By default \texttt{gene\_wise} limits to variants passing the clause:
\begin{verbatim}
is_exonic = 1 AND impact_severity !='LOW'
\end{verbatim}

but this can be changed with the \texttt{--where} clause, e.g.:

\begin{verbatim}
--where "((is_exonic = 1 or is_splicing = 1) AND impact_severity !='LOW')"
\end{verbatim}

\textbf{pathways: Map genes and variants to KEGG pathways.}

Mapping genes to biological pathways is useful in understanding the function/role played by a gene. Likewise, genes involved in common pathways is helpful in understanding heterogeneous diseases. We have integrated the KEGG pathway mapping for gene variants, to explain/annotate variation. This requires your VCF be annotated with either snpEff/VEP.

Examples:

\begin{verbatim}
$ gemini pathways -v 68 example.db  
chrom start end ref alt impact sample genotype gene  \n--transcript pathway  
chr10 52004314 52004315 T C intron M128215 C/C ASAH2  
˓→ ENST00000395526 hsa00600:Sphingolipid_metabolism,hsa01100:Metabolic_pathways
\end{verbatim}
Here, -v specifies the version of the Ensembl genes used to build the KEGG pathway map. Hence, use versions that match the VEP/snpEff versions of the annotated vcf for correctness. For e.g VEP v2.6 and snpEff v3.1 use Ensembl 68 version of the genomes.

We currently support versions 66 through 71 of the Ensembl genes

--lof

By default, all gene variants that map to pathways are reported. However, one may want to restrict the analysis to LoF variants using the --lof option.

### interactions: Find genes among variants that are interacting partners.

Integrating the knowledge of the known protein-protein interactions would be useful in explaining variation data. Meaning to say that a damaging variant in an interacting partner of a potential protein may be equally interesting as the protein itself. We have used the HPRD binary interaction data to build a p-p network graph which can be explored by GEMINI.

Examples:

```bash
$ gemini interactions -g CTBP2 -r 3 example.db
sample gene order_of_interaction interacting_gene
M128215 CTBP2 0_order: CTBP2
M128215 CTBP2 1_order: RAI2
M128215 CTBP2 2_order: RB1
M128215 CTBP2 3_order: TGM2,NOTCH2NL
```

Return CTBP2 (-g) interacting gene variants till the third order (-r)

### lof_interactions

Use this option to restrict your analysis to only LoF variants.

```bash
$ gemini lof_interactions -r 3 example.db
sample lof_gene order_of_interaction interacting_gene
M128215 TGM2 1_order: RB1
M128215 TGM2 2_order: none
M128215 TGM2 3_order: NOTCH2NL,CTBP2
```

Meaning to say return all LoF gene TGM2 (in sample M128215) interacting partners to a 3rd order of interaction.
An extended variant information (chrom, start, end etc.) for the interacting gene may be achieved with the `--var` option for both the `interactions` and the `lof_interactions`

```bash
$ gemini interactions -g CTBP2 -r 3 --var example.db
```

<table>
<thead>
<tr>
<th>sample</th>
<th>gene</th>
<th>order_of_interaction</th>
<th>interacting_gene</th>
<th>var_id</th>
<th>chrom</th>
<th>start</th>
<th>end</th>
<th>impact</th>
<th>biotype</th>
<th>in_dbsnp</th>
<th>clinvar_sig</th>
<th>clinvar_disease_name</th>
<th>aaf_1kg_all</th>
<th>aaf_esp_all</th>
</tr>
</thead>
<tbody>
<tr>
<td>M128215</td>
<td>CTBP2</td>
<td>0</td>
<td>CTBP2</td>
<td>5</td>
<td>chr10</td>
<td>126678091</td>
<td>126678092</td>
<td>stop_</td>
<td>protein_coding</td>
<td>1</td>
<td>None</td>
<td>None</td>
<td>None</td>
<td>None</td>
</tr>
<tr>
<td>M128215</td>
<td>CTBP2</td>
<td>1</td>
<td>RA12</td>
<td>9</td>
<td>chrX</td>
<td>17819376</td>
<td>17819377</td>
<td>none_</td>
<td>protein_coding</td>
<td>1</td>
<td>None</td>
<td>1</td>
<td>0.000473</td>
<td></td>
</tr>
<tr>
<td>M128215</td>
<td>CTBP2</td>
<td>2</td>
<td>RB1</td>
<td>7</td>
<td>chr13</td>
<td>48873834</td>
<td>48873835</td>
<td>none_</td>
<td>protein_coding</td>
<td>1</td>
<td>None</td>
<td>0.94</td>
<td>None</td>
<td></td>
</tr>
<tr>
<td>M128215</td>
<td>CTBP2</td>
<td>3</td>
<td>NOTCH2NL</td>
<td>1</td>
<td>chr1</td>
<td>145273344</td>
<td>145273345</td>
<td>none_</td>
<td>protein_coding</td>
<td>1</td>
<td>None</td>
<td>None</td>
<td>None</td>
<td>None</td>
</tr>
<tr>
<td>M128215</td>
<td>CTBP2</td>
<td>3</td>
<td>TGM2</td>
<td>8</td>
<td>chr20</td>
<td>36779423</td>
<td>36779424</td>
<td>stop_</td>
<td>protein_coding</td>
<td>0</td>
<td>None</td>
<td>None</td>
<td>None</td>
<td>None</td>
</tr>
</tbody>
</table>

`lof_sieve`: Filter LoF variants by transcript position and type

Not all candidate LoF variants are created equal. For e.g, a nonsense (stop gain) variant impacting the first 5% of a polypeptide is far more likely to be deleterious than one affecting the last 5%. Assuming you’ve annotated your VCF with snpEff v3.0+, the `lof_sieve` tool reports the fractional position (e.g. 0.05 for the first 5%) of the mutation in the amino acid sequence. In addition, it also reports the predicted function of the transcript so that one can segregate candidate LoF variants that affect protein_coding transcripts from processed RNA, etc.

```bash
$ gemini lof_sieve chr22.low.exome.snpeff.100samples.vcf.db
```

<table>
<thead>
<tr>
<th>chrom</th>
<th>start</th>
<th>ref</th>
<th>alt</th>
<th>highest_impact</th>
<th>aa_change</th>
<th>var_trans_pos</th>
<th>trans_aa_length</th>
<th>var_trans_pct</th>
<th>sample</th>
<th>genotype</th>
<th>gene</th>
<th>transcript</th>
<th>trans_type</th>
</tr>
</thead>
<tbody>
<tr>
<td>chr22</td>
<td>17072346</td>
<td>17072347</td>
<td>C</td>
<td>T</td>
<td>stop_gain</td>
<td>W365*</td>
<td>365</td>
<td>557</td>
<td>0.655296229803</td>
<td>NA19327</td>
<td>C</td>
<td>T</td>
<td>CCT8L2</td>
</tr>
<tr>
<td>chr22</td>
<td>17072346</td>
<td>17072347</td>
<td>C</td>
<td>T</td>
<td>stop_gain</td>
<td>W365*</td>
<td>365</td>
<td>557</td>
<td>0.655296229803</td>
<td>NA19375</td>
<td>T</td>
<td>C</td>
<td>CCT8L2</td>
</tr>
<tr>
<td>chr22</td>
<td>17129539</td>
<td>17129540</td>
<td>C</td>
<td>T</td>
<td>splice_donor</td>
<td>None</td>
<td>None</td>
<td>None</td>
<td>None</td>
<td>NA18964</td>
<td>T</td>
<td>C</td>
<td>TPTPE1</td>
</tr>
<tr>
<td>chr22</td>
<td>17129539</td>
<td>17129540</td>
<td>C</td>
<td>T</td>
<td>splice_donor</td>
<td>None</td>
<td>None</td>
<td>None</td>
<td>None</td>
<td>NA19675</td>
<td>T</td>
<td>C</td>
<td>TPTPE1</td>
</tr>
</tbody>
</table>
amend: updating / changing the sample information

Occasionally one may need to update the sample information stored in the samples table. The amend tool allows one to provide an updated PED file as input and it will update each sample_id in the PED file that matches a sample_id.

For example, assume you have already loaded a GEMINI database with a samples.ped where mom and dad are unaffected and kid is affected:

```
$ cat samples.ped
1 dad 0 0 1 1
1 mom 0 0 1 1
1 kid dad mom 2 2
$ gemini load -v my.vcf -p samples.ped -t VEP my.db
```

Now, let’s say you realized that the dad is also affected and you want to correct the samples table accordingly. You would first edit the PED file and then run the amend tool using the updated PED.

```
$ cat samples.ped
1 dad 0 0 1 2
1 mom 0 0 1 1
1 kid dad mom 2 2
$ gemini amend --sample samples.ped my.db
```

annotate: adding your own custom annotations

It is inevitable that researchers will want to enhance the gemini framework with their own, custom annotations. gemini provides a sub-command called annotate for exactly this purpose. As long as you provide a tabix’ed annotation file in BED or VCF format, the annotate tool will, for each variant in the variants table, screen for overlaps in your annotation file and update a one or more new column in the variants table that you may specify on the command line. This is best illustrated by example.

Let’s assume you have already created a gemini database of a VCF file using the load module.

```
$ gemini load -v my.vcf -t snpEff my.db
```

Now, let’s imagine you have an annotated file in BED format (important.bed) that describes regions of the genome that are particularly relevant to your lab’s research. You would like to annotate in the gemini database which variants overlap these crucial regions. We want to store this knowledge in a new column in the variants table called important_variant that tracks whether a given variant overlapped (1) or did not overlap (0) intervals in your annotation file.

To do this, you must first TABIX your BED file:

```
$ bgzip important.bed
$ tabix -p bed important.bed.gz
```

-a boolean Did a variant overlap a region or not?

Note: Formerly, the -a option was the -t option.
Now, you can use this TABIX’ed file to annotate which variants overlap your important regions. In the example below, the results will be stored in a new column called “important”. The \(-t\) boolean option says that you just want to track whether (1) or not (0) the variant overlapped one or more of your regions.

```bash
$ gemini annotate -f important.bed.gz -c important -a boolean my.db
```

Since a new columns has been created in the database, we can now directly query the new column. In the example results below, the first and third variants overlapped a crucial region while the second did not.

```bash
$ gemini query
    -q "select chrom, start, end, variant_id, important from variants"
    my.db
    | head -3
chr22 100 101 1 1
chr22 200 201 2 0
chr22 300 500 3 1
```

**-a count How many regions did a variant overlap?**

Instead of a simple yes or no, we can use the \(-t\) count option to count how many important regions a variant overlapped. It turns out that the 3rd variant actually overlapped two important regions.

```bash
$ gemini annotate -f important.bed.gz -c important -a count my.db
$ gemini query
    -q "select chrom, start, end, variant_id, crucial from variants"
    my.db
    | head -3
chr22 100 101 1 1
chr22 200 201 2 0
chr22 300 500 3 2
```

**-a extract Extract specific values from a BED file**

Lastly, we may also extract values from specific fields in a BED file (or from the INFO field in a VCF) and populate one or more new columns in the database based on overlaps with the annotation file and the values of the fields therein. To do this, we use the \(-a\) extract option.

This is best described with an example. To set this up, let’s imagine that we have a VCF file from a different experiment and we want to annotate the variants in our GEMINI database with the allele frequency and depth tags from the INFO fields for the same variants in this other VCF file.

```bash
# bgzip and tabix the vcf for use with the annotate tool. $ bgzip other.vcf $ tabix other.vcf.gz
```

Now that we have a proper TABIX’ed VCF file, we can use the \(-a\) extract option to populate new columns in the GEMINI database. In order to do so, we must specify:

1. its type (e.g., text, int, float) \((-t\)
2. the field in the INFO column of the VCF file that we should use to extract data with which to populate the new column \((-e\)
3. what operation should be used to summarize the data in the event of multiple overlaps in the annotation file \((-o\)
4. (optionally) the name of the column we want to add \((-c\), if this is not specified, it will use the value from \(-e\).
For example, let’s imagine we want to create a new column called “other_allele_freq” using the AF field in our VCF file to populate it.

```
$ gemini annotate -f other.vcf.gz \
  -a extract \
  -c other_allele_freq \
  -t float \
  -e AF \
  -o mean \
  my.db
```

This create a new column in my.db called other_allele_freq and this new column will be a FLOAT. In the event of multiple records in the VCF file overlapping a variant in the database, the average (mean) of the allele frequencies values from the VCF file will be used.

At this point, one can query the database based on the values of the new other_allele_freq column:

```
$ gemini query -q "select * from variants where other_allele_freq < 0.01" my.db
```

- **-t TYPE** Specifying the column type(s) when using `-a extract`  
The annotate tool will create three different types of columns via the `-t` option:

1. Floating point columns for annotations with decimal precision as above (`-t float`)  
2. Integer columns for integral annotations (`-t integer`)  
3. Text columns for string columns such as “valid”, “yes”, etc. (`-t text`)  

**Note:** The `-t` option is only valid when using the `-a extract` option.

- **-o OPERATION** Specifying the summary operations when using `-a extract`  
In the event of multiple overlaps between a variant and records in the annotation file, the annotate tool can summarize the values observed with multiple options:

1. `-o mean`. Compute the average of the values. **They must be numeric.**  
2. `-o median`. Compute the median of the values. **They must be numeric.**  
3. `-o min`. Compute the minimum of the values. **They must be numeric.**  
4. `-o max`. Compute the maximum of the values. **They must be numeric.**  
5. `-o mode`. Compute the maximum of the values. **They must be numeric.**  
6. `-o first`. Use the value from the first record in the annotation file.  
7. `-o last`. Use the value from the last record in the annotation file.  
8. `-o list`. Create a comma-separated list of the observed values. **-t must be text**  
9. `-o uniq_list`. Create a comma-separated list of the distinct (i.e., non-redundant) observed values. **-t must be text**  
10. `-o sum`. Compute the sum of the values. **They must be numeric.**
Note: The -o option is only valid when using the -a extract option.

Annotating with VCF

Most of the examples to this point have pulled a column from a tabix indexed bed file. It is likewise possible to pull from the INFO field of a tabix index VCF. The syntax is identical but the -e operation will specify the names of fields in the INFO column to pull. By default, those names will be used, but that can still be specified with the -c column. Here are some example uses:

```bash
# put a DP column in the db:
gemini annotate -f anno.vcf.gz -o list -e DP -t integer my.db

# ... and name it 'depth'
gemini annotate -f anno.vcf.gz -o list -e DP -c depth -t integer my.db

# use multiple columns

gemini annotate -f anno.vcf.gz -o list,mean -e DP,Qmeter -c depth,qmeter -t integer my.db
```

Missing values are allowed since we expect that in some cases an annotation VCF will not have all INFO fields specified for all variants.

Note: We recommend decomposing and normalizing variants before annotating. See Step 1. split, left-align, and trim variants for a detailed explanation of how to do this.

Extracting and populating multiple columns at once.

One can also extract and populate multiple columns at once by providing comma-separated lists (no spaces) of column names (-c), types (-t), numbers (-e), and summary operations (-o). For example, recall that in the VCF example above, we created a TABIX’ed BED file containing the allele frequency and depth values from the INFO field as the 4th and 5th columns in the BED, respectively.

Instead of running the annotate tool twice (once for each column), we can run the tool once and load both columns in the same run. For example:

```bash
$ gemini annotate -f other.bed.gz
  -a extract
  -c other_allele_freq,other_depth
  -t float,integer
  -e 4,5
  -o mean,max
  my.db
```

We can then use each of the new columns to filter variants with a GEMINI query:

```bash
$ gemini query -q "select * from variants
  where other_allele_freq < 0.01
  and other_depth > 100" my.db
```
region: Extracting variants from specific regions or genes

One often is concerned with variants found solely in a particular gene or genomic region. `gemini` allows one to extract variants that fall within specific genomic coordinates as follows:

```
--reg
$ gemini region --reg chr1:100-200 my.db
```

```
--gene
Or, one can extract variants based on a specific gene name.
$ gemini region --gene PTPN22 my.db
```

```
--columns
By default, this tool reports all columns in the variants table. One may choose to report only a subset of the columns using the `--columns` option. For example, to report just the gene, chrom, start, end, ref, alt, impact, and impact_severity columns, one would use the following:
```
$ gemini region --gene DHODH
  --columns "chrom, start, end, ref, alt, gene, impact"
my.db

<table>
<thead>
<tr>
<th>chr1</th>
<th>start</th>
<th>end</th>
<th>ref</th>
<th>alt</th>
<th>gene</th>
<th>impact</th>
</tr>
</thead>
<tbody>
<tr>
<td>chr16</td>
<td>72057281</td>
<td>72057282</td>
<td>A</td>
<td>G</td>
<td>DHODH</td>
<td>intron</td>
</tr>
<tr>
<td>chr16</td>
<td>72057434</td>
<td>72057435</td>
<td>C</td>
<td>T</td>
<td>DHODH</td>
<td>non_syn_coding</td>
</tr>
<tr>
<td>chr16</td>
<td>72059268</td>
<td>72059269</td>
<td>T</td>
<td>C</td>
<td>DHODH</td>
<td>downstream</td>
</tr>
</tbody>
</table>
```

```
--filter
By default, this tool will report all variants regardless of their putative functional impact. In order to apply additional constraints on the variants returned, one can use the `--filter` option. Using SQL syntax, conditions applied with the `--filter` option become WHERE clauses in the query issued to the GEMINI database. For example, if we wanted to restrict candidate variants to solely those with a HIGH predicted functional consequence, we could use the following:
```
$ gemini region --gene DHODH
  --columns "chrom, start, end, ref, alt, gene, impact"
  --filter "alt='G'"
my.db

<table>
<thead>
<tr>
<th>chr1</th>
<th>start</th>
<th>end</th>
<th>ref</th>
<th>alt</th>
<th>gene</th>
<th>impact</th>
</tr>
</thead>
<tbody>
<tr>
<td>chr16</td>
<td>72057281</td>
<td>72057282</td>
<td>A</td>
<td>G</td>
<td>DHODH</td>
<td>intron</td>
</tr>
</tbody>
</table>
```

```
--json
Reporting query output in JSON format may enable HTML/Javascript apps to query GEMINI and retrieve the output in a format that is amenable to web development protocols.

To report in JSON format, use the `--json` option. For example:
```
```
windower: Conducting analyses on genome “windows”.

gemini includes a convenient tool for computing variation metrics across genomic windows (both fixed and sliding). Here are a few examples to whet your appetite. If you’re still hungry, contact us.

Compute the average nucleotide diversity for all variants found in non-overlapping, 50Kb windows.

```bash
$ gemini windower -w 50000 -s 0 -t nucl_div -o mean my.db
```

Compute the average nucleotide diversity for all variants found in 50Kb windows that overlap by 10kb.

```bash
$ gemini windower -w 50000 -s 10000 -t nucl_div -o mean my.db
```

Compute the max value for HWE statistic for all variants in a window of size 10kb

```bash
$ gemini windower -w 10000 -t hwe -o max my.db
```

stats: Compute useful variant statistics.

The stats tool computes some useful variant statistics like

Compute the transition and transversion ratios for the snps

```bash
$ gemini stats --tstv my.db
ts  tv  ts/tv
ts  tv  ts/tv
4  5  0.8
```

```
--tstv-coding
```

Compute the transition/transversion ratios for the snps in the coding regions.

```
--tstv-noncoding
```

Compute the transition/transversion ratios for the snps in the non-coding regions.

Compute the type and count of the snps.

```bash
$ gemini stats --snp-counts my.db
type  count
A->G  2
C->T  1
G->A  1
```

Calculate the site frequency spectrum of the variants.
Gemini is a powerful tool for analyzing variant data, and the `gemini stats` command provides several ways to extract insights from your data. Here are some examples of how to use it:

- To compute the pair-wise genetic distance between each sample:
  ```sh
  $ gemini stats --mds my.db
  sample1 sample2 distance
  M10500 M10500 0.0
  M10475 M10478 1.25
  M10500 M10475 2.0
  M10500 M10478 0.5714
  ``

- To return a count of the types of genotypes per sample:
  ```sh
  $ gemini stats --gts-by-sample my.db
  sample num_hom_ref num_het num_hom_alt num_unknown total
  M10475 4 1 3 1 9
  M10478 2 2 4 1 9
  ``

- To return the total variants per sample (sum of homozygous and heterozygous variants):
  ```sh
  $ gemini stats --vars-by-sample my.db
  sample total
  M10475 4
  M10478 6
  ``

- To summarize the variants per sample of an arbitrary query using the `--summarize` flag:
  ```sh
  $ gemini stats --summarize "select * from variants where in_dbsnp=1 and chrom='chr1'" --my.db
  sample total num_het num_hom_alt
  M10475 1 1 0
  M128215 1 1 0
  M10478 2 2 0
  M10500 2 1 1
  ``

## burden: perform sample-wise gene-level burden calculations

The `burden` tool provides a set of utilities to perform burden summaries on a per-gene, per sample basis. By default, it outputs a table of gene-wise counts of all high impact variants in coding regions for each sample:

```sh
$ gemini burden test.burden.db
  gene  M10475  M10478  M10500  M128215
  WDR37 2 2 2 2
  CTBP2 0 0 0 1
  DHODH 1 0 0 0
```

---

**5.7. Built-in analysis tools**
--nonsynonymous

If you want to be a little bit less restrictive, you can include all non-synonymous variants instead:

```
$ gemini burden --nonsynonymous test.burden.db

<table>
<thead>
<tr>
<th>gene</th>
<th>M10475</th>
<th>M10478</th>
<th>M10500</th>
<th>M128215</th>
</tr>
</thead>
<tbody>
<tr>
<td>SYCE1</td>
<td>0</td>
<td>1</td>
<td>1</td>
<td>0</td>
</tr>
<tr>
<td>WDR37</td>
<td>2</td>
<td>2</td>
<td>2</td>
<td>2</td>
</tr>
<tr>
<td>CTBP2</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>ASAH2C</td>
<td>2</td>
<td>1</td>
<td>1</td>
<td>0</td>
</tr>
<tr>
<td>DHODH</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
</tbody>
</table>
```

--calpha

If your database has been loaded with a PED file describing case and control samples, you can calculate the c-alpha statistic for cases vs. control:

```
$ gemini burden --calpha test.burden.db

<table>
<thead>
<tr>
<th>gene</th>
<th>T</th>
<th>c</th>
<th>Z</th>
<th>p_value</th>
</tr>
</thead>
<tbody>
<tr>
<td>SYCE1</td>
<td>-0.5</td>
<td>0.25</td>
<td>-1.0</td>
<td>0.841344746069</td>
</tr>
<tr>
<td>WDR37</td>
<td>-1.0</td>
<td>1.5</td>
<td>-0.816496580928 0.792891910879</td>
<td></td>
</tr>
<tr>
<td>CTBP2</td>
<td>0.0</td>
<td>0.0</td>
<td>nan</td>
<td>nan</td>
</tr>
<tr>
<td>ASAH2C</td>
<td>-0.5</td>
<td>0.75</td>
<td>-0.57735026919</td>
<td>0.718148569175</td>
</tr>
<tr>
<td>DHODH</td>
<td>0.0</td>
<td>0.0</td>
<td>nan</td>
<td>nan</td>
</tr>
</tbody>
</table>
```

To calculate the P-value using a permutation test, use the --permutations option, specifying the number of permutations of the case/control labels you want to use.

--min-aaf and --max-aaf for --calpha

By default, all variants affecting a given gene will be included in the C-alpha computation. However, one may establish alternate allele frequency boundaries for the variants included using the --min-aaf and --max-aaf options.

```
$ gemini burden --calpha test.burden.db --min-aaf 0.0 --max-aaf 0.01
```

--cases and --controls for --calpha

If you do not have a PED file loaded, or your PED file does not follow the standard PED phenotype encoding format you can still perform the c-alpha test, but you have to specify which samples are the control samples and which are the case samples:

```
$ gemini burden --controls M10475 M10478 --cases M10500 M128215 --calpha test.burden.db

<table>
<thead>
<tr>
<th>gene</th>
<th>T</th>
<th>c</th>
<th>Z</th>
<th>p_value</th>
</tr>
</thead>
<tbody>
<tr>
<td>SYCE1</td>
<td>-0.5</td>
<td>0.25</td>
<td>-1.0</td>
<td>0.841344746069</td>
</tr>
<tr>
<td>WDR37</td>
<td>-1.0</td>
<td>1.5</td>
<td>-0.816496580928 0.792891910879</td>
<td></td>
</tr>
<tr>
<td>CTBP2</td>
<td>0.0</td>
<td>0.0</td>
<td>nan</td>
<td>nan</td>
</tr>
<tr>
<td>ASAH2C</td>
<td>-0.5</td>
<td>0.75</td>
<td>-0.57735026919</td>
<td>0.718148569175</td>
</tr>
<tr>
<td>DHODH</td>
<td>0.0</td>
<td>0.0</td>
<td>nan</td>
<td>nan</td>
</tr>
</tbody>
</table>
```
--nonsynonymous --calpha

If you would rather consider all nonsynonymous variants for the C-alpha test rather than just the medium and high impact variants, add the --nonsynonymous flag.

**ROH: Identifying runs of homozygosity**

Runs of homozygosity are long stretches of homozygous genotypes that reflect segments shared identically by descent and are a result of consanguinity or natural selection. Consanguinity elevates the occurrence of rare recessive diseases (e.g. cystic fibrosis) that represent homozygotes for strongly deleterious mutations. Hence, the identification of these runs holds medical value.

The ‘roh’ tool in GEMINI returns runs of homozygosity identified in whole genome data. The tool basically looks at every homozygous position on the chromosome as a possible start site for the run and looks for those that could give rise to a potentially long stretch of homozygous genotypes.

For e.g. for the given example allowing 1 HET genotype (h) and 2 UKW genotypes (u) the possible roh runs (H) would be:

<table>
<thead>
<tr>
<th>genotype_run</th>
<th>roh_run1</th>
<th>roh_run2</th>
<th>roh_run3</th>
<th>roh_run4</th>
</tr>
</thead>
<tbody>
<tr>
<td>H H H H h H H H H u H H H H u H H H H H H H H h H H H H H u H H H H H H H</td>
<td>H H H H h H H H H u H H H H H u H H H H H H H</td>
<td>H H H H H H</td>
<td>H H H H H H</td>
<td></td>
</tr>
</tbody>
</table>

roh returned for --min-snps = 20 would be:

<table>
<thead>
<tr>
<th>roh_run1</th>
<th>roh_run2</th>
</tr>
</thead>
<tbody>
<tr>
<td>H H H H H H H H u H H H H H u H H H H H H H</td>
<td>H H H H u H H H H H u H H H H H H H</td>
</tr>
</tbody>
</table>

As you can see, the immediate homozygous position right of a break (h or u) would be the possible start of a new roh run and genotypes to the left of a break are pruned since they cannot be part of a longer run than we have seen before.

Return roh with minimum of 50 snps, a minimum run length of 1 mb and a minimum sample depth of 20 for sample S138 (with default values for allowed number of HETS, UNKS and total depth).

```bash
$ gemini roh --min-snps 50 --min-gt-depth 20 --min-size 1000000 -s S138 roh_run.db
```

<table>
<thead>
<tr>
<th>chrom</th>
<th>start</th>
<th>end</th>
<th>sample</th>
<th>num_of_snps</th>
<th>density_per_kb</th>
<th>run_length_in_bp</th>
</tr>
</thead>
<tbody>
<tr>
<td>chr2</td>
<td>233336080</td>
<td>234631638</td>
<td>S138</td>
<td>2583</td>
<td>1.9953</td>
<td>1295558</td>
</tr>
<tr>
<td>chr2</td>
<td>238341281</td>
<td>239522281</td>
<td>S138</td>
<td>2899</td>
<td>2.4555</td>
<td>1181000</td>
</tr>
</tbody>
</table>

**set_somatic: Flag somatic variants**

Somatic mutations in a tumor-normal pair are variants that are present in the tumor but not in the normal sample.

**Note:** 1. This tool requires that you specify the sample layout via a PED file when loading your VCF into GEMINI via:

```bash
gemini load -v my.vcf -p my.ped my.db
```

**Example PED file format for GEMINI**
**default behavior**

By default, *set_somatic* simply marks variants that are genotyped as homozygous reference in the normal sample and non-reference in the tumor. More stringent somatic filtering criteria are available through tunable command line parameters.

```
$ gemini set_somatic
  --min-depth 30
  --min-qual 20
  --min-somatic-score 18
  --min-tumor-depth 10
  --min-norm-depth 10
tumor_normal.db
```

<table>
<thead>
<tr>
<th>tumor</th>
<th>tum_gt</th>
<th>tum_alt_freq</th>
<th>tum_alt_depth</th>
<th>tum_depth</th>
<th>nrm_name</th>
<th>nrm_gt</th>
<th>nrm_alt_freq</th>
<th>nrm_alt_depth</th>
<th>nrm_depth</th>
<th>ref</th>
<th>alt</th>
<th>gene</th>
</tr>
</thead>
<tbody>
<tr>
<td>GAAAAAAAGGTGAAAATT/GAAAAAAAAAAAGGTGAAAATT</td>
<td>0.217391304348</td>
<td>5</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>23 normal GAAAAAAAGGTGAAAATT/GAAAAAAAAAAAGGTGAAAATT</td>
<td>0.0</td>
<td>0</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>25 chrX 132838304 132838328 GAAAAAAAAAAAGGTGAAAATT</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>GAAAAAAAAAAAGGTGAAAATT GPC3</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>tumor CTGCTATTTTG/CG 0.22 11 50 normal CTGCTATTTTG/CTGCTATTTTG 0.0</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>0 70 chr17 59861630 59861641 GAAAAAAAAAAAGGTGAAAATT</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>25 chrX 132838304 132838328 GAAAAAAAAAAAGGTGAAAATT</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>GAAAAAAAAAAAGGTGAAAATT GPC3</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>tumor C/A 0.555555555556 10 18 normal C/C 0.0</td>
<td>0</td>
<td>17</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>chr17 7578460 7578461 C A TP53</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>tumor C/T 0.1875 12 64 normal C/C 0.0</td>
<td>0</td>
<td>30</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>128046288 128046289 C T ERCC3</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Identified and set 4 somatic mutations

--- **--min-depth [None]**

The minimum required combined depth for tumor and normal samples.

--- **--min-qual [None]**

The minimum required variant quality score.

--- **--min-somatic-score [None]**

The minimum required somatic score (SSC). This score is produced by various somatic variant detection algorithms including SpeedSeq, SomaticSniper, and VarScan 2.

--- **--max-norm-alt-freq [None]**

The maximum frequency of the alternate allele allowed in the normal sample.
gemini Documentation, Release 0.20.1

--max-norm-alt-count [None]
The maximum count of the alternate allele allowed in the normal sample.

--min-norm-depth [None]
The minimum depth required in the normal sample.

--min-tumor-alt-freq [None]
The minimum frequency of the alternate allele required in the tumor sample.

--min-tumor-alt-count [None]
The minimum count of the alternate allele required in the tumor sample.

--min-tumor-depth [None]
The minimum depth required in the tumor sample.

--chrom [None]
A specific chromosome on which to flag somatic mutations.

--dry-run
Don’t set the is_somatic flag, just report what _would_ be set. For testing purposes.

**actionable_mutations**: Report actionable somatic mutations and drug-gene interactions

Actionable mutations are somatic variants in COSMIC cancer census genes with medium or high impact severity predictions. This tool reports actionable mutations as well as their known drug interactions (if any) from DGIdb. Current functionality is only for SNVs and indels.

**Note:**

1. This tool requires somatic variants to have been flagged using **set_somatic**

```bash
$ gemini actionable_mutations tumor_normal.db
```

```
actionable_mutations

| tum_name | chrom | start | end | ref   |.alt| gene | impact   | is_somatic | dgidb_info   |
|----------|-------|-------|-----|-------+----|------|----------|-----------|-------------|
| tumor    | chr2  | 128046288 | 128046289 | C    | T  | ERCC3 | non_syn_  |            |
|          |       |       |     |       |    |      | coding 1 | 1          |             |
| tumor    | chr17 | 7578460  | 7578461 | C    | A  | TP53  | non_syn_coding | 1          |             |
```

- **TP53**, 'geneCategories': ['CLINICALLY ACTIONABLE', 'DRUGGABLE', 'GENOME', 'TUMOR SUPPRESSOR', 'TRANSCRIPTION FACTOR COMPLEX', 'DRUG RESISTANCE', 'HISTONE MODIFICATION', 'DNA REPAIR', 'TRANSCRIPTION FACTOR BINDING'], 'geneName': 'TP53', 'geneLongName': 'tumor protein p53', 'interactions': [{'source': 'DrugBank', 'interactionId': '71cdec2-439b-4d46-963a-39b35bbbd801', 'interactionType': 'n/a', 'drugName': '1-(9-ETHYL-9H-CARBAZOL-3-YL)-N-METHYLMETHANAMINE'}, {'source': 'PharmGKB', 'interactionId': '8234d9b9-085d-49b1-aac2-cf3375d91477', 'interactionType': 'n/a', 'drugName': 'FLUOROURACIL'}, {'source': 'PharmGKB', 'interactionId': '605d7bca-7ed9-428e-aa7c-f76aaf66db54', 'interactionType': 'n/a', 'drugName': 'PACLITAXEL'}, {'source': 'TTD', 'interactionId': '1fe9db63-3581-435b-b22a-1245c89864', 'interactionType': 'activator', 'drugName': 'CURAXIN CBL102'}, {'source': 'TALC', 'interactionId': 'd59e14bc-b9a5-4c9f-a5aa-7ba322f0fa0e', 'interactionType': 'vaccine', 'drugName': 'MUTANT P53 PEPTIDE PULSED DENDRITIC CELL'}, {'source': 'TALC', 'interactionId': '79256b6e-9a16-4fbe-a237-28dbca28bc2a', 'interactionType': 'vaccine', 'drugName': 'AD.P53-DC'}

5.7. **Built-in analysis tools**
fusions: Report putative gene fusions


Note:

1. This tool requires somatic variants to have been flagged using `set_somatic`

default behavior

By default, `fusions` reports structural variants that are flagged as somatic, join two different genes, and preserve transcript strand orientation. These may be further filtered using tunable command line parameters.

```bash
$ gemini fusions
  --min_qual 5
  --in_cosmic_census
tumor_normal.db
  chromA breakpointA_start breakpointA_end chromB breakpointB_start breakpointB_end var_id qual strandA strandB sv_type geneA geneB tool evidence_type is_precise sample
tumor chr17 59861630 59861641 chr17 59861630 59861641 CTGCTATTTG CG BRIP1 -
inframe_codon_loss 1 1 None
  tumor chrX 132838304 132838328 chrX 132838304 132838328 GAAAAAAAAAAAAGGTGAAAATT GPC3 splice_region 1 1 None

--min_qual [None]

The minimum required variant quality score.

--evidence_type STRING

The required supporting evidence types for the variant from LUMPY ("PE", "SR", or "PE,SR").

--in_cosmic_census

Require at least one of the affected genes to be in the COSMIC cancer gene census.

db_info: List the gemini database tables and columns

Because of the sheer number of annotations that are stored in gemini, there are admittedly too many columns to remember by rote. If you can’t recall the name of particular column, just use the `db_info` tool. It will report all of the tables and all of the columns / types in each table:
$ gemini db_info test.db

<table>
<thead>
<tr>
<th>table_name</th>
<th>column_name</th>
<th>type</th>
</tr>
</thead>
<tbody>
<tr>
<td>variants</td>
<td>chrom</td>
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<td>integer</td>
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<tr>
<td>variant_impacts</td>
<td>impact</td>
<td>text</td>
</tr>
</tbody>
</table>
The GEMINI browser interface

Currently, the majority of GEMINI’s functionality is available via a command-line interface. However, we are developing a browser-based interface for easier exploration of GEMINI databases created with the `gemini load` command.

Ironically, as of now, one must launch said browser from the command line as follows (where `my.db` should be replaced with the name of the GEMINI database you would like to explore).

```
$ gemini browser my.db --use builtin
```

At this point, the GEMINI browser is running on port 8088 on your local machine. Open a web browser to http://localhost:8088/query You should see something like:
NOTE: The internal (builtin) browser is in maintenance mode, please look at alternative browser support below for now.

## Alternative browser support

GEMINI also allows to third-party genome browsers such as [puzzle](https://github.com/robinandeer/puzzle), the successor of [SciLifeLab](http://www.scilifelab.se)’s [scout](https://github.com/robinandeer/scout).

If you want to spawn GEMINI-compatible browsers, you can use the following commandline:

```
$ gemini browser my.db [--use puzzle]
```
The GEMINI database schema

The variants table

Core VCF fields

<table>
<thead>
<tr>
<th>column_name</th>
<th>type</th>
<th>notes</th>
</tr>
</thead>
<tbody>
<tr>
<td>chrom</td>
<td>STRING</td>
<td>The chromosome on which the variant resides (from VCF CHROM field).</td>
</tr>
<tr>
<td>start</td>
<td>INTEGER</td>
<td>The 0-based start position. (from VCF POS field, but converted to 0-based coordinates)</td>
</tr>
<tr>
<td>end</td>
<td>INTEGER</td>
<td>The 1-based end position. (from VCF POS field, yet inferred based on the size of the variant)</td>
</tr>
<tr>
<td>vcf_id</td>
<td>STRING</td>
<td>The VCF ID field.</td>
</tr>
<tr>
<td>variant_id</td>
<td>INTEGER</td>
<td>PRIMARY_KEY</td>
</tr>
<tr>
<td>anno_id</td>
<td>INTEGER</td>
<td>Variant transcript number for the most severely affected transcript</td>
</tr>
<tr>
<td>ref</td>
<td>STRING</td>
<td>Reference allele (from VCF REF field)</td>
</tr>
<tr>
<td>alt</td>
<td>STRING</td>
<td>Alternate allele for the variant (from VCF ALT field)</td>
</tr>
<tr>
<td>qual</td>
<td>INTEGER</td>
<td>Quality score for the assertion made in ALT (from VCF QUAL field)</td>
</tr>
<tr>
<td>filter</td>
<td>STRING</td>
<td>A string of filters passed/failed in variant calling (from VCF FILTER field)</td>
</tr>
</tbody>
</table>
## Variant and PopGen info

<table>
<thead>
<tr>
<th>Field</th>
<th>Type</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>type</strong></td>
<td>STRING</td>
<td>The type of variant. Any of: [snp, indel]</td>
</tr>
<tr>
<td><strong>sub_type</strong></td>
<td>STRING</td>
<td>The variant sub-type. If type is snp: [ts, (transition), tv (transversion)]</td>
</tr>
<tr>
<td></td>
<td></td>
<td>If type is indel: [ins, (insertion), del (deletion)]</td>
</tr>
<tr>
<td><strong>call_rate</strong></td>
<td>FLOAT</td>
<td>The fraction of samples with a valid genotype</td>
</tr>
<tr>
<td><strong>num_hom_ref</strong></td>
<td>INTEGER</td>
<td>The total number of homozygotes for the reference (ref) allele</td>
</tr>
<tr>
<td><strong>num_het</strong></td>
<td>INTEGER</td>
<td>The total number of heterozygotes observed.</td>
</tr>
<tr>
<td><strong>num_hom_alt</strong></td>
<td>INTEGER</td>
<td>The total number of homozygotes for the reference (alt) allele</td>
</tr>
<tr>
<td><strong>num_unknown</strong></td>
<td>INTEGER</td>
<td>The total number of unknown genotypes</td>
</tr>
<tr>
<td><strong>aaf</strong></td>
<td>FLOAT</td>
<td>The observed allele frequency for the alternate allele</td>
</tr>
<tr>
<td><strong>hwe</strong></td>
<td>FLOAT</td>
<td>The Chi-square probability of deviation from HWE (assumes random mating)</td>
</tr>
<tr>
<td><strong>inbreedingCoeff</strong></td>
<td>FLOAT</td>
<td>The inbreeding coef-ficient that expresses the likelihood of effects due to inbreeding</td>
</tr>
<tr>
<td><strong>pi</strong></td>
<td>FLOAT</td>
<td>The computed nucleotide diversity (pi) for the site</td>
</tr>
</tbody>
</table>
## Genotype information

<table>
<thead>
<tr>
<th>Field</th>
<th>Type</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>gts</td>
<td>BLOB</td>
<td>A compressed binary vector of sample genotypes (e.g., “A/A”, “A</td>
</tr>
<tr>
<td>gt_types</td>
<td>BLOB</td>
<td>A compressed binary vector of numeric genotype “types” (e.g., 0, 1, 2) - Inferred from the VCF GT genotype tag.</td>
</tr>
<tr>
<td>gt_phases</td>
<td>BLOB</td>
<td>A compressed binary vector of sample genotype phases (e.g., False, True, False) - Extracted from the VCF GT genotype tag’s allele delimiter e.g., A/G means an unphased genotype. Value is FALSE. e.g., A</td>
</tr>
<tr>
<td>gt_depths</td>
<td>BLOB</td>
<td>A compressed binary vector of the depth of aligned sequence observed for each sample - Extracted from the VCF DP genotype tag.</td>
</tr>
<tr>
<td>gt_ref_depths</td>
<td>BLOB</td>
<td>A compressed binary vector of the depth of reference alleles observed for each sample - Extracted from the VCF AD genotype tag.</td>
</tr>
<tr>
<td>gt_alt_depths</td>
<td>BLOB</td>
<td>A compressed binary vector of the depth of alternate alleles observed for each sample - Extracted from the VCF AD genotype tag.</td>
</tr>
<tr>
<td>gt_alt_freqs</td>
<td>BLOB</td>
<td>A compressed binary (float) vector of the frequency of alternate alleles observed for each sample - equivalent to gt_alt_depths / (gt_alt_depths + gt_ref_depths)</td>
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</table>

### 5.9. The GEMINI database schema
### Gene information

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<th>Type</th>
<th>Description</th>
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<td>Corresponding gene name of the highly affected transcript</td>
</tr>
<tr>
<td>transcript</td>
<td>STRING</td>
<td>The variant transcript that was most severely affected (for two equally affected transcripts, the protein_coding biotype is prioritized (SnpEff/VEP)</td>
</tr>
<tr>
<td>is_exonic</td>
<td>BOOL</td>
<td>Does the variant affect an exon for &gt;= 1 transcript?</td>
</tr>
<tr>
<td>is_coding</td>
<td>BOOL</td>
<td>Does the variant fall in a coding region (excl. 3’ &amp; 5’ UTRs) for &gt;= 1 transcript?</td>
</tr>
<tr>
<td>is_lof</td>
<td>BOOL</td>
<td>Based on the value of the impact col, is the variant LOF for &gt;= transcript?</td>
</tr>
<tr>
<td>is_splicing</td>
<td>BOOL</td>
<td>Does the variant affect a canonical or possible splice site? That is, set to TRUE if the SO term is any of splice_acceptor_variant, splice_donor_variant, or splice_region_variant.</td>
</tr>
<tr>
<td>exon</td>
<td>STRING</td>
<td>Exon information for the severely affected transcript</td>
</tr>
<tr>
<td>codon_change</td>
<td>STRING</td>
<td>What is the codon change?</td>
</tr>
<tr>
<td>aa_change</td>
<td>STRING</td>
<td>What is the amino acid change (for a snp)?</td>
</tr>
<tr>
<td>aa_length</td>
<td>STRING</td>
<td>Has the format pos/len when biotype=protein_coding, is empty otherwise. len=protein length. pos = position of the amino acid change when is_coding=1 and is_exonic=1, ‘-’ otherwise.</td>
</tr>
<tr>
<td>biotype</td>
<td>STRING</td>
<td>The ‘type’ of the severely affected transcript (e.g., protein-coding, pseudogene, rRNA etc.) (only SnpEff)</td>
</tr>
<tr>
<td>impact</td>
<td>STRING</td>
<td>The consequence of the most severely affected transcript</td>
</tr>
<tr>
<td>impact_so</td>
<td>STRING</td>
<td>The Sequence ontology term for the most severe consequence</td>
</tr>
<tr>
<td>impact_severity</td>
<td>STRING</td>
<td>Severity of the highest order observed for the variant</td>
</tr>
<tr>
<td>polyphen_pred</td>
<td>STRING</td>
<td>Polyphen predictions for the snps for the most severely affected transcript</td>
</tr>
<tr>
<td>polyphen_score</td>
<td>FLOAT</td>
<td>Polyphen scores for the severely affected transcript (only VEP)</td>
</tr>
<tr>
<td>sift_pred</td>
<td>STRING</td>
<td>SIFT predictions for the snp’s for the most severely affected transcript (only VEP)</td>
</tr>
<tr>
<td>sift_score</td>
<td>FLOAT</td>
<td>SIFT scores for the predictions (only VEP)</td>
</tr>
<tr>
<td>pfam_domain</td>
<td>STRING</td>
<td>Pfam protein domain that the variant affects</td>
</tr>
</tbody>
</table>
## Optional VCF INFO fields

<table>
<thead>
<tr>
<th>Field</th>
<th>Type</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>anc_allele</td>
<td>STRING</td>
<td>The reported ancestral allele if there is one.</td>
</tr>
<tr>
<td>rms_bq</td>
<td>FLOAT</td>
<td>The RMS base quality at this position.</td>
</tr>
<tr>
<td>cigar</td>
<td>STRING</td>
<td>CIGAR string describing how to align an alternate allele to the reference allele.</td>
</tr>
<tr>
<td>depth</td>
<td>INTEGER</td>
<td>The number of aligned sequence reads that led to this variant call</td>
</tr>
<tr>
<td>strand_bias</td>
<td>FLOAT</td>
<td>Strand bias at the variant position. From the “SB” tag.</td>
</tr>
<tr>
<td>rms_map_qual</td>
<td>FLOAT</td>
<td>RMS mapping quality, a measure of variance of quality scores</td>
</tr>
<tr>
<td>in_hom_run</td>
<td>INTEGER</td>
<td>Homopolymer runs for the variant allele</td>
</tr>
<tr>
<td>num_mapq_zero</td>
<td>INTEGER</td>
<td>Total counts of reads with mapping quality equal to zero</td>
</tr>
<tr>
<td>num_alleles</td>
<td>INTEGER</td>
<td>Total number of alleles in called genotypes</td>
</tr>
<tr>
<td>num_reads_w_dels</td>
<td>FLOAT</td>
<td>Fraction of reads with spanning deletions</td>
</tr>
<tr>
<td>haplotype_score</td>
<td>FLOAT</td>
<td>Consistency of the site with two segregating haplotypes</td>
</tr>
<tr>
<td>qual_depth</td>
<td>FLOAT</td>
<td>Variant confidence or quality by depth</td>
</tr>
<tr>
<td>allele_count</td>
<td>INTEGER</td>
<td>Allele counts in genotypes</td>
</tr>
<tr>
<td>allele_bal</td>
<td>FLOAT</td>
<td>Allele balance for hets</td>
</tr>
<tr>
<td>info</td>
<td>BLOB</td>
<td>Stores the INFO field of the VCF</td>
</tr>
</tbody>
</table>

## Population information

<table>
<thead>
<tr>
<th>Field</th>
<th>Type</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>in_dbsnp</td>
<td>BOOL</td>
<td>Is this variant found in dbSNP?</td>
</tr>
<tr>
<td></td>
<td></td>
<td>0 : Absence of the variant in dbsnp</td>
</tr>
<tr>
<td></td>
<td></td>
<td>1 : Presence of the variant in dbsnp</td>
</tr>
<tr>
<td>rs_ids</td>
<td>STRING</td>
<td>A comma-separated list of rs ids for variants present in dbSNP</td>
</tr>
<tr>
<td>in_hm2</td>
<td>BOOL</td>
<td>Whether the variant was part of HapMap2.</td>
</tr>
<tr>
<td>in_hm3</td>
<td>BOOL</td>
<td>Whether the variant was part of HapMap3.</td>
</tr>
<tr>
<td>in_esp</td>
<td>BOOL</td>
<td>Presence/absence of the variant in the ESP project data</td>
</tr>
<tr>
<td>in_1kg</td>
<td>BOOL</td>
<td>Presence/absence of the variant in the 1000 genome project data (phase 3)</td>
</tr>
<tr>
<td>aaf_esp_ea</td>
<td>FLOAT</td>
<td>Minor Allele Frequency of the variant for European Americans in the ESP project</td>
</tr>
</tbody>
</table>

Continued on next page
Table 5.3 – continued from previous page

<table>
<thead>
<tr>
<th>Variable</th>
<th>Type</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>aaf_esp_aa</td>
<td>FLOAT</td>
<td>Minor Allele Frequency of the variant for African Americans in the ESP project</td>
</tr>
<tr>
<td>aaf_esp_all</td>
<td>FLOAT</td>
<td>Minor Allele Frequency of the variant w.r.t both groups in the ESP project</td>
</tr>
<tr>
<td>aaf_1kg_amr</td>
<td>FLOAT</td>
<td>Allele frequency of the variant in AMR population based on AC/AN (1000g project, phase 3)</td>
</tr>
<tr>
<td>aaf_1kg_eas</td>
<td>FLOAT</td>
<td>Allele frequency of the variant in EAS population based on AC/AN (1000g project, phase 3)</td>
</tr>
<tr>
<td>aaf_1kg_sas</td>
<td>FLOAT</td>
<td>Allele frequency of the variant in SAS population based on AC/AN (1000g project, phase 3)</td>
</tr>
<tr>
<td>aaf_1kg_afr</td>
<td>FLOAT</td>
<td>Allele frequency of the variant in AFR population based on AC/AN (1000g project, phase 3)</td>
</tr>
<tr>
<td>aaf_1kg_eur</td>
<td>FLOAT</td>
<td>Allele frequency of the variant in EUR population based on AC/AN (1000g project, phase 3)</td>
</tr>
<tr>
<td>aaf_1kg_all</td>
<td>FLOAT</td>
<td>Global allele frequency (based on AC/AN) (1000g project - phase 3)</td>
</tr>
<tr>
<td>in_exac</td>
<td>BOOL</td>
<td>Presence/absence of the variant in ExAC (Exome Aggregation Consortium) data (Broad)</td>
</tr>
<tr>
<td>aaf_exac_all</td>
<td>FLOAT</td>
<td>Raw allele frequency (population independent) of the variant based on ExAC exomes (AF)</td>
</tr>
<tr>
<td>aaf_adj_exac_all</td>
<td>FLOAT</td>
<td>Adjusted allele frequency (population independent) of the variant based on ExAC (Adj_AC/Adj_AN)</td>
</tr>
<tr>
<td>aaf_adj_exac_afr</td>
<td>FLOAT</td>
<td>Adjusted allele frequency of the variant for AFR population in ExAC (AC_AFR/AN_AFR)</td>
</tr>
<tr>
<td>aaf_adj_exac_amr</td>
<td>FLOAT</td>
<td>Adjusted allele frequency of the variant for AMR population in ExAC (AC_AMR/AN_AMR)</td>
</tr>
<tr>
<td>aaf_adj_exac_eas</td>
<td>FLOAT</td>
<td>Adjusted allele frequency of the variant for EAS population in ExAC (AC_EAS/AN_EAS)</td>
</tr>
<tr>
<td>aaf_adj_exac_fin</td>
<td>FLOAT</td>
<td>Adjusted allele frequency of the variant for FIN population in ExAC (AC_FIN/AN_FIN)</td>
</tr>
<tr>
<td>aaf_adj_exac_nfe</td>
<td>FLOAT</td>
<td>Adjusted allele frequency of the variant for NFE population in ExAC (AC_NFE/AN_NFE)</td>
</tr>
<tr>
<td>aaf_adj_exac_oth</td>
<td>FLOAT</td>
<td>Adjusted allele frequency of the variant for OTH population in ExAC (AC_OTH/AN_OTH)</td>
</tr>
</tbody>
</table>

Continued on next page
<table>
<thead>
<tr>
<th>Field Name</th>
<th>Type</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>aaf_adj_exac_sas</td>
<td>FLOAT</td>
<td>Adjusted allele frequency of the variant for SAS population in ExAC (AC_SAS/AN_SAS)</td>
</tr>
<tr>
<td>max_aaf_all</td>
<td>FLOAT</td>
<td>the maximum of aaf_gnomad[afr,amr,eas,nfe,sas],aaf_gnomad_esp_ea,aaf_gnomad_esp_aa,aaf_1kg_amr,aaf_1kg_eas,aaf_1kg_sas,aaf_1kg_afr,aaf_1kg_eur,aaf_adj_exac_afr,aaf_adj_exac_amr,aaf_adj_exac_eas,aaf_adj_exac_nfe,aaf_adj_exac_sas. and -1 if none of those databases/populations contain the variant.</td>
</tr>
<tr>
<td>exac_num_het</td>
<td>INTEGER</td>
<td>The number of heterozygote genotypes observed in ExAC. Pulled from the ExAC AC_Het INFO field.</td>
</tr>
<tr>
<td>exac_num_hom_alt</td>
<td>INTEGER</td>
<td>The number of homozygous alt. genotypes observed in ExAC. Pulled from the ExAC AC_Het INFO field.</td>
</tr>
<tr>
<td>exac_num_chroms</td>
<td>INTEGER</td>
<td>The number of chromosomes underlying the ExAC variant call. Pulled from the ExAC AN_Adj INFO field.</td>
</tr>
<tr>
<td>aaf_gnomad_all</td>
<td>FLOAT</td>
<td>Allele frequency (population independent) of the variant in gnomad,</td>
</tr>
<tr>
<td>aaf_gnomad_afr</td>
<td>FLOAT</td>
<td>Allele frequency (AFR population) of the variant in gnomad</td>
</tr>
<tr>
<td>aaf_gnomad_amr</td>
<td>FLOAT</td>
<td>Allele frequency (AMR population) of the variant in gnomad</td>
</tr>
<tr>
<td>aaf_gnomad_asj</td>
<td>FLOAT</td>
<td>Allele frequency (ASJ population) of the variant in gnomad</td>
</tr>
<tr>
<td>aaf_gnomad_eas</td>
<td>FLOAT</td>
<td>Allele frequency (EAS population) of the variant in gnomad</td>
</tr>
<tr>
<td>aaf_gnomad_fin</td>
<td>FLOAT</td>
<td>Allele frequency (FIN population) of the variant in gnomad</td>
</tr>
<tr>
<td>aaf_gnomad_nfe</td>
<td>FLOAT</td>
<td>Allele frequency (NFE population) of the variant in gnomad</td>
</tr>
<tr>
<td>aaf_gnomad_oth</td>
<td>FLOAT</td>
<td>Allele frequency (OTH population) of the variant in gnomad</td>
</tr>
<tr>
<td>aaf_gnomad_sas</td>
<td>FLOAT</td>
<td>Allele frequency (SAS population) of the variant in gnomad</td>
</tr>
<tr>
<td>gnomad_num_het</td>
<td>INTEGER</td>
<td>Number of het genotypes observed in gnomad</td>
</tr>
<tr>
<td>gnomad_num_hom_alt</td>
<td>INTEGER</td>
<td>Number of hom_alt genotypes observed in gnomad</td>
</tr>
<tr>
<td>gnomad_num_chroms</td>
<td>INTEGER</td>
<td>Number of chromosomes genotyped in gnomad</td>
</tr>
</tbody>
</table>
### Disease phenotype info (from ClinVar).

<table>
<thead>
<tr>
<th>Field</th>
<th>Type</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>in_omim</td>
<td>BOOL</td>
<td>0 : Absence of the variant in OMIM database</td>
</tr>
<tr>
<td></td>
<td></td>
<td>1 : Presence of the variant in OMIM database</td>
</tr>
<tr>
<td>clinvar_causal_allele</td>
<td>STRING</td>
<td>The allele(s) that are associated or causal for the disease.</td>
</tr>
<tr>
<td>clinvar_sig</td>
<td>STRING</td>
<td>The clinical significance scores for each variant according to ClinVar:</td>
</tr>
<tr>
<td></td>
<td></td>
<td><em>unknown</em>, <em>untested</em>, <em>non-pathogenic</em>, <em>probable-non-pathogenic</em>,</td>
</tr>
<tr>
<td></td>
<td></td>
<td><em>probable-pathogenic</em>, <em>pathogenic</em>, <em>drug-response</em>, <em>histocompatibility</em>,</td>
</tr>
<tr>
<td></td>
<td></td>
<td><em>other</em></td>
</tr>
<tr>
<td>clinvar_disease_name</td>
<td>STRING</td>
<td>The name of the disease to which the variant is relevant</td>
</tr>
<tr>
<td>clinvar_dbsource</td>
<td>STRING</td>
<td>Variant Clinical Channel IDs</td>
</tr>
<tr>
<td>clinvar_dbsource_id</td>
<td>STRING</td>
<td>The record id in the above database</td>
</tr>
<tr>
<td>clinvar_origin</td>
<td>STRING</td>
<td>The type of variant.</td>
</tr>
<tr>
<td></td>
<td></td>
<td><em>unknown</em>, <em>germline</em>, <em>somatic</em>, <em>inherited</em>, <em>paternal</em>, <em>maternal</em>,</td>
</tr>
<tr>
<td></td>
<td></td>
<td><em>de-novo</em>, <em>biparental</em>, <em>uniparental</em>, <em>not-tested</em>, <em>tested-inconclusive</em>,</td>
</tr>
<tr>
<td></td>
<td></td>
<td><em>other</em></td>
</tr>
<tr>
<td>clinvar_dsdb</td>
<td>STRING</td>
<td>Variant disease database name</td>
</tr>
<tr>
<td>clinvar_dsdbid</td>
<td>STRING</td>
<td>Variant disease database ID</td>
</tr>
<tr>
<td>clinvar_disease_acc</td>
<td>STRING</td>
<td>Variant Accession and Versions</td>
</tr>
<tr>
<td>clinvar_in_locus_spec_db</td>
<td>BOOL</td>
<td>Submitted from a locus-specific database?</td>
</tr>
<tr>
<td>clinvar_on_diag_assay</td>
<td>BOOL</td>
<td>Variation is interrogated in a clinical diagnostic assay?</td>
</tr>
<tr>
<td>clinvar_gene_phenotype</td>
<td>STRING</td>
<td>‘</td>
</tr>
<tr>
<td></td>
<td></td>
<td>variant in the same gene in clinvar not just the current variant).</td>
</tr>
<tr>
<td>geno2mp_hpo_ct</td>
<td>INTEGER</td>
<td>Value from geno2mp indicating count of HPO profiles. Set to -1 if missing</td>
</tr>
</tbody>
</table>
### Structural variation columns

<table>
<thead>
<tr>
<th>Column</th>
<th>Type</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>sv_cipos_start_left</td>
<td>INTEGER</td>
<td>The leftmost position of the leftmost SV breakpoint confidence interval.</td>
</tr>
<tr>
<td>sv_cipos_end_left</td>
<td>INTEGER</td>
<td>The rightmost position of the leftmost SV breakpoint confidence interval.</td>
</tr>
<tr>
<td>sv_cipos_start_right</td>
<td>INTEGER</td>
<td>The leftmost position of the rightmost SV breakpoint confidence interval.</td>
</tr>
<tr>
<td>sv_cipos_end_right</td>
<td>INTEGER</td>
<td>The rightmost position of the rightmost SV breakpoint confidence interval.</td>
</tr>
<tr>
<td>sv_length</td>
<td>INTEGER</td>
<td>The length of the structural variant in base pairs.</td>
</tr>
<tr>
<td>sv_is_precise</td>
<td>BOOL</td>
<td>Is the structural variant precise (i.e., to 1-bp resolution)?</td>
</tr>
<tr>
<td>sv_tool</td>
<td>STRING</td>
<td>The name of the SV discovery tool used to find the SV.</td>
</tr>
<tr>
<td>sv_evidence_type</td>
<td>STRING</td>
<td>What type of alignment evidence supports the SV?</td>
</tr>
<tr>
<td>sv_event_id</td>
<td>STRING</td>
<td>A unique identifier for the SV.</td>
</tr>
<tr>
<td>sv_mate_id</td>
<td>STRING</td>
<td>The ID for the “other end” of the SV.</td>
</tr>
<tr>
<td>sv_strand</td>
<td>STRING</td>
<td>The orientations of the SV breakpoint(s).</td>
</tr>
</tbody>
</table>
### Genome annotations

<table>
<thead>
<tr>
<th>Annotation</th>
<th>Type</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>exome_chip</td>
<td>BOOL</td>
<td>Whether a SNP is on the Illumina HumanExome Chip</td>
</tr>
<tr>
<td>cyto_band</td>
<td>STRING</td>
<td>Chromosomal cytobands that a variant overlaps</td>
</tr>
<tr>
<td>rmsk</td>
<td>STRING</td>
<td>A comma-separated list of RepeatMasker annotations that the variant overlaps. Each hit is of the form: name_class_family</td>
</tr>
<tr>
<td>in_cpg_island</td>
<td>BOOL</td>
<td>Does the variant overlap a CpG island?</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Based on UCSC: Regulation &gt; CpG Islands &gt; cpgIslandExt</td>
</tr>
<tr>
<td>in_segdup</td>
<td>BOOL</td>
<td>Does the variant overlap a segmental duplication?</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Based on UCSC: Variation&amp;Repeats &gt; Segmental Dups &gt; genomicSuperDups track</td>
</tr>
<tr>
<td>is_conserved</td>
<td>BOOL</td>
<td>Does the variant overlap a conserved region?</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Based on the 29-way mammalian conservation study</td>
</tr>
<tr>
<td>gerp_bp_score</td>
<td>FLOAT</td>
<td>GERP conservation score. Only populated if the --load-gerp-bp option is used when loading.</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Higher scores reflect greater conservation. At base-pair resolution.</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Details: <a href="http://mendel.stanford.edu/SidowLab/downloads/gerp/">http://mendel.stanford.edu/SidowLab/downloads/gerp/</a></td>
</tr>
<tr>
<td>gerp_element_pval</td>
<td>FLOAT</td>
<td>GERP elements P-val. Lower P-values scores reflect greater conservation. Not at base-pair resolution.</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Details: <a href="http://mendel.stanford.edu/SidowLab/downloads/gerp/">http://mendel.stanford.edu/SidowLab/downloads/gerp/</a></td>
</tr>
<tr>
<td>recomb_rate</td>
<td>FLOAT</td>
<td>Returns the mean recombination rate at the variant site</td>
</tr>
</tbody>
</table>
Note: CADD scores (http://cadd.gs.washington.edu/) are Copyright 2013 University of Washington and Hudson-Alpha Institute for Biotechnology (all rights reserved) but are freely available for all academic, non-commercial applications. For commercial licensing information contact Jennifer McCullar (mccullaj@uw.edu).

## Variant error assessment

<table>
<thead>
<tr>
<th>Field</th>
<th>Type</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>gms_solid</td>
<td>FLOAT</td>
<td>Genome Mappability Scores with SOLiD error models</td>
</tr>
<tr>
<td>gms_iontorrent</td>
<td>FLOAT</td>
<td>Genome Mappability Scores with IonTorrent error models</td>
</tr>
<tr>
<td>in_cse</td>
<td>BOOL</td>
<td>Is a variant in an error prone genomic position, using CSE: Context-Specific Sequencing Errors <a href="https://code.google.com/p/discovering-cse/">https://code.google.com/p/discovering-cse/</a> <a href="http://www.biomedcentral.com/1471-2105/14/S5/S1">http://www.biomedcentral.com/1471-2105/14/S5/S1</a></td>
</tr>
</tbody>
</table>
## ENCODE information

<table>
<thead>
<tr>
<th>Field</th>
<th>Type</th>
<th>Description</th>
</tr>
</thead>
</table>
| encode_tfb                 | STRING  | Comma-separated list of transcription factors that were observed by ENCODE to bind DNA in this region. Each hit in the list is constructed as `TF_CELLCOUNT`, where:  
|                             |          | `TF` is the transcription factor name  
|                             |          | `CELLCOUNT` is the number of cells tested that had nonzero signals. Provenance: wgEncodeRegTfbsClusteredV2 UCSC table |  
| encode_dnasel_cell_count   | INTEGER | Count of cell types that were observed to have DnaseI hypersensitivity.                                                                                                                                  |
| encode_dnasel_cell_list     | STRING  | Comma-separated list of cell types that were observed to have DnaseI hypersensitivity. Provenance: Thurman, et al, *Nature*, 489, pp. 75-82, 5 Sep. 2012 |  
| encode_consensus_gm12878   | STRING  | ENCODE consensus segmentation prediction for GM12878. CTCF: CTCF-enriched element E: Predicted enhancer PF: Predicted promoter flanking region R: Predicted repressed or low-activity region TSS: Predicted promoter region including TSS T: Predicted transcribed region WE: Predicted weak enhancer or open chromatin cis-regulatory element |  
| encode_consensus_h1hesc     | STRING  | ENCODE consensus segmentation prediction for h1HESC. See encode_consseg_gm12878 for details.                                                                                                           |
| encode_consensus_helas3     | STRING  | ENCODE consensus segmentation prediction for Helas3. See encode_consseg_gm12878 for details.                                                                                                            |
| encode_consensus_hepg2      | STRING  | ENCODE consensus segmentation prediction for HEPG2. See encode_consseg_gm12878 for details.                                                                                                             |
| encode_consensus_huvec      | STRING  | ENCODE consensus segmentation prediction for HuVEC. See encode_consseg_gm12878 for details.                                                                                                              |
| encode_consensus_k562       | STRING  | ENCODE consensus segmentation prediction for k562. See encode_consseg_gm12878 for details.                                                                                                             |
| vista_enhancers             | STRING  | Experimentally validated human enhancers from VISTA (http://enhancer.lbl.gov/frnt_page_n.shtml)                                                                                                           |
## Cancer related columns

<table>
<thead>
<tr>
<th>Column</th>
<th>Type</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>is_somatic</td>
<td>BOOL</td>
<td>Whether the variant is somatically acquired.</td>
</tr>
<tr>
<td>cosmic_ids</td>
<td>STRING</td>
<td>A list of known COSMIC ids for this variant.</td>
</tr>
<tr>
<td></td>
<td></td>
<td>See: <a href="http://cancer.sanger.ac.uk/cancergenome/projects/cosmic/">http://cancer.sanger.ac.uk/cancergenome/projects/cosmic/</a></td>
</tr>
</tbody>
</table>
## The variant_impacts table

<table>
<thead>
<tr>
<th>column_name</th>
<th>type</th>
<th>notes</th>
</tr>
</thead>
<tbody>
<tr>
<td>variant_id</td>
<td>INTEGER</td>
<td>PRIMARY_KEY (Foreign key to variants table)</td>
</tr>
<tr>
<td>anno_id</td>
<td>INTEGER</td>
<td>PRIMARY_KEY (Based on variant transcripts)</td>
</tr>
<tr>
<td>gene</td>
<td>STRING</td>
<td>The gene affected by the variant.</td>
</tr>
<tr>
<td>transcript</td>
<td>STRING</td>
<td>The transcript affected by the variant.</td>
</tr>
<tr>
<td>is_exonic</td>
<td>BOOL</td>
<td>Does the variant affect an exon for this transcript?</td>
</tr>
<tr>
<td>is_coding</td>
<td>BOOL</td>
<td>Does the variant fall in a coding region (excludes 3' &amp; 5' UTR's of exons)?</td>
</tr>
<tr>
<td>is_lof</td>
<td>BOOL</td>
<td>Based on the value of the impact col, is the variant LOF?</td>
</tr>
<tr>
<td>exon</td>
<td>STRING</td>
<td>Exon information for the variants that are exonic</td>
</tr>
<tr>
<td>codon_change</td>
<td>STRING</td>
<td>What is the codon change?</td>
</tr>
<tr>
<td>aa_change</td>
<td>STRING</td>
<td>What is the amino acid change?</td>
</tr>
<tr>
<td>aa_length</td>
<td>STRING</td>
<td>The length of CDS in terms of number of amino acids (SnpEff only)</td>
</tr>
<tr>
<td>biotype</td>
<td>STRING</td>
<td>The type of transcript (e.g., protein-coding, pseudogene, rRNA etc.) (SnpEff only)</td>
</tr>
<tr>
<td>impact</td>
<td>STRING</td>
<td>Impacts due to variation (ref.impact category)</td>
</tr>
<tr>
<td>impact_so</td>
<td>STRING</td>
<td>The sequence ontology term for the impact</td>
</tr>
<tr>
<td>impact_severity</td>
<td>STRING</td>
<td>Severity of the impact based on the impact column value (ref.impact category)</td>
</tr>
<tr>
<td>polyphen_pred</td>
<td>STRING</td>
<td>Impact of the SNP as given by PolyPhen (VEP only) benign, possibly_damaging, probably_damaging, unknown</td>
</tr>
<tr>
<td>polyphen_scores</td>
<td>FLOAT</td>
<td>Polyphen score reflecting severity (higher the impact, higher the score) (VEP only)</td>
</tr>
<tr>
<td>sift_pred</td>
<td>STRING</td>
<td>Impact of the SNP as given by SIFT (VEP only) neutral, deleterious</td>
</tr>
<tr>
<td>sift_scores</td>
<td>FLOAT</td>
<td>SIFT prob. scores reflecting severity (Higher the impact, lower the score) (VEP only)</td>
</tr>
</tbody>
</table>
### Details of the impact and impact_severity columns

<table>
<thead>
<tr>
<th>impact severity</th>
<th>impacts</th>
<th>SO_impacts</th>
</tr>
</thead>
</table>
| HIGH            | • exon_deleted  
• frame_shift  
• splice_acceptor  
• splice_donor  
• start_loss  
• stop_gain  
• stop_loss  
• non_synonymous_start  
• transcript_codon_change  
• rare_amino_acid  
• chrom_large_del  | • exon_loss_variant  
• frameshift_variant  
• splice_acceptor_variant  
• splice_donor_variant  
• start_lost  
• stop_gained  
• stop_lost  
• initiator_codon_variant  
• initiator_codon_variant  
• rare_amino_acid_variant  
• chromosomal_deletion  |
| MED             | • non_syn_coding  
• inframe_codon_gain  
• inframe_codon_loss  
• inframe_codon_change  
• codon_change_del  
• codon_change_ins  
• UTR_5_del  
• UTR_3_del  
• splice_region  
• mature_miRNA  
• regulatory_region  
• TF_binding_site  
• regulatory_region_ablation  
• regulatory_region_amplification  
• TFBS_ablation  
• TFBS_amplification  | • missense_variant  
• inframe_insertion  
• inframe_deletion  
• coding_sequence_variant  
• disruptive_inframe_deletion  
• disruptive_inframe_insertion  
• 5_prime_UTR_truncation  
• exon_loss_variant  |
| LOW             | • synonymous_stop  
• synonymous_coding  
• UTR_5_prime  
• UTR_3_prime  
• intron  
• CDS  
• upstream  
• downstream  
• intergenic  
• intragenic  
• gene  
• transcript  
• exon  
• start_gain  
• synonymous_start  
• intron_conserves  
• nc_transcript  
• NMD_transcript  
• incomplete_terminal_codon  | • stop_retained_variant  
• synonymous_variant  
• 5_prime_UTR_variant  
• 3_prime_UTR_variant  
• intron_variant  
• coding_sequence_variant  
• upstream_gene_variant  
• downstream_gene_variant  
• intergenic_variant  
• intragenic_variant  
• gene_variant  
• transcript_variant  
• exon_variant  
• start_retained_variant  
• conserved_intron_variant  
• nc_transcript_variant  
• NMD_transcript_variant  |

5.9. The GEMINI database schema
The samples table

<table>
<thead>
<tr>
<th>column name</th>
<th>type</th>
<th>notes</th>
</tr>
</thead>
<tbody>
<tr>
<td>sample_id</td>
<td>INTEGER</td>
<td>PRIMARY_KEY</td>
</tr>
<tr>
<td>name</td>
<td>STRING</td>
<td>Sample names</td>
</tr>
<tr>
<td>family_id</td>
<td>INTEGER</td>
<td>Family ids for the samples [User defined, default: NULL]</td>
</tr>
<tr>
<td>paternal_id</td>
<td>INTEGER</td>
<td>Paternal id for the samples [User defined, default: NULL]</td>
</tr>
<tr>
<td>maternal_id</td>
<td>INTEGER</td>
<td>Maternal id for the samples [User defined, default: NULL]</td>
</tr>
<tr>
<td>sex</td>
<td>STRING</td>
<td>Sex of the sample [User defined, default: NULL]</td>
</tr>
<tr>
<td>phenotype</td>
<td>STRING</td>
<td>The associated sample phenotype [User defined, default: NULL]</td>
</tr>
<tr>
<td>ethnicity</td>
<td>STRING</td>
<td>The ethnic group to which the sample belongs [User defined, default: NULL]</td>
</tr>
</tbody>
</table>

The resources table

Establishes provenance of annotation resources used to create a GEMINI database.

<table>
<thead>
<tr>
<th>column name</th>
<th>type</th>
<th>notes</th>
</tr>
</thead>
<tbody>
<tr>
<td>name</td>
<td>STRING</td>
<td>Name of the annotation type</td>
</tr>
<tr>
<td>resource</td>
<td>STRING</td>
<td>Filename of the resource, with version information</td>
</tr>
</tbody>
</table>

The version table

Establishes which version of gemini was used to create a database.

<table>
<thead>
<tr>
<th>column name</th>
<th>type</th>
<th>notes</th>
</tr>
</thead>
<tbody>
<tr>
<td>version</td>
<td>STRING</td>
<td>What version of gemini was used to create the DB.</td>
</tr>
</tbody>
</table>

The gene_detailed table

Built on version 75 of Ensembl genes
<table>
<thead>
<tr>
<th>column_name</th>
<th>type</th>
<th>notes</th>
</tr>
</thead>
<tbody>
<tr>
<td>uid</td>
<td>INTEGER</td>
<td>PRIMARY_KEY (unique identifier for each entry in the table)</td>
</tr>
<tr>
<td>chrom</td>
<td>STRING</td>
<td>The chromosome on which the gene resides</td>
</tr>
<tr>
<td>gene</td>
<td>STRING</td>
<td>The gene name</td>
</tr>
<tr>
<td>is_hgnc</td>
<td>BOOL</td>
<td>Flag for gene column: 0 for non HGNC symbol and 1 for HGNC symbol = TRUE</td>
</tr>
<tr>
<td>ensembl_gene_id</td>
<td>STRING</td>
<td>The ensembl gene id for the gene</td>
</tr>
<tr>
<td>transcript</td>
<td>STRING</td>
<td>The ensembl transcript id for the gene</td>
</tr>
<tr>
<td>biotype</td>
<td>STRING</td>
<td>The biotype (e.g., protein coding) of the transcript</td>
</tr>
<tr>
<td>transcript_status</td>
<td>STRING</td>
<td>The status of the transcript (e.g. KNOWN, PUTATIVE etc.)</td>
</tr>
<tr>
<td>ccds_id</td>
<td>STRING</td>
<td>The consensus coding sequence transcript identifier</td>
</tr>
<tr>
<td>hgnc_id</td>
<td>STRING</td>
<td>The HGNC identifier for the gene if HGNC symbol is TRUE</td>
</tr>
<tr>
<td>entrez_id</td>
<td>STRING</td>
<td>The entrez gene identifier for the gene</td>
</tr>
<tr>
<td>cds_length</td>
<td>STRING</td>
<td>The length of CDS in bases</td>
</tr>
<tr>
<td>protein_length</td>
<td>STRING</td>
<td>The length of the transcript as the number of amino acids</td>
</tr>
<tr>
<td>transcript_start</td>
<td>STRING</td>
<td>The start position of the transcript in bases</td>
</tr>
<tr>
<td>transcript_end</td>
<td>STRING</td>
<td>The end position of the transcript in bases</td>
</tr>
<tr>
<td>strand</td>
<td>STRING</td>
<td>The strand of DNA where the gene resides</td>
</tr>
<tr>
<td>synonym</td>
<td>STRING</td>
<td>Other gene names (previous or synonyms) for the gene</td>
</tr>
<tr>
<td>rvis_pct</td>
<td>FLOAT</td>
<td>The RVIS percentile values for the gene</td>
</tr>
</tbody>
</table>

**The gene_summary table**

Built on version 75 of Ensembl genes

5.9. The GEMINI database schema
Using the GEMINI API

The GeminiQuery class

Extracting the VCF INFO tags with GEMINI API

The GEMINI API is useful to extract the individual tags within the INFO field of a VCF (stored as a compressed dictionary in the variants table). This would be of particular interest to those who want to add custom annotations to their VCF and still be able to access the individual tags programmatically. Here is an example where we try to extract the dbNSFP fields from the ‘INFO’ tag of a VCF, using the API.

```python
#!/usr/bin/env python
import sys
from gemini import GeminiQuery
```
from sys import argv

database = argv[1]
gq = GeminiQuery(database)
query = "SELECT variant_id, chrom, start, end, ref, alt, info FROM variants"

gq.run(query)

for row in gq:
    try:
        print '\t'.join([str(row['chrom']), str(row['start']), str(row['end']), str(row['ref']), str(row['alt']), str(row.info['dbNSFP_SIFT'])])
    except KeyError:
        pass

# yields
chr1 906272 906273 C T P|D|P
chr1 906273 906274 C A D|D|D
chr1 906276 906277 T C D|D|D
chr1 906297 906298 G T B|B|B
chr1 1959074 1959075 A C D
chr1 1959698 1959699 G A B
chr1 1961452 1961453 C T P
chr1 2337953 2337954 C T D

**Speeding genotype queries**

**Design of Gemini**

Gemini stores all genotypes information in a single column. E.g. \textit{gt\_depths} is a single column in the sqlite database that contains information on all samples in a compressed and serialized array. So, in order to do a query involving \texttt{--gt\_filter gemini} must iterate over each row, decompress and de-serialize the array and then evaluate the genotype filter. Even if the filter involves only a single sample, we must deserialize the entire array. This design works quite well and we have improved performance greatly, but for complex queries, it is quite slow.

We provide the means to index genotype fields and \textit{optionally} use those indexes to perform the genotype filtering. These are external to gemini in that they do not change the behavior of \textit{gemini} when used without the engine.

**bcolz indexes**

We have a implemented indexes using \texttt{bcolz}.

These can be used for existing databases by creating the external indexes:

```
gemini bcolz\_index $db
```

This is easily parallelized by specifying a column per process, e.g.

```
gemini bcolz\_index $db --cols gt\_types
```

```
gemini bcolz\_index $db --cols gt\_depths # in another process
```

Which can index about 20K variants / second for 17 samples in our testing. Due to a limitation in the string type in \texttt{bcolz}, indexing the \texttt{gts} column is slower so we recommend not doing those queries with \texttt{bcolz} (e.g. \texttt{gts\_sample1 == 'A/C'}). See note below.

5.11. Speeding genotype queries 95
It is recommended to only index the columns you’ll be using in the \textit{--gt-filter}.

Indexing is done only once; after that, add \textit{--use-bcolz} to an existing gemini query command. e.g.

\begin{verbatim}
gemini query -q "select variant_id, gts.1719PC0016 from variants" \ 
   --gt-filter "gt_types.1094PC0012 == HET and gt_types.1719PC0016 == HET and gts.\rightarrow1094PC0012 == 'A/C'" \ 
   --use-bcolz \test/test.query.db
\end{verbatim}

This query will return identical results with or without using bcolz. It is likely only beneficial to use \textit{bcolz} on complex queries that are slow with the default gemini apparatus. Queries that are more selective—e.g. return fewer rows are the best target for speed improvement with \textit{bcolz}.

An example of the types of improvements (usually 20X to 50X) with various queries is here.

\textbf{limitations}

As the number of samples grows, it becomes less beneficial to use \textit{--gt-filter} s that touch all samples. For example: \texttt{(gt_types).(\!HOM\_REF).(all)} will become slower as samples are added since it must test every sample. However, a query like: \texttt{(gt_types.DAD == HOM\_REF and gt_types.MOM == HOM\_REF and gt_types.KID \neq HOM\_REF)} will be much less affected by the number of samples in the database because they are only touching 3 samples.

The image below shows the time to perform the filters on a database with 1.8 million variants and varying sample size:

1. \texttt{(gt_types).(\!HOM\_REF).(all)} which is affected by sample size
2. \texttt{(gt_types.DAD == HOM\_REF and gt_types.MOM == HOM\_REF and gt_types.KID \neq HOM\_REF)} which is less affected by sample-size

Note that at 2500 samples, using bcolz is slower than the standard gemini query, however using bcolz is consistently 30 to 50 times faster for the 2nd query. (This is up to 1000 times faster than versions of gemini before 0.12). The y-axis is log10-scaled.
Note: indexing the ‘gts’ column will be much slower (only about 350 variants per second instead of up to 25K per second) as it must be stored as an object rather than a fixed-size numeric type. It will be a much larger index. So only create an index on ‘gts’ if necessary.

Usage

First, make sure you have gemini version 0.15 or greater:

```
gemini --version
```

Then, you can index an existing database ($db) with:
Then, wherever you have a slow query that’s using the \texttt{--gt-filter}, you can add \texttt{--use-bcolz} to the query command.

\begin{verbatim}
gemini query -q "select chrom, start, end from variants" $db \  
    --use-bcolz \  
    --gt-filter "gt_depth.samples1 >= 20 and gt_depth.samples2 >= 20 and gt_depth. samples3 >= 20 \  
    and (gt_types.sample1 == HOM_REF and gt_types.sample2 == HOM_REF and gt_ \  
    --types.sample3 != HOM_REF)"
\end{verbatim}

Note that nothing has changed except that \texttt{--use-bcolz} is added to the query.

\section*{Design of Genotype Query Engines}

This sections is for those wishing to create their own genotype query engines to plug in to gemini and will not be needed for most users.

Genotype Query Engines can be plugged in to \texttt{gemini}. They must be exposed with a single function:

\begin{verbatim}
def filter(db_path, gt_filter, user_dict)
\end{verbatim}

where \texttt{db_path} is the path to the gemini sqlite database, \texttt{gt_filter} is the genotype query string. \texttt{user_dict} will be pre-filled with things like \texttt{user_dict} contains things like HET, UNKNOWN, etc. used in gemini.

The \texttt{filter} function must return a list of integer variant_ids that meet the specified filter. If it can not perform the query, it must return \texttt{None}.

\texttt{gemini} will internally use the returned variant_ids to modify the sqlite query to select only those rows.

The \texttt{filter} function only needs to worry about which variant_ids to return, not how to integrate with the rest of \texttt{gemini}.

\section*{Acknowledgements}

GEMINI is developed by Uma Paila and Aaron Quinlan in the Quinlan laboratory at the University of Virginia. Substantial contributions to the design, functionality, and code base have been made by the following:

- Brad Chapman, HSPH
- Rory Kirchner, HSPH
- Oliver Hofmann, HSPH
- Björn Grüning, University of Freiburg

\section*{Release History}

\subsection*{0.21.0 (future)}

1. Use vcfanno for faster, more generalized variant annotation and database creation.
0.20.1

1. Allow `–gt-filter` to specify the gt_type, e.g. `(gt_quals).=(HET).(<20).all` (thanks Jessica for suggesting)
2. Add `–gene-where` argument to `comp_hets` tool to dictate how variants are restricted to genes. (thanks Jessica for suggesting)
3. genewise: don’t show any variants that don’t meet the required number of optional filters.
4. Fix AF for decomposed gnomad variants (thanks @mbootwalla).

0.20.0

1. Include gnomad exome allele frequencies for all populations and include these in calc of max_aaf_all.
2. Initial support for python 3.
3. Support annotating with bcftools BCSQ (http://biorxiv.org/content/early/2016/12/01/090811) in addition to VEP and SnpEff.
4. Fix bug that resulted in the message: `UnsupportedOperation: IOSTream has no fileno.` when loading. (Thanks Andrew O. and Ryan R.)
5. Fix bug where gemini annotate would never finish (see: arq5x/gemini#809 (Thanks @mmoisse)
6. Add extra columns from VEP when -t all is used (previously extra VEP columns were ignored under -t all. thanks @hoppman #816).
7. Fix gemini annotate (would stop after updating 100K variants)
8. Fix browser query report (thanks @Nmael see #818).
9. Allow specifying a `–where` clause for `gene_wise` to filter considered variants. (Thanks Uma for suggestion).
10. Fix bug in propagating vep-extras to variant_impacts table (Thanks Sergey for reporting).
11. Update clinvar annotation.
12. Add new `gemini load` flag `–skip-pls` which can give size reductions for the database and speed Improvements for querying and loading. If you never use `gt_phred_ll_homalt/het/homref` then you can load with this flag.
13. Add new `gt_alt_freqs` column which is `gt_alt_depths / (gt_ref_depths + gt_alt_depths)`.
14. Update CADD to v1.3

0.19.1

1. Fix bug in `comp_het` where candidates were not removed when either parent was homozygous at both sites (thanks Jessica Chong). (inheritance v0.1.0)
2. Update `x-linked-dominant` with the following rules (thanks Jessica Chong): + mothers of affected males must be het (and affected) + at least 1 parent of affected females must be het (and affected).
3. Store extra vep fields for each transcript in variant_impacts.
4. Update geneimpacts module to handle new snpEff annotations of SVs (Brad Chapman).
5. Update to dbsnp 147.
0.19.0

1. Use SQLAlchemy for table definitions to support different RDBMS back-ends.
2. Several optimizations to loading.
3. X-linked recessive, dominant, and de novo tools.
4. Raise exceptions rather than sys.exit() to facilitate use as library. (thanks @brainstorm).
5. The comp_het tool is as much as 20X faster for large cohorts.

0.18.3

1. handle rare VEP annotation with ‘’ or “?” as impact
2. fix handling of multiple values for list/uniq_list in gemini_annotate
3. handle snpEff with unknown impact (TODO: bump geneimpacts req to 0.1.0)
4. fix builtin browser and add support for puzzle browser
5. add --gt-filter-required to gemini gene_wise tool that must pass for each variant.
6. fix bug in lof_sieve when transcript pos is not specified. (thanks @mmoisse)

0.18.2

1. Update the Clinvar annotation file to the February 3, 2016 release.

0.18.1

1. Add clinvar_gene_phenotype column which can be used to limit candidates to those that are in the same gene as a gene with the known phenotype from clinvar. Phenotypes are all lower case. Likely usage is: --filter "... and clinvar_gene_phenotype LIKE '%dysplasia%' where the ‘%’ wildcard is needed because the column is a ‘|’ delimited list of all disease for that gene.
2. Add geno2mp_hpo_ct column which will be > 0 if that variant is present in geno2mp (http://geno2mp.gs.washington.edu/Geno2MP/#/)
3. Fix comp_hets tool when exac_num_hom_alt was requested. (Thanks to @davemcg for reporting).
4. Change splice_region_variant from LOW to MED priority in the interest of reducing false negatives in the study of rare disease.
5. Set missing AF to -1 (instead of NULL) for unknown for all ESP, 1KG, ExAC allele frequency columns.
6. Don’t user order by when not needed for built in tools. Speeds up queries when min-kindreds is None or 1.
7. Document the --min-gq option
8. Fix gemini load with -t all when only VEP is present.
9. Fix bug in gemini annotate where numeric operations on integers did not work for VCF.
10. Fix bug in comp_hets tool where attempting to phase a “.” genotype would result in an error. (Thanks to Aparna and Martina for reporting)
11. Fix bug in mendel_errors tool when specifying --families. Thanks to @davemcg for reporting.
0.18.0

0. Improved installation and update via conda (via Brad Chapman!).

1. Update support for SO term variant impact predictions via VEP and SnpEff and support newest snpEff version. If both snpEff and VEP annotations are present use `gemini load -t all` to save all annotations and the most deleterious will be saved in the variants table.

2. Add an `is_splicing` column.

3. Add `exac_num_het`, `exac_num_hom_alt` and `exac_num_chroms` columns. (See #568).

4. Fix issues with cyvcf2 handling of haploid calls (from X chromosome from GATK; thanks Athina for reporting)

5. Fix handling of VEP extra fields (thanks @jsh58).

6. Remove pygraph dependency (networkx is easier to install). Allow specifying custom edges to interactions tools.

7. Fix some edge-cases in compound het tool. Thanks to Jamie Kwok for reporting.

8. Allow requiring a minimum genotype quality in inheritance tools with –min-gq option.

9. Fix bug in cyvcf2 for parsing hemizyous (0/.) variants from decomposition.

10. `comp_het` tool now has a priority 2 level for unphased singletons where both sites are HET. Candidates where both parents and HET, along with the affected are now priority 3.

0.17.2

1. Fix bcolz dependencies owing to issues with bcolz 0.11.0

0.17.1

1. Change handling of missing values in PL/GL (–gt-pl-max) so that missing values are set to int32 max. (Thanks Karyn for reporting).

2. Fix distributed loading of VEP with extra columns (@chapmanb) [Regression since 0.17.0]

3. Fix comp_het test and improve efficiency (thanks Bianca for reporting)


0.17.0

1. switch to cyvcf2 to speed loading

2. per-sample depths are calculated using AD (GATK) or AO+RO (Freebayes). This makes depth filters more conservative.

3. extra VEP annotations are loaded with loading machinery, not as an extra step as before.

4. add `max_aaf_all` column ([https://github.com/arq5x/gemini/issues/520](https://github.com/arq5x/gemini/issues/520)) as an aggregate of a number of population filters.

5. use –families to limit queries before any work is done. Thanks to Bianca for reporting.

6. No longer create bcolz indicies by default. Users can create them with `gemini bcolz_index`.

7. New `genewise` tool. See docs.
8. Gemini load: --skip-info-string has been replaced with --save-info-string and the INFO field is no longer saved by default.

9. Comp_hets: default to only showing confident (priority 1) candidates. Show all candidates with --max-priority 3.

0.16.3

1. Fix bug in comp_het with reporting same pair multiple times.
2. Handle UNKOWN genotypes in comp_het tool
3. Fix cyvcf dependency in requirements
4. Only run tests that require bgzip/tabix/bedtools if they are available on PATH
5. Limit ipython version to 3<version<4

0.16.2

1. Hone rules for unphased and partially-phased compound hets.
2. Remove --lenient argument for comp_hets and add --pattern-only to find compound_hets regardless of affection status.
3. The --lenient argument to the autosomal_dominant tool has been relaxed to allow parents with unknown phenotypes.
4. Re (vt) decompose data files for 1000 genomes and ExAC (thanks Julien and Xiaolin for reporting).

0.16.1

1. Fix regression in loading when AAF is None
2. Fix handing in mendelian error tool where all genotype likelihoods are low (thanks Bianca)
3. Don’t phase de-novo’s (caused error in comp_het tool). (thanks Bianca)
4. Fix regression in loading VEP with multicores (thanks Andrew)

0.16.0

1. The built-in inheritance model tools (auto_rec, etc.) have been modified to be more restrictive in order to remove false positive candidates. The strictness can be reduced by using the --lenient option.
2. Leverage bcolz indexing for the built-in inheritance model tools to dramatically improve speed.
3. Support for multi-generational pedigrees for the built-in inheritance model tools. (thanks to Jessica, Andrew, and jmcelwee for extensive discussion https://github.com/arq5x/gemini/issues/388)
4. Leverage genotype likelihoods in tools other than mendel_errors as a means to filter variants.
5. Automatically phase genotypes by transmission on the fly for the comp_hets tool.
6. Further performance improvements for bcolz queries
7. The --affected-only option has been made the default and it’s opposing replacement named --allow-unaffected to revert.
8. Fixed a reporting error for the inheritance tools (i.e., family_id was mis-specified in output).
9. Annotate the variants table with impact even if there is not severe impact. Thanks to @mjsduncan for reporting.
10. Reduce memory requirements when loading. Thanks to @mjsduncan for reporting.

0.15.1

1. Fix regression in grabix. Thanks to Sven-Eric Shelhorn for reporting.
2. Fix handling of samples with "-". Thanks to Uma Paila for reporting.

0.15.0

1. Use external index to speed genotype queries (this is created by default on load unless –no-bcolz is specified)
2. Match on ref and alternate alleles (not just position) when annotating with VCF. Thanks Jeremy Goecks.
3. Related to matching, we now load extra annotation, e.g. VEP as VCF and require ref and alt matching. Previously was done with bed overlap.
4. Faster queries due to lazy loading of genotype columns.
5. Read gt* columns from the database for better backward compatibility.

0.14.0

1. Standardized the output from the built-in tools into a common, BED+ format. Thanks to feedback from Jessica Chong and Daniel Gaston.
2. Release of mendel_errors tool which also outputs the type of error and the probability (based on PL’s)
3. Improvements to the load tool when running on large compute clusters using PBS, SGE, SLURM, etc. Also provide a workaround for NFS locking issues. Many thanks to Ben Weisburd in Daniel Macarthur’s lab.
4. Improve preprocess script to support varscan, platypus (https://gist.github.com/brentp/4db670df147cbd5a2b32)
5. Performance improvements for many of the built-in tools (pre-compile evals)

0.13.1 (2015-Apr-09)

1. Major query speed improvements. For example, the following query goes from 43 seconds in version 0.12.2 to 11 seconds in 0.13.0. All queries involving gt_* fields should be substantially faster.

```
$ gemini query \
    -q "select chrom, start, (gts).(*) from variants" data/tmaster.db \ 
    --gt-filter "((gt_depths).(*).>=20).(all)" > /dev/null
```

2. Speed improvements to load. The following went from 7 minutes 9 seconds to 6 minutes 21 seconds.

```
$ gemini load -t VEP -v data/v100K.vcf.gz data/tmaster.db --cores 4
```
3. We added the `gt_phred_ll_homref`, `gt_phred_ll_het`, `gt_phred_ll_homalt` columns to database. These are the genotype likelihoods pulled from the GL or PL columns of the VCF if available. They can all be queried and filtered in the same way as existing `gt_*` columns. In future releases, we are planning to use genotype likelihood to assign likelihoods to de novo mutations, mendelian violations, and variants meeting other inheritance patterns.

4. Fixed bugs related to splitting multiple alts (thanks to @jdh237)

5. We are working to improve development and release testing. This is ongoing, but now we support `gemini_install.py --version unstable` so that users can try out the latest changes and help with testing before releases. `gemini_update` is still limited to `master` as the most recent version.

6. Update cyvcf so it doesn’t error when AD tag is used for non-list data.

7. Fix regression in cyvcf to handle Flags in info field. (Thanks to Jon for reporting)

8. Improvements to install related to PYTHONHOME and other env variables (@chapmanb & @bw2)

0.12.2

Corrected a stale .c file in the cyvcf library. This is effectively a replacement for the 0.12.1 release.

0.12.1

1. Support for input VCF files containing variants with multiple alternate alleles. Thanks to Brent Pedersen.

2. Updated, decomposed, and normalized the ExAC, Clinvar, Cosmic, dbsNP, and ESP annotation files to properly support variants with multiple alternate alleles.

3. Updated the logic for the clinvar significance column to retain all documented significances.

4. Support for VCF annotation files in the `annotate` tool.

5. Improved the speed of loading by 10-15%. Thanks to Brent Pedersen.

6. Added `--only-affected` and `--min-kindreds` options to the compound heterozygotes tool.

7. Added a `--format vcf` option to the `query` tool to output query results in VCF format.

8. Added the `--families` option to the `auto_*`, `de_novo`, and `comp_hets` tools. Thanks to Mark Cowley and Tony Roscioli.

9. Added the `--only-affected` option to the `de_novo` tool.

10. Allow the `--sample-filter` to work with `--format TPED`. Thanks to Rory Kirchner.

11. Add `--format sampledetail` option that provides a melted/tidy/flattened version of samples along with `--showsample` and includes information from samples table. Thanks to Brad Chapman.

12. Add ‘not’ option to `--in` filtering. Thanks to Rory Kirchner.

13. Fixed a bug in the `de_novo` tool that prevented proper function when families have affected and unaffected children. Thanks to Andrew Oler.

14. Fixed a bug in cyvcf that falsely treated ‘.l’ genotypes as homozygous alternate. Thanks to Xiao Xu.

15. GEMINI now checks for and warns of old grabix index files. Thanks to Andrew Oler and Brent Pedersen.

16. Fixed a bug that added newlines at the end of tab delimited PED files. Thanks to Brad Chapman.
0.11.0

1. Integration of ExAC annotations (v0.2): http://exac.broadinstitute.org/
2. New tools for cancer genome analysis. Many thanks to fantastic work from Colby Chiang.
   - `gemini set_somatic`
   - `gemini actionable_mutations`
   - `gemini fusions`
3. Improved support for structural variants. New columns include:
   - `sv_cipos_start_left`
   - `sv_cipos_end_left`
   - `sv_cipos_start_right`
   - `sv_cipos_end_right`
   - `sv_length`
   - `sv_is_precise`
   - `sv_tool`
   - `sv_evidence_type`
   - `sv_event_id`
   - `sv_mate_id`
   - `sv_strand`
4. Updated the 1000 Genomes annotations to the Phase variant set.
5. Added `clinvar_causal_allele` column.
6. Fixed a bug in grabix that caused occasional duplicate and missed variants.

0.10.1

1. Add fitCons <http://biorxiv.org/content/early/2014/09/11/006825> scores as an additional measure of potential function in variants of interest, supplementing existing CADD and dbNSFP approaches.
2. Updated Clinvar, COSMIC, and dbSNP to their latest versions.

0.10.0

1. Provide an `--annotation-dir` argument that specifies the path the annotation databases, to overwrite configured data inputs. Thanks to Björn Grüning,
2. Support reproducible versioned installs of GEMINI with Python dependencies. Enables Galaxy integration. Thanks to Björn Grüning,

0.8.0

1. Support arbitrary annotation supplied to VEP, which translate into queryable columns in the main variant table.
2. Improve the power of the genotype filter wildcard functionality.
0.7.1

1. Suppress openpyxl/pandas warnings (thanks to @chapmanb)
2. Fix unit tests to account for cases where a user has not downloaded the CADD or GERP annotation files. Thanks to Xiaoloin Zhu and Daniel Swensson for reporting this and to Uma Paila for correcting it.

0.7.0

1. Added support for CADD scores via new cadd_raw and cadd_scaled columns.
2. Added support for genotype wildcards in query select statements. E.g., SELECT chrom, start, end (gts).(phenotype==2) FROM variants. See http://gemini.readthedocs.org/en/latest/content/querying.html#selecting-sample-genotypes-based-on-wildcards.
4. Added support for the VCF INFO field both in the API and as a column that can be SELECT’ed.
5. Upgraded to the latest version of ClinVar.
7. Automatically add indexes to custom, user-supplied annotation columns.
8. Improvements to the installation script.
10. Upgraded the gene_summary and gene_detailed tables to version 75 of Ensembl.
11. Added support for the MPI Mouse Phenotype database via the mam_phenotype_id column in the gene_summary table.
12. Enhanced security.
13. Corrected the ESP allele frequencies to be based report _alternate_ allele frequency instead of _minor_ allele frequency.
14. VEP version support updated (73-75) Support for aa length and bio type in VEP.
15. The lof_sieve tool support has been extended to VEP annotations.
16. Added the ccds_id and entrez_id columns to the gene_detailed table.

0.6.6

1. Added COSMIC mutation information via new cosmic_ids column.

0.6.4 (2014-Jan-03)

1. New annotation: experimentally validated human enhancers from VISTA.
2. Installation improvements to enable isolated installations inside of virtual machines and containers without data. Allow data-only upgrades as part of update process.
3. Fix for gemini query error when --header specified (#241).
0.6.3.2 (2013-Dec-10)

1. Fixed a bug that caused --gt-filter to no be enforced from query tool unless a GT* column was selected.
2. Support for ref and alt allele depths provided by FreeBayes.

0.6.3.1 (2013-Nov-19)

1. Fixed undetected bug preventing the comp_hets tool from functioning.
2. Added unit tests for the comp_hets tool.

0.6.3 (2013-Nov-7)

1. Addition permutation testing to the c-alpha test via the --permutations option.
2. Addition of the --passonly option during loading to filter out all variants with a filter flag set.
3. Fixed bug with parallel loading using the extended sample table format.
4. SLURM support added.
5. Refactor of loading options to remove explosion of xxx-queue options. Now load using --scheduler on --queue.
6. Refactor of Sample class to handle the expanded samples table.
7. Addition of --carrier-summary-by-phenotype for summarizing the counts of carriers and non-carriers stratified by the given sample phenotype column.
8. Added a --nonsynonymous option to the C-alpha test.
9. Added gemini amend to edit an existing database. For now only handles updating the samples table.
10. Fixed a bug that prevented variants that overlapped with multiple 1000G variants from having AAF info extracted from 1000G annotations. This is now corrected such that multiple overlaps with 1000G variants are tolerated, yet the logic ensures that the AAF info is extracted for the correct variant.
11. Fixed installation issues for the GEMINI browser.
12. Added --show-families option to gemini query.

0.6.2 (2013-Oct-7)

1. Moved –tped and –json options into the more generic –format option.
2. Fixed bug in handling missing phenotypes in the sample table.
4. API change: GeminiQuery.run takes an optional list of predicates that a row must pass to be returned.
5. --sample-filter option added to allow for restricting variants to samples that pass the given sample query.
6. ethnicity removed as a default PED field.
7. PED file format extended to allow for extra columns to be added to the samples table under the column named in the header.
8. The autosomal_recessive and autosomal_dominant tools now warn, but allow for variants to be detected in the absence of known parent/child relationships.
0.6.1 (2013-Sep-09)

1. Corrected bug in de_novo tool that was undetected in 0.6.0. Unit tests have been added to head this off in the future. Thanks to Jessica Chong

2. Added the \texttt{-d} option (minimum sequence depth allowed for a genotype) to the \texttt{autosomal_recessive} and \texttt{autosomal_dominant} tools.

3. New \texttt{-tped} option in the \texttt{query} tool for reporting variants in TPED format. Thanks to Rory Kirchner.

4. New \texttt{-tfam} option in the \texttt{dump} tool for reporting sample infor in TFAM format. Thanks to Rory Kirchner.

0.6.0 (2013-Sep-02)

1. Add the \texttt{--min-kindreds} option to the \texttt{autosomal_recessive} and \texttt{autosomal_dominant} tools to restrict candidate variants/genes to those affecting at least \texttt{--min-kindreds}. Thanks to Jessica Chong

2. Addition of a new \texttt{burden} tool for gene or region based burden tests. First release supports the C-alpha test. Thanks to Rory Kirchner.

3. Use of Continuum Analytics Anaconda python package for the automated installer. Thanks to Brad Chapman.

4. Enhancements to the \texttt{annotate} tool allowing one to create new database columns from values in custom \texttt{BED+} annotation files. Thanks to Jessica Chong and Graham Ritchie.

5. Addition of the \texttt{--column}, \texttt{--filter}, and \texttt{--json} options to the \texttt{region} tool.

6. Improvements to unit tests.

7. Allow alternate sample delimiters in the \texttt{query} tool via the \texttt{--sample-delim} option. Thanks to Jessica Chong.

8. Provide a REST-like interface to the gemini browser. In support of future visualization tools.

9. Allow the \texttt{query} tool to report results in JSON format via the \texttt{--json} option.

10. Various minor improvements and bug fixes.

0.5.0b (2013-Jul-23)

1. Tolerate either -9 or 0 for unknown parent or affected status in PED files.

2. Refine the rules for inheritance and parental affected status for autosomal dominant inheritance models.

3. The \texttt{autosomal_dominant}, \texttt{autosomal_recessive}, and \texttt{de_novo} mutation tools have received the following improvements.
   - improved speed (especially when there are multiple families)
   - by default, all columns in the variant table are reported and no conditions are placed on the returned variants. That is, as long as the variant meets the inheritance model, it will be reported.
   - the addition of a \texttt{--columns} option allowing one to override the above default behavior and report a subset of columns.
   - the addition of a \texttt{--filter} option allowing one to override the above default behavior and filter reported variants based on specific criteria.

4. The default minimum aligned sequencing depth for each variant reported by the \texttt{de_novo} tool is 0. Greater stringency can be applied with the \texttt{-d} option.
**0.4.0b (2013-Jun-12)**

1. Added new `gt_ref_depths`, `gt_alt_depths`, and `gt品質s` columns.
2. Added a new `--show-samples` option to the query module to display samples with alternate allele genotypes.
3. Improvements and bug fixes for installation.

**0.3.0b**

1. Improved speed for adding custom annotations.
2. Added GERP conserved elements.
3. Optionally addition of GERP conservation scores at base pair resolution.
4. Move annotation files to Amazon S3.

**F.A.Q.**

**Does GEMINI work with non-human genomes?**

Currently, no. However, we recognize that the GEMINI framework is suitable to genetic research in other organisms. This may be a focus of future work.

**What versions of the human genome does GEMINI support?**

Currently, we support solely build 37 of the human genome (a.k.a, hg19). We intend to support forthcoming versions of the human genome in future releases.

**How can I use PLINK files with GEMINI?**

Many datasets, especially those derived from GWAS studies, are based on SNP genotyping arrays, and are thus stored in the standard PLINK formats. While GEMINI only supports VCF input files, it is relatively straightforward to convert PLINK datasets to VCF with the PLINK/SEQ toolkit.

1. First, load the PLINK BED file into a new PLINK/SEQ project using the instructions found in the “Load a PLINK binary files” section here.
2. Next, use PLINK/SEQ to convert the project to VCF using the instructions found here.

At this point, you should have a VCF file that is compatible with GEMINI.

Alternatively, in his bcbio project, Brad Chapman has written a convenient script for directly converting PLINK files to VCF. Below is an example of how to use this script.

```
$ plink_to_vcf.py <ped file> <map file> <UCSC reference file in 2bit format>
```
Other information

Besides the GEMINI manuscript itself, see this presentation for a high-level overview of GEMINI. Additionally, this video provides more details about GEMINI’s aims and utility.