The VESPA (Very large-scale Evolution and Selective Pressure analyses) toolkit is a collection of commands designed to simplify molecular evolutionary analyses. The major motivation behind the development of VESPA was minimizing potential sources of error in large-scale selective pressure analyses using codeML from the PAML package [Yang 2007].

See the *Introduction to VESPA* for more information or *Installation* to get started.
1.1 Introduction to VESPA

The VESPA (Very large-scale Evolution and Selective Pressure analyses) toolkit is a collection of commands designed to simplify molecular evolutionary analyses. The major motivation behind the development of VESPA was minimizing potential sources of error in large-scale selective pressure analyses using codeML from the PAML package [Yang 2007].

Assessing selective pressure variation on a large scale using protein coding DNA sequences requires a complex pipeline composed of numerous independent analyses, including: ortholog identification, multiple sequence alignment, phylogenetic reconstruction, and assessment of codon-based models of evolution. The pipeline requires multiple data manipulation steps to combine the output of each of these different stages (such as parsing BLAST result files to identify homologs and assessment of the suitability of the phylogenetic tree for selective pressure analysis). For researchers new to bioinformatics, manual data manipulation is prone to error, is potentially unstandardized, and is difficult to reproduce. VESPA eliminates the need for manual data manipulation by creating functions that automatically complete the majority of data manipulation steps using a standardized approach. In addition, the use of VESPA should minimize the requirements for users to create their own programs.

While the procedures within each stage of a selective pressure analysis are independent, there are requirements on the order in which the phases are carried out. VESPA creates a standardize pipeline of analyses with a specific ordering of phases in the process. In addition, the package encompasses multiple specialized pipelines to accurately assess selective pressure and reduce potential false positives (such as those caused by alignment error [Fletcher and Yang, 2010]).

Lastly, VESPA was designed to increase user productivity by automating labor intensive or highly repetitive tasks: i) automation by recursion – used to repeat an analysis on a number of files (e.g. cleaning and translating a directory of genomes), and ii) automation of analysis methods – used to complete tasks that are normally demanding but invariable in execution (e.g. identifying homologs within BLAST output data). Automating these procedures within VESPA has created an analysis package that is highly scalable (i.e. from single gene to whole genome analyses) and that flexible enough to be useful for many levels of expertise and many alternative purposes/endpoints.
1.1.1 Phase and pipeline structure

The VESPA toolkit is separated into five separate analysis phases. The rationale behind the ‘phase’ system was primarily to aid users in understanding the distinct procedures involved in selective pressure analysis and to provide more advanced users with a flexible and adaptable pipeline. Functions within a phase also analyze the same input type (e.g. sequences, BLAST output, etc.).

The output of each phase in the VESPA tool requires an analysis step that must be completed by the user with third-party software. These analyses are not automated by VESPA for three reasons: i) these analyses are far too computationally intensive, and ii) the submission process for these programs may differ from user to user, and iii) software updates may create bugs within the pipeline.

The software package also incorporates two analysis pipelines, a basic pipeline for single gene orthologs (SGOs) and an advanced pipeline for both SGOs and multi-gene families (MGFs). The basic pipeline was designed to bypass the phylogenetic reconstruction techniques (phase 3) by inferring a gene phylogeny from a user-defined species phylogeny. Usage of the basic pipeline is only recommended if the genes are confirmed SGOs.
1.1. Introduction to VESPA

- **Phase 1**: Ensembl Clean or Clean
  - Translate
  - Database
  - or Gene Selection
- **Phase 2**: Similarity Groups or Reciprocal Groups or Best Reciprocal Groups
  - Identify SGOs

**Sequence Alignment**

- **Phase 3**: metaL Compare or metaL Compare
  - ProtTest Setup
  - Model Selection
  - ProtTest Reader
  - MrBayes Setup

**Phylogenetic Reconstruction**

- **Phase 4**: Map Alignments or Create Sub-Trees
  - Infer Gene Trees or MrBayes Reader
  - or Branch-site Setup
  - CodeML Setup

**CodeML Analysis**

- **Phase 5**: CodeML Reader
  - CodeML Alignments
Note: Each phase indicates the functions (white boxes) and the order in which they are invoked. Optional functions are indicated by 'or' and may be skipped. Dark boxes indicate third-party programs. (a) Phase 1 (Section 1.6) is the data preparation phase and includes the functions: ensembl_clean/clean (Section 1.6.1), translate (Section 1.6.2), create_database (Section 1.6.3), and gene_selection (Section 1.6.4). This phase ends with the requirements for sequence similarity searching. (b) Phase 2 (Section 1.7) is the similarity group creation phase and includes the following functions: similarity_groups (Section 1.7.2), reciprocal_groups (Section 1.7.2) and best_reciprocal_groups (Section 1.7.3). This phase results in the creation of requirements for multiple sequence alignment (MSA). (c) Phase three (Section 1.8) is the alignment assessment stage and includes both a basic pipeline (on the left) for MSA files that contain only single gene orthologous (SGOs) and an advanced pipeline (on the right) for unconfirmed MSA files. The phase includes the following functions: metal_compare (Section 1.8.1), protest_setup (Section 1.8.2), protest_reader (Section 1.8.2), and mrbayes_setup (Section 1.8.3). This phase results in either: i) a phylogenetic trees of the MSAs for the advanced pipeline or ii) selected MSAs for the basic pipeline. (d) Phase four (Section 1.9) is the selective pressure phase and continues the basic pipeline and advanced pipeline of the previous phase. The phase four basic pipeline includes: map_alignment (Section 1.9.1), infer_genetree (Section 1.9.2), setup_codeml (Section 1.9.3), and create_branch (Section 1.9.6). The phase four advanced pipeline includes: mrbayes_reader (Section 1.9.4), create_subtrees (Section 1.9.5), create_branch (Section 1.9.6), and setup_codeml (Section 1.9.3). This phase results in the input requirements for selective pressure analysis by codeML. (e) The final phase (Section 1.10) includes the function codeml_reader (Section 1.10.1) that analyzes the results of the codeML analysis.

1.1.2 Command structure

The VESPA software package was written in python (v2.7) and requires a UNIX environment to operate. VESPA may be invoked as follows:

$ python vespa.py

The VESPA help screen will then be displayed by default. If desired, the help screen may also be displayed using the following commands.

$ python vespa.py help

In addition to the basic help screen, VESPA has the option to display basic help information for each VESPA command. If desired, the help information may be displayed by specifying the command of interest subsequent to the help screen call (please note the space):

$ python vespa.py help translate

Commands in VESPA are specified after the program call (i.e. python vespa.py) on the UNIX command-line. Please note a space is required between the program call and the desired command. For example, the translate command would be invoked as shown below:

$ python vespa.py translate

Commands also require specific options to be invoked to function correctly. Options are specified after the command and begin with a dash symbol (-) and end with an equal sign (=) followed by either a user-specified file or Boolean value (i.e. True/False). For example, the translate command requires the user to specify the input (here 'user_data.txt') as follows:

$ python vespa.py translate -input=user_data.txt

Please note the space between the command and option, it should also be noted that there is no space separating the option (i.e. -input=) and the user-specification (i.e. user_data.txt). Multiple options may be invoked on the same command-line as shown below and are separated by a space:

$ python vespa.py translate -input=user_data.txt -cleave_terminal=False
1.1.3 Basic and required options

Commands in VESPA (see this manual Pg. 10) use two categories of options: basic and command-specific. Basic options may be invoked alongside any command, whereas command-specific options are limited to particular commands. This version of VESPA incorporates two basic options: ‘input’ and ‘output’. The ‘input’ option: This option is invoked by the user to indicate the desired input file or directory for a command. As indicated, this option is designed to function with either: i) an individual file or ii) a directory housing multiple files. Please note that the ‘input’ option is a REQUIRED option and therefore is required by all commands to function. Not specifying the input option will result in VESPA printing a warning message. Please note that ‘USR_INPUT’ is a placeholder for the input defined by the user.

```bash
$ python vespa.py temp_command -input=USR_INPUT
```

For example, if a user wanted to analyze the directory ‘Genomes’ they would type:

```bash
$ python vespa.py temp_command -input=Genomes
```

The ‘output option: This option indicates the desired name the user supplies for the output of a command. Depending on the input used, the option will either specify: i) the output filename (if an individual file was the input), or ii) the output directory name (if a directory was the input). It should be noted that some commands have specialized output, in these cases the desired name will be applied where possible.

```bash
$ python vespa.py command -input=USR_INPUT -output=USR_DEF
```

1.1.4 VESPA commands

<table>
<thead>
<tr>
<th>Phase one</th>
<th>Phase two</th>
<th>Phase three</th>
<th>Phase four</th>
<th>Phase five</th>
</tr>
</thead>
<tbody>
<tr>
<td>clean</td>
<td>similarity_groups</td>
<td>metal_compare</td>
<td>map_alignments</td>
<td>codeml_reader</td>
</tr>
<tr>
<td>clean_ensembl</td>
<td>reciprocal_groups</td>
<td>prottest_setup</td>
<td>infer_genetree</td>
<td></td>
</tr>
<tr>
<td>rev_complement</td>
<td>best_reciprocal_groups</td>
<td>prottest_reader</td>
<td>mrbayes_reader</td>
<td></td>
</tr>
<tr>
<td>translate</td>
<td></td>
<td>mrbayes_setup</td>
<td>codeml_setup</td>
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</tr>
<tr>
<td>create_database</td>
<td></td>
<td>create_subtrees</td>
<td></td>
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<td>gene_selection</td>
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<tr>
<td>individual_sequences</td>
<td></td>
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<tr>
<td>split_sequences</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

1.2 Installation

```bash
$ curl -o VESPA.tar.gz https://github.com/aewebb80/VESPA/archive/1.0.0.tar.gz
$ tar -xzf VESPA.tar.gz
$ cd VESPA
$ chmod +x vespa.py
$ sudo mv vespa.py /usr/local/bin
```

Perl Dependencies

VESPA requires users to install multiple Perl scripts and modules to be fully operational. These may be found alongside vespa.py within the program tarball.

Once downloaded, it can be installed as follows:

```bash
$ tar -xvzf VESPA.tar.gz
$ cd VESPA
$ chmod +x *Codeml*.pl
```
Note: Replace 5.XX with the version of Perl used by your system. (determined by executing $ perl -v)

DendroPy

VESPA requires users to install the DendroPy python library (version 4.0). Instructions to installing DendroPy can be found at the following link: https://pythonhosted.org/DendroPy/#installing

### 1.2.1 Third party software

The VESPA software package is designed to interface with a number of third-party programs. It should be noted that some functions were designed to interface with specific versions of these third-party programs and future updates may require updates to VESPA as well. Details on these third-party programs can be found below.

<table>
<thead>
<tr>
<th>Program</th>
<th>Version</th>
<th>URL</th>
</tr>
</thead>
<tbody>
<tr>
<td>DendroPy</td>
<td>4.0</td>
<td><a href="https://pythonhosted.org/DendroPy/#installing">https://pythonhosted.org/DendroPy/#installing</a></td>
</tr>
<tr>
<td>MetAL</td>
<td>1.1</td>
<td><a href="http://kumiho.smith.man.ac.uk/blog/whelanlab/?page_id=396">http://kumiho.smith.man.ac.uk/blog/whelanlab/?page_id=396</a></td>
</tr>
<tr>
<td>MrBayes</td>
<td>3.2.3</td>
<td><a href="http://mrbayes.sourceforge.net/">http://mrbayes.sourceforge.net/</a></td>
</tr>
<tr>
<td>NoRMD</td>
<td>1.3</td>
<td>ftp://ftp-igbmc.u-strasbg.fr/pub/NORMD/</td>
</tr>
<tr>
<td>PAML</td>
<td>4.4e</td>
<td><a href="http://abacus.genet.ucl.ac.uk/software/paml.html">http://abacus.genet.ucl.ac.uk/software/paml.html</a></td>
</tr>
<tr>
<td>ProtTest3</td>
<td>3.4</td>
<td><a href="https://github.com/ddarriba/prottest3">https://github.com/ddarriba/prottest3</a></td>
</tr>
</tbody>
</table>

### 1.3 Phase 1: data preparation

The data preparation phase was included for users new to bioinformatics. The phase prepares downloaded genomes for homology searching using the two VESPA supported homology search tools: BLAST [Altschul et al., 1990] and HMMER [Eddy, 1998]. This phase also includes supplementary functions not required for either pipeline shown in Fig. 1.1 but rather to aid users in homology searching.

#### 1.3.1 clean and clean_ensembl functions

The VESPA toolkit incorporates two quality control functions: clean and clean_ensembl.

**clean**

This basic function was designed as a QC filter for downloaded nucleotide sequences and/or genomes (Fig. 1.1a). Each sequence is confirmed as protein coding by using a conditional statement to verify that the nucleotide sequence contains only complete codons (i.e. the length of the sequence is exactly divisible by 3) (Fig. 1.1b). This is an essential step to confirm gene annotation quality and permit the codon substitution models of codeML [Yang, 2007]. Only sequences that pass QC are retained for further analysis (Fig. 1.1c).

$ python vespa.py clean -input=<user_input>
clean_ensembl

This more advanced function was designed to identify the longest nucleotide (canonical) transcript within an Ensembl nucleotide genome that passed the QC step detailed in the clean function. This is achieved by exploiting the pattern of ensembl sequence identifiers, which consistently begin with the gene identifier followed by the transcript identifier (Fig. 1.1d). The longest transcript is then identified for each ensembl gene identifier and saved within the output file.

```
$ python vespa.py clean_ensembl -input=<user_input>
```

Note: Supported file format(s): -input: fasta formatted files

Command-specific options: Both clean functions incorporate a single enabled option (rm_internal_stop) and two disabled options (label_filename and infer_ensembl_species) that may be manually configured by the user. The option rm_internal_stop will remove sequences if they contain an internal stop codon (Fig. 1.1g), those removed will be reported in the command log file. It should be noted that while rm_internal_stop is configurable, codeML does not permit nonsense mutations and this option should be enabled if the toolkit is being used for that purpose. The options label_filename and infer_ensembl_species alter sequence headers (i.e. Ensembl gene and transcript identifiers) by adding an additional identifier at the beginning of the header: infer_ensembl_species adds the common species name of the respective Ensembl identifier (Fig. 1.1e) and label_filename adds the filename (without the file extension) (Fig. 1.1f). It should be noted that executing a labeling option is required for enabling VESPA to automate the creation of gene trees and setup of the codeML branch-site models (for details see Section 1.9.6).

```
$ python vespa.py clean -input=USR_INPUT -rm_internal_stop=False
$ python vespa.py clean -input=USR_INPUT -label_filename=True
$ python vespa.py clean -input=USR_INPUT -infer_ensembl_species=True
```

Overview of `clean` and `clean_ensembl` functions

1.3.2 translate

The translate function translates nucleotide sequences that passed the QC filter of either clean function into amino acid sequences in the first reading frame forward only (Fig. 1.2a). The function operates by splitting the nucleotide sequence into codons and then translating them into their respective amino acids (Fig. 1.2b). Translation is a mandatory step to produce alignments permitted by the codon substitution models of codeML (see Section 1.9.1) [Yang, 2007]. The resulting protein sequences are then saved (Fig. 1.2c). If non-coding sequences (incomplete codons or internal stop codons) were not removed prior to invoking translate, the function will produce a warning message. The warning reports that the function is designed to only translate protein-coding sequences and terminates the function.

```
usr$ python vespa.py translate –input=USR_INPUT Command-specific options: translate incorporates a single unique option cleave_terminal and the previously described options of the clean functions (Section 1.6.1). The cleave_terminal option is enabled by default and is designed to cleave the terminal stop codon of each sequence (Fig. 1.2d). The function and default status of the remaining options are detailed in Section 1.6.1.

```

```
$ python vespa.py translate -input=USR_INPUT -cleave_terminal=False
```

Note: Supported file format(s): input: fasta formatted files

Overview of translate

Editing gene headers

1.3. Phase 1: data preparation
Fig. 1.1: FastA formatted files are shown as grey boxes and the associated white boxes show the filename. Data confirmation steps shown as readout beneath each example indicates if the results passed the check. The following QC checks are illustrated here: (a) Cleaning an input file, (b) initiates with codon confirmation, (c) only sequences that pass are saved in the output. If the `ensembl_clean` function is invoked, in addition to codon confirmation, each transcript of an ensembl gene undergoes (d) a longest transcript confirmation and only the longest transcript is saved in the output. Two options are available to append a prefix to sequence headers: (e) `infer_ensembl_species` to append the Ensembl genome, or (f) `label_filename` to append the input filename. Invoking (g) `rm_internal_stop` will remove genes that fail stop codon confirmation.
Fig. 1.2: Fasta formatted files are shown as grey boxes and their filenames are given in white boxes. (a) Translating an input file using translate initiates the translation procedure by separating the sequence (as in (b)) into each codon to determine the respective amino acid, (c) translated sequences are saved in the Translated output file. (d) If the cleave_terminal option is invoked, terminal stop codons will be removed from each applicable sequence.
To keep the headers uniform throughout the process, edit the nucleotide versions of the sequence files after the `clean` or `ensembl_clean` step but before the ‘translate’ step.

```bash
# Ensembl genomes have the headers in the format
>ENS(3 sp specific characters)G(11 digit gene ID)|ENS(3 sp specific characters)T(11 digit transcript ID which may or may not be identical to gene ID).

For eg
>ENSMODG0000000014|ENSMODT0000000012

# Need to insert the species common name immediately after the > sign. So use sed command
sed 's/>/>(species common name|/g` input filename > output filename

# For Ensembl genomes, I used
sed 's/>/>Opossum|/g` Translated_Ensembl_Cleaned_Genomes/Cleaned_Opossum.txt >

# To shorten the Ensembl headers to 30 characters or less (necessary later at the CodeML step) - remove the transcript ID from the gene headers - do this only after the :code:`clean` or :code:`ensembl_clean` step!!

sed -ic '/|ENS/s/...................$//` INPUT FILE (Ensembl genome)

# Finds the pattern |ENS, and in that line, substitutes the last 19 characters (can change the number of characters here if needed) with nothing. -ic means it modifies the files and makes a backup copy of the original file.

# NCBI genomes are more complex. They have very long gene headers, for example
>XM_007934499.1 PREDICTED: Orycteropus afer afer serum deprivation response (SDPR),

# Need to insert the species common name after the > sign - same as for ensembl genome. But also need to truncate the header after the XM id, i.e., after the first white space. So use ".*" which is the wildcard after a space, and say substitute everything that comes after a space to nothing. s/ .*/g

sed 's/>/>Tenrec|/g; s/ .*/g` Cleaned_Tenrec.fa > Tenrec_edit.txt
```

### 1.3.3 create_database

The `create_database` function was designed for users to concatenate multiple genomes into the single database required for homology searching. The function operates by building the database a single sequence at a time (Fig. 1.3a and Fig. 1.3b). The command-line version of BLAST requires additional commands to create a BLAST-formatted database. If the user enables the option `format_blast` and BLAST is installed on the system the function will attempt to automate the additional steps required for producing a BLAST-ready database (Fig. 1.3c). If `create_database` is unable to create the BLAST-formatted database, a warning message will be produced (see Section 1.12 BLAST version requirements).

```
usr$ python vespa.py create_database -input=USR_INPUT
```

**Note:** Supported file format(s): input: fasta formatted files

```
# To set up the directory structure for the BLAST database
mkdir BlastdbAfr
cp database.fas BlastdbAfr
```
Overview of `create_database`

1.3.4 gene_selection

If the user is only interested in a subset of genes, the gene_selection function was designed to enable the user to search a database for gene identifiers specified in a separate file. The function operates by searching the sequence headers of the database for matches with the user specified gene identifiers (Figure 5a). The matching process only requires the user-specified identifiers to match a portion of the database sequence headers (Figure 5b). The function saves a single sequence file for each matched identifier (Figure 5c). If a user-specified identifier matches more than a single sequence header in the database, or indeed no sequence in the database, the function will produce a warning message. It should be noted that the gene_selection function requires the option `selection_csv` to operate.

```
$ python vespa.py gene_selection -input=USR_INPUT -selection_csv=USR_INPUT
```

Note: Supported file format(s): input: fasta formatted files; selection_csv: csv, tsv, and unformatted.

Overview of `gene_selection` function

1.3.5 Supplementary functions

The VESPA toolkit also incorporates three supplementary functions that were designed to aid users in potential data manipulations required for homology searching: `rev_complement`, `individual_sequences`, and `split_sequences`. The `rev_complement` function: This function was designed for users to return the reverse...
Fig. 1.3: Fasta formatted files are shown as grey boxes and their filenames in white boxes. Invoking the `create_database` function (a) combines numerous sequence files into (b) a single sequence database file. (c) Shows the `format_blast=Yes` option that will generate the required database files for BLAST [Altschul et al., 1990].
VESPA Documentation, Release 1.0.0

Fig. 1.4: FastA formatted files are shown as grey boxes and their filenames in white boxes. Data confirmation steps indicate if the results passed the check. (a) The `gene_selection` function requires two files to operate: a database (Human.fasta) and a user specified gene identifiers file (genes.csv). (b) The function operates using header confirmation to identify sequences in the database that match to those specified by the user. (c) The output of the function is a single sequence file for each user specified genes found.

1.3. Phase 1: data preparation
complement of nucleotide sequences. Depending on the desired use, it is recommended that the user run the QC filter of the clean functions either preceding or proceeding the `rev_complement` function.

```
$ python vespa.py rev_complement -input=USR_INPUT
```

**Note:** Supported file format(s): input: fasta formatted files

Command-specific options: The `rev_complement` function incorporates the two labeling options of the clean functions (previously described in Section 1.6.1). It should be noted that the option `rm_internal_stop` was not included in this function.

The `individual_sequences` function: This function was designed for users to separate files/directories housing large collections of sequences (i.e. genome file(s) and database files) into individual sequence files.

```
$ python vespa.py individual_sequences -input=USR_INPUT
```

**Note:** Supported file format(s): input: fasta formatted files

The `split_sequences` function: This function was designed for users to separate files/directories housing large collections of sequences (i.e. genome file(s) and database files) into sequence files that house a specified number of sequences. The number of sequences in each output file may be specified using the `split_number` option; otherwise the default value of 100 is used.

```
$ python vespa.py split_sequences -input=USR_INPUT -split_number=USR_DEF
```

**Note:** Supported file format(s): input: fasta formatted files

### 1.4 Phase 2: homology search

The second phase of VESPA is concerned with identifying groups of similar sequences from either BLAST [Altschul et al., 1990] or HMMER [Eddy, 1998] homology searches. Three types of sequence similarity are recognized by VESPA: non-reciprocal (unidirectional), reciprocal (bidirectional), and best-reciprocal. ‘Non-reciprocal similarity’ is characterized by sequence similarity that is only detected by one of the pair of sequences, commonly resultant of an E-value near the threshold. Non-reciprocal similarity is generally distantly related sequences. ‘Reciprocal similarity’ is similarity identified by both sequences in the pair. Reciprocal similarity is typically closely related orthologs or paralogs. “Best-reciprocal similarity” requires that the sequences pass two criteria: (i) they are sequences from different species, and (ii) in the pair-wise connection each sequence finds no other sequence in the respective species with a lower E-value. These requirements limit identification to orthologs (non-orthologs may be identified due to identical E-values or the absence of a true ortholog). Each type of similarity connection is invoked using a separate function and will generate the families specific to that connection type. Each function is required to be linked to a protein sequence database (see `create_database`). The database is used to produce an output file of each similarity group containing the protein sequences of each member. Each protein sequence files then undergoes multiple sequence alignment.

#### 1.4.1 Core options

The `-input` argument of each function within the second phase is designed to accept the modular output of BLAST and the standard output of HMMER (i.e. USR_HOMOLOGY). In addition, each function also requires both the
database and format arguments. The database argument is used to specify the protein sequence database created by earlier in the VESPA pipeline (i.e. ‘USR_DB’) (see create_database) whereas the -format argument is used to specify the input format as either blast or hmmer. Please note that commands below are written on a single line.

```
$ python vespa.py similarity_groups -input=USR_HOMOLOGY -format=blast -database=USR_DB
```

Each function also includes three optional threshold options that are disabled by default: e_value, alignment_length, and percent_identity. The three options enable the user to define threshold values for the E-value, alignment length, and percentage identity of each homology connection. Enabled thresholds must be passed for a pair-wise homology connection to be used in creating similarity groups. If an E-value threshold is not enabled, each function is designed to only accept E-values < 1, otherwise warning message is printed.

```
$ python vespa.py similarity_groups -input=USR_HOMOLOGY -format=blast -database=USR_DB -e_value=0.001
$ python vespa.py similarity_groups -input=USR_HOMOLOGY -format=blast -database=USR_DB -alignment_length=75
$ python vespa.py similarity_groups -input=USR_HOMOLOGY -format=blast -database=USR_DB -percent_identity=75
```

### 1.4.2 similarity_groups and reciprocal_groups

The ‘similarity_groups’ and ‘reciprocal_groups’ functions both construct sequence similarity groups using a similar approach. Both functions iteratively read a single line of input (BLAST or HMMER output) and record only the name of the query and subject if they pass enabled thresholds. Limiting the recorded data of the homology search to sequence names and their respective role (query or subject) results in reduced computational requirements, increased function speed, and permits the function to parse larger BLAST or HMMER input files. Both functions are able to recognize and record input that denotes reciprocal homology of a previously recorded entry. Once each function has completed processing the input, the pair-wise homologs are used to build families. The ‘similarity_groups’ function allows both non-reciprocal and reciprocal connections within a sequence group (Fig. 1.5a) whereas ‘reciprocal_groups’ is restricted to reciprocal connection within a sequence group (Fig. 1.5b).

```
$ python vespa.py similarity_groups -input=USR_HOMOLOGY -format=blast -database=USR_DB
$ python vespa.py reciprocal_groups -input=USR_HOMOLOGY -format=blast -database=USR_DB
```

**Note:** Supported file format(s): ‘input’: BLAST tabular output format and HMMER standard output.

```
# My scripts for this step
# Concatenate the Blast output files
cat *.txt.out > BlastOutput_AllAfr.txt.out

# Make a similarity groups directory and copy the concatenated file there
mkdir Similarity_GroupsAfr
cp BlastOutput_AllAfr.txt.out Similarity_GroupsAfr/

# Copy the database file from the Blastdb into the Similarity groups folder
cp database_Afrotheria.fas ../Similarity_GroupsAfr/

# Create a submission script in the scripts folder
emacs similarity_groupsAfr.sh

# Use previous script as a template, give the full 96 hours and 128 GB, specify the filenames and paths.
```
## 1.4.3 Best-reciprocal similarity group (species-based) function

The `best_reciprocal_groups` function constructs sequence homology groups by iteratively reading each line of input and storing the record within a database in reference to the query sequence. Once the function has completed parsing the input, the database is used to determine the best-homolog for each query sequence. This is achieved by identifying which subject sequence has the best E-value for each designated species. The designated best-hit for each query are then parsed to determine if the relationship is reciprocal (i.e. the subject sequence [as a query] identifies the query [as a subject]). If a query and subject are identified as best-reciprocal homology hits, they are used to create families (Fig. 1.5c).

```bash
$ python vespa.py best_reciprocal_groups -input=USR_HOMOLOGY -format=blast -
--database=USR_DB
```

**Note:** Supported file format(s): ‘input’: BLAST tabular output format and HMMER standard output.

### Similarity groups created by functions

**Get SGOs**

```python
# Python script for sorting all the similarity groups into single gene orthologs and
# paralogs. Save as GetSGO.py in the Scripts folder,

import glob, os
import subprocess as unix

try:
    os.mkdir('./SGO/')
except:
    print("SGO folder already here")

try:
    os.mkdir('./paralogs/')
```
Fig. 1.5: The families created using (a) ‘similarity_groups’, (b) ‘reciprocal_groups’, and (c) ‘best_reciprocal_groups’. Shorter lines represent better E-values between two sequences (circles). Lines with a single arrow represent non-reciprocal or unidirectional similarity connections. Lines with arrows on both sides represent reciprocal or bidirectional similarity connections. Sequence identifiers are shown for each sequence, different species are designated in this figure by lowercase letter at the beginning of each sequence identifier – h (human), m (mouse), r (rat), g (gorilla). (a) the ‘similarity_groups’ function connects all sequences as they are connected by either unidirectional or bidirectional similarity connections. (b) the ‘reciprocal_groups’ function creates two groups because the sequences mGY and rGY only exhibit a bidirectional similarity connection with each other. (c) the ‘best_reciprocal_groups’ function creates a three groups as the gorilla GX2 (gGX2) exhibits a stronger (i.e. lower e-value) bidirectional similarity connection with human GX2 (hGX2) than human GX (hGX).
except:
    print("paralogs folder already here")
for file in glob.glob('*.fasta'):
    paraStatus=0
    fileDict=dict()
    with open(file, 'rU') as f:
        for line in f:
            if line.startswith('>'):
                name=line.strip()
                species=name.strip('>').split('|')[0][0:6]
                if fileDict.has_key(species):
                    unix.call(['cp', file, './paralogs/' + file])
                    paraStatus=1
                    continue
                else:
                    fileDict[species]=species
            else:
                continue
    if paraStatus==0:
        unix.call(['cp', file, './SGO/' + file])
    else:
        continue

# Submission script saved as GetSGO.sh
#$ -cwd
#$ -V
#$ -l h_rt=48:00:00
#$ -l h_vmem=12G
#$ -m be
#$ -M fbsisi@leeds.ac.uk

python ../../Scripts_Afrotheria/GetSGO.py
# Go to the folder that has all the similarity groups fasta files. And submit the .sh script that calls python and the .py script.
nohup qsub ../Scripts_Afrotheria/Get_SGOAfr.sh &

# Creates 2 folders within the similarity groups folder - paralogs and SGO.
Sort SGOs to keep gene families with 7 or more members and remove uninformative gene families
# Count the number of headers ">" there are in each similarity groups fasta file
grep -c ">" *.fasta > GenecountsSGO.txt
# Export this file to desktop, view in MSExcel and filter it to retain only the files that contain 7 or more members. Copy the list of gene families with less than 7 members, and then on the command line, move those files to a separate folder called "uninformative". You can use a script, or simply the "mv" command. I just pasted the list into a text file, removed line breaks, and then used the 'mv' command.

1.5 Phase 3: alignment assessment and phylogeny reconstruction

The third phase of VESPA combines multiple third-party programs (i.e. MetAl [Blackburne and Whelan, 2012] and NorMD [Thompson et al., 2001]) to automate the assessment and choice of protein Multiple Sequence Alignments
(MSAs). In addition this phase enables simplified large-scale phylogenetic reconstruction. Alignment error is reported to cause high rates of false positives in a selective pressure analysis [Fletcher and Yang, 2010]. Therefore, VESPA incorporates third-party programs for MSA comparison and scoring. A complete analysis of the MSAs from each method is recommended. The next step in this phase is the selection of the empirical model of evolution that best-fits each MSA [Darriba et al., 2011; Keane et al., 2006]. The third phase concludes with an automated method for generating the files necessary for phylogenetic reconstruction using the previously selected MSA and model of evolution. The functions of this phase are primarily designed to interface with selected third-party programs. However, each step of this phase has been made optional if the user has other preferences or needs.

# To do Muscle alignment, use the following script, saved as a muscle_script.sh file
for i in *.fasta
do
  muscle -in $i -out $i.mu
done
#$ -cwd
#$ -V
#$ -l h_rt=2:00:00

# Run from inside the informative SGO fasta file folder, completes within seconds, →maybe a couple of minutes. Check the output files once done, saved in the same →folder with a .mu suffix. make a new directory - Muscle_Output. Move *.mu file into →it.

# For Mafft alignment, use the following script
#$ -cwd
#$ -V
#$ -l node_type=48core-3T
#$ -l h_rt=48:00:00
#$ -l h_vmem=32G
#$ -m be
#$ -M fbsisi@leeds.ac.uk
#$ -o mafft.out
#$ -e mafft.err

mkdir Mafft_Output
for i in *.fasta
do
  mafft --auto --thread 2 $i > Mafft_Output/$i.mft
done

# Run from inside the SGO fasta file folder, completes within a couple of minutes. →Check the output files once done, saved in the Mafft_Output folder created by the →script.

# For Prank alignment, first we need to remove the pipe between the species common →name and the gene ID and replace it with a space - in the gene headers. This is →because Ali figured out that Prank does not recognise short names in gene headers →that don’t have a space between the species common name and the gene ID. Or it uses →the space to truncate the names. Even underscore doesn’t work. And if there is an →underscore or |, running prank produces no outputs, but also no errors!!.
# So create a copy of all the .fasta files
mkdir Prank_Input
cp *.fasta Prank_Input
cd Prank_Input

# Use sed to replace the 1st pipe “|” in every header with a space
sed -i 's/|/ /' *.fasta

# -i overwrites the .fasta files inside the Prank_Input folder. Leaving out the 'g',
--from the command makes sure the placement is not global. i.e., it will replace only
--the 1st instance in every line. So you get

> Tenrec XM_004699007.1
MPSLSCRFYQHKFPEVEDVVMNVNRVSIAEMAYSLYEYNNIEGMILLSELSSRRRISIN
KLIRIGNECVVIRDKEKKYIDLSKRRSVPPEAIKCEDKFTKSHTVYLHVAEVELE
YTKDSEQLESFLQRTANWFDYKRRPGYGDAYAFKHAVSDPAILDSLDNEENERRVLDNIN
NRRLTPQAVKIRAGIDAVKEALRAGLCSTENMPKINLIAPPYVMTIITLTEREGLSV
LNQAMAVIKEEKEKRGFVNVQEMPKVVTIDETELARQLERLERENAEVDGDADAEEMAKAED

>Dolphin ENSTTRG00000015805|ENSTTRT00000015805
MPGLSCRFYQHKFPEVEDVVMNVNRVSIAEMAYSLYEYNNIEGMILLSELSSRRRISIN
KLIRIGNECVVIRDKEKKYIDLSKRRSVPPEAIKCEDKFTKSHTVYLHVAEVELE
YTKDSEQLESFLQRTANWFDYKRRPGYGDAYAFKHAVSDPAILDSLDNEENERRVLDNIN
NRRLTPQAVKIRAGIDAVKEALRAGLCSTENMPKINLIAPPYVMTTITLTEREGLSV
LNQAMAVIKEEKEKRGFVNVQEMPKVVTIDETELARQLERLERENAEVDGDADAEEMAKAED

>Macaque ENSMUG00000021419|ENSMUT00000031052
MPGLSCRFYQHKFPEVEDVVMNVNRVSIAEMAYSLYEYNNIEGMILLSELSSRRRISIN
KLIRIGNECVVIRDKEKKYIDLSKRRSVPPEAIKCEDKFTKSHTVYLHVAEVELE
YTKDSEQLESFLQRTANWFDYKRRPGYGDAYAFKHAVSDPAILDSLDNEENERRVLDNIN
NRRLTPQAVKIRAGIDAVKEALRAGLCSTENMPKINLIAPPYVMTIITLTEREGLSV
LNQAMAVIKEEKEKRGFVNVQEMPKVVTIDETELARQLERLERENAEVDGDADAEEMAKAED

# For Prank alignments using a guide tree, put a copy of the nested parentheses tree
--in the Prank_Input folder. It is very very important that the names used in the
--guide tree are identical to the ones used in the sequence files!! Then run the
--script

SAMPLES=*.fasta
COMMANDS=()
for S in $SAMPLES; do COMMANDS+=("prank -d=${S} -o=${S}.pk -t=19MammalsTree.txt -
--prunetree -shortnames -once -nobppa 2> prank_job.${SGE_TASK_ID}.std.err 1> prank_
--job.${SGE_TASK_ID}.std.out"); done
#$ -cwd
#$ -V
#$ -l h_rt=48:00:00
#$ -l h_vmem=8G
#$ -t 1-426
#$ -tc 426
#$ -o prank.err
#$ -e prank.err
eval ${COMMANDS[$SGE_TASK_ID-1]}

# Make sure the number in -t and -tc is the total number of .fasta files you will use
--as input - changes for similarity and reciprocal groups. Make sure the name of the
--guide tree is correct. I tried paying the whole text of the nested parentheses tree
--in the command - the help section of Prank says it should be possible. But didn’t
--work. I think the parentheses in the guide tree were confusing for the script.
# The whole run takes a few minutes. Make a Prank_Ouput directory in the SGO foldern
--and move all the *.fas files from Prank_Input to Prank_Output.
# For alignments without a guide tree, use the following script
SAMPLES=*.fasta
COMMANDS=()
for S in $SAMPLES; do COMMANDS+=("prank -d=${S} -o=${S}.pk -prunetree -shortnames - → nobppa 2> prank_job.$SGE_TASK_ID.stdout 1> prank_job.$SGE_TASK_ID.stderr"); done
#$ -cwd
#$ -V
#$ -l h_rt=48:00:00
#$ -l h_vmem=8G
#$ -t 1-426
#$ -tc 426
#$ -o prank.
#$ -e prank.err
eval ${COMMANDS[$SGE_TASK_ID-1]}

# Put all the output files into a folder Prank_Output_notree in the SGO folder.
# Prank outputs have short names, need to add the gene ID headers again, use Ray's
→ python script (vespa_ChangeNamesToOriginalLongFormat.py)
import glob
for file in glob.glob('*.fasta'):
    # make map
    with open(file, 'r') as f1:
        longSpNames={}
        for line1 in f1:
            if line1.startswith(">"):
                spName=line1.strip().split(" ")[0]
                longSpNames[spName]=line1.strip()
            else:
                continue
                shortAli_name=file+".pk.best.fas"
                newName=shortAli_name+".longNames"
                with open(shortAli_name, 'r') as shortAli, open(newName, 'w') as f2:
                    for line2 in shortAli:
                        oriName=longSpNames[line2.strip()]+'\n'
                        f2.write(oriName.replace(" ", "\|"))
                    else:
                        f2.write(line2)

# Copy the original similarity_group_*.fasta files into the Prank Output folder. Then
→ run python vespa_ChangeNamesToOriginalLongFormat.py in the folder where all of
→ these files are, you should get all the files written with the extension ".longNames"
# CodeML cannot take headers that are longer than 30 character. To shorten the
→ Ensembl headers to 30 characters or less (needed by CodeML) - remove the transcript
→ ID
sed -ic '/|ENS/s/...................$//' similarity_group_0028.fasta.mu
# Finds the pattern |ENS, and in that line, substitutes the last 19 characters with
→ nothing. -ic means it modifies the files and makes a backup copy of the original
→ file. I found NCBI gene ID headers to be smaller than 30 characters (once the gene
→ description was removed) so there was no need to shorten those again.

1.5. Phase 3: alignment assessment and phylogeny reconstruction 23
1.5.1 Alignment comparison function

The `metal_compare` function is designed to fully automate MSA comparison and scoring. The function operates using the third-party program MetAl [Blackburne and Whelan, 2012] to compare two protein MSAs. If MetAl indicates that the two MSAs are dissimilar, the function employs the third-party program NorMD [Thompson et al., 2001] to score each protein MSA using column-based similarity. The MSA with the highest NorMD (i.e. column-based similarity) score is then selected for subsequent analysis. It should be noted that the `metal_compare` function requires the option `-compare` to operate.

```bash
$ python vespa.py metal_compare -input=USR_INPUT -compare=USR_INPUT
```

Command-specific options: The `metal_compare` function incorporates one additional option (`-metal_cutoff`) that may be configured by the user. The `-metal_cutoff` option assigns the numeric threshold determining MSA dissimilarity and by default is fixed at 5%. Alignment methods that yield MetAl scores lower than defined value are considered comparable and the function will select the MSA from the first alignment method (indicated using the `-input` option).

```bash
$ python vespa.py metal_compare -input=USR_INPUT -compare=USR_INPUT -metal_cutoff=0.10
```

**Note:** Supported file format(s): `-input` and `-compare`: fasta formatted files (nexus to be added in a future release).

**Note:** Vespa metAl works if you make sure headers are identical between all the different alignments. (reintroduce long gene ID headers into the Prank alignments). But not for MAFFT – somehow it doesn’t recognize the Mafft alignment. Tried Mafft single line fasta file as well, didn’t work. Does not recognize the input format as an alignment

```bash
python vespa.py metal_compare -input=Prank_Output_Longnames/ -compare=Muscle_Output/
```

1.5.2 Empirical model selection functions

The `prottest_setup` function: This function is designed to automate the process of identifying the best-fit model of amino acid replacement for a specified protein alignment using the third-party program ProtTest3 [Darriba et al., 2011]. The function is designed to test each amino acid substitution model in both the absence and presence of invariant sites, gamma categories, and a combination of the two.

```bash
$ python vespa.py prottest_setup -input=USR_INPUT
```

**Note:** Supported file format(s): `-input` fasta formatted files (nexus to be added in a future release).

The `prottest_reader` function: This function automates the process of reading the output of ProtTest3. The function creates two output files: `best_models.csv` and `best_supported_models.csv`. The best models file reports the best-fit model of amino acid replacement (± rate-heterogeneity) reported by ProtTest3 whereas the best supported file reports the best-fit model of amino acid replacement (± rate-heterogeneity) supported by the third-party phylogenetic reconstruction program MrBayes [Ronquist and Huelsenbeck, 2003]. The two output files are given to enable the user to use different phylogenetic reconstruction software if desired.

```bash
usr$ python vespa.py prottest_reader -input=USR_INPUT
```
1.5.3 mrbayes_setup

The mrbayes_setup function (Fig. 1.5.3) is designed to simplify the process of phylogenetic reconstruction using the third-party program MrBayes [Ronquist and Huelsenbeck, 2003]. The function begins by converting each protein MSA into the nexus format (Fig. 1.5.3a). Each nexus-formatted MSA is then appended with a standardized MrBayes command block that defines the variables required for phylogenetic reconstruction (Fig. 1.5.3b-d), they include the number of MCMC generations, the number of chains (trees) to be examined per generation, the temperature of the heated chain, the burn-in percentage, and the best-fit model of amino acid replacement (see Empirical model selection functions). Please note that the mrbayes_setup function requires the option -model_list to operate. The model_list option is used to target the ‘best_supported_models.csv’ output file generated by the protest_reader function (see Empirical model selection functions).

```
$ python vespa.py mrbayes_setup -input=USR_INPUT -model_list=MODEL_DATA
```

Note: Supported file format(s): input: prottest3 standard output format.

Command-specific options: The mrbayes_setup function incorporates multiple options (-mcmc_gen, -mcmc_chains, -mcmc_temp, -mcmc_burnin) for permitting the user to alter variables within the MrBayes command block (Fig. 1.5.3b-d). The mcmc_gen option sets the number of generations for the phylogenetic reconstruction and should be increased from the default value of 200,000 if previous attempts failed to converge. The remaining options have the following recommended settings by default: mcmc_chains i.e. the number of chains (default = 4), mcmc_temp i.e. the temperature of the heated chain (default = 0.2), and mcmc_burnin, i.e. the burn-in percentage respectfully (default = 0.25).

```
$ python vespa.py mrbayes_setup -input=USR_INPUT -model_list=MODEL_DATA -mcmc_gen=100000
$ python vespa.py mrbayes_setup -input=USR_INPUT -model_list=MODEL_DATA -mcmc_chains=6
$ python vespa.py mrbayes_setup -input=USR_INPUT -model_list=MODEL_DATA -mcmc_temp=0.3
$ python vespa.py mrbayes_setup -input=USR_INPUT -model_list=MODEL_DATA -mcmc_burnin=0.3
```

Overview of mrbayes_setup.

The MrBayes input file is described as follows: (a) The NEXUS file is separated into two blocks, a sequence alignment block and a MrBayes command block. (b) The specific commands within the MrBayes command block are each assigned default values (in bold) based on recommend values and previous commands. (c) The commands lset and prset by default are automatically assigned by VESPA from the best_supported_models.csv file (see Empirical model selection functions) specified by the model_list option. (d) The remaining commands are assigned based on recommended values, but may configured by the user is desired.

1.6 Phase 4: selection analysis preparation

The fourth phase of VESPA automates large-scale selective pressure analysis using codeML from the PAML package [Yang, 2007]. Phase four is characterized by specific commands for the basic and advanced pipeline options (Fig. 1.1). These pipeline-associated functions are designed to process the specific input of each pipeline into a standardized file format for the common functions used by both pipelines. Following standardization, VESPA automates the normally
a

```nexus
#NEXUS
BEGIN DATA;
DIMENSIONS NTRAX=12 NCCHAR=276;
FORMAT DATATYPE=PROTEIN MISSING=- INTERLEAVE;

MATRIX
Mouse[TLR3] MKGCSSEYMY SPOGLLSLWLL PVSSSTWQCT VREYVADCSS
Human[TLR3] MIQIGFPCYV VREGLVRGML CASYTRVCTY SLK-VALKSN
Dog TLR3 SQGEEHYYT SFQSLYFVI LCTQSTINCV VREYVADCSS

Mouse[TLR3] MKGCSSEYMY SPOGLLSLWLL PVSSSTWQCT VREYVADCSS
Human[TLR3] MIQIGFPCYV VREGLVRGML CASYTRVCTY SLK-VALKSN
Dog TLR3 SQGEEHYYT SFQSLYFVI LCTQSTINCV VREYVADCSS
;
END;
```

Alignment Block

MiBayes Command Block

b

```plaintext
begin mrbayes;
log start filename=TLR3_nexus.log replace;
set autoclose=yes;
1set applyto=(all) nst=4 rates=gamma;
  preset amodelpr=fixed(jones);
mcmcp ngen=200000 printfreq=2000 samplefreq=200 nchains=4
temp=0.2 savebrlens=yes relburnin=yes burninfrac=0.25;
mcmc;
  sumt;
sump;
  log stop;
end;
```

c

```plaintext
1set applyto=(all) nst=4 rates=gamma;
  preset amodelpr=fixed(jones);
```

Assigned by ProtTest

d

```plaintext
mcmcp ngen=200000 printfreq=2000 samplefreq=200 nchains=4
temp=0.2 savebrlens=yes relburnin=yes burninfrac=0.25;
```

Assigned by command ‘mcmc_gen’

nchains=4 Assigned by command ‘mcmc_chains’

temp=0.2 Assigned by command ‘mcmc_temp’

burninfrac=0.25 Assigned by command ‘mcmc burnin’
labor-intensive process of creating the necessary files and directory structures for codeML. Phase four also incorporates
a single optional function branch-label table (see create_branch) that may be invoked to enable the branch-
site models of codeML [Yang, 2007].

1.6.1 Alignment mapping function

The map_alignments function is designed to automate the conversion of protein MSAs to nucleotide MSAs (Fig.
1.6). This process is mandatory as the codon substitution models of codeML require nucleotide alignments. Protein-
MSA guided nucleotide MSAs are generated rather than directly generating nucleotide MSAs because: i) each column
within the protein MSA represents aligned codons and therefore avoids aligning incomplete codons or frame-shift
mutations, and ii) protein MSAs represent a comparison of the phenotype-producing elements of protein-coding se-
quen ces (Fig. 1.6a). The function begins by reading the protein MSA to map the non-gap position of each codon within
the inferred nucleotide alignment (Fig. 1.6b). The sequence of the mapped codons is then inferred using the nucleotide
dataset (preferably as a database) from earlier in the pipeline (Fig. 1.6c). If the mapping process results in no errors,
the respective nucleotide MSA is created (Fig. 1.6d). All errors detected by the function will be returned within a
separate log file. Please note that the map_alignments function requires the option -database to indicate the
nucleotide dataset for correct sequence inference.

```bash
$ python vespa.py map_alignments -input=USR_INPUT -database=USR_DB
```

Note: Supported file format(s): -input: fasta formatted files (nexus and phylip formats to be added in a future
release); -database: fasta formatted files.

Overview of the map_alignments function

1.6.2 Gene tree inference function

The -infer_genetree function is designed to automate the creation of the corresponding gene tree for a user-
specified MSA. This is achieved by associating the taxa specified on a user-defined species tree with the headers created
by 'label_filename' and infer_ensembl_species (see clean and clean_ensembl functions) within the MSA.
The function operates by first creating a copy of the species tree with the species names (Fig. 1.7). The species tree
is designated using the required species_tree option. The species names are then replaced with their associated
MSA headers (Fig. 1.7b). If any species names remain after this phase, the taxa and their respective branches are
removed from the tree to create the finished gene tree (Fig. 1.7c). It should be noted that the infer_genetree
function incorporates the non-standard python library dendropy [Sukumaran et al., 2010]. Further details can be found
at Third party software.

```bash
$ python vespa.py infer_genetree -input=USR_INPUT -species_tree=USR_INPUT
```

Note: Command-specific options: The infer_genetree function incorporates a single option
-allow_paralogs that is disabled by default. Normally, infer_genetree is designed to only allow a sin-
gle MSA header to associate with a species name (Fig. 1.7d). If multiple headers are found to associate with a species
name, VESPA will produce a warning message. The -allow_paralogs may be enabled in these situations if the
association error(s) are caused by within-species paralogs, in this case a gene tree will be created with associated
headers shown as within-species paralogs (Fig. 1.7e).

```bash
$ python vespa.py infer_genetree -input=USR_INPUT -species_tree=USR_INPUT -allow_˓→paralogs=True
```
Fig. 1.6: Sequence files are shown above as grey boxes indicating the sequences and white boxes indicating the filename. The ‘map_alignments’ function requires (a) two files to operate: a protein alignment (Alignment.fasta) and a nucleotide sequence database (Database.fasta). The function initiates by (b) mapping the gaps of the nucleotide alignment. (c) The nucleotide sequence of each alignment is then mapped using the sequence database to produce (d) the completely mapped output file.
Overview of the `infer_genetree` function

Fig. 1.7: The goal here is to determine the phylogenetic relationship of the sequences within the alignment in relation to the species phylogeny. (a) The ‘infer_genetree’ function requires two files to operate: a nucleotide alignment (Sequence_group_00.fasta) and a species phylogeny. (b) The function begins by replacing each species name within the phylogeny with their respective gene identifier (i.e. Human → Human|TLR2) located in the nucleotide alignment. (c) The function then creates the gene phylogeny by removing the species that have not been replaced by a gene identifier. (d) If the nucleotide alignment specified by the user contains paralogs (Chicken TLR2A and TLR2B) VESPA will produce an error message. (e) If the ‘allow_paralogs’ option is enabled the function will create a new branch to house the paralogs with the original species acting as an ancestral node.

1.6.3 `codeml_setup` function

The `codeml_setup` function is designed to simplify the creation of the complex codeML directory structure. This is achieved by incorporating previously written in-house software `GenerateCodemlWorkspace.pl` written by Dr. Thomas Walsh to produce the codeML directory structure [Walsh, 2013]. The purpose of automating the program `GenerateCodemlWorkspace.pl` via `setup_codeml` was to simplify input requirements and enable high-throughput analyses. The function requires only a protein-inferred nucleotide MSA (see Alignment mapping function) and an associated phylogenetic tree (See Gene tree inference function or MrBayes reader function) to construct the directory structure for the CodeML site-specific models [Walsh, 2013].

1.6. Phase 4: selection analysis preparation
$ python vespa.py codeml_setup -input=USR_INPUT

**Note:** Supported file format(s): `input`: newick formatted files (nexus tree format to be added in a future release).

Command-specific options: If the user has created the optional branch-label table (see `create_branch`) and enabled the `-label_table` option the function will create the directory structure for the codeML branch-site models. Automating the branch-site models requires a specific directory for each species and/or lineage specified by the user in the optional branch-label table (Fig. 1.6.3a). Next the `setup_codeml` function will produce a codeML taskfile that contains each codeML command line command to be computed (Fig. 1.6.3b). Following creation of the taskfile, a separate log file reporting the branch-site models that cannot be tested (due to missing taxa) is produced.

$ python vespa.py codeml_setup -input=USR_INPUT -label_table=USR_INPUT

Overview of the ‘codeml_setup’ function.

### 1.6.4 MrBayes reader function

If phylogenetic reconstruction has been performed by MrBayes then the ‘mrbayes_reader’ function is designed to replace ‘infer_genetree’ [Ronquist and Huelsenbeck, 2003]. The function operates by converting the nexus-formatted phylogeny into the newick format supported by VESPA and codeML [Yang 2007]. If the function is unable to locate the original amino acid fasta-formatted MSA required by `mrbayes_setup` the nexus-formatted MSA will be converted and placed with the newick-formatted phylogeny. It should be noted that ‘mrbayes_reader’ is unable to check phylogenies for convergence. Instead users are directed to confirm convergence using third party software such as Tracer [Rambaut et al., 2014].

$ python vespa.py mrbayes_reader -input=USR_INPUT

**Note:** Supported file format(s): `input` MrBayes standard output format.

### 1.6.5 Subtree function

The `create_subtrees` function is designed for high-throughput tree pruning. This optional step is often required to prune very large multigene family phylogenies into smaller sub-phylogenies. Larger phylogenies may require this pruning step due to feasibility concerns and as subfamilies decrease computational requirements whilst making data easier to manage we have included this optional function. Users may require this option for pruning out SGOs for selection analyses that are focused on single genes. The function operates by displaying the current phylogeny with a set of pruning commands/options. The user is then prompted to select one of the four commands: ‘select subtree’, ‘remove subtree’, ‘remove leaf’, or ‘keep original’. If either ‘select subtree’ or ‘remove subtree’ is selected, the user is prompted to select a single node (numbered on the displayed phylogeny) for selection or removal respectively (Fig. 1.8a-b). If ‘remove leaf’ is selected, the user is prompted to select a leaf label (sequence header) for removal (Fig. 1.8c). If ‘keep original’ is selected the tree manipulation step is skipped. The ‘create_subtrees’ function will produce a protein sequence file of the remaining nodes in the phylogeny (Fig. 1.8d). The protein sequence file is then required to undergo re-alignment and it proceeds from Phase 3 through the remainder of the pipeline (Fig. 1.1). The ‘create_subtrees’ function will also produce a separate log file of the original phylogeny, the selected command, and the resulting phylogeny. The ‘create_subtrees’ function incorporates the non-standard python library dendropy [Sukumaran et al., 2010] (see CodeML results assessment).

usr$ python vespa.py create_subtree -input=USR_INPUT
1.6. Phase 4: selection analysis preparation

a. branch_table.txt

Mouse
Human
Eglières: Human, Mouse

b. codeml_taskfile.txt

```
cd CodeML/TLR2/TLR2_Human/modelA/Omega0/; codeml
cd CodeML/TLR2/TLR2_Human/modelA/Omega10/; codeml
cd CodeML/TLR2/TLR2_Human/modelA/Omega1/; codeml
cd CodeML/TLR2/TLR2_Human/modelA/Omega2/; codeml
cd CodeML/TLR2/TLR2_Mouse/modelA/Omega1/; codeml
cd CodeML/TLR2/TLR2_Mouse/modelA/Omega2/; codeml
cd CodeML/TLR2/TLR2_Mouse/modelA/Omega0/; codeml
cd CodeML/TLR2/TLR2_Eglières/modelA/Omega10/; codeml
cd CodeML/TLR2/TLR2_Eglières/modelA/Omega2/; codeml
cd CodeML/TLR2/TLR2_Eglières/modelA/Omega1/; codeml
cd CodeML/TLR2/TLR2_Eglières/modelA/Omega0/; codeml
```
Overview of \texttt{create_subtrees} function

(a) If the user specifies the ‘select subtree’ option along with a node, the function creates the subtree by dissociating the specified node from its ancestral node and returning the requested subtree. (b) The ‘remove subtree’ options functions similarly to ‘select subtree’ except that requested subtree is discarded and the subtree containing the remaining leaves is returned. (c) The ‘remove leaf’ option will remove the specified taxa from the phylogeny. (d) The function terminates by creating sequence files for each pruned phylogeny.

\textbf{1.6.6 create_branch}

The \texttt{create_branch} function is designed to simplify the creation of the branch-label table required for the branch-site models of codeML [Yang 2007]. The branch-label table (previously shown in Fig. 1.6.3a) indicates the lineages or ‘branches’ that will undergo lineage-specific selection analysis, i.e. designation of the ‘foreground lineages’ for
codeML. Each line indicates one lineage, either a species or an ancestral node. Ancestral nodes (uniquely named by user [i.e. Eglires]) are followed by a list of descendant (extant) species (Fig. 1.6.3a). The function operates by displaying a user-specified species phylogeny and promoting the user to select the species and/or ancestral nodes (numbered on the displayed phylogeny) of interest for the study (identical display methodology as described in MrBayes reader function - see phylogeny in Fig. 1.8 for example). When the user has finished their selection, the function will automatically produce the branch-label table. It should be noted that this function is completely optional as the branch-label table may be easily created by hand. The `create_branch` function incorporates the non-standard python library dendropy [Sukumaran et al., 2010] (see CodeML results assessment).

```
$ python vespa.py create_branch -input=USR_INPUT
```

**Note:** Supported file format(s): `input`: newick formatted files (nexus tree format to be added in a future release)

## 1.7 Phase 5: selection analysis assessment

### 1.7.1 CodeML results assessment

The `codeml_reader` function is designed to parse the complex CodeML directory structure and create simplified results for inexperienced users. This is achieved by incorporating in-house software `CreateSummaryReport.pl` written by Dr. Thomas Walsh [Walsh, 2013] to produce the majority of the codeML results. In addition to automating `CreateSummaryReport.pl`, `codeml_reader` produces supplementary output files (Fig. 1.9) and specialized MSAs that are designed to aid in the detection of false positives (Fig. 1.10). If the user specifies a branch-label table (see `create_branch`) `codeml_reader` will produce CodeML MSAs, these MSAs are characterized by the addition of 

- i) the putative positively selected sites, and 
- ii) the codons/amino acids that are positively selected in the respective lineage(s).

```
$ python vespa.py codeml_reader -input=USR_INPUT
```

**Note:** Supported file format(s): `input`: VESPA formatted codeML standard output.

Sample supplementary output file created by `codeml_reader`

Sample specialized MSA created by `codeml_reader`

## 1.8 References


Fig. 1.9: The supplementary output file includes information for each site-specific and branch-specific model of codeML. The following information is provided for each model: the tree tested; the type of model (i.e. site-specific or branch-specific) being tested; number of free parameters in the $\omega$ distribution that are estimated by codeML, the initial $\omega$ value used by codeML; the resulting log likelihood (lnL) of the analysis; the resulting model of the likelihood ratio test (LRT); the parameter estimates of codeML; if positive selection was detected; and the positively selected sites (if positive selection was detected).

<table>
<thead>
<tr>
<th>Model</th>
<th>Tree</th>
<th>Model Type</th>
<th>$D$</th>
<th>lnL</th>
<th>LRT Result</th>
<th>Parameter Estimates</th>
<th>Positive Selection</th>
<th>Positively Selected Sites in BEVS Alignment (FW=0.1)</th>
</tr>
</thead>
<tbody>
<tr>
<td>m0</td>
<td>Sample MSA</td>
<td>Homogeneous</td>
<td>1</td>
<td>-5.72,966,594</td>
<td>N/A</td>
<td>$\omega_i=0.5192$</td>
<td>No</td>
<td>Alignment (FW=0.1)</td>
</tr>
<tr>
<td>m1Neutral</td>
<td>Sample MSA</td>
<td>Site-specific</td>
<td>1</td>
<td>-6,517,231</td>
<td>N/A</td>
<td>$\omega_0=0.3560$, $\sigma_1=0.6491$, $\omega_0=0.0840$, $\sigma_1=0.0161$</td>
<td>No</td>
<td></td>
</tr>
<tr>
<td>m2Selection</td>
<td>Sample MSA</td>
<td>Site-specific</td>
<td>2</td>
<td>-6,517,231</td>
<td>mlNeutral</td>
<td>$\omega_0=0.3709$, $\sigma_1=0.6239$, $\omega_0=0.0310$, $\sigma_1=0.0161$</td>
<td>No</td>
<td></td>
</tr>
<tr>
<td>n1Disent2</td>
<td>Sample MSA</td>
<td>Site-specific</td>
<td>3</td>
<td>-6,517,231</td>
<td>mlDisent2</td>
<td>$\omega_0=0.3709$, $\sigma_1=0.6239$, $\omega_0=0.0310$, $\sigma_1=0.0161$</td>
<td>Yes</td>
<td>Alignment (FW=0.1)</td>
</tr>
<tr>
<td>m1Disent3</td>
<td>Sample MSA</td>
<td>Site-specific</td>
<td>5</td>
<td>-6,517,231</td>
<td>mlDisent2</td>
<td>$\omega_0=0.3709$, $\sigma_1=0.6239$, $\omega_0=0.0310$, $\sigma_1=0.0161$</td>
<td>No</td>
<td></td>
</tr>
<tr>
<td>m7</td>
<td>Sample MSA</td>
<td>Site-specific</td>
<td>2</td>
<td>-6,517,231</td>
<td>N/A</td>
<td>$\omega_0=0.1520$, $\sigma_1=0.0999$, $\omega_0=0.0000$, $\sigma_1=0.0000$</td>
<td>No</td>
<td></td>
</tr>
<tr>
<td>m8</td>
<td>Sample MSA</td>
<td>Site-specific</td>
<td>4</td>
<td>-6,517,231</td>
<td>m7, mBA</td>
<td>$\omega_0=0.1520$, $\sigma_1=0.0999$, $\omega_0=0.0000$, $\sigma_1=0.0000$</td>
<td>No</td>
<td></td>
</tr>
<tr>
<td>n8A</td>
<td>Sample MSA</td>
<td>Site-specific</td>
<td>4</td>
<td>-6,517,231</td>
<td>N/A</td>
<td>$\omega_0=0.1520$, $\sigma_1=0.0999$, $\omega_0=0.0000$, $\sigma_1=0.0000$</td>
<td>No</td>
<td></td>
</tr>
<tr>
<td>m4A</td>
<td>Sample MSA</td>
<td>Site-specific</td>
<td>3</td>
<td>-6,517,231</td>
<td>m1Neutral</td>
<td>$\omega_0=0.3560$, $\sigma_1=0.6491$, $\omega_0=0.0840$, $\sigma_1=0.0161$</td>
<td>No</td>
<td></td>
</tr>
<tr>
<td>m4Dnull</td>
<td>Sample MSA</td>
<td>Site-specific</td>
<td>3</td>
<td>-6,517,231</td>
<td>N/A</td>
<td>$\omega_0=0.3560$, $\sigma_1=0.6491$, $\omega_0=0.0840$, $\sigma_1=0.0161$</td>
<td>No</td>
<td></td>
</tr>
<tr>
<td>m4Chimp</td>
<td>Sample MSA</td>
<td>Site-specific</td>
<td>3</td>
<td>-6,517,231</td>
<td>m1Neutral</td>
<td>$\omega_0=0.3560$, $\sigma_1=0.6491$, $\omega_0=0.0840$, $\sigma_1=0.0161$</td>
<td>No</td>
<td></td>
</tr>
<tr>
<td>m4null</td>
<td>Sample MSA</td>
<td>Site-specific</td>
<td>3</td>
<td>-6,517,231</td>
<td>N/A</td>
<td>$\omega_0=0.3560$, $\sigma_1=0.6491$, $\omega_0=0.0840$, $\sigma_1=0.0161$</td>
<td>No</td>
<td></td>
</tr>
<tr>
<td>m4null</td>
<td>Sample MSA</td>
<td>Site-specific</td>
<td>3</td>
<td>-6,517,231</td>
<td>N/A</td>
<td>$\omega_0=0.3560$, $\sigma_1=0.6491$, $\omega_0=0.0840$, $\sigma_1=0.0161$</td>
<td>No</td>
<td></td>
</tr>
</tbody>
</table>

Fig. 1.10: The specialized MSA shown above includes data on the location of positively selected codons or residues. Depending on the type of model being explored, the MSA will include additional information. For all models (site-specific or branch-specific), the header PS_Sites indicates the position of the positively selected codons (shown as NNN) or residues (shown as X). For branch-specific, the characters under positive selection are shown for each relevant lineage using the header PS_Characters followed by the lineage of interest (i.e. PS_Characters|Chimp above).


### 1.9 Roadmap

**Forthcoming improvements to VESPA:**

- Simplified operation
- Streamlined installation
- Improved documentation
- Parallelised computation for some steps
- nexus and phylip support for `metal_compare`

Contribute to [VESPA on GitHub](http://vespa.github.io/)