Origins and losses of parasitism an analysis of the phylogenetic tree of life with a parsimony-like algorithm

Abstract

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1 Introduction

This paper is about the further development of parsimony algorithms for non-binary trees, applied to the currently largest phylogeny synthesis tree of Open Tree Of Life, with the application to the ancestral state reconstruction of parasitism.

Researchers of the phylogenies have been dealt with the ancenstral state reconstruction in the 60s. The first methods were only brute force TODO: Quelle, siehe Fitch: Camin and Sokal 1965

. Next came a set of parsimony algorithms such as: Fitch-parsimony [Fit71], Wagner-parsimony [SM87] ... TODO: weitere?

With more and more data, there is now the possibility to use more information to calculate the probabilities of the ancestral states. In addition to the states of the leaves, algorithms could also use branch lengths. The likelihood based algorithms came more in interest.

Our focus came with another 'data extension'. We wanted to work with the biggest phylogenetic tree that exists at this moment, which goes over all observed species. For most TODO: most? species there is no phylogeny, but only a taxonomic classification. So the biggest 'phylogenetic tree' is a synthesis of phylogenetic trees filled with a taxonomic tree given by Open Tree of Life [HSA+15]. This tree is not binary and therefore the developed algorithms are not directly applicable.

In this work, we have looked at the algorithms that are generally suited to our data, to develop them further for the not binary case, and finally to compare their usability with our sythesis tree.

We have decided to consider only parsimony algorithms since we have no information on branch lengths and no other additional information like different transition probabilities of our states.

2 Methods

This work consist of two aspects. One is the application of the algorithm to our question

Diese Arbeit besteht aus zwei Fragestellungen gute Algorithmen zu finden und sie auf unser Anwendungsbeispiel anzuwenden.

Zur Bewertung der gefundenen / entwickelten algorithmen haben wir eine Simulation ausgeführt.

Auf der Anderen Seite haben wir unser reales Problem aufgestellt, für die Algorithmen vorbereitet und diese angewendet.

Metadata -> Simulation -> real Data

2.1 Metadata analysis

Properties of real Data - Metadata analysis

There are some Parameters to find out or notice:

- transition probabilities of tags
- multifurcation
- nr of parasites to free-living
- nr of unknown nodes

2.2 Simulation

- build random binary trees, tag these (parameters: parasites vs free-living, beta-distribution)
- run fitch-parsimony, wagner-parsimony, our parsimony like algorithm
- build not binary tree (poisson distribution?)
- run new algorithms
- compare trees (distances)
- i) build random binary trees
- ii) tag tree
- iii) multifurcate tree
- iv) run maximum parsimony algorithms
 - Fitch
 - Sankoff (Castor package)
 - my algorithm
- v) Evaluation

2.2.1 random binary tree

To get a random binary tree, I used the Phylo package from biopython. They offer a randomized function which returns a BaseTree ¹:

```
from Bio import Phylo
Phylo.BaseTree.Tree.randomized(number_leafnodes)
```

From the BaseTree class:

¹https://github.com/biopython/biopython/blob/master/Bio/Phylo/BaseTree.py

TODO: Zitat von BaseTree und buildTree.py

2.2.2 tag tree

At this point we want one fully tagged tree, and one less tagged tree which looks like our real data.

Let's say the first specie (the root node) was free-living (start with a parasite without a host makes no sence). For every transition from a node to his child, we take a random number from the father distribution. We decided that from the biological perspective a beta distribution reflects our transition probabilities best (see Figure 2.1 TODO: ref einfügen).

For example when our father node was free-living, then we take from the free-living beta distribution. Is the number under the threshold for beeing parasite, we get a change and tag the current node as parasite. Otherwise we tag it as free-living.

With this procedure we traverse through the tree from the root to every leave node. A part of this code you see here:

```
from numpy import random
if father_tag == 0:
    # freeliving_distribution:
    new_random = random.beta(a=A_FL, b=B_FL)
else:
    # parasite_distribution:
```

Figure 2.1: 60% Free-living - 40% Parasites red: parasites, blue: free-living

TODO: Bessere Beschriftung, Plot neu erstellen! U.a. mit threshold

unknowns.. forget information.. nodelist

2.2.3 multifurcate tree

2.3 real data analysis

- Import tree
- Import interactions
- run castor algorithm / and others?
- interprete results (leave one out)

2.4 Implementation

3 Results

4 Discussion

Wie gut ist der randomisiert erstellte Baum?

Wie gut kommt unsere Simulation an die echte Datenlage heran.

Fehlerqoute der Daten an sich?

Wie gut ist unsere Datenlage? 3 mio knoten, 1.8 named species (leaf nodes), 200.000 leaf nodes mit Information.

Simulation von subtrees

Welche Teile des Baumes sind gut, an welchen muss noch viel geforscht werden.

Wieviele Origins haben wir gefunden, was bedeutet diese Zahl?

Parameter der Simulation:

- Wie ist die Verteilung der vergessenen internen Knoten? Zum Wurzelknoten hin mehr vergessen?
- Wie sehen die übergangswahrscheinlichkeiten aus von P->FL und andersherum?
- Verteilung Parasiten zu Freilebend zu keine Information

Selecting of the 'right' / best Distribution

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