An Evaluation of Tree-to-Tree Distance Metrics used to Visualize Phylogenetic Tree Landscapes

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INTRODUCTION

Phylogenetic analyses of large and diverse data sets generally result in large sets of competing phylogenetic trees. Consensus tree methods used to summarize sets of competing trees discard important information regarding the similarity and distribution of competing trees. A more fine-grain approach is to use a dimensionality reduction method to project tree-to-tree distances in 2D or 3D space [1]. In this study, we evaluate several tree-to-tree distance metrics using trees obtained from independent nonparametric bootstrap analyses of genes from a mitochondrial genome alignment.

Study Goals

1. Visually and analytically evaluate projections of four commonly used tree-to-tree distance metrics.
2. Estimate the intrinsic dimensionality of tree-to-tree distance metrics.

Methods

Aligned whole salamander mitochondrial DNA (mtDNA) genomes were obtained from Zhang et al. [2]. The software package PAUP* 4.0b10 [3] was used to perform 100-replicate nonparametric bootstrap analyses [4] on each of 15-gene partitions contained within the mtDNA alignment. The maximum likelihood (ML) criterion and a heuristic search [neighbor joining starting tree, Sub-tree Pruning and Regrafting (SPR) branch swapping with a reconstruction limit of 10] were used to select optimal phylogenetic trees for each bootstrap replicate. Parameters of the ML model (i.e., nucleotide substitution rates, base frequencies [5] and an among site rate heterogeneity parameter [6]) were independently optimized for each gene partition on a neighbor joining tree constructed from each gene partition. A special purpose script by JCW (available upon request) was used to distribute phylogenetic analyses in parallel on FSU’s shared HPC system.

The program TreeScaper [7] was used to evaluate several dimensionality reduction cost functions and optimization algorithms. The Curvilinear Components Analysis (CCA) cost function and the stochastic gradient decent (SGD) optimization algorithm provided the best fit to the original tree-to-tree distances according to several goodness of fit measures [8, 9] (Fig. 1).

Efficiency of Cost Function and Optimization

Distribution of Tree-to-Tree Distances

Efficiency of Cost Function and Optimization

FIGURE 1. Goodness of fit measure plotted as a function of iteration, where 1NN = 1 Nearest Neighbour [8], CON = Continuity [9] and TRU = Trustworthiness [9].

Intrinsic Dimensionality

Distribution of Tree-to-Tree Distances

Efficiency of Cost Function and Optimization

FIGURE 1. Goodness of fit measure plotted as a function of iteration, where 1NN = 1 Nearest Neighbour [8], CON = Continuity [9] and TRU = Trustworthiness [9].

### Observations

1. Tree-to-tree distance metrics can qualitatively and quantitatively influence projections of “tree landscapes.”
2. The projection of the RF-distances shows groups of related mtDNA gene trees best.
3. The Match Distance metric discriminates among bootstrap trees better than other metrics.
4. Fewer dimensions are required to optimally project the “Match Distance” tree-to-tree distance metric.

References


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FIGURE 2. Two-dimensional projections of 1921 non-parametric bootstrap trees from the salamander data set using four tree-to-tree distance metrics (Robinson Foulds [10], Match Distance [11, 12], Agd1 [13], and Agreement Subtree [13]). The colors represent the underlying genes used to generate the trees. Projections were made using TreeScaper [7] with the cost function set to CCA and the optimization algorithm set to Stochastic Gradient Decent.

FIGURE 3. A projection of the relationship among 2D matrices, similar to those displayed in Figure 2. Ten projections for each of the four tree-to-tree distance metrics were generated based on different initialization conditions. The 40 2D projections were compared using a Procrustes analysis and the program TreeScaper [7] with the cost function set to CCA and the optimization algorithm set to Stochastic Gradient Decent.

FIGURE 4. Raw stress of CCA plotted as a function of the projection dimensionality. More than 4 dimensions does not greatly improve the raw stress for the Match Distance metric.

FIGURE 5. A projection of the relationship among 2D matrices, similar to those displayed in Figure 2. Ten projections for each of the four tree-to-tree distance metrics were generated based on different initialization conditions. The 40 2D projections were compared using a Procrustes analysis and the program TreeScaper [7] with the cost function set to CCA and the optimization algorithm set to Stochastic Gradient Decent.