De Oliveria Martins & Posada

Centre national de la recherche scientifique (CNRS) (French National Center for Scientific Research) Station d'Ecologie Expérimentale du CNRS à Moulis

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Introduction

Introduction

- Materials and Methods
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Essentially, the paper is presentation of the software guenomu that is used to predict a set of likely species trees, while estimating uncertainty

 computationally reasonable (447 gene families $\Longrightarrow \sim 6$ hours on a single processor)

associated with the input gene trees [Martins et al., 2014]

- Gene trees can be of any input format (i.e. CAT family of models)
 - Posterior distributions of unrooted gene tree topologies
- Allows for multiple leaves on a species branch

- Gene tree evolution is not an exact representation of a species tree:
 DL- duplication/loss, ILS- incomplete lineage sorting, HGT- horiz. gene transfer
- Classes of species tree inference based on gene trees/alignments:
 - Supermatrix also a 'supergene' approach where genes are concatenated into a single large alignment
 - Supertree phylogenetic analysis for each alignment followed by species tree inference
 - Model-based Probabilistic modeling of the incongruence between species and gene trees (subclass of other classes)
- Robinson-Foulds (RF) supertree finds a species tree that minimizes disagreement with gene tree collections without explicitly taking into account biological phenomena [Bansal and Eulenstein, 2013]
- Gene tree parsimony (GTP) Finding the species tree that minimizes the reconciliation cost [Guigó et al., 1996]



Model-based inference of species trees

- Most supertree approaches neglect gene tree branch lengths
- Multispecies coalescent Another class of methods closely related to supertree methods that try to reconstruct a species tree based on a matrix of distances between species (for review [Liu et al., 2009])
- Gene trees are often assumed to be known in other words supertree methods generally do not account for uncertainty

guenomu approach

Input: a set of unrooted gene tree distributions (one per family) * and a list of species names

Output: posterior distribution of the rooted species as well as a posterior distribution of gene trees for each gene family

^{*} Posteriors of gene trees from programs like PhyloBayes [Lartillot and Philippe, 2004] can be used as input



- Let gene family i be a set of homologous sequences that compose an alignment D_i , which can comprise paralogs and orthologs
- Each member of the gene family needs to have a map to a species
- Gene tree (G_i) any phylogenetic tree connecting all sampled members of gene family i

as a reminder

Likelihood: $P(X|\theta)$ is the probability of the evidence given the parameters Posterior: $P(\theta|X)$ is the probability of the parameters given the evidence Prior: $P(\theta)$ is the probability describing the uncertainty about the parameters before evidence is taken into account

Bayesian Hierarchical Model

A Phylogenetic likelihood is

$$P(D|G,\theta) \tag{1}$$

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TABLE 1. Parameter values used in the simulations

	Description	Symbol	Distribution
Species tree			
(Dendropy)	Number of species		Uniform(10, 80)
	Number of generations (total tree height)		Uniform(10 ² , 10 ⁴)
	Expected number of duplications	E_{dup}	Uniform(10-3, 4)
	Number of gene families	N	Uniform(2, 50)
	Rate heterogeneity multiplier	H_s	Gamma(1, 1)
Locus tree ^a			
(SimPhy)	Gene duplication rate b	β	Exponential(E_{duv}/σ)
	Gene loss rate		Uniform $(0, 0.75 \times \beta)$
	Rate heterogeneity multiplier	H_{l}	Gamma(1, 1)
Gene tree ^a			
(SimPhy)	Effective population size	N_e	2000× LogNormal(0, 0.25
	Number of individuals per species c		Uniform(1, 10)
	Substitution rate (per time unit)		0.001
	Rate heterogeneity multiplier	H_g	Gamma(1, 1)
Gene tree			
Uncertainty	Maximum number of generated trees		160
(in-house)	Tree dispersion term	L_T	Uniform(2, 5)
	Tree location term	D_T	Uniform(3, 6)
	Frequency of trees with uncertainty	p_T	$1.1 \times \text{Beta}(L_T \times D_T, D_T)$
	Branch location term	L_B	Uniform(1, 5)
	Branch-wise error probability	p_B	$1.5 \times \text{Beta}(L_B, 1)$

Notes: Each simulation replicate was parameterized with values sampled from predefined statistical distributions. Given a species tree, simulated with *Dendropy*, the program SimPhy was used to simulate gene family trees. An in-house program was then used to mimic gene tree uncertainty, transforming each gene tree in a distribution of gene tree topologies in an attempt to emulate the effect of alignment-based phylogenetic inference in practice.

a terminology used by SimPhy

 $^{^{}b}$ the term σ is the total sum of branch lengths (in generations) in the species tree

c common for all gene families, emulating perfect sampling of individuals

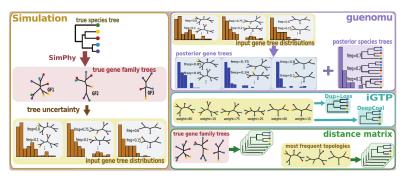


Figure 1. Simulation workflow. On the left a single data set is produced using Dendropy and SimPhy. The true species tree (rooted, with branch lengths) simulated by Dendropy is used as input by SimPhy to generate several (rooted) phylogenies, one per gene family. Then, uncertainty is added such that we have a distribution of topologies (unrooted, no branch lengths) per gene family. These collections will be used as input to the inference programs in the right panel. Our software guenomu estimates the posterior distribution of species trees (rooted, without branch lengths) and the posterior distributions of gene family topologies, based on all input gene tree distributions. The software inCTP also uses the input gene trees after transforming the frequencies into integer values representing weights, and estimates two rooted species trees: one under the Duplication and Loss cost, and another under the Deep Coalescence cost. For the distance-based species tree inference algorithms only one tree per gene family is used, and two alternative choices were attempted: one was to use the true gene families, with branch lengths; and the other was to use the most frequent gene trees (topologies only) after introducing phylogenetic uncertainty. In the later case it was assumed that all branches had the same length.

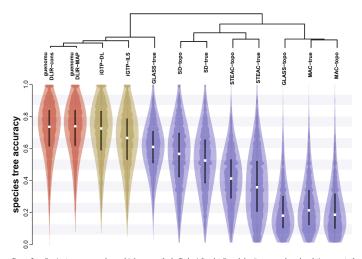
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Species tree accuracy of several inference methods. Each violin plot (kernel density curves plus a boxplot) represents the distribution of tree accuracy values for the different methods evaluated. From each posterior distribution estimated by guenomu under the DLIR parameterization we obtained two point estimates of the species tree: the MAP tree and the consensus tree (labeled 'MAP' and "cons"). We also obtained two estimates of species trees using iGTP, running the program under the DL cost and under the ILS cost ("iGTP-DL" and "iGTP-ILS"). We also ran four distance matrix approaches (GLASS, SD, STEAC, and MAC) using two types of data sets, just the topologies with uncertainty ("topo") or the true simulated gene family trees with branch lengths ("true"). At the top, we show the hierarchical clustering of the different methods based on their tree accuracy values for all replicates.

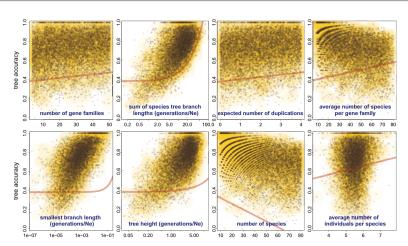
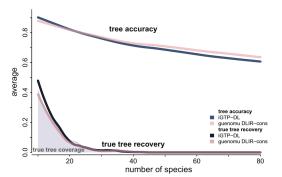


FIGURE 3. Guenomu's tree accuracy with respect to simulation parameters. Here, the species tree was estimated as the consensus tree from the posterior distribution assuming a DLIR parameterization. Based on a multiple linear regression analysis, these are the parameters that most significantly affected tree accuracy: at the top we have, from left to right, the total number of gene families, the total species tree length (in coalescent units), the expected number of duplications per gene family and the average number of species represented by gene family; at the bottom we have the length of the smallest branch in the species tree (in coalescent units), the height of the species tree from root to tips (scaled by the effective population size), the total number of species on the species tree and the average number of individuals from same species per gene family. Over each panel we show the regression line overlaid.

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Tree accuracy and true tree recovery for guenomu and iGTP according to the number of species. The lower lines show the true tree recoveries pooled over the number of species, whereas the higher lines show the average tree accuracy under different conditions. The gray area represents the true tree coverage under guenomu DLIR model (i.e., the proportion of data sets where the true species tree was within the 95% credible set). Each line is actually a smooth regression line over the individual values. For iGTP the species tree was estimated under the DL cost, whereas for guenomu we plot the Bayesian consensus under the DLIR parameterization.

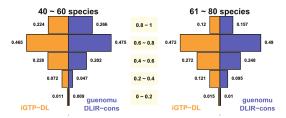


FIGURE 5. Tree accuracies of *iGTP* and *guenomu* for large data sets. On the left we have all data sets where the number of species was between 40 and 60, and on the right we have the pooled replicates with species tree sizes between 61 and 80 leaves. Each bar represents the proportion of data sets with tree accuracies between the ranges described in the middle, such that at the top we have the fraction of simulations where more than 80% of the splits were recovered, while the inferences whereas less than 20% of the true splits were found are the bottom.

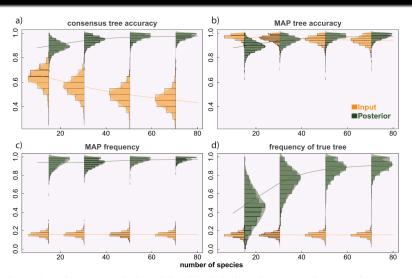


FIGURE 6. Input and posterior gene tree distributions. Each panel shows the distribution of input gene trees (after generation of tree uncertainty using our algorithm) and their posterior counterparts (resampled by guenomu) for several ranges of species tree sizes, together with a smooth regression line over all samples. The panels at the top show the accuracies of the consensus (a) and MAP (b) estimates, when compared to the true gene trees simulated by SimPlny, whereas the bottom panels display the frequencies of the MAP (c) and true (d) gene trees. All values are averages over all gene families from each replicate.

 Usually different sources of gene tree disagreement are considered separately (e.g. DL,ILS), however unrecognized processes adversely affect gene tree estimation [Rasmussen and Kellis, 2012]

Simulations

- guenomu does well, GTP methods do well, but coalescence methods perform poorly under the scenarios considered
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