Taxon-Specific Barcode Statistics For Amplicon Metagenomics

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Introduction

Methods

Database consolidation and standardization

In planning a amplicon metabarcoding experiment it is sometimes nessesary to combine multiple reference database with different formats so they can be used in the same pipeline. This is especially the case when studing multiple kingdoms within a single community or an abscure group of organisms requiring specilized databases. Unfortunatly most databases have their own custom FASTA header formats and few have an explicit taxonomy specified, so their taxonomys must be downloaded from NCBI by referencing an accession or taxonomy ID. This means that the first step in any pipline that seeks to use sequences from multiple reference databases must be able to read in multiple format. Therefore, I made a script called reformat_fasta.py that reformats the FASTA headers of a varity of common reference database formats into a format similar to that used by UNITE (Figure 1). A single format also makes it easier for multiple scirpts to communicate. The header format contians pipe-delimited positional fields for the organism name, genbank ID, refere database-specific ID, and taxonomy. The levels of the taxonomy field (Figure 2) are semi-colon-delimited and each is in the form "l_name", where "l" is a one or two character abbreviation of the taxonomic level and "name" is the taxon (e.g. "p_Ascomycota"). Currently the reformat_fasta.py can read in the formats of the following database: UNITE, ITS1, Genbank, RDP, Phyto ID, and Phyto DB. Of these, only Genbank, RDP, and UNITE have full taxonomy information included. However the others do have either genbank accession or taxonomy IDs, allowing their full taxonomy to be downloaded. A possible future imporvement to reformat_fasta.py could be an ability to automatically download the taxonomov information for these databases and insert it into their headers.

Initial database statistics

In order to make downsteam scripts less memory intensive, I made a script called fasta_taxon_statistics.py to caluculate taxon-specific statistics before the distance matrix is calculatd. This is principally done to aid in randomly subsampling the database without loading it all into memory, as is described in the next section.

Statistics are calculated for each taxon at each taxonomic level. Taxa are differntiated by their linage (i.e. the name of the taxon, as well all of the higher level taxa it is a part of). Using the lineage to define a taxon instead of just its name and level allows for correct handleing of multiple taxa at the same level with the same name, but part of different higher level taxa (Figure 2). This is important for correct summary statistics and downstream tree visualization. Statistics calculated

```
RDP: >S000415306 Sparassis crispa; MAFF 238626 Lineage=Root;rootr...

UNITE: >Lachnum|JF690990|SH189789.06FU|reps|k_Fungi;p_Ascomycota;c_Le...

ITS1: >EF093575_ITS1_GB|Caloplaca albopruinosa|tax_id:413252|represen...

Phyto DB: >PD_00795_ITS (Phytophthora infestans )

Phyto ID: >AF271224.1|Pythium vexans
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Figure 1: Format of reference database FASTA headers.

```
d__Fungi;p__Glomeromycota
d__Fungi;p__Ascomycota;c__Dothideomycetes
d__Fungi;p__Ascomycota;c__Dothideomycetes;o__Pleosporales;f__Phaeosphaeriaceae;g__Phoma
d__Fungi;p__Ascomycota;c__Dothideomycetes;o__Pleosporales;f__Didymellaceae;g__Phoma
```

Figure 2: Format of taxon identifiers and FASTA taxonomy strings.

for each taxon include: the number of sequences, the taxonomic level, the parent taxon, and the child taxa.

Database subsampling

In order to focus on a specific subset of an entire reference database, I made a python script that subsamples a FASTA reference database, taking into consideration the taxonomy encoded in the headers. This is often nessesary since reference databases are very large and the time it takes to calculate a pair-wise distance matrix is proportional to the square of the number of sequences. Some taxa have much more sequences than are nessary. On the other hand, many taxa have too few sequences to characterize their sequence variation. These should be also removed so computational time can be devoted to more informative taxa. There also can be inconsienties in the number of taxonomic levels specified, so some filtering is required to remove sequences with low-resuluton taxonomies.

Therefore, I have made a script called fasta_taxon_subsample.py that has the ability to subsample a reference database in a taxonomy-specific manner. Taxa with too many sequences can be randomly subsampled and taxa with too few can be removed. It uses the output of fasta_taxon_statistics.py (Figure 3) to decide which sequences are to be filtered out before it reads the reference database. This allows it to reads the database one sequences at a time and thus it is able to process nearly any size of reference database of data with minimal RAM. It can also filter out sequences that are not identified to a specific set of taxonomic levels. Specific taxa can also be required or removed, so that the user can focus on specific groups of organisms.

Distance matrix calculation

One of the most important and computationally intensive steps in this procedure is that calculation of a pairwise distance matrix between all sequences. Any program or distance metric can be used; the choice depends of the type of sequence data available and the characterics of the locus analyzed. If the reference database is a multiple sequences alignment, then a program such as fdnadist from the EMBOSS package can be used. If the sequences are unaligned and all of the same locus, than a pairwise global alignment tool, such as the pair.seqs command from the Mothur package can be

id	taxon	level	name	count	parent	children
1	dFungi	d	Fungi	8000		2;3;4;5;6;7
2	dFungi;pAscomycota	p	Ascomycota	4390	1	8;9;10;11

Figure 3: Format of fasta_taxon_statistics.py output.

used. If the sequences are unaligned and contain the locus of interest, but are not globally alignable, it might be possible to use *in silico* PCR, using a program such as <code>isPCR</code>, to extract amplicon sequenes so they can be globally aligned. This method adds a whole new level of complexity and will be explained further in the discussion. Finally, if the sequences do not share any genes it might still be possible to obtain a rough distance using tetranucleotide frequencey or GC content.

After a distance matrix is obtained using one the methods described above it might be nessesary to rename the rows and columns as the sequence taxonomy headers. To do this I have written a simple script called reformat_matrix.py. Currently this script only accepts the Phylip format outputted by fdnadist, but I plan to add support for other formats as needed.

Calculation of taxon-specific distance statistics

once a correctly formatted distance matrix is produced, it can be subsampled for each taxon and statistics calculated. The most important of these are the optimal clustering threshold and the expected error rate at that threshold. The optimal clustering threshold can be calculated for any taxonomic level (e.g. clustering species vs clustering genera).

To do this and visulize the results, I wrote an R script. For clustering a given taxonomic level, it subsamples the entire distance matrix for each taxon and calucluats the distribution of inter-taxon and intra-taxon distances, the optimal clustering threshold, and error rate. The clustering threshold optimaization and error rate calculations are done using the Spider package function threshOpt. The optimal clustering threshold is determined by clustering over a range of distance thresholds and pciking the threshold that minimizes the sum of the flase positive and false negative error rates. False positives result from too low of a threshold, causing sequences to be classified differntly when they are actually the same for a given taxonomic level. False negatives result from too high of a threshold, causing sequences to be classified as the same when they are differnt.

Results

Discussion