Quantifying Uncertainty in Evolutionary Analyses

Alexandros Stamatakis

ERA Chair, Institute of Computer Science, Foundation for Research and Technology - Hellas Research Group Leader, Heidelberg Institute for Theoretical Studies Full Professor, Dept. of Informatics, Karlsruhe Institute of Technology

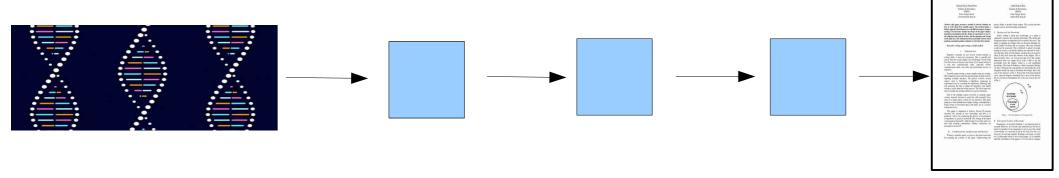
www.biocomp.gr (Crete lab)

www.exelixis-lab.org (Heidelberg lab)

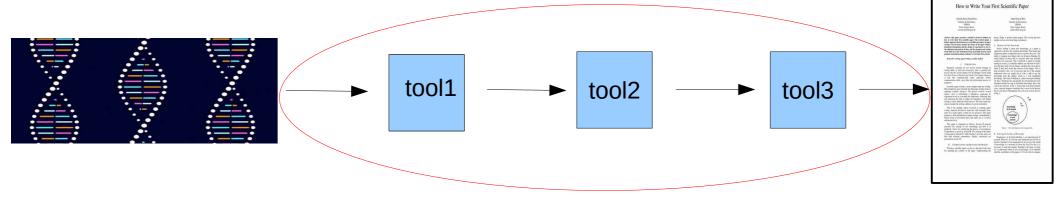
Outline

- Our Approach to Bioinformatics
- Introduction to Phylogenetic Inference
- Sources of Uncertainty
- Phylogenetic Difficulty
- Other stuff we are working on

Bioinformatics

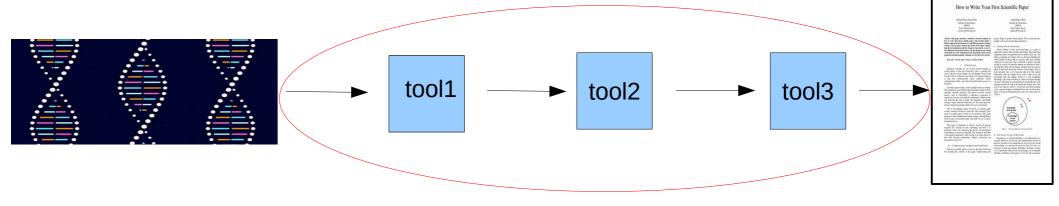


Bioinformatics

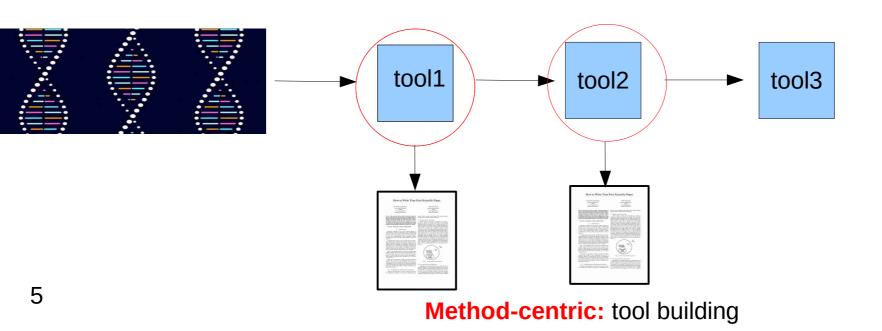


Data-centric: pipeline building

Bioinformatics



Data-centric: pipeline building



Our Approach

- Focus on core tool, model, algorithm, and method development
- Method development better fits the research interests of a computer scientist
- Goal: Enable Research in Evolutionary Biology

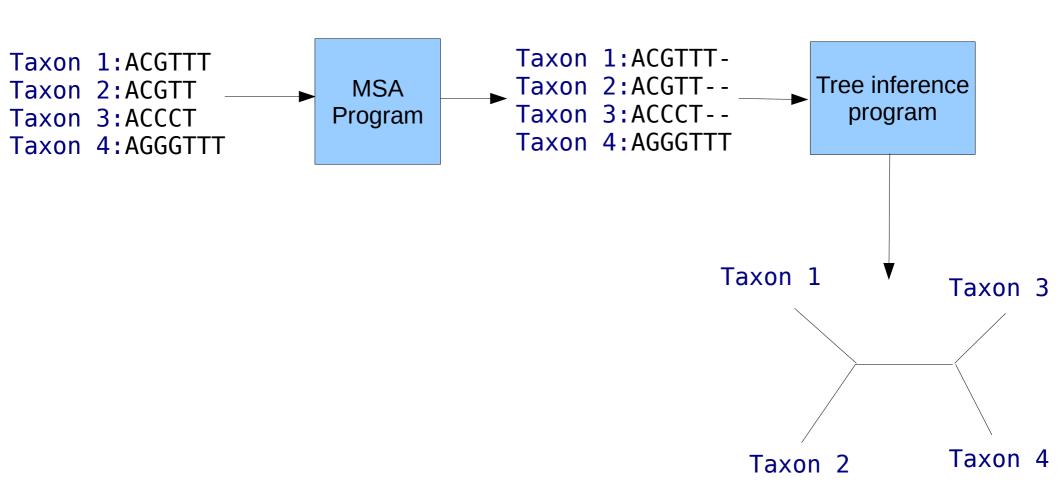
Our Approach

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- Method development better fits the research interests of a computer scientist
- Goal: Enable Research in Evolutionary Biology
- Nonetheless, we often conduct data centric research in side projects

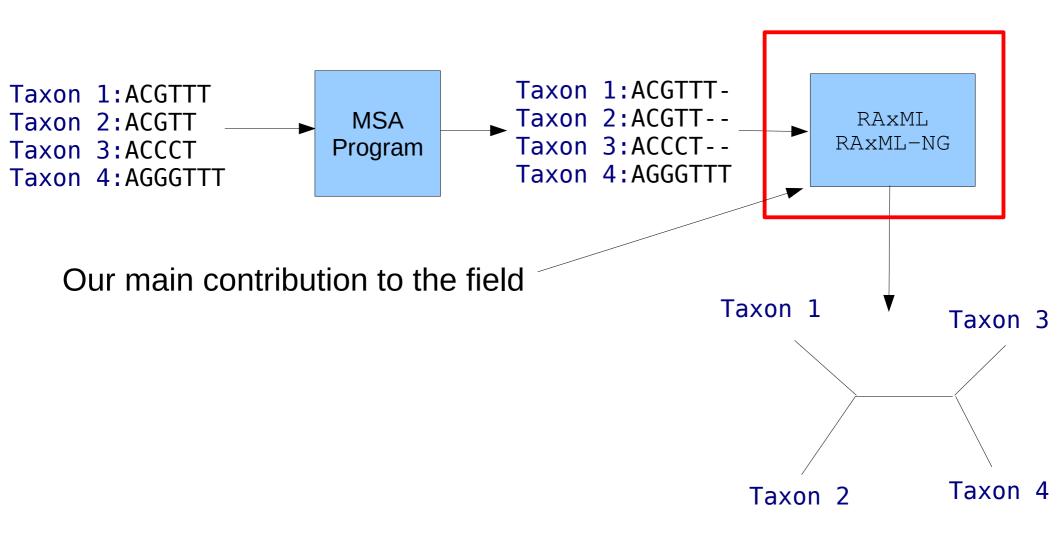
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Tree Inference Pipeline

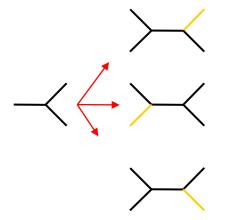


Tree Inference Pipeline

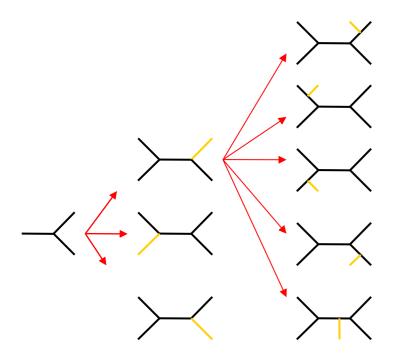




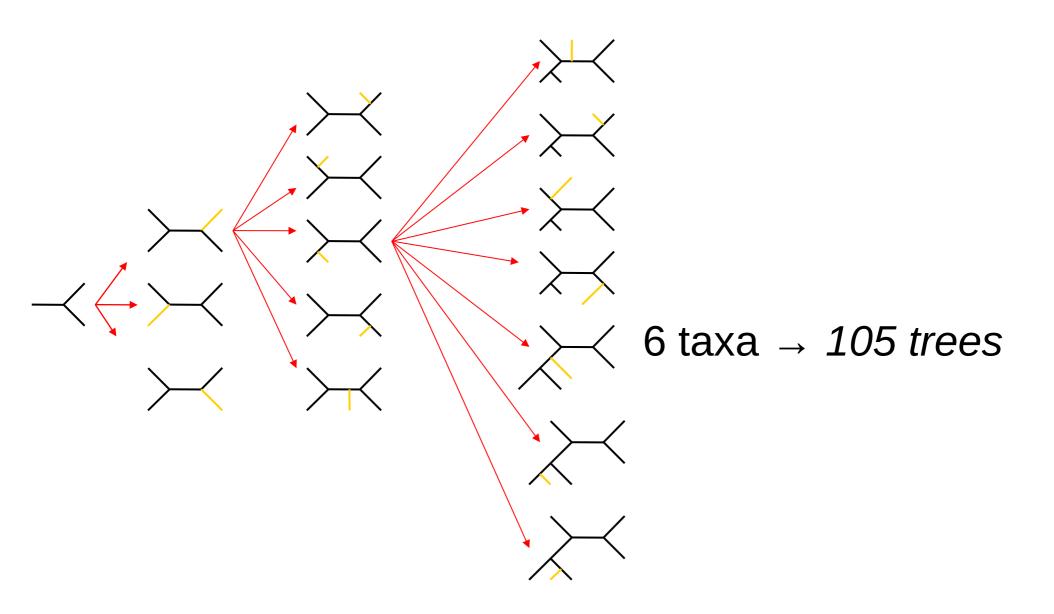
 $3 taxa \rightarrow 1 tree$



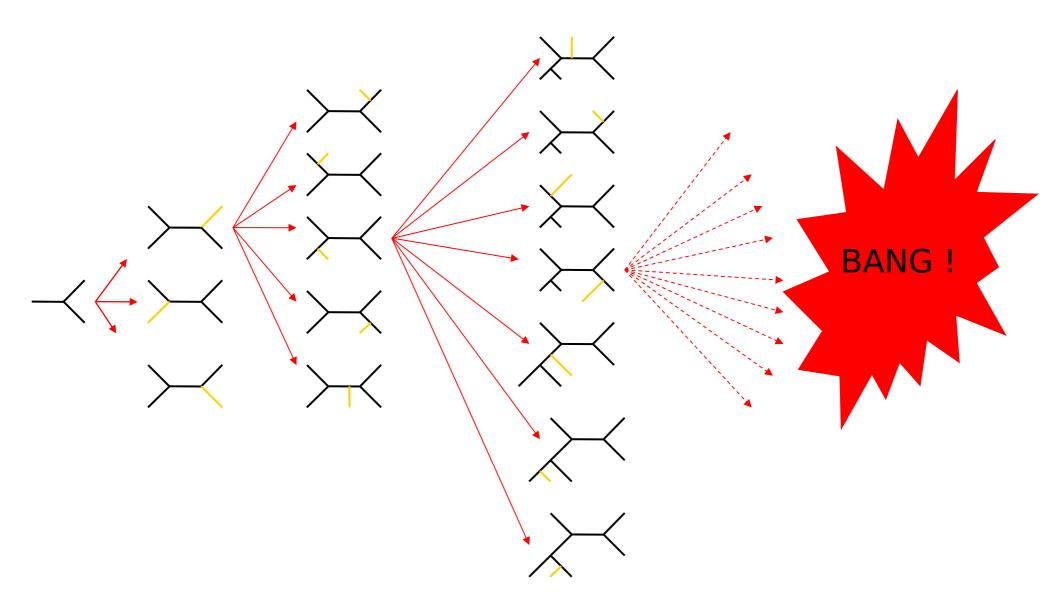
4 taxa → 3 trees



 $5 taxa \rightarrow 15 trees$



The number of trees explodes!



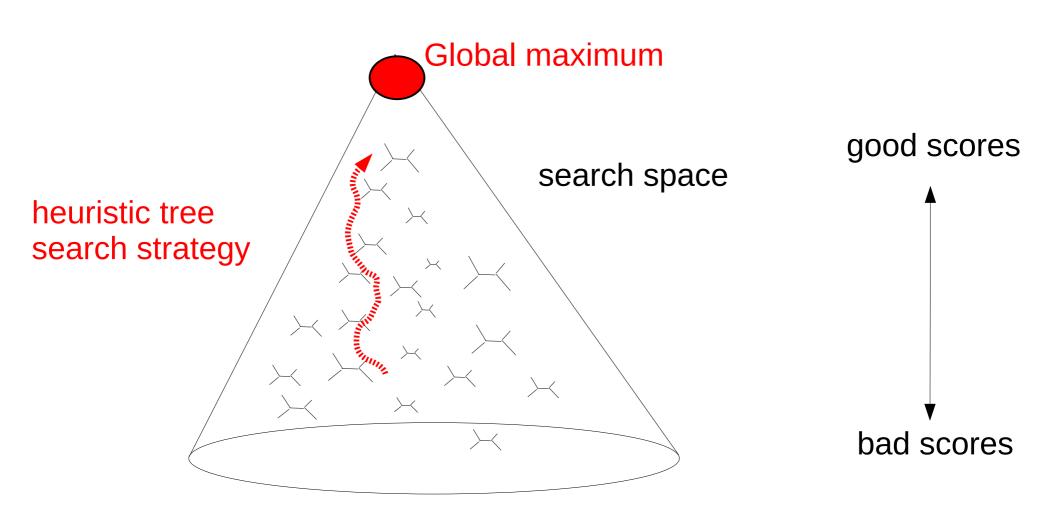
possible trees with 2000 taxa

stamatak@exelixis:~/Desktop/GIT/TreeCounter\$./treeCounter -n 2000

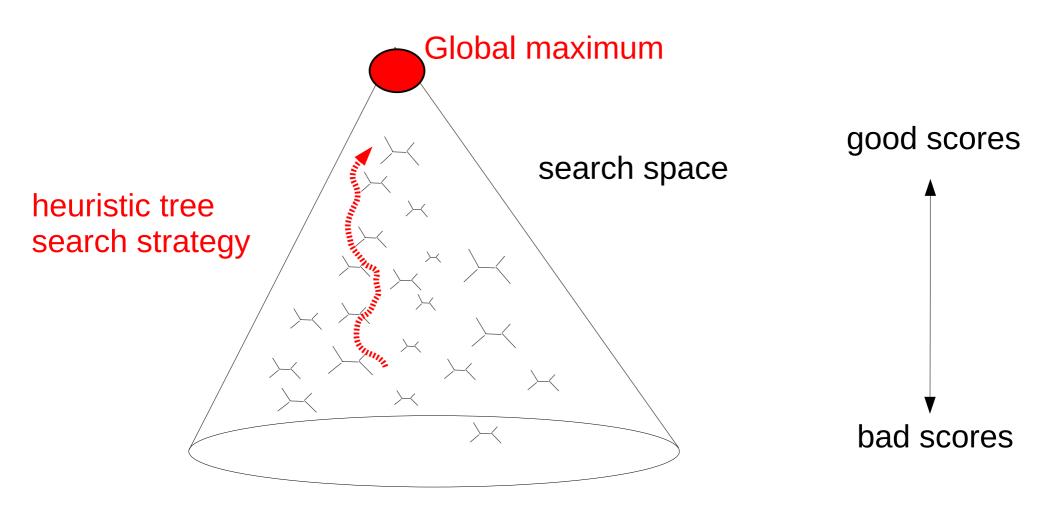
GNU GPL tree number calculator released June 2011 by Alexandros Stamatakis

Number of unrooted binary trees for 2000 taxa: 30049638174211656151632910065681814981377232074237013089504954043012636525258308210827685996688247000464352735 350341796875

Problem Complexity

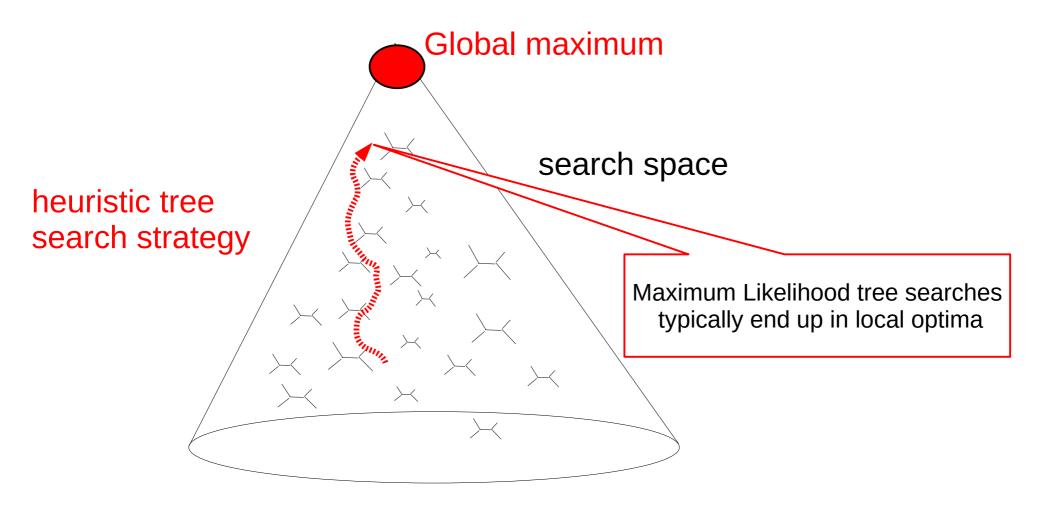


Problem Complexity

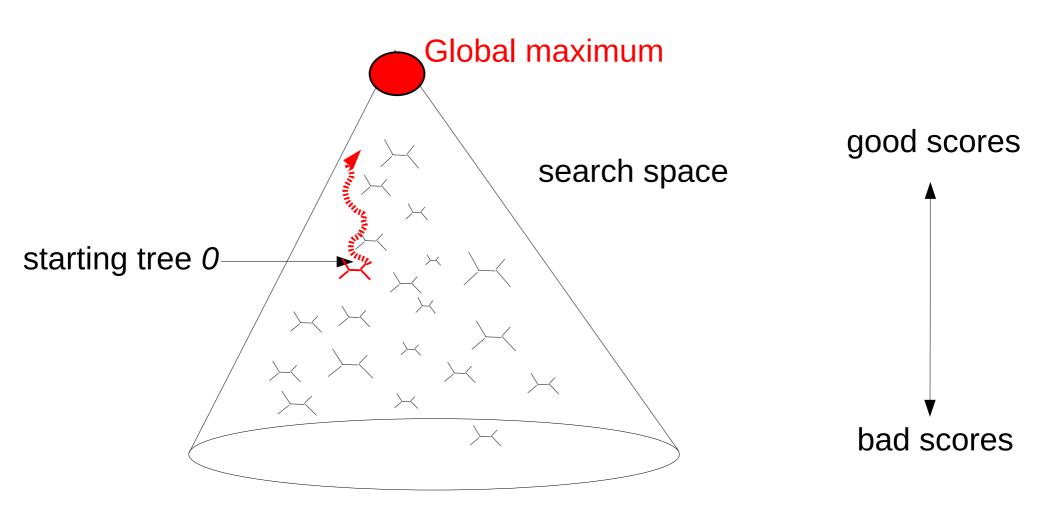


Finding the best tree under Maximum Likelihood is NP-hard!

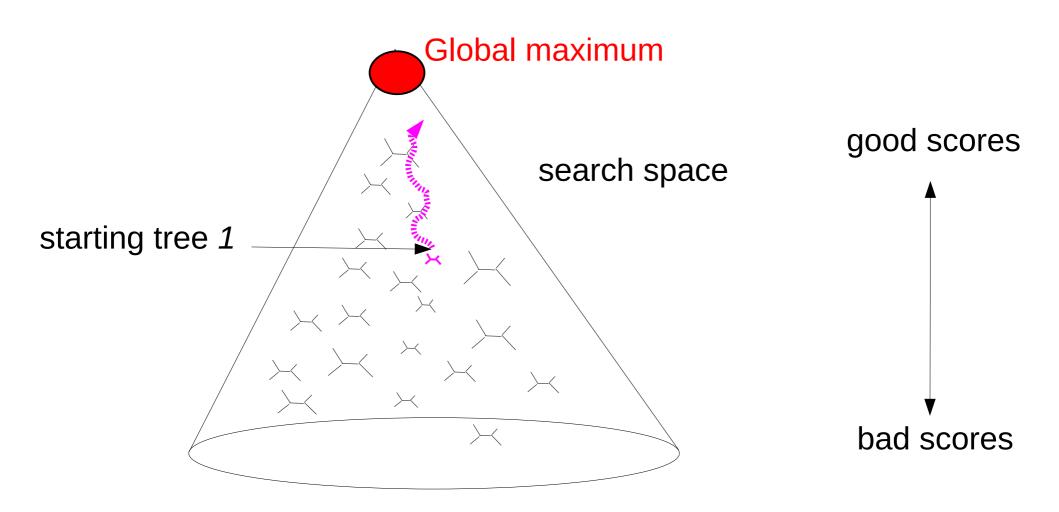
Problem Complexity



Starting Trees



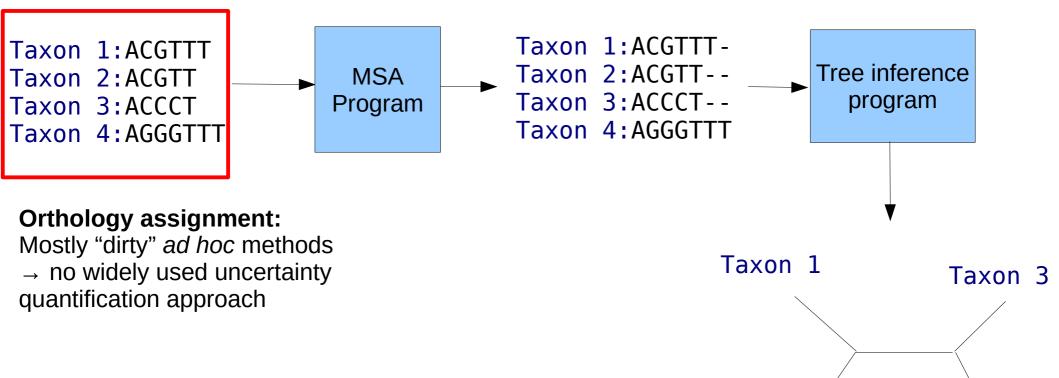
Starting Trees



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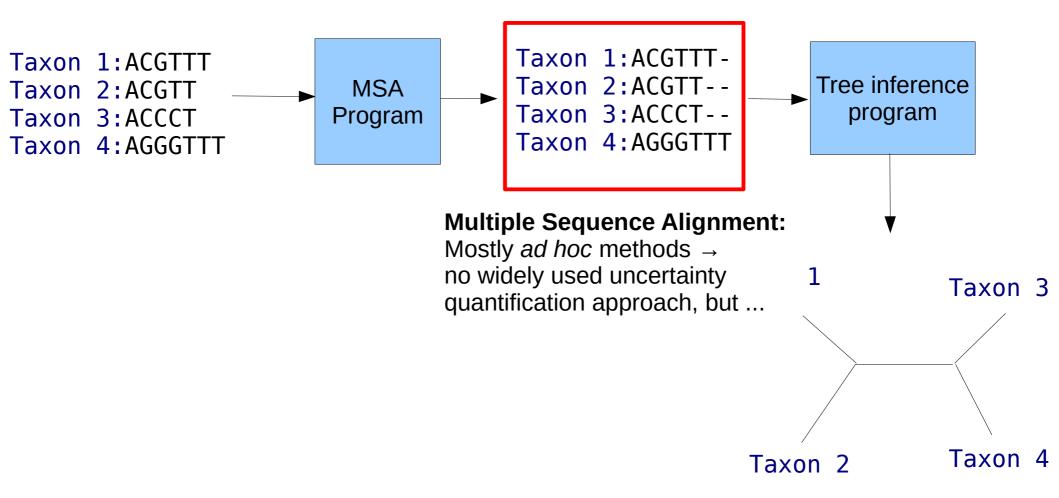
Tree Inference Pipeline



Taxon 4

Taxon 2

Tree Inference Pipeline



Muscle5

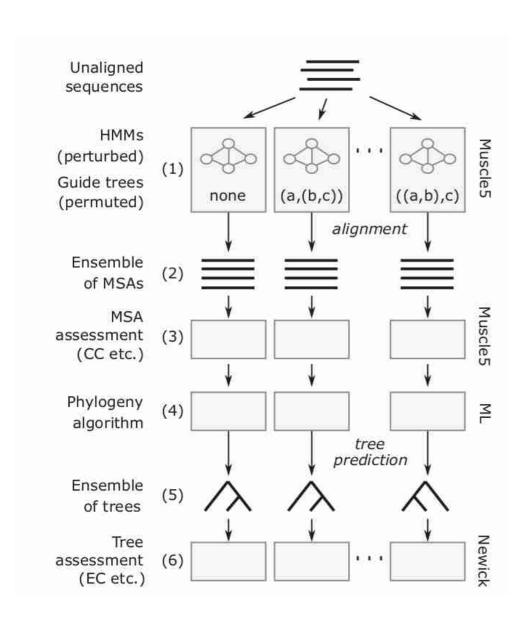
Article Open Access Published: 15 November 2022

Muscle5: High-accuracy alignment ensembles enable unbiased assessments of sequence homology and phylogeny

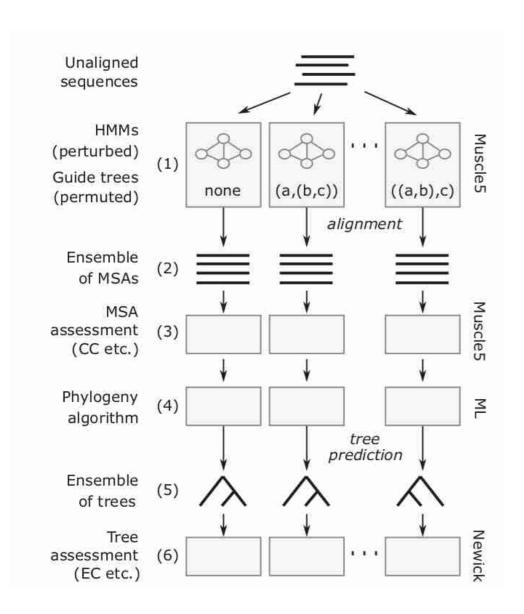
Robert C. Edgar

Nature Communications 13, Article number: 6968 (2022) | Cite this article

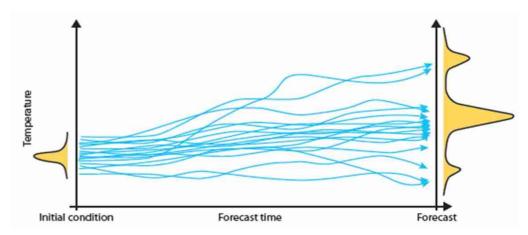
Muscle5



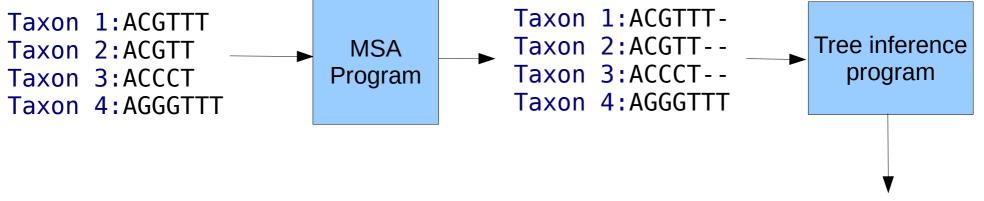
Muscle5



Temperature Ensemble Forecast

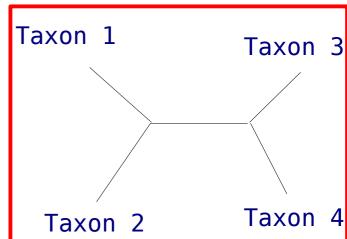


Tree Inference Pipeline

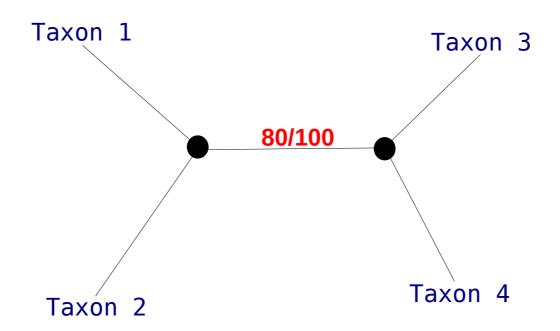


Phylogenetic Inference:

A long history of explicit uncertainty models Bootstrap Methods for Maximum Likelihood Posterior Probabilities for Bayesian Inference using MCMC



A Tree with Support Values



Sources of Uncertainty

- 1) Orthology Assignment
- 2) Multiple Sequence Alignment
- 3) Tree Inference
- 4) **BUT**

Software Issues

- Bugs & Software Quality
- Numerical Instability
- Reproducibility
- We re-designed & optimized numerous tools the Next Generation (NG) tools series
 - RAxML-NG
 - ModelTest-NG
 - EPA-NG
 - Lagrange-NG

Sources of Uncertainty

- 1) Orthology Assignment
- 2) Multiple Sequence Alignment
- 3) Tree Inference
- 4) Software issues
- 5) BUT

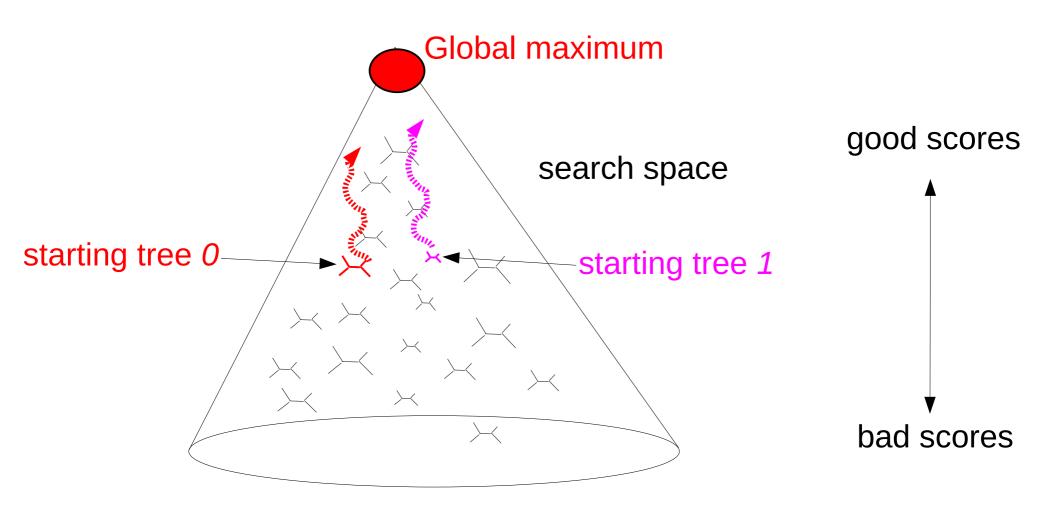
Propagating Uncertainty

- Assume
 - 10 alternative orthology assignments
 - 10 x 10 alternative MSAs
 - 10 x 10 x 10 alternative trees
 - → exponential explosion with increasing pipeline length
 - → intelligent ways to explore parameter space in pipelines needed

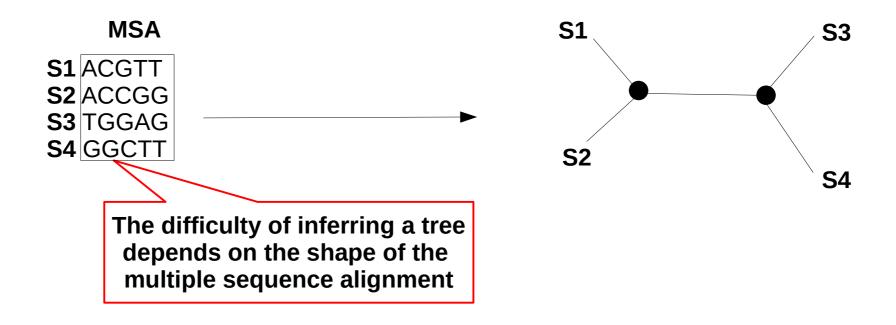
Outline

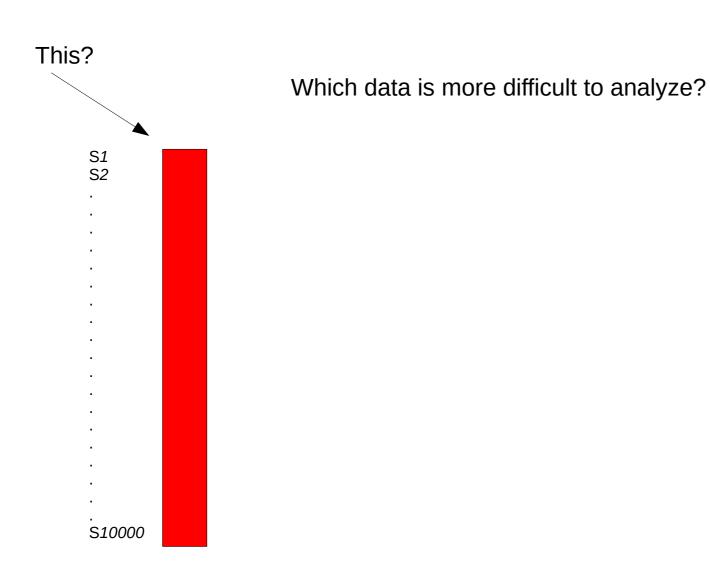
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Can we predict how difficult a phylogenetic analysis will be?

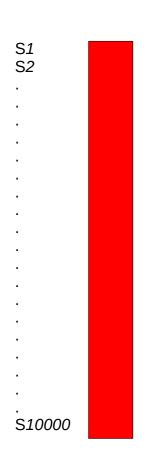


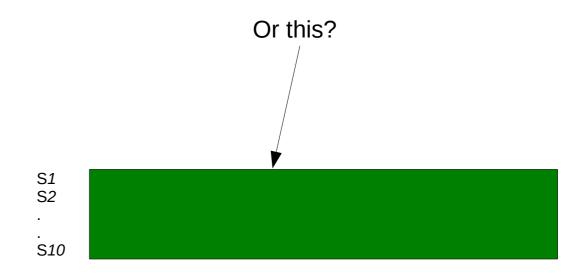
Phylogenetic Inference



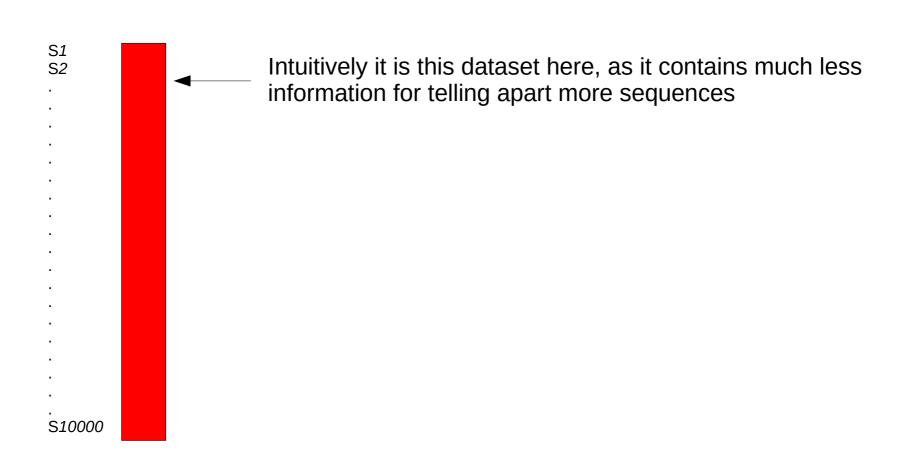


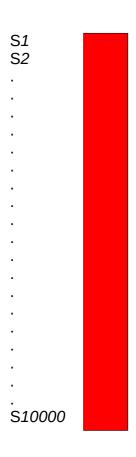
Which data is more difficult to analyze?





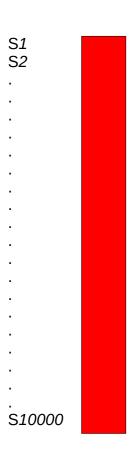
Few sequences, long sequence length





Intuitively it is this dataset here, as it contains much less information for telling apart more sequences

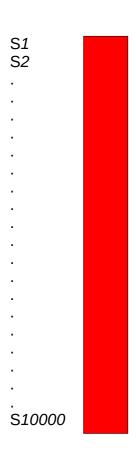
SARS-CoV-2 is such a difficult dataset; it even exhibits some additional difficulties:



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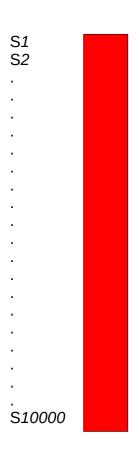
1. Due to the low mutation rate (rate at which nucleotides change) sequences are very similar to each other



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SARS-CoV-2 is such a difficult dataset; it even exhibits some additional difficulties:

- 1. Due to the low mutation rate (rate at which nucleotides change) sequences are very similar to each other
- 2. The genome is \approx 30,000 nucleotides long, but the sequences differ in only 1500-2000 positions \rightarrow highly similar



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- 1. Due to the low mutation rate (rate at which nucleotides change) sequences are very similar to each other
- 2. The genome is \approx 30,000 nucleotides long, but the sequences differ in only 1500-2000 positions \rightarrow highly similar
- 3. The input sequences are **not from distinct species**!

Consequences

- SARS-CoV-2 data
 - Extremely hard to infer a reliable tree
 - Numerical issues with tree inference tools because the sequences are so closely related
 - Post-analyzing the tree (e.g., determining the root, identifying virus sub-classes) appears to not be feasible using computational tools

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For details, see: Benoit Morel, Pierre Barbera, Lucas Czech, Ben Bettisworth, Lukas Hübner, Sarah Lutteropp, Dora Serdari, Evangelia-Georgia Kostaki, Ioannis Mamais, Alexey Kozlov, Pavlos Pavlidis, Dimitrios Paraskevis, Alexandros Stamatakis. "Phylogenetic analysis of SARS-CoV-2 data is difficult", Molecular Biology and Evolution 2021

Phylogenetic Inference

- Assembled 4 distinct input datasets
- Per input dataset → executed 100 tree searches
- As we use likelihood models, we determined the trees that are not statistically significantly different from each other per set of 100 trees

Results

• For all input datasets about 70 out of 100 trees are not significantly different from each other with respect to their likelihood scores

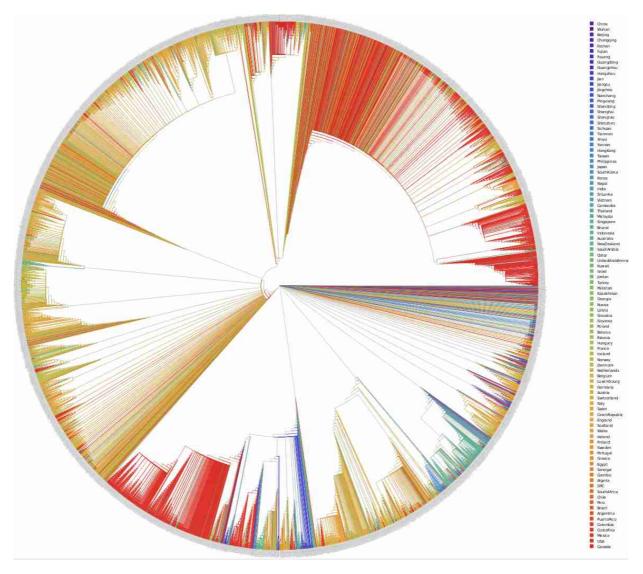
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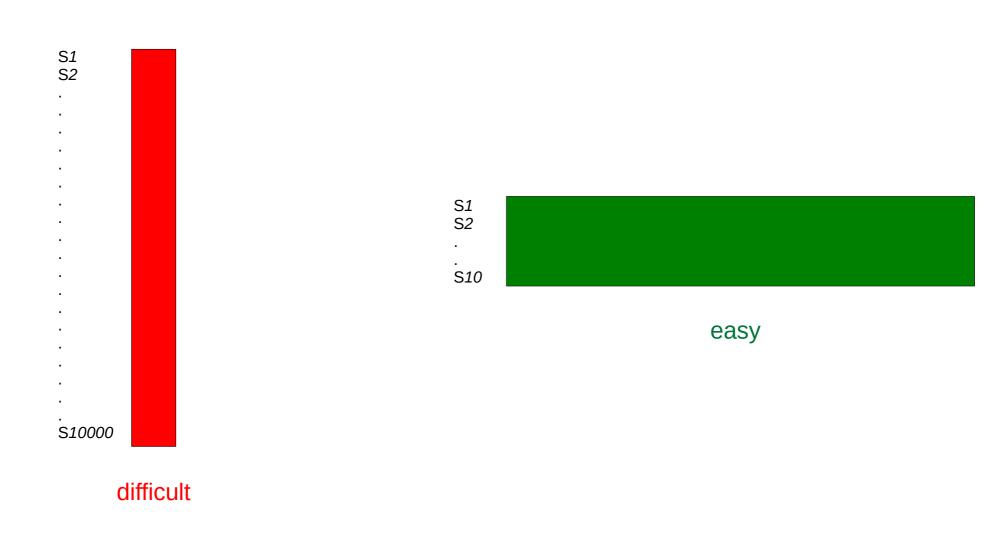
- For all input datasets about 70 out of 100 trees are not significantly different from each other with respect to their likelihood scores
- But, their pair-wise topological differences (difference in tree shapes) amount on average to 70%!
 - → extremely weak signal
 - → don't draw conclusions from a single tree!
 - → try to summarize the trees via summary statistics!

Summarized Trees



SARS-CoV-2 consensus tree colored by country

Difficulty of an MSA



Difficulty of an MSA

This is very hand-wavy → can we quantify & predict this



Difficulty Prediction

JOURNAL ARTICLE

From Easy to Hopeless—Predicting the Difficulty of Phylogenetic Analyses 8

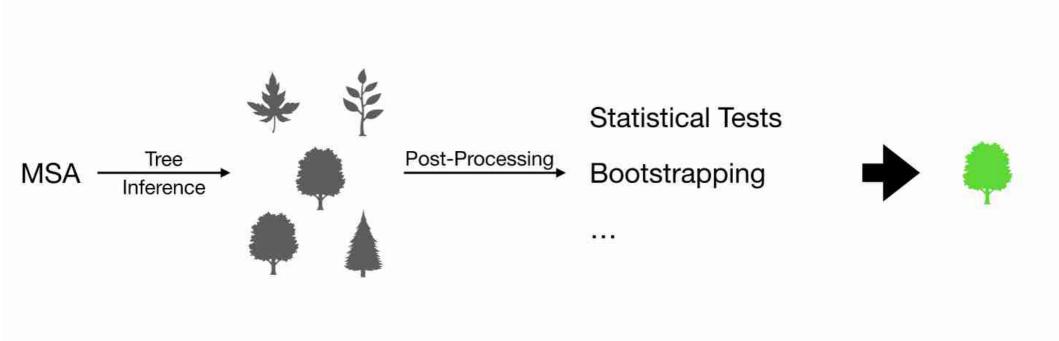
Julia Haag █, Dimitri Höhler, Ben Bettisworth, Alexandros Stamatakis

Molecular Biology and Evolution, Volume 39, Issue 12, December 2022, msac254,

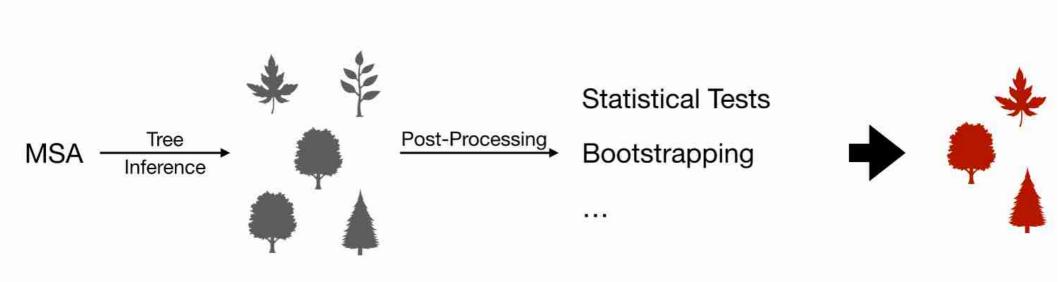
https://doi.org/10.1093/molbev/msac254

Published: 17 November 2022

Easy



Difficult



What does Difficulty mean?

Difficulty = ruggedness of the tree space

Easy

- Few highly similar tree topologies
- Single likelihood peak

Difficult

- Highly distinct topologies, statistically indistinguishable
- Multiple likelihood peaks

Predicting Difficulty with Pythia

- Pythia = Boosted Tree Regressor
- Supervised Regression Task
 - Predict difficulty between 0.0 (easy) and 1 (difficult)
 - Ground truth difficulty as training target based on 100 distinct Maximum Likelihood tree inferences
- Trained on 4K empirical MSAs
 - Mean absolute error: 2.5%

Definition of Difficulty

MSA
$$\frac{\text{Tree Inference}}{(\text{RAxML-NG})}$$
 $\frac{N_{\text{all}} = 100}{N_{\text{pl}}}$ $\frac{\text{Statistical Tests}}{(\text{IQ-Tree})}$ $\frac{N_{\text{pl}}}{N_{\text{pl}}}$ plausible trees $\frac{RF_{\text{pl}}}{N_{\text{pl}}}$ $\frac{RF_{\text{pl}}}{N_{\text{pl}}}$ difficulty(MSA) = $\frac{1}{5} \cdot \left[RF_{\text{all}} + \frac{N_{\text{all}}^*}{N_{\text{all}}} + RF_{\text{pl}} + \frac{N_{\text{pl}}^*}{N_{\text{pl}}} + \left(1 - \frac{N_{\text{pl}}}{N_{\text{all}}} \right) \right]$

Prediction Features

- Eight Features
 - 4 MSA attributes
 - sites-over-taxa, patterns-over-taxa, % gaps, % invariant sites
 - 2 MSA information metrics
 - Shannon entropy, Bollback multinomial test statistic
 - 2 Parsimony-tree-based features
 - Infer 100 parsimony trees
 - → average RF-Distance
 - → % unique topologies

SARS-CoV-2 Example

"Phylogenetic Analysis of SARS-CoV-2 Data Is Difficult" (https://doi.org/10.1093/molbev/msaa314)

The predicted difficulty for MSA examples/covid.fasta is: 0.84.

FEATURES: num_taxa: 4869 num_sites: 28361 [...] num_sites/num_taxa: 5.82 [...] avg_rfdist_parsimony: 0.79 proportion_unique_topos_parsimony: 1.0 Feature computation runtime: 1830.182 seconds

[...]

Using Pythia

- Prior to tree inference
 - → determine analysis & post-analysis setup
 - → adjust/modify MSA
 - → explore data filtering & assembly strategies
 - → adjust user expectations about data

Pythia developments

- New release (yesterday)
 - Trained on 12K datasets
 - 11,108 DNA MSAs
 - 979 Protein MSAs
 - 460 Morphological MSAs
 - Two new features
 - The new predictor shows an improved accuracy:
 - Mean absolute error: 0.07 (previously 0.09)
 - Mean absolute percentage error: 1.7% (previously 2.5%)
- Follow-up projects
 - A Simulation study
 - Deploy to inform tree search heuristics

Simulation Study





New Results

Follow this preprint

A representative Performance Assessment of Maximum Likelihood based Phylogenetic Inference Tools

Dimitri Höhler, Julia Haag, O Alexey M. Kozlov, Alexandros Stamatakis doi: https://doi.org/10.1101/2022.10.31.514545

This article is a preprint and has not been certified by peer review [what does this mean?].

Accuracy as Function of Difficulty

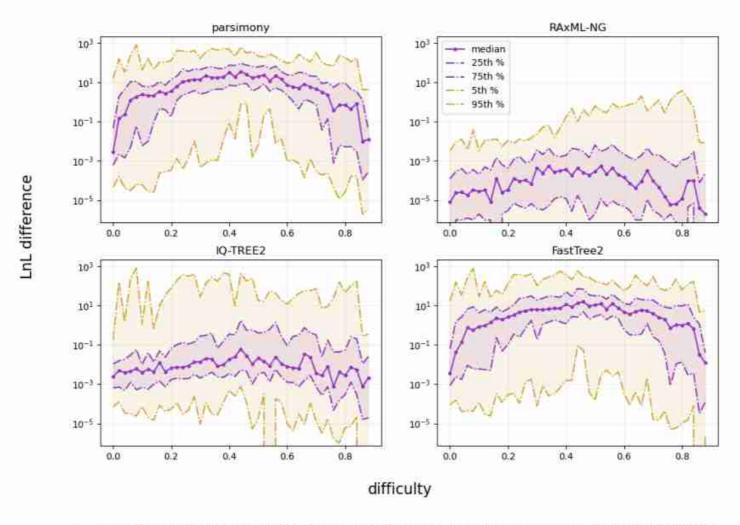


Fig. 3. Absolute log-likelihood (LnL) score differences (log scale) from the best-known ML tree on TreeBASE data.

Adaptive RAxML-NG

New Results

Adaptive RAxML-NG: Accelerating Phylogenetic inference under Maximum Likelihood using dataset difficulty

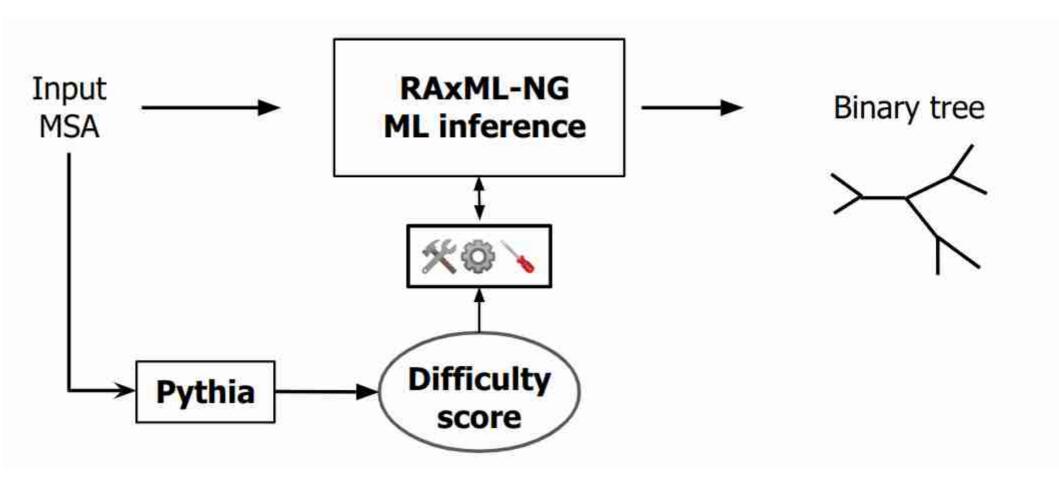
Anastasis Togkousidis,
Alexandros Stamatakis

doi: https://doi.org/10.1101/2023.05.15.540873

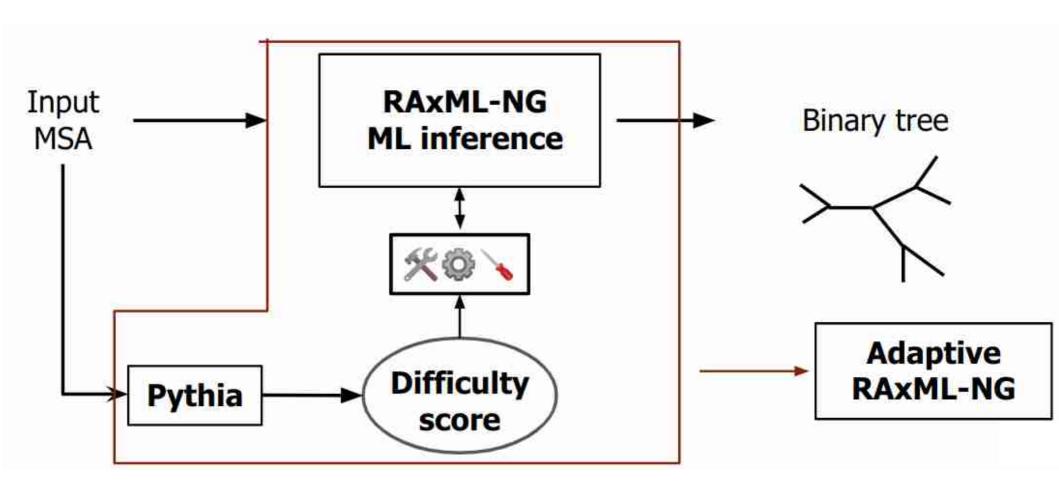
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Preprint published two days ago - May 18

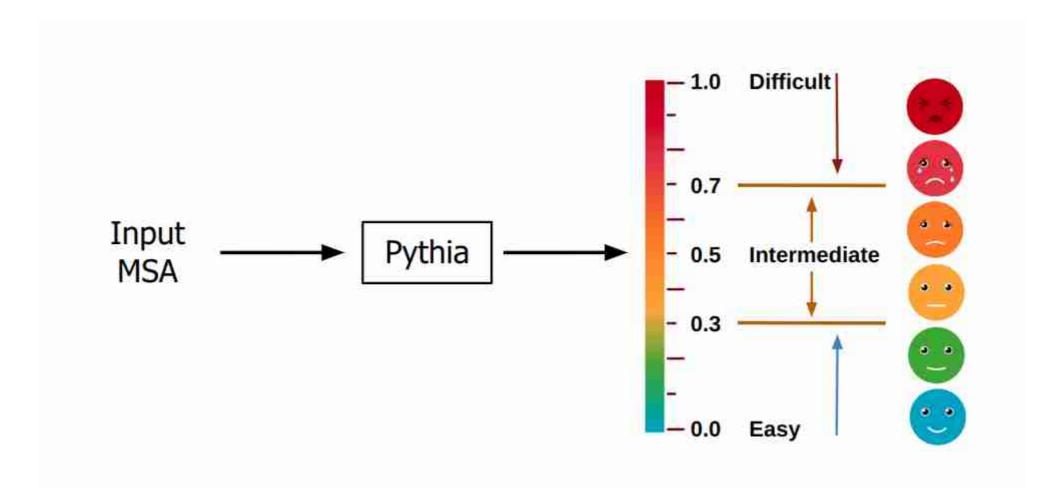
Adaptive RAxML-NG



Adaptive RAxML-NG



Pythia



Adaptive RAXML-NG Heuristics

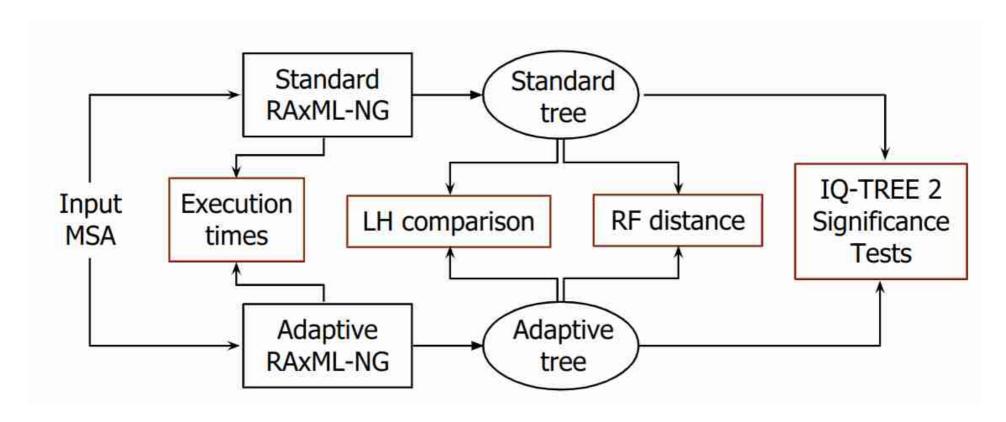
- Modify as a function of difficulty
 - the number of ML tree searches
 - the thoroughness of the search
- And introduce an additional tree search mechanism

Test Data & Setup

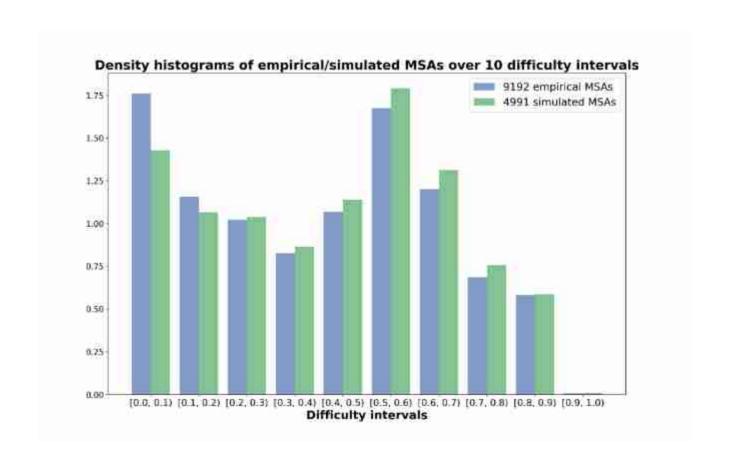
- 10K empirical MSAs from TreeBase
 - → 9192 MSAs after filtering

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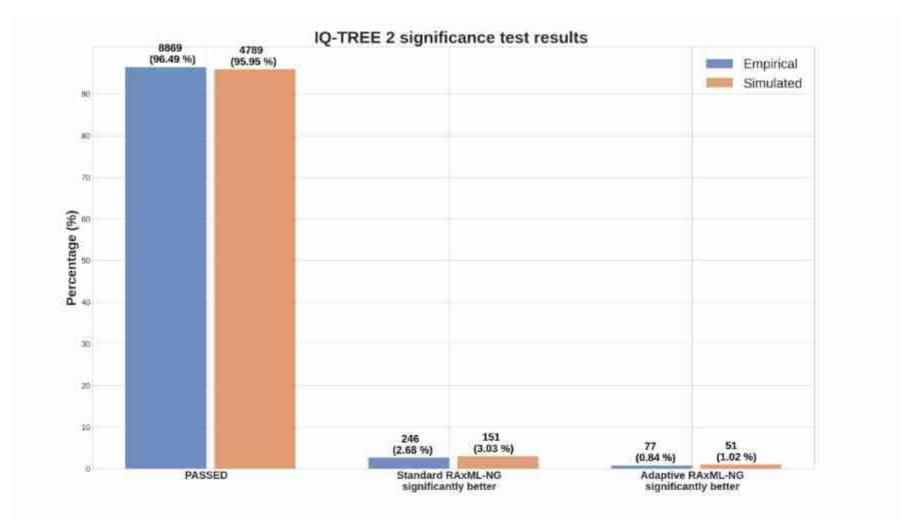
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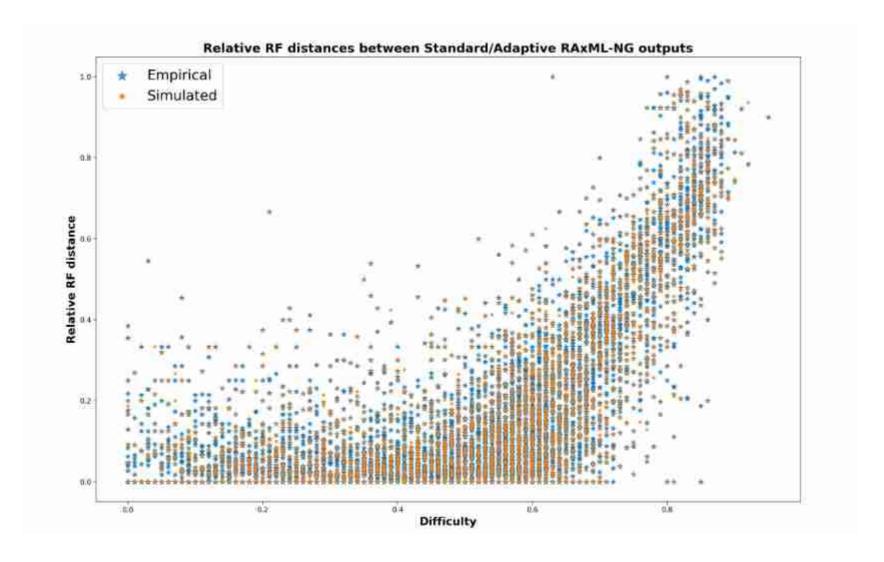
Difficulty Score Distribution

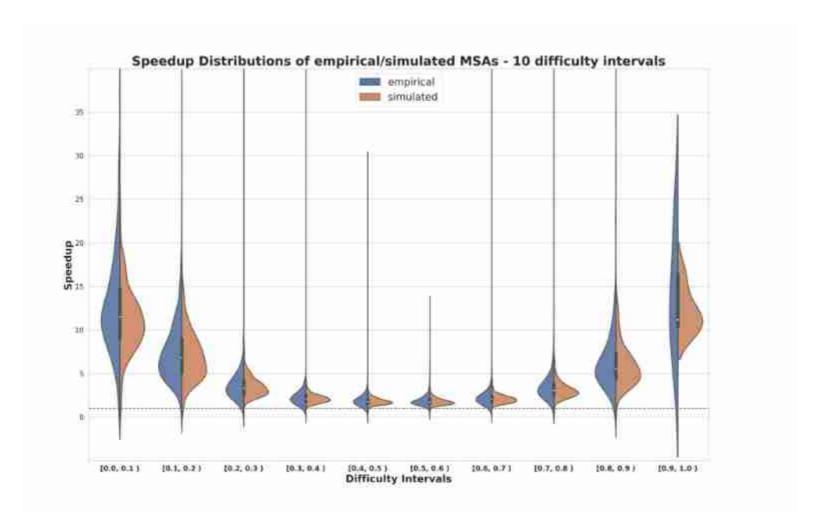


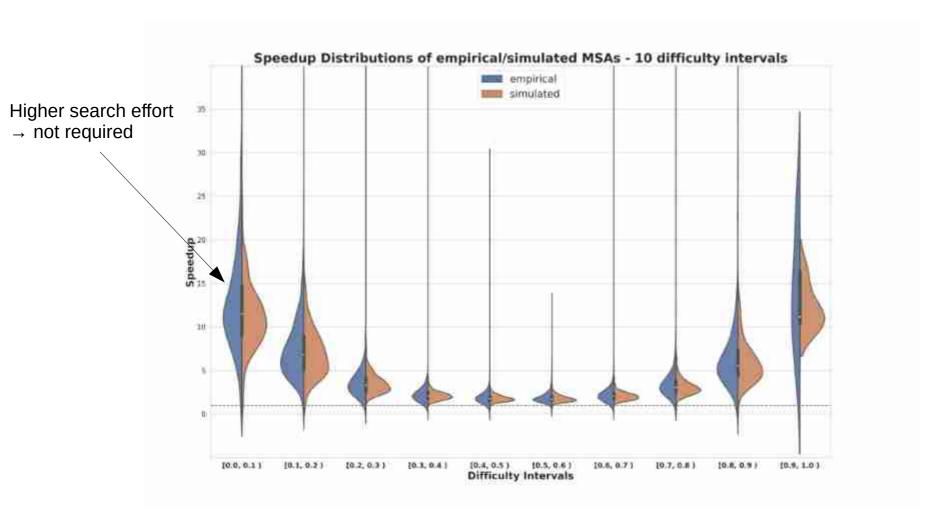
Significance Tests

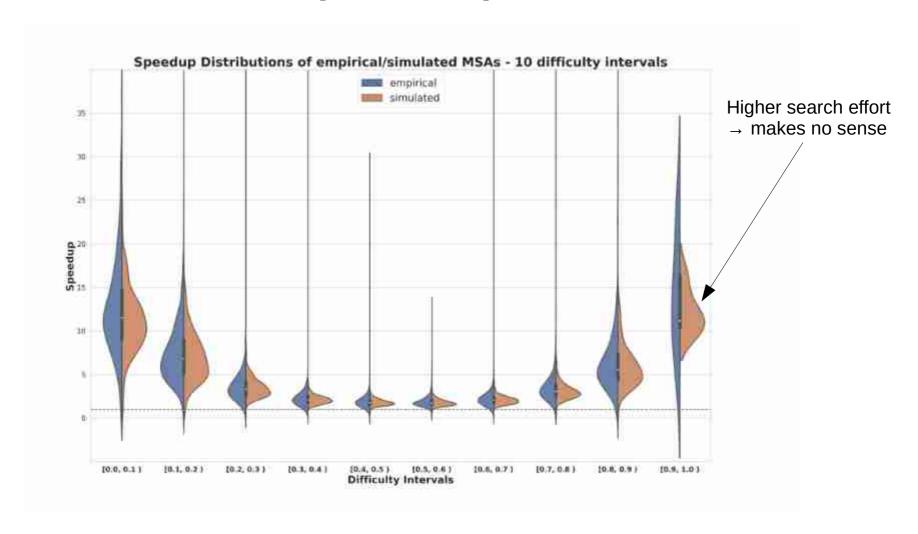


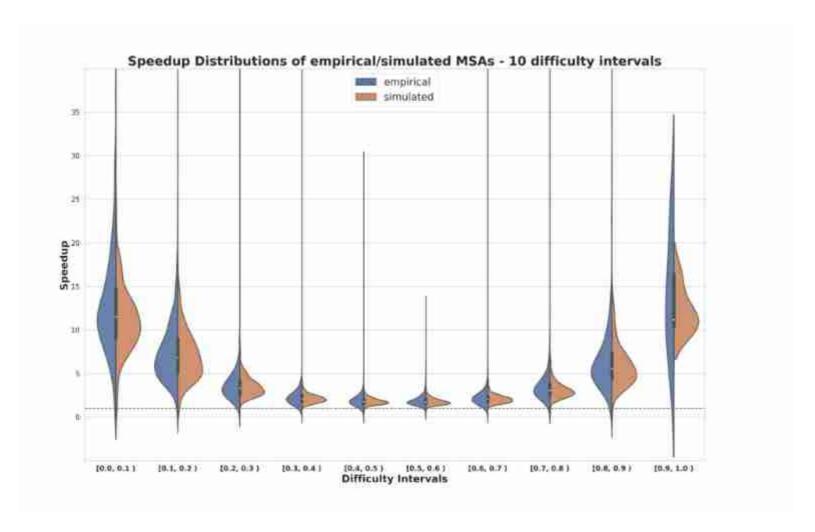
Distances between trees











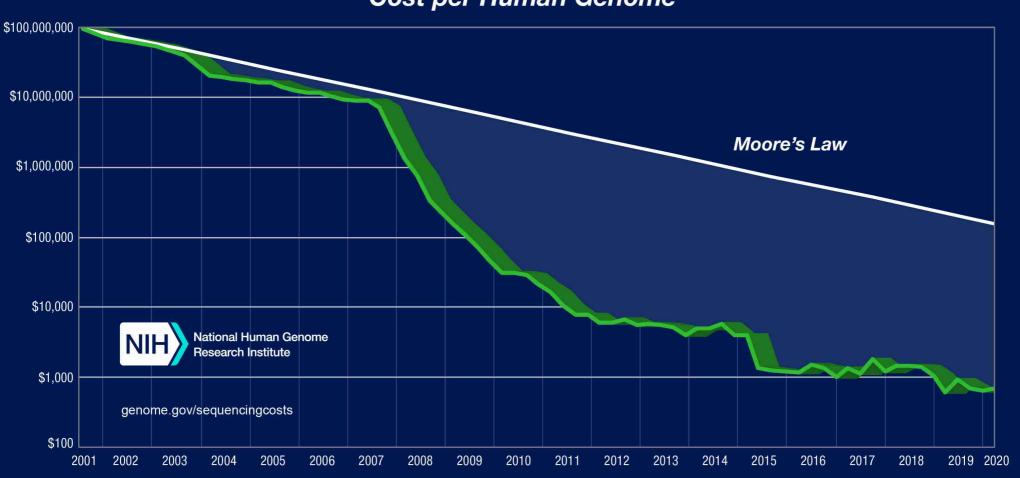
Overall accumulated speedup: approx. 3 on empirical data

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Scalability

Cost per Human Genome



Single Cell Evolution

- Reconstructing the evolution, e.g., of cancer cells in a single patient is challenging
 - Noisy data
 - Erroneous data
 - Little signal
 - Few & simplistic models

Eleven grand challenges in single-cell data science

David Lähnemann, Johannes Köster, [...] Alexander Schönhuth

Genome Biology 21, Article number: 31 (2020) | Cite this article

32k Accesses | 16 Citations | 281 Altmetric | Metrics | New Results |

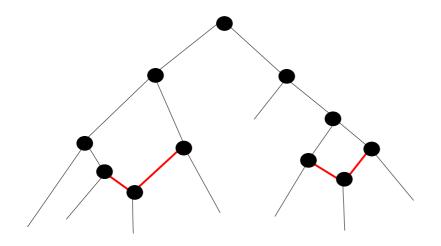
CellPhy: accurate and fast probabilistic inference of single-cell phylogenies from scDNA-seq data

Alexey Kozlov, David Posada doi: https://doi.org/10.1101/2020.07.31.230292

This article is a preprint and has not been certified by peer review [what does this mean?].

Phylogenetic Networks

- Evolution does not need to occur in a tree-like manner due to recombination events
- We can model this via so-called phylogenetic networks



Phylogenetic Networks

- Evolution does not need to occur in a tree-like manner due to recombination events
- We can model this via so-called phylogenetic networks
- The likelihood of such a network is substantially more difficult to compute than on a tree
 - → computational challenges



Gene Tree Species Tree Reconciliation

- There are other phenomena that complicate evolution
 - Gene loss
 - Gene transfer
 - Gene duplication
 - → gene tree ≠ species tree
- Infer & correct trees under a joint likelihood model comprising the phylogenetic likelihood and a reconciliation likelihood model

GeneRax

 First full and efficient Maximum Likelihood implementation to infer gene family trees using a given rooted species tree under a joint phylogenetic & reconciliation likelihood model

GeneRax: A Tool for Species-Tree-Aware Maximum Likelihood-Based Gene Family Tree Inference under Gene Duplication, Transfer, and Loss &

Benoit Morel ➡, Alexey M Kozlov, Alexandros Stamatakis, Gergely J Szöllősi

Molecular Biology and Evolution, Volume 37, Issue 9, September 2020, Pages 2763–2774, https://doi.org/10.1093/molbev/msaa141

Published: 05 June 2020

SpeciesRax

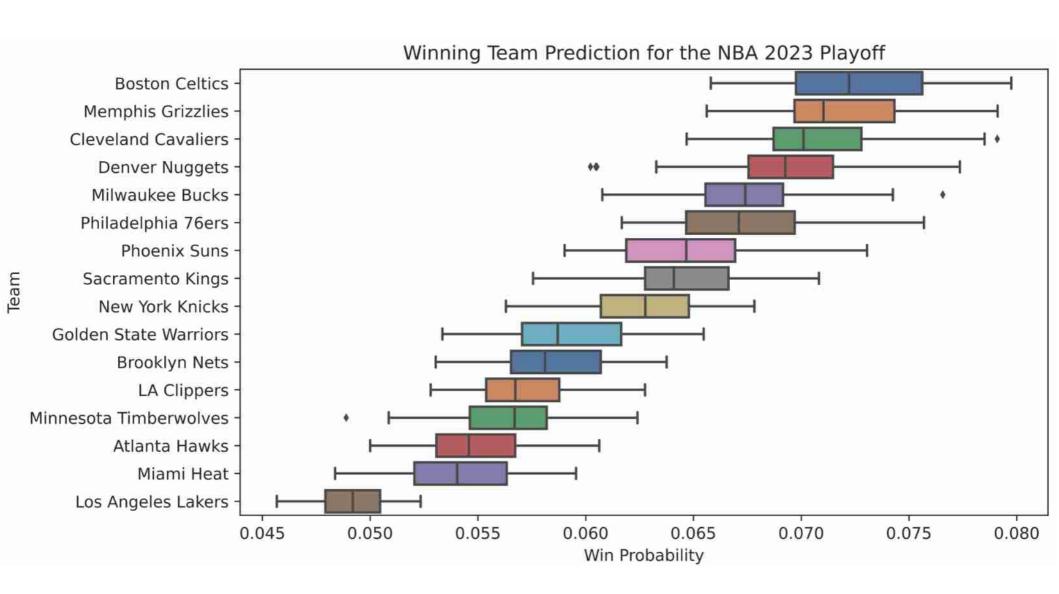
 Goal: Simultaneously infer the gene family trees and the species tree under a joint phylogenetic/reconciliation likelihood model

Species Rax: A Tool for Maximum Likelihood
Species Tree Inference from Gene Family Trees
under Duplication, Transfer, and Loss ∂
Benoit Morel ☒, Paul Schade, Sarah Lutteropp, Tom A Williams, Gergely J Szöllősi,
Alexandros Stamatakis

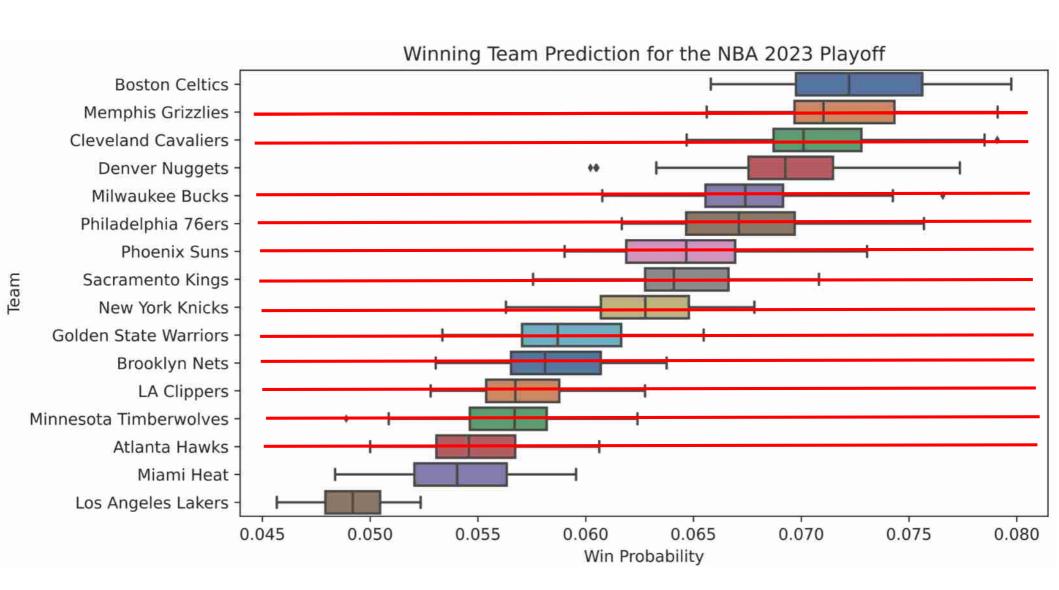
Molecular Biology and Evolution, Volume 39, Issue 2, February 2022, msab365,
https://doi.org/10.1093/molbev/msab365

Published: 11 January 2022

Tournament Prediction



Tournament Prediction



Software Quality Assessment

 SoftWipe tool for automatic scientific software quality assessment (C and C++)

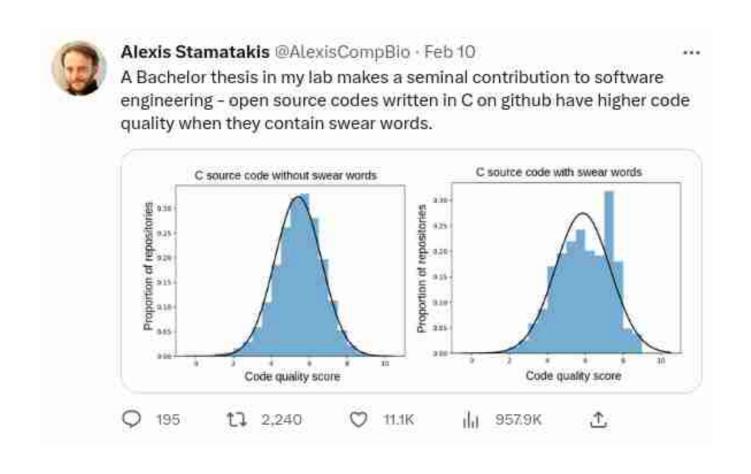
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The SoftWipe tool and benchmark for assessing coding standards adherence of scientific software

Adrian Zapletal, Dimitri Höhler, Carsten Sinz & Alexandros Stamatakis 

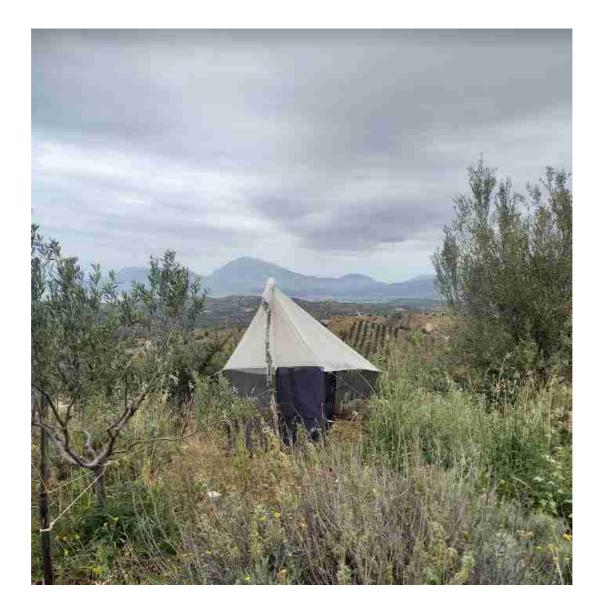
Scientific Reports 11, Article number: 10015 (2021) | Cite this article

4270 Accesses | 1 Citations | 115 Altmetric | Metrics
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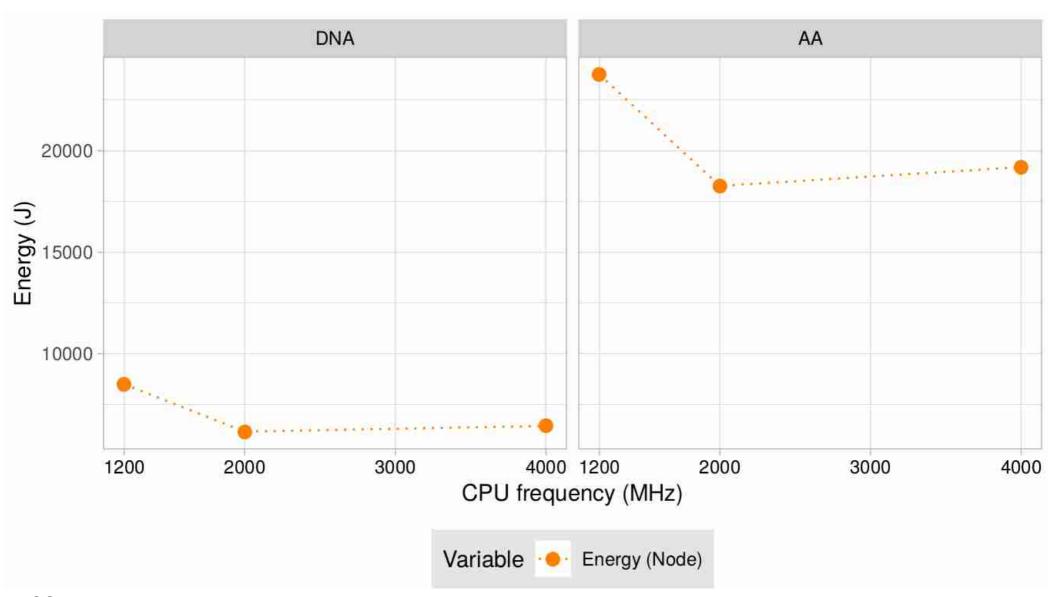
Empirical Software Engineering with SoftWipe



Biological Field Work

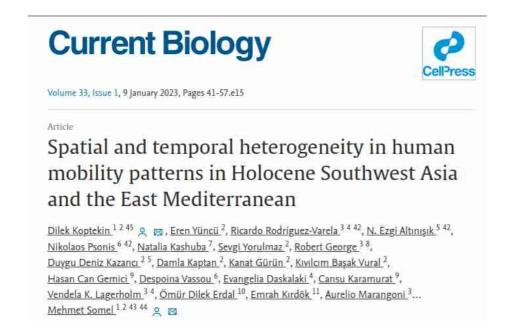


Energy Efficiency

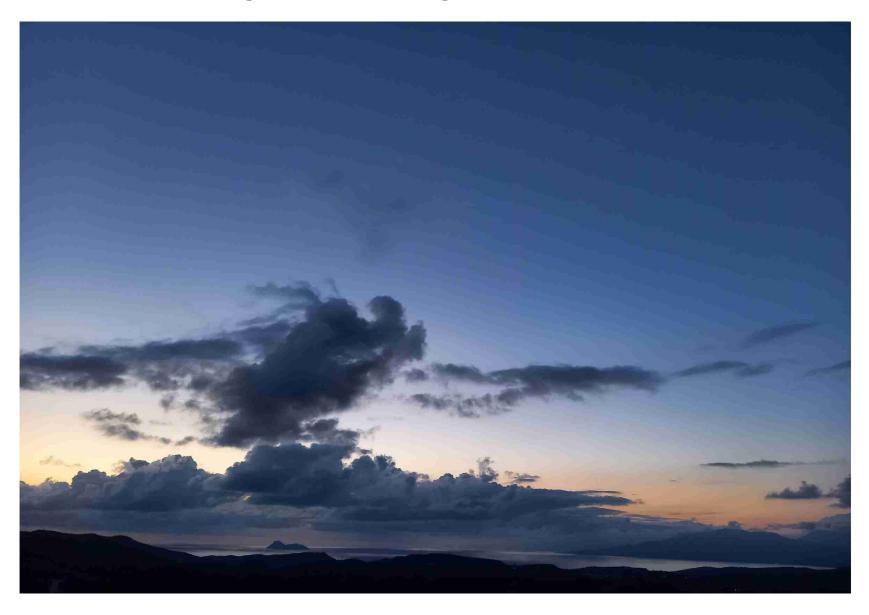


Ancient DNA

- Better tools for ancient DNA analyses
- Classic aDNA data analyses



Thank you for your attention



Pipeline Complexity

the good old days

Sequence

```
T1 ACGT
```

T2 ACC

T3 ACGG

T4 AAGC

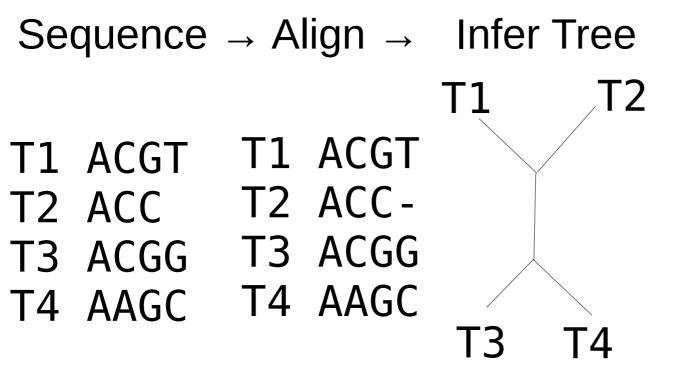
single gene & few species

the good old days

Sequence → Align

```
T1 ACGT T1 ACGT
T2 ACC T2 ACC-
T3 ACGG T3 ACGG
T4 AAGC T4 AAGC
```

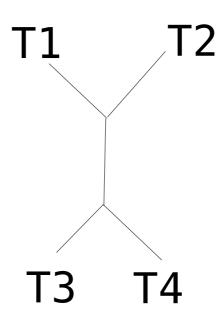
the good old days



the good old days

Sequence → Align → Infer Tree → Publish

T1 ACGT T1 ACGT
T2 ACC T2 ACCT3 ACGG T3 ACGG
T4 AAGC T4 AAGC



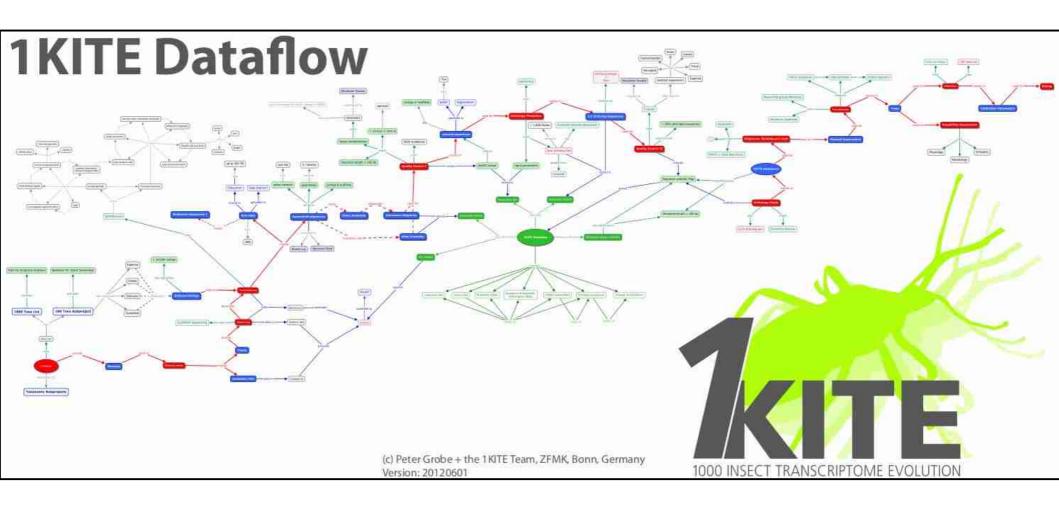


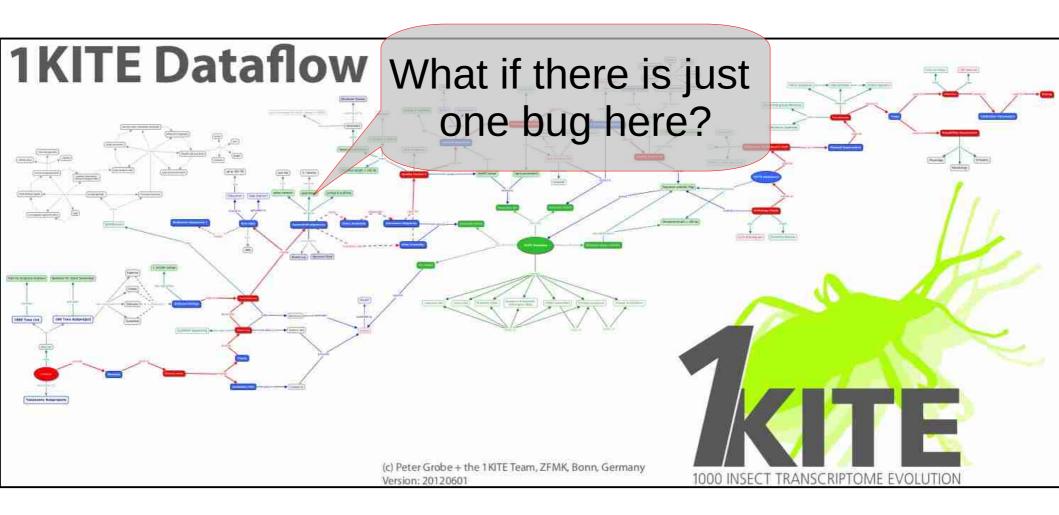


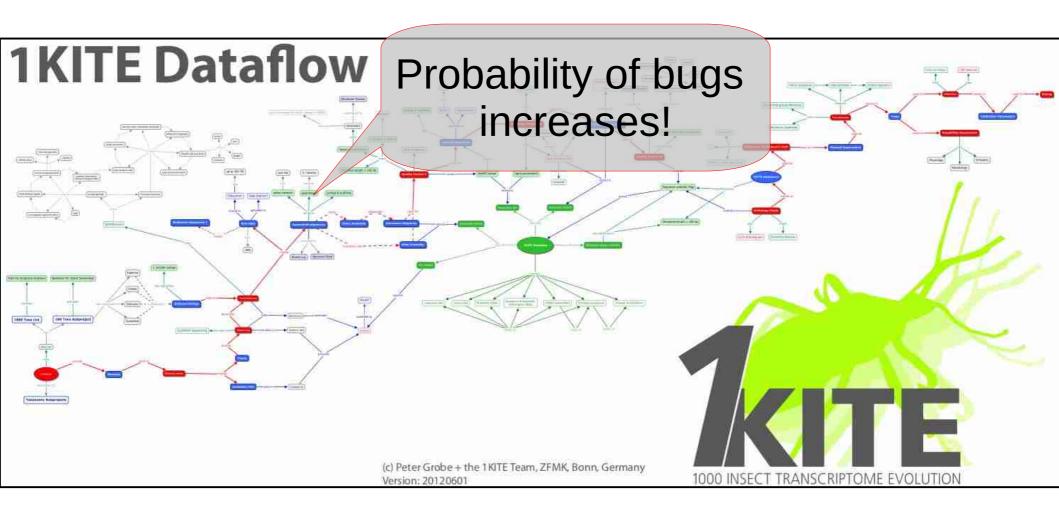
150 insect transcriptomes

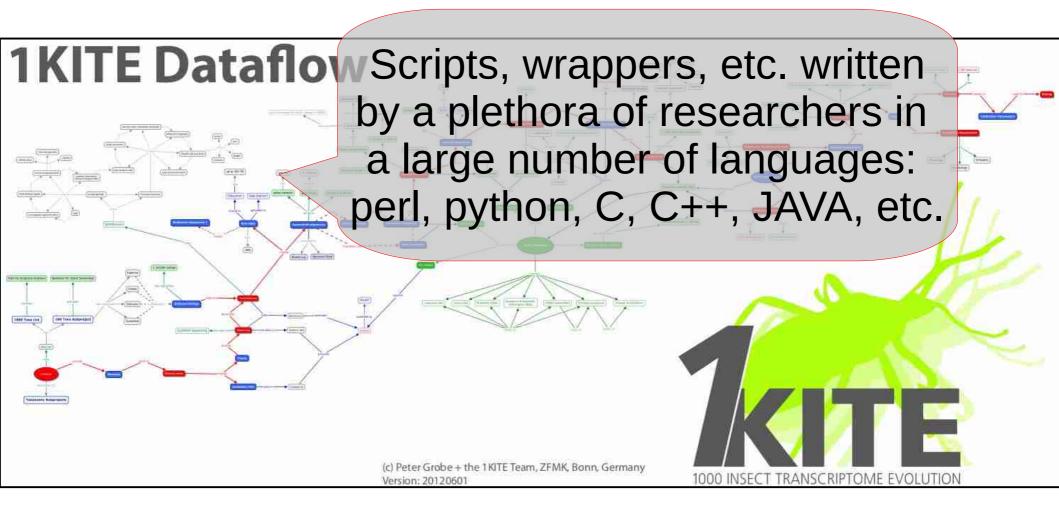


50 bird genomes









The 'crappy' software project

- Analyzed 15 widely-used evolutionary biology tools ≈ 65,000 citations
- Analyses performed
 - Compiled with gcc and clang with all warnings enabled
 - Memory check with valgrind
 - Checked if assertions are used via assert()
 - Analyzed degree of code duplication
- Caution: "bad" quality does not induce that a tool is faulty, but the probability of it being faulty is higher!

The State of Software for Evolutionary Biology ∂ Diego Darriba, Tomáš Flouri, Alexandros Stamatakis Molecular Biology and Evolution, Volume 35, Issue 5, May 2018, Pages 1037–1046, https://doi.org/10.1093/molbev/msy014 Published: 29 January 2018

SoftWipe

- Discussion with Science Journalist "Can this process be automated?"
- Development of SoftWipe An automated tool and benchmark for relative quality ranking of scientific software
- Ranking of 51 open source tools written in C or C++ from a wide range of research areas
 - Astrophysics
 - Computer Science
 - Bioinformatics

New Results

O Comment on this paper

SoftWipe - a tool and benchmark to assess scientific software quality

Adrian Zapletal, Dimitri Hoehler, Carsten Sinz, Alexandros Stamatakis doi: https://doi.org/10.1101/2020.10.07.330621

program name	absolute score	
genesis	8.6	8.8
hyperphylo	8.6	8.6
kahypar	8.4	8.5
candy-kingdom	8.2	8.2
bindash-1.0	8.0	7.9
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axe-0.3.3	7.5	7.5
virulign-1.0.1	7.4	7.4
naf-1.1.0/unnaf	7.4	7.5
naf-1.1.0/ennaf	7.4	7.4
ExpansionHunter	7.3	7.5
glucose-3-drup	7.1	7.0
raxml-ng	7.0	7.0
dawg	6.8	6.9
ntEdit-1.2.3	6.4	6.2
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lemon	6.1	6.0
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BGSA_CPU-1.0	5.9	5.4
emeraLD	5.8	5.5
dr_sasa_n	5.7	6.0
copmem-0.2	5.7	5.7
samtools	5.6	5.6
seq-gen	5.6	5.6
dna-nn-0.1	5.3	5.2
sf	5.2	5.2
cryfa-18.06	5.1	5.1
ngsLD	5.1	5.0
HLA-LA	4.9	4.5
igtree 1.6.10	4.9	4.9
vsearch	4,6	4.6
prank	4.6	4.5
prequal	4.5	4.4
minimap	4.5	4.4
phyml	4.4	4.4
clustal	4.2	4.3
mrbayes	4.1	4.1
tcoffee	4.1	4.2
gadget	4.1	4.0
gaoget crisflash	4.0	4.0
erisnasn PopLDdecay	3.8	
	100000	3.8
cellcoal	3.8	3.6
bpp	3.8	3.6
ms	3.7	3.7
mafft	3.3	3.1
athena	2.9	2.8
covid-sim-0.13.0	2.5	2.4
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SoftWipe Benchmark

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SoftWipe Benchmark

Does not change over time as more tools are added \rightarrow can easily be referenced

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and continuing the	3.8	3.8
PopLDdecay		
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mafft	3.3	3.1
athena	2.9	2.8
covid-sim-0.13.0	2.5	2.4
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Does change over time as more tools are added \rightarrow Difficult to be referenced

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	5.9	5.4
BGSA_CPU-1.0		
emeraLD	5.8	5.5
dr_sasa_n	5.7	6.0
copmem-0.2	5,7	5.7
samtools	5.6	5.6
seq-gen	5.6	5.6
dna-nn-0.1	5.3	5.2
sf	5,2	5.2
cryfa-18.06	5.1	5.1
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phyml	4.4	4.4
clustal	4.2	4.3
mrbayes	4.1	4.1
coffee	4.1	4.2
gadget	4.1	4.0
crisflash	4.0	4.0
PopLDdecay	3.8	3.8
cellcoal	3.8	3.6
opp	3.8	3.6
ms	3.7	3.7
mafft	3.3	3.1
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Written by computer scientists

SoftWipe Benchmark

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repeatscounter	7.5	7.7
ixe-0.3.3	7.5	7.5
rirulign-1.0.1	7.4	7.4
af-1.1.0/unnaf	7.4	7.5
naf-1.1.0/ennaf	7.4	7.4
ExpansionHunter	7.3	7.5
glucose-3-drup	7.1	7.0
axml-ng	7.0	7.0
lawg	6.8	6.9
ntEdit-1.2.3	6.4	6.2
lefor	6.3	6.4
warm	6.2	6.2
emon	6.1	6.0
reerecs	6.1	6.1
Q-TREE-2.0-rc1	6.1	5.7
BGSA_CPU-1.0	5.9	5.4
meraLD	5.8	5.5
ir sasa n	5.7	6.0
opmem-0.2	5.7	5.7
samtools	5.6	5.6
eq-gen	5.6	5.6
ina-nn-0.1	5.3	5.2
ana-un-o, r	5.2	5.2
ryfa-18.06	5.1	5.1
	5.1	
ngsLD		5.0
HLA-LA	4.9	4.5
qtree 1.6.10	4.9	4.9
search	4,6	4.6
orank	4.6	4.5
requal	4.5	4.4
ninimap	4.5	4.4
ohyml	4.4	4.4
lustal	4.2	4.3
nrbayes	4.1	4.1
coffee	4.1	4.2
gadget	4.1	4.0
risflash	4.0	4.0
PopLDdecay	3.8	3.8
ellcoal	3.8	3.6
opp	3.8	3.6
ns	3.7	3.7
nafft	3.3	3.1
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ovid-sim-0.13.0	2.5	2.4
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My lab

SoftWipe Benchmark

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naf-1.1.0/unnaf	7.4	7.5
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swarm	6.2	6.2
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treerecs	6.1	6.1
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BGSA_CPU-1.0	5.9	5.4
emeraLD	5.8	5.5
dr_sasa_n	5.7	6.0
copmem-0.2	5.7	5.7
samtools	5.6	5.6
seq-gen	5.6	5.6
dna-nn-0.1	5.3	5.2
sf	5,2	5.2
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phyml	4.4	4.4
clustal	4.2	4.3
mrbayes	4.1	4.1
tcoffee	4.1	4.2
gadget	4.1	4.0
crisflash	4.0	4.0
PopLDdecay	3.8	3.8
cellcoal	3.8	3.6
bpp	3.8	3.6
ms	3.7	3.7
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Astrophysics

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BGSA_CPU-1.0	5.9	5.4
emeraLD	5.8	5.5
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copmem-0.2	5.7	5.7
samtools	5.6	5.6
seq-gen	5.6	5.6
dna-nn-0.1	5.3	5.2
sf	5.2	5.2
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phyml	4.4	1/1
clustal	4.2	4.3
mrbayes	4.1	4.1
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gadget	4.1	4.0
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Tools with highly similar functionality

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swarm	6.2	6.2
lemon	6.1	6.0
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IQ-TREE-2.0-rc1	6.1	5.7
BGSA CPU-1.0	5.9	5.4
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samtools	5.6	5.6
seq-gen	5.6	5.6
dna-nn-0.1	5.3	5.2
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tcoffee	4.1	4.2
gadget	4.1	4.0
crisflash	4.0	4.0
PopLDdecay	3.8	3.8
cellcoal	3.8	3.6
bpp	3.8	
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IQ-TREE-2.0-rc1	6.1	5.7
BGSA_CPU-1.0	5.9	5.4
emeral_D	5.8	5.5
The state of the s	5.7	6.0
dr_sasa_n	5.7	
copmem-0.2		5.7
samtools	5.6	5.6
seq-gen	5.6	5.6
dna-nn-0.1	5.3	5.2
sf	5,2	5.2
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mrbayes	4.1	4.1
tcoffee	4.1	4.2
gadget	4.1	4.0
crisflash	4.0	4.0
PopLDdecay		7.0
cellcoal	3.8	3.8
bpp	3.8	3.6
	3.8 3.8	3.6 3.6
ms	3.8 3.8 3.7	3.6
ms mafft	3.8 3.8 3.7 3.3	3.6 3.6
	3.8 3.8 3.7	3.6 3.6 3.7
mafft	3.8 3.8 3.7 3.3	3.6 3.6 3.7 3.1



Covid simulation tool

SoftWipe in Practice

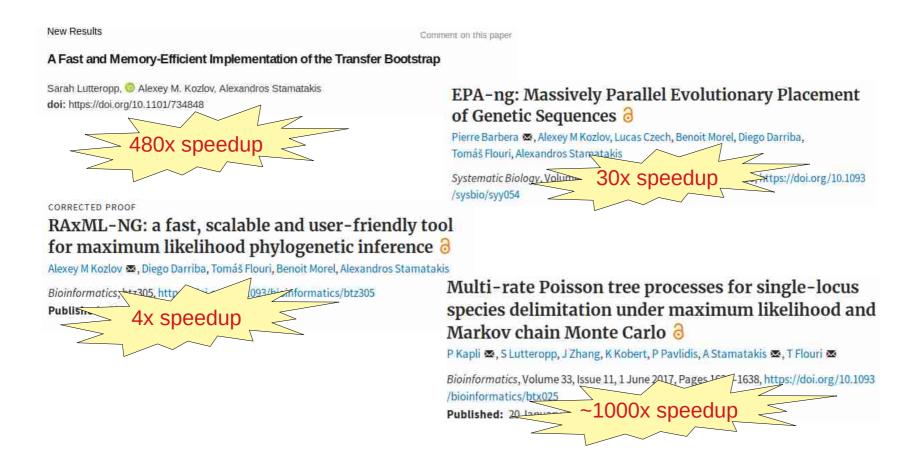
- Leads to healthy competition among lab members → everyone wants to write the cleanest code
- Used by researchers inside and outside of the lab during the development process → potential bugs identified and avoided
- Used as teaching tool in programming practicals
- SoftWipe score already used by us and others in Bioinformatics software paper submissions
- Vision: Establish software quality indicators as a necessary prerequisite for software paper submissions

Software Quality and Maintainability

- The Next Generation (-NG) projects:
 - Re-design, re-factoring, from scratch re-implementation of flagship tools to ensure maintainability, sustainability, and extensibility & increase scalability/performance
 - ModelTest-NG model testing of evolutionary models for phylogenetic inference
 - RAxML NG phylogenetic inference
 - EPA NG phylogenetic placement of environmental reads

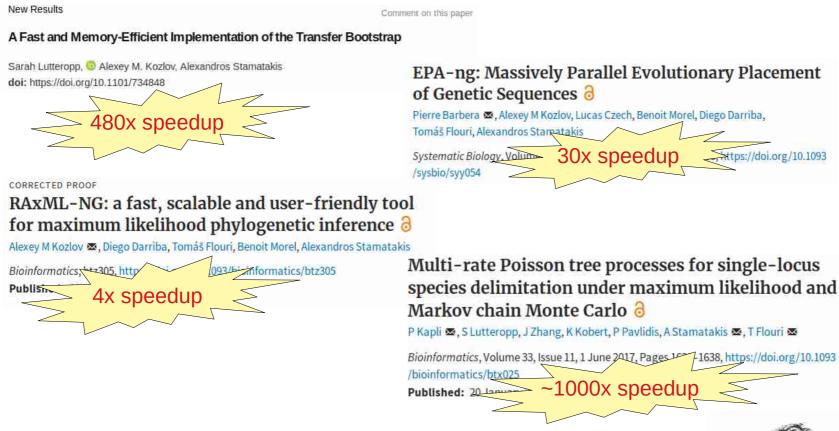
Energy Efficiency

Innovations



Oh, wow, this will help save a lot of energy!

The Jevons Paradox

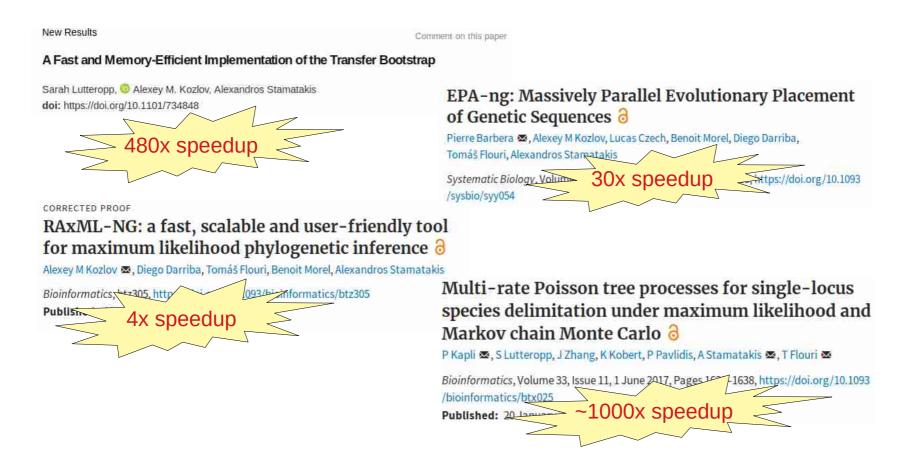


W. S. Jevons "The Coal Question" (1865)





The Jevons Paradox



We need an alternative solution!

Energy monitoring: RAxML - NG

New in RAxML-NG v1.0: energy usage report

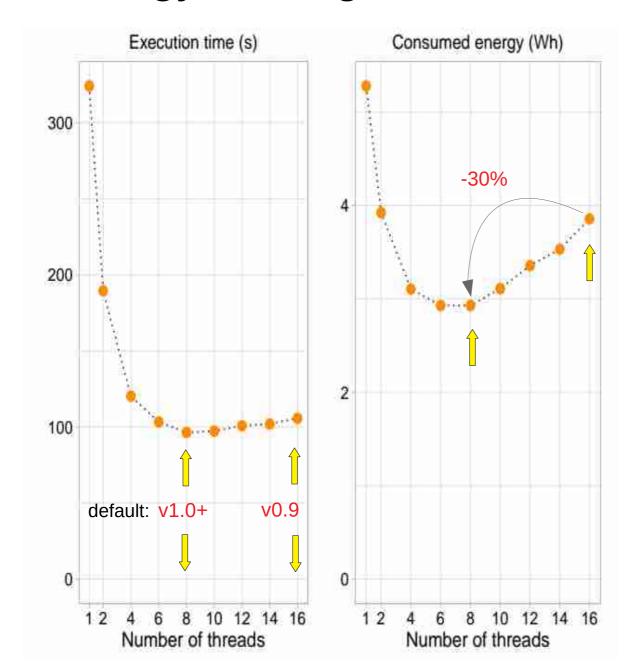
```
Elapsed time: 42846.287 seconds

Consumed energy: 162370.469 Wh (= 812 km in an electric car, or 4059 km with an e-scooter!)
```

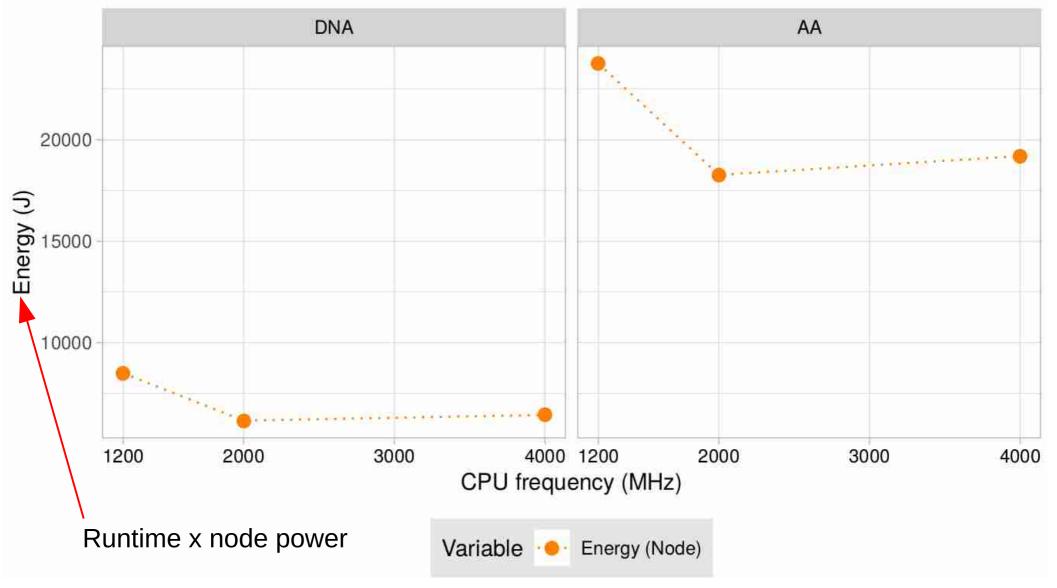
Single tree search (96 nodes x 12h):

>160 kWh

Energy Saving Mode in RAxML-NG v1.0



Phylogenetic Inference: Energy as a function of CPU clock frequency



Phylogenetic Inference: Energy as a function of CPU clock frequency

