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# **Generating Decisive Sub-matrices for Phylogenetic Tree Construction**

## Background

A large problem when building phylogenetic trees is data incompleteness. A significant amount of missing data can lead to ambiguity in the structure of tree, leaving biologists unsure of the accuracy of the tree. However, even with incomplete data, an unambiguous phylogenetic tree can still be built depending on the makeup of data. Data can be represented using a data availability matrix which indicates whether sequencing data is available for a give taxon at a given loci. Each row represents an organism, and each column represents a loci.

Sanderson and Steele found that these matrices can be used to check if an unambiguous tree can be built. They showed that if a matrix is decisive, then you can confidently build a phylogenetic tree from that data. However, if a matrix is indecisive, we can find a decisive submatrix, which would correspond to a subset of organisms for which we can build a phylogenetic tree. Checking decisiveness of a matrix is equivalent to solving the no rainbow 4-coloring problem on the matrix's hypergraph. We can formulate this NP-complete problem as both an integer linear programming instance and satisfiability instance for efficient solving in practice. Once we find a decisive submatrix, a biologist can confidently construct a phylogenetic tree for those taxa.

# SAT Formulation

To check for check for decisiveness, we formulate the no rainbow 4coloring problem as a SAT problem. We use the following constraints and their corresponding Boolean formulas to check for a coloring. Let  $x_{ia} = true \leftrightarrow x_i$  has color a.  $x_i$  represents one of the *n* taxa. (1) Each taxa has only 1 color.

 $\bigwedge x_{i1} \oplus x_{i2} \oplus x_{i3} \oplus x_{i4}$ 

(2) Each color appears at least once in the coloring.

#### $\bigwedge \bigvee x_{iq}$ $q \in [4] i \in [n]$

(3) No loci is rainbow colored. Let k be the number of loci and  $Y_i$  be the set of all taxa who have data for loci *j*.

 $j \in [k] q \in [4] i : a_i \in Y_i$ 

We also use an existing ILP formulation of the no rainbow 4-coloring.

# Software Pipeline

We built a software pipeline that generates a decisive submatrix from an indecisive data availability matrix. The pipeline works as follows:

while the matrix is not decisive:

- add/remove rows using heuristic
- 2. Generate an ILP or CNF Boolean formula
- 3. Solve the formulation

We solve the ILP instances using the Gurobi ILP solver, a paid software available under academic license. We tested several SAT solvers from the PySAT library.

V	e	n	

Data Set	# Loci	# Taxa	
Allium	6	57	Tolle
Asplenium	6	133	
Caryophyllaceae	7	224	
Eucalyptus	6	136	
Euphorbia	7	131	
Ficus	5	112	R
Iris	6	137	
Meredith.mammals	6	169	tog Rab
Miadlikowska.fungi	9	1317	ata Miediji
Primula	6	185	C Mared
Rabosky.scincids	6	213	Wicicu
Ranunculus	7	170	
Rhododendron	7	117	
Rosaceae	7	529	Car
Shi.bats	9	815	Car
Solanum	7	187	
Soltis.saxifragales	1	946	
Szygium	5	106	
Tolley.chameleons	6	202	

Table 1: Sizes of initial data sets

	Remove Method			
Data Set	Final Size	Gurobi time (sec)	Glucose 4 time (se	
Allium	6	0.316	0.146	
Asplenium	32	0.162	0.136	
Caryophyllaceae	69	0.209	0.199	
Eucalyptus	3	0.17	0.14	
Euphorbia	82	0.167	0.142	
Ficus	26	0.183	0.143	
Iris	25	0.172	0.147	
Meredith.mammals	168	0.217	0.17	
Miadlikowska.fungi	320	0.231	0.17	
Primula	29	0.17	0.138	
Rabosky.scincids	207	0.174	0.135	
Ranunculus	15	0.173	0.148	
Rhododendron	63	0.167	0.133	
Rosaceae	64	0.2	0.15	
Shi.bats	35	4.769	6.137	
Solanum	27	0.212	0.167	
Soltis.saxifragales	16	3.319	2.766	
Szygium	85	0.18	0.169	
Tolley.chameleons	188	0.199	0.186	

**Table 2:** Timing and size results for all experiments

## Experiment

We ran the pipeline on real data provided by Dobrin, Zwickl, and Sanderson. The pipeline was always run on the same computer with four cores and a Linux operating system with no background programs running. From these experiments, we wanted to know:

1: Are the theoretical results feasible for use in practice?

2: What problem formulation should be used?

3: What heuristic yields the best sub-matrices?

The heuristics were tested were basic:

**Remove method**: Repeatedly remove a taxon that has the fewest number of non-zero entries until a decisive submatrix is found.

Add method: Remove taxa as described in the remove method and add back sets of taxa in the powerset of initially removed taxa until the largest decisive sub-matrix is found.

We ran four types of experiments: add method with ILP, add method with SAT, remove method with ILP, and remove method with SAT.



0.194

0.184

0.136

0.2

0.181

0.361

0.164

0.27

1.769

# method Size Results

time.

**Timing Results** 

The add method consistently produces larger decisive submatrices when it finished. Several of the submatrices produced by the remove method for organisms like fungi and chameleons were relatively large submatrices.

#### **Other Results**

When analyzing the resulting submatrices, we saw that all the matrices produced using the add method were trivial, meaning they all shared one gene in common. Many of the matrices produced by the add results were also trivial, but several were non-trivial. Some of the nontrivial matrices had good coverage over several families of organisms.

### Formulation

Use the SAT formulation in future software - SAT and ILP run in comparable time, and SAT solving software is available open source and free to anyone.

### Heuristic

Avoid the add method – this method works poorly on large matrices, runs slower, and produces trivial matrices. Remove method shows feasibility – though remove method often finds trivial matrices, it shows it is possible to find non-trivial submatrices with good coverage using these ideas.

### **Future Work**

Find a better heuristic – we need a theoretically backed way to choose rows to eliminate to get more non-trivial results.

Alexey Ignatiev and Antonio Morgado and Joao Marques-Silva(2018). PySAT: A Python Toolkit for Prototyping with SAT Oracles. In SAT (pp. 428–437). B. H. Dobrin, D. J. Zwickl, and M. J. Sanderson, "The prevalence of terraced treescapes in analyses of phylogenetic data sets," BMC Evolutionary Biology, vol. 18, p. 46, 2018. Parvini, G., Braught, K., Fernández-Baca, D.: Checking phylogenetic decisiveness in theory and in practice (February 2020), arXivpreprinthttps://arxiv.org/abs/2002.09722 Steel, M., Sanderson, M.J.: Characterizing phylogenetically decisive taxon coverage. Applied Mathematics Letters 23(1), 82–86(2010)

164

87

415

151

94

194

# Results

We tested several SAT solvers and found that Glucose 4 was generally the fastest, so we use those results to compare to the ILP results. We see that in most cases, ILP and SAT formulations run in comparable

We also see that on smaller matrices, the add and remove methods run in similar time, but more programs time out when using the add

# Conclusion

## References