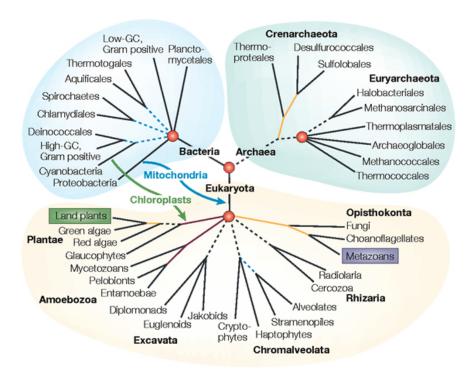
Computational Phylogenomics and Metagenomics

Tandy Warnow
Departments of Bioengineering and
Computer Science
The University of Illinois at Urbana-Champaign
http://tandy.cs.illinois.edu

The "Tree of Life"

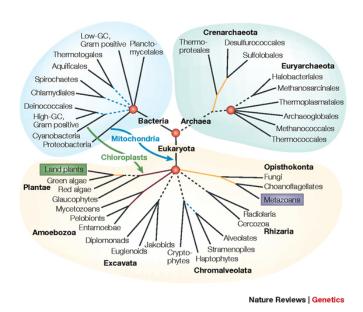


Applications of phylogenies to:
 protein structure and function
 population genetics
 human migrations

Nature Reviews | Genetics

Phylogenomics

(Phylogenetic estimation from whole genomes)





Challenges: gene tree conflict due to incomplete lineage sorting, horizontal gene transfer, or gene duplication and loss.

The NIH Human Microbiome Project

25,000 human genes, 1,000,000 bacterial genes



Basic Questions

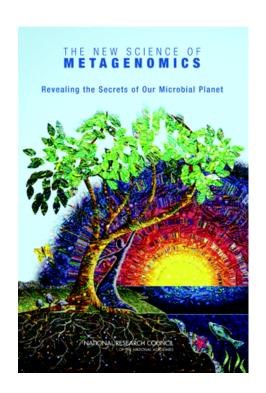
1. What is this fragment? (Classify each fragment as well as possible.)

2. What is the taxonomic distribution in the dataset? (Note: helpful to use marker genes.)

3. What are the organisms in this metagenomic sample doing together?

Computational Phylogenetics and Metagenomics





Courtesy of the Tree of Life project

Phylogenomic pipeline

- Select taxon set and markers
- Gather and screen sequence data, possibly identify orthologs
- Compute multiple sequence alignments for each locus
- Compute species tree or network:
 - Compute gene trees on the alignments and combine the estimated gene trees, OR
 - Perform "concatenation analysis" (aka "combined analysis")
- Get statistical support on each branch (e.g., bootstrapping)
- Estimate dates on the nodes of the phylogeny
- Use species tree with branch support and dates to understand biology

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Avian Phylogenomics Project (100+ people)

Erich Jarvis, HHMI



MTP Gilbert, Copenhagen



G Zhang, BGI



T. Warnow UT-Austin



S. Mirarab Md. S. Bayzid UT-Austin UT-Austin





- Approx. 50 species, whole genomes
- 8000+ genes, UCEs
- Gene sequence alignments and trees computed using SATé (Science 2009, Systematic Biology 2012)
- Concatenation analysis (multi-million site) using ExaML (new version of RAxML for very long alignments)

CHALLENGE:

Massive gene tree incongruence suggestive of incomplete lineage sorting, but gene trees had poor resolution due to low phylogenetic signal

Coalescent-based estimation methods had poor accuracy

1kp: Thousand Transcriptome Project (~40 people)

G. Ka-Shu Wong U Alberta J. Leebens-Mack U Georgia N. Wickett Northwestern N. Matasci iPlant T. Warnow, UT-Austin

S. Mirarab, UT-Austin N. Nguyen, UT-Austin Md. S.Bayzid UT-Austin

















- Plant Tree of Life based on transcriptomes of ~1200 species
- More than 13,000 gene families (most not single copy)

Challenges:

- 1. Alignment of datasets with > 100,000 sequences, with many fragmentary sequences
- 2. Gene tree incongruence due to ILS, too many species for standard methods

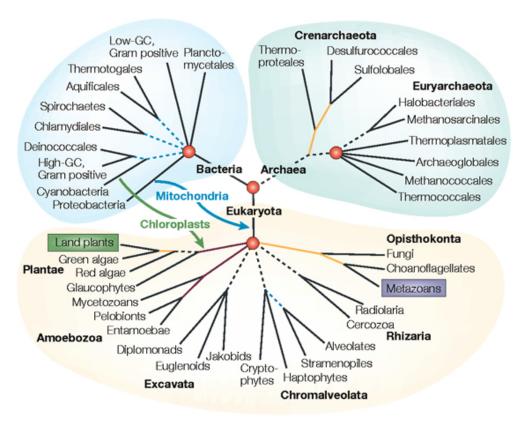
Current Research Projects

- Ultra-large multiple-sequence alignment (briefly today)
- Metagenomic taxon identification and abundance profiling
- Species tree estimation from multiple genes in the presence of gene tree conflict (Math Colloquium, Oct 2)

Other Research Projects

- Alignment-free phylogeny estimation
- Supertree estimation
- Genome rearrangement phylogeny
- Reticulate evolution detection and reconstruction
- Data mining techniques to explore multiple optima
- Theoretical guarantees under Markov models of evolution
- Historical linguistics

Phylogenetic Estimation: Big Data Challenges



Nature Reviews | Genetics

NP-hard problems

Large datasets:

100,000+ sequences 10,000+ genes

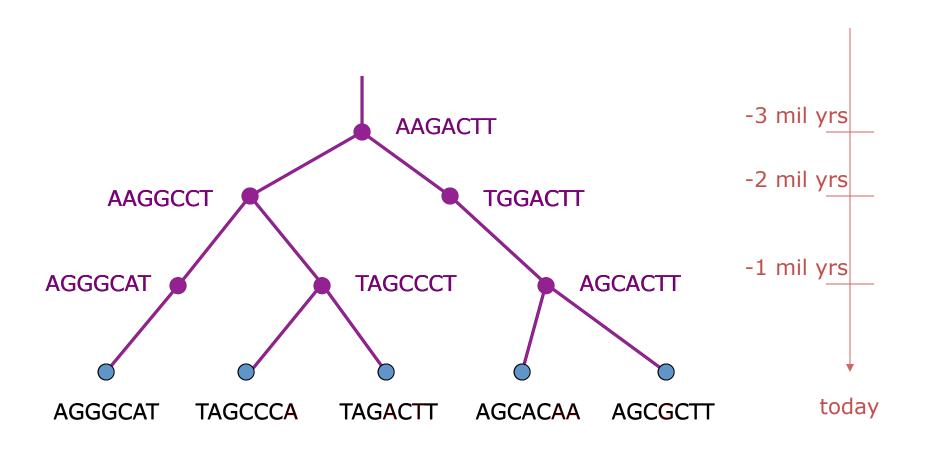
"BigData" complexity
missing data
mixture models
errors in input data
model misspecification
streaming data

This talk

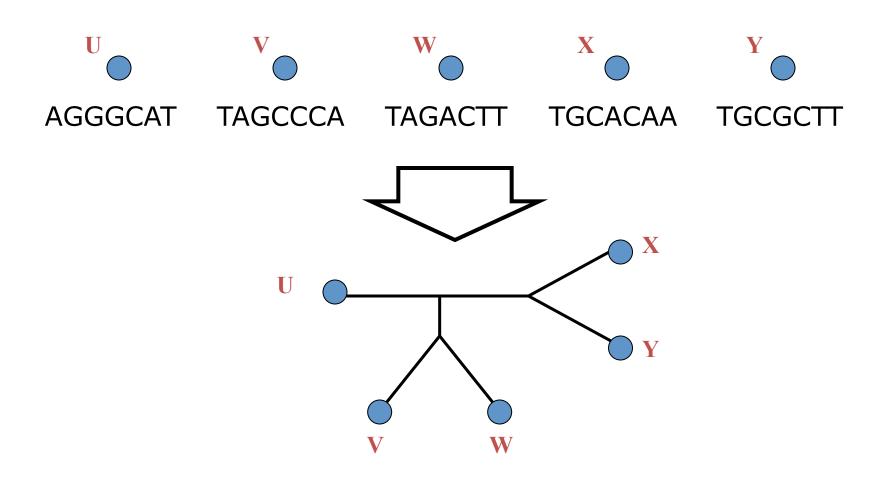
Multiple sequence alignment methods, <u>SATé</u>
(2009), <u>PASTA</u> (2014), and <u>UPP</u> (in preparation),
and applications to metagenomic taxon
identification (TIPP, submitted)

Multiple Sequence Alignment

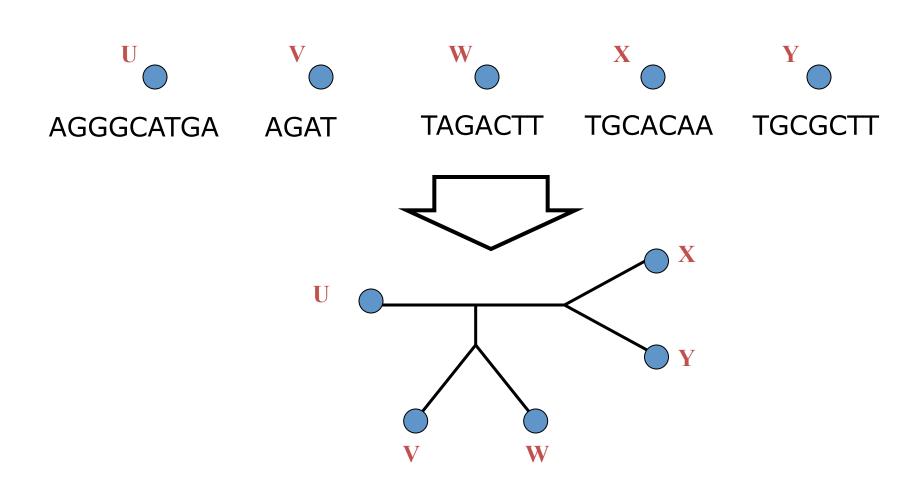
DNA Sequence Evolution



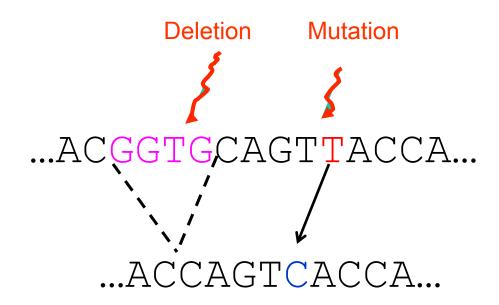
Phylogeny Problem

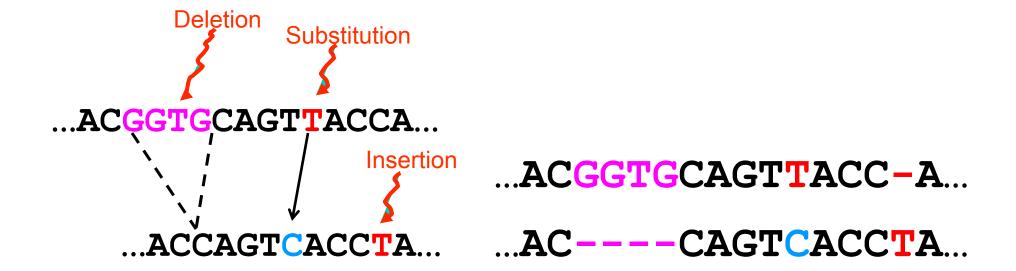


The "real" problem



Indels (insertions and deletions)





The true multiple alignment

- Reflects historical substitution, insertion, and deletion events
- Defined using transitive closure of pairwise alignments computed on edges of the true tree

Input: unaligned sequences

```
S1 = AGGCTATCACCTGACCTCCA
```

S2 = TAGCTATCACGACCGC

S3 = TAGCTGACCGC

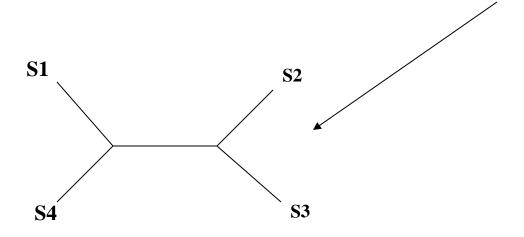
S4 = TCACGACCGACA

Phase 1: Alignment

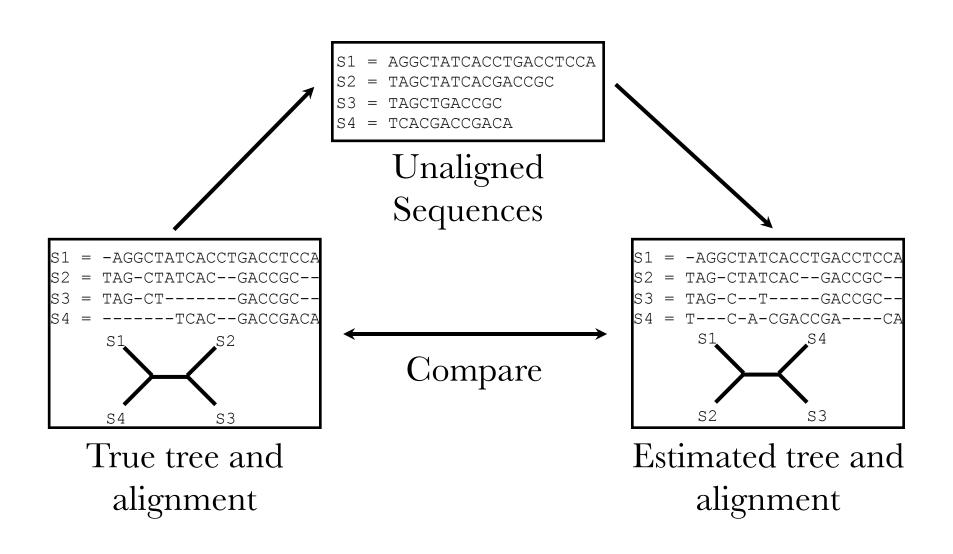
```
S1 = AGGCTATCACCTGACCTCCA
S2 = TAGCTATCACGACCGC
S3 = TAGCTGACCGC
S4 = TCACGACCGACA
S1 = -AGGCTATCACCTGACCTCCA
S2 = TAG-CTATCAC--GACCGC--
S3 = TAG-CT-----GACCGC--
S4 = -----TCAC--GACCGACA
```

Phase 2: Construct tree

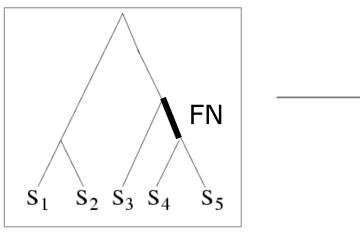
```
S1 = AGGCTATCACCTGACCTCCA
S2 = TAGCTATCACGACCGC
S3 = TAGCTGACCGC
S4 = TCACGACCGACA
S1 = -AGGCTATCACCTGACCTCCA
S2 = TAG-CTATCAC--GACCGC--
S3 = TAG-CT------GACCGC--
S4 = -----TCAC--GACCGACA
```



Simulation Studies



Quantifying Error



TRUE TREE

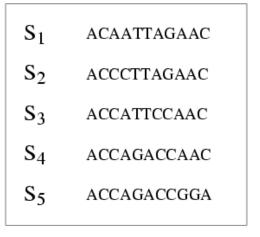
FN: false negative

(missing edge)

FP: false positive

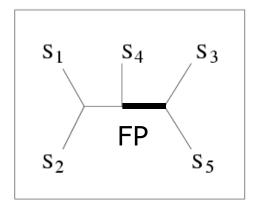
(incorrect edge)

50% error rate



DNA SEQUENCES





INFERRED TREE

Two-phase estimation

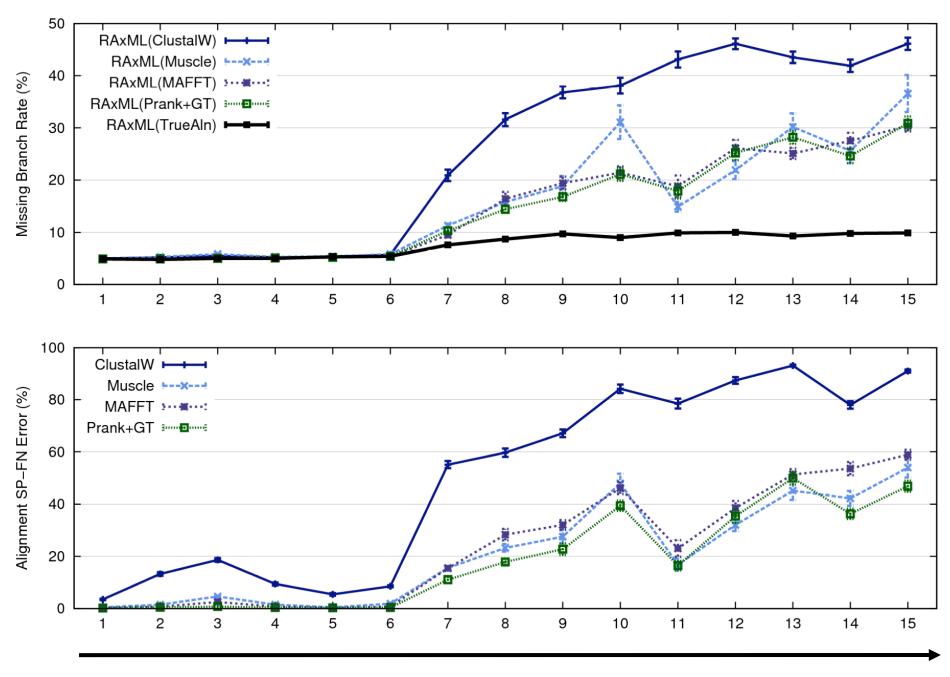
Alignment methods

- Clustal
- POY (and POY*)
- Probcons (and Probtree)
- Probalign
- MAFFT
- Muscle
- Di-align
- T-Coffee
- Prank (PNAS 2005, Science 2008)
- Opal (ISMB and Bioinf. 2007)
- FSA (PLoS Comp. Bio. 2009)
- Infernal (Bioinf. 2009)
- Etc.

Phylogeny methods

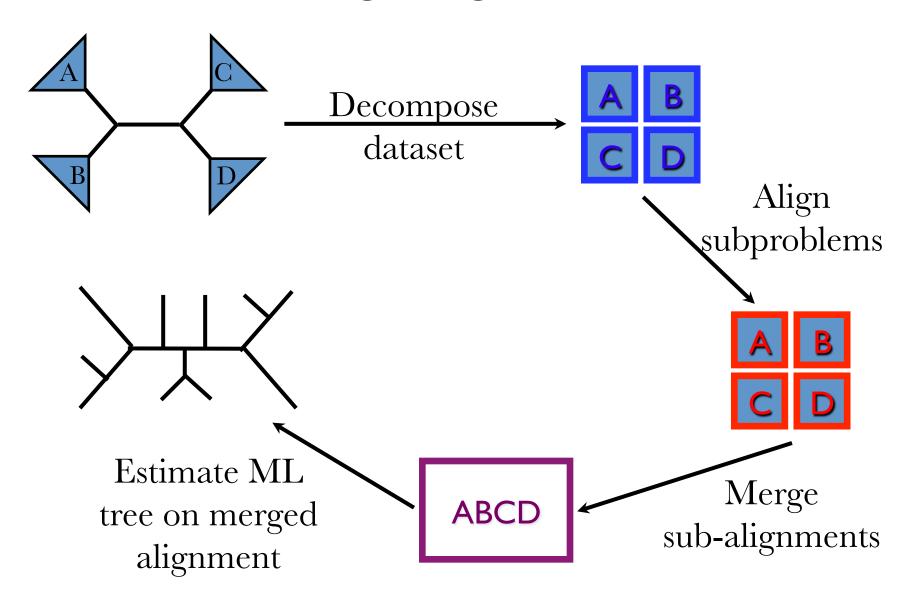
- Bayesian MCMC
- Maximum parsimony
- Maximum likelihood
- Neighbor joining
- FastME
- UPGMA
- Quartet puzzling
- Etc.

RAXML: heuristic for large-scale ML optimization



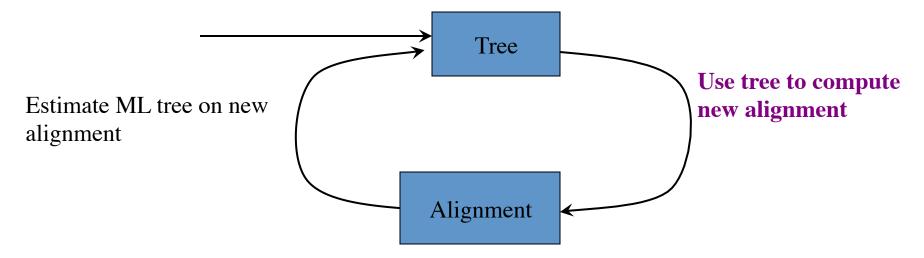
1000-taxon models, ordered by difficulty (Liu et al., 2009)

Re-aligning on a tree



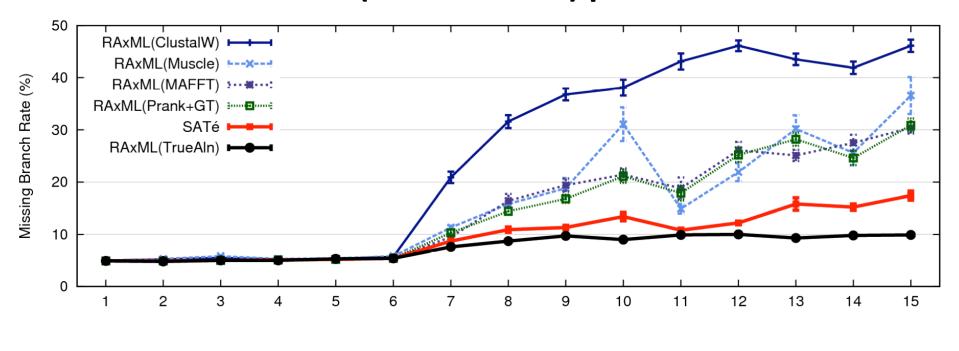
SATé and PASTA Algorithms

Obtain initial alignment and estimated ML tree



If new alignment/tree pair has worse ML score, realign using a different decomposition Repeat until termination condition (typically, 24 hours)

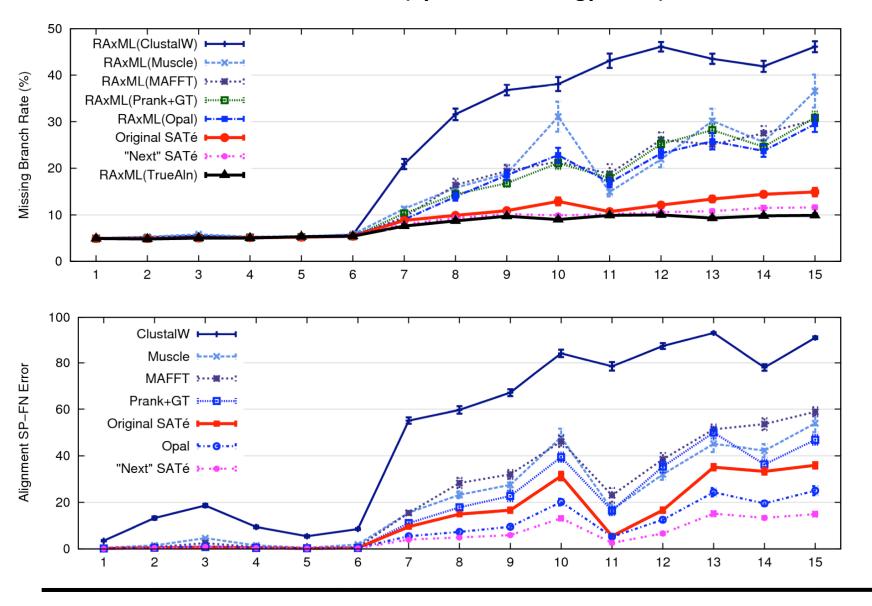
SATé-1 (Science 2009) performance



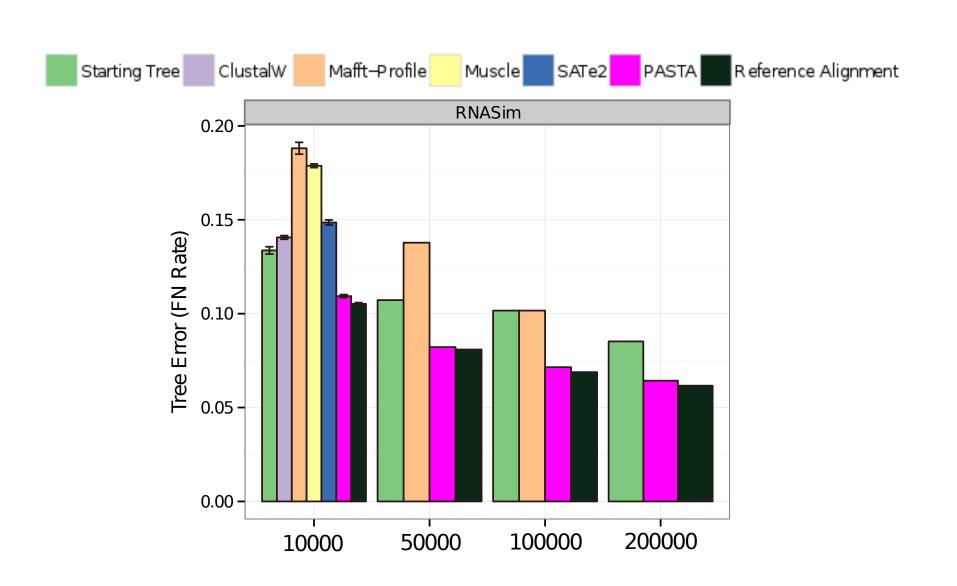
1000 taxon models, ordered by difficulty

SATé-1 24 hour analysis, on desktop machines (Similar improvements for biological datasets)
SATé-1 can analyze up to about 30,000 sequences.

SATé-1 and SATé-2 (Systematic Biology, 2012)



PASTA (2014): even better than SATé-2



1kp: Thousand Transcriptome Project

G. Ka-Shu Wong U Alberta J. Leebens-Mack U Georgia N. Wickett Northwestern N. Matasci iPlant

T. Warnow, UT-Austin

S. Mirarab, UT-Austin N. Nguyen, UT-Austin Md. S.Bayzid UT-Austin















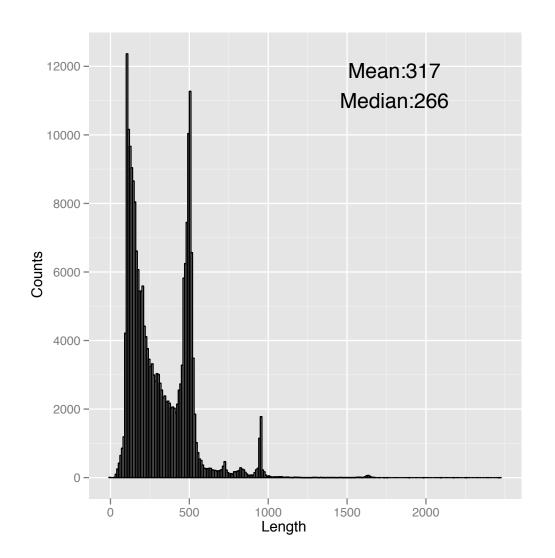


Plus many many other people...

- Plant Tree of Life based on transcriptomes of ~1200 species
- More than 13,000 gene families (most not single copy)

Challenge:

Alignment of datasets with > 100,000 sequences with many fragmentary sequences



1KP dataset: more than 100,000 p450 sequences, many fragmentary

All standard alignment methods we tested performed poorly.

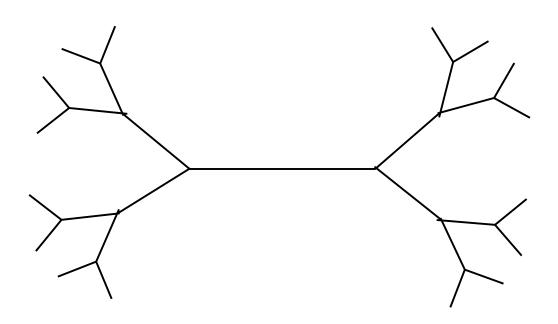
UPP

Basic approach:

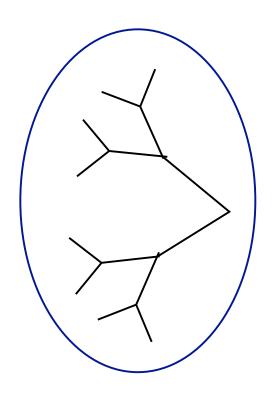
- Select random subset of sequences, and build "backbone alignment"
- Construct a "Family of Hidden Markov Models" on the backbone alignment
- Add all remaining sequences to the backbone alignment using the Family of HMMs

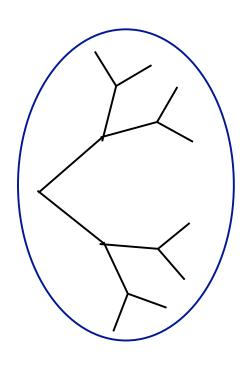
UPP can analyze much larger datasets than PASTA, and is highly robust to fragmentary sequences

One Hidden Markov Model for the entire alignment?

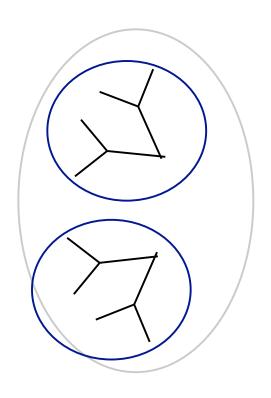


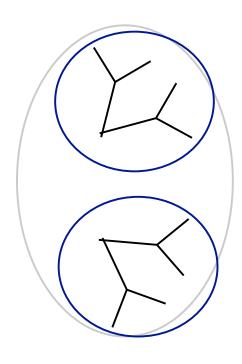
Or 2 HMMs?



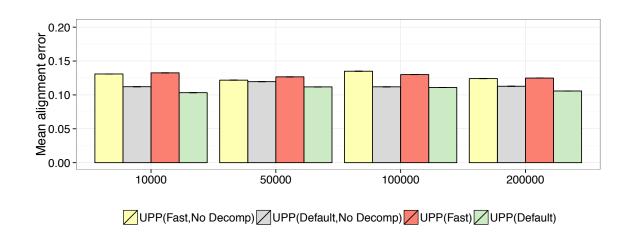


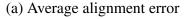
Or 4 HMMs?

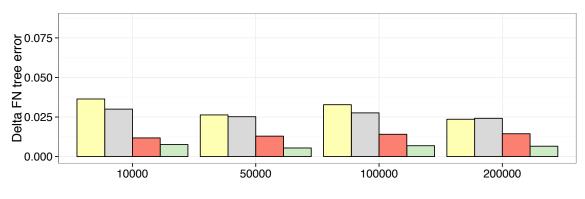




Impact of backbone size and use of HMM Family technique







(b) Average tree error

UPP(Fast,No Decomp) UPP(Default,No Decomp) UPP(Fast) UPP(Default)

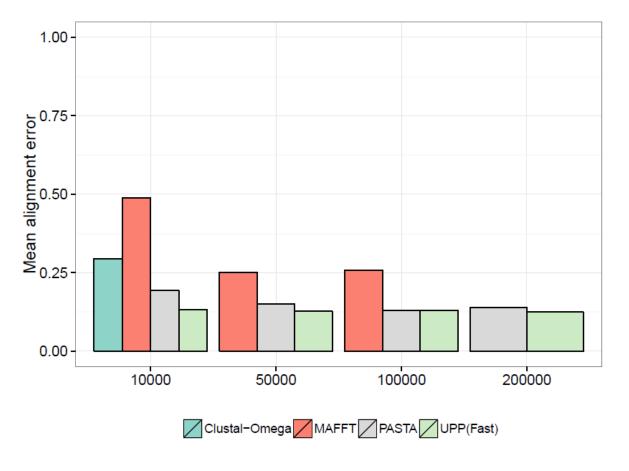
Notes:

Relative performance under standard alignment criteria is not predictive of relative performance for tree estimation.

For alignment estimation, a large backbone is important.

For tree estimation, the use of the HMM Family is most important, but large backbones also help.

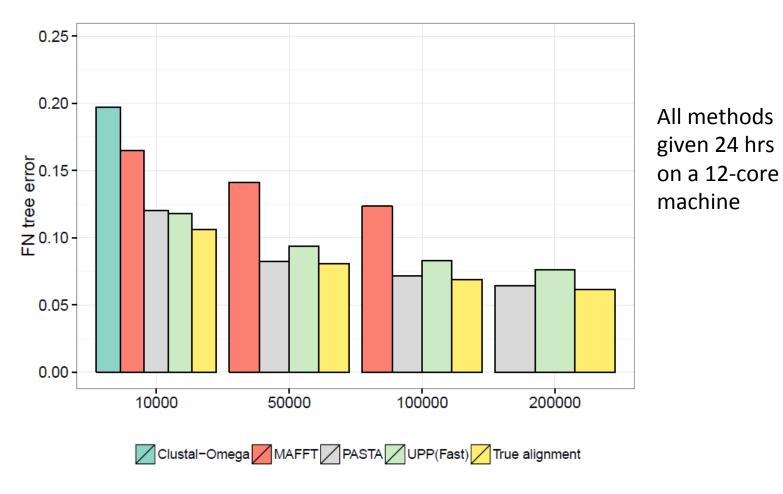
RNASim: alignment error



All methods given 24 hrs on a 12-core machine

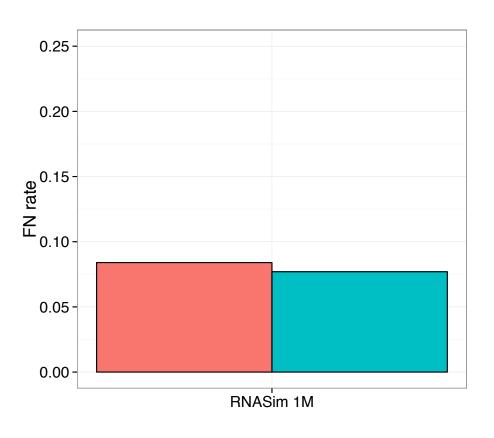
Note: Mafft was run under default settings for 10K and 50K sequences and under Parttree for 100K sequences, and fails to complete under any setting For 200K sequences. Clustal-Omega only completes on 10K dataset.

RNASim: tree error



Note: Mafft was run under default settings for 10K and 50K sequences and under Parttree for 100K sequences, and fails to complete under any setting For 200K sequences. Clustal-Omega only completes on 10K dataset.

RNASim Million Sequences: tree error

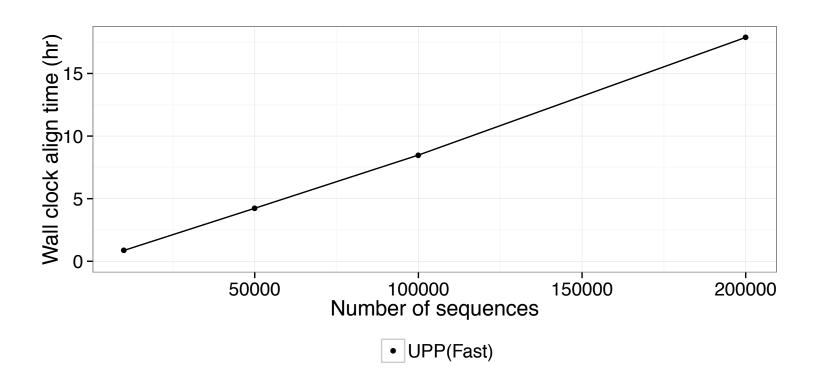


Notes:

- PASTA cannot run on this dataset size,
- UPP(Fast,NoDecomp) took 2.2 days, and
- UPP(Fast) took 11.9 days (both using 12 processors).

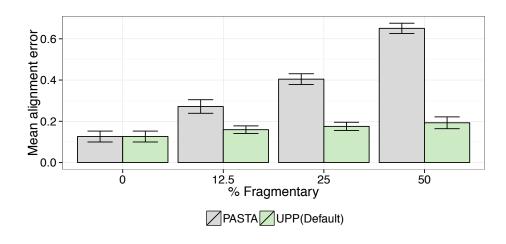
Method UPP(Fast,No Decomp) UPP(Fast)

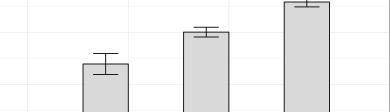
Running Time



Wall-clock time used (in hours) given 12 processors

UPP vs. PASTA: impact of fragmentation





Delta FN tree error 0.0 12.5 50 % Fragmentary

(a) Average alignment error

(b) Average tree error

PASTA UPP(Default)

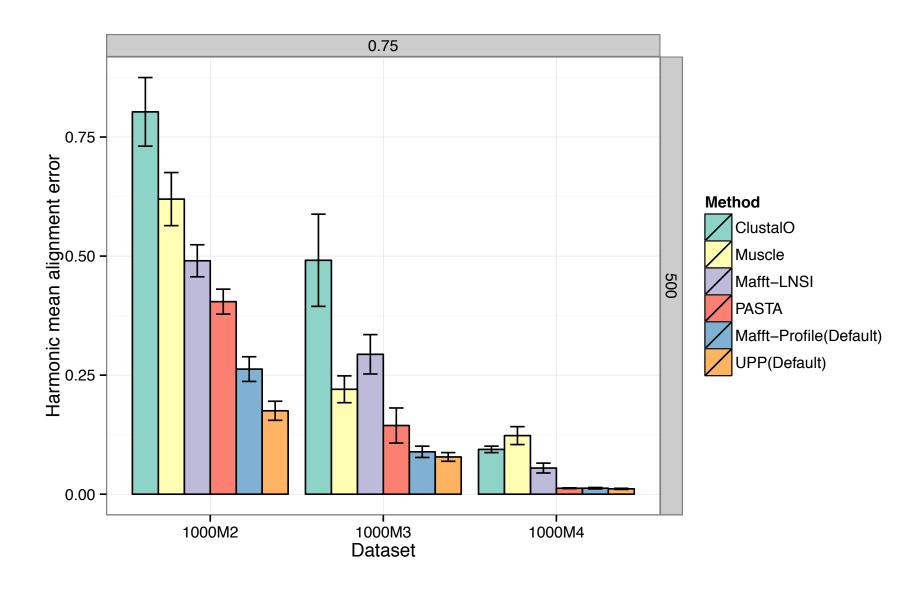
Under high rates of evolution, PASTA is badly impacted by fragmentary sequences (the same is true for other methods).

Under low rates of evolution, PASTA can still be highly accurate (data not shown).

UPP continues to have good accuracy even on datasets with many fragments under all rates of evolution.

Performance on fragmentary datasets of the 1000M2 model condition

Impact of rate of evolution on performance on fragmentary sequence data



New Alignment Methods

- SATé-1 (Science 2009) can analyze 10,000 sequences
- SATé-2 (Systematic Biology 2012) can analyze 50,000 sequences, is faster and more accurate than SATé-1
- PASTA (RECOMB 2014) can analyze 200,000 sequences, and is faster and more accurate than both SATé versions.
- UPP (in preparation) uses a very different technique than SATé and PASTA.
 - Fast
 - Scalable (up to 1,000,000 sequences)
 - Robust to fragmentary data

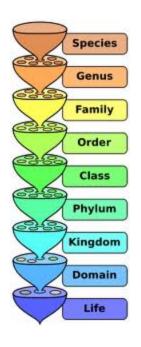
The NIH Human Microbiome Project

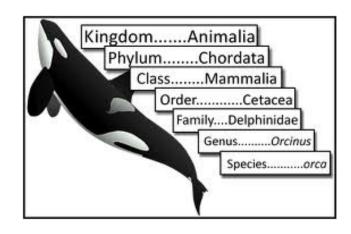
25,000 human genes, 1,000,000 bacterial genes



Metagenomic Taxon Identification

Objective: classify short reads in a metagenomic sample





Abundance Profiling

Objective: Distribution of the species (or genera, or families, etc.) within the sample.

For example: The distribution of the sample at the species-level is:

```
50% species A
```

20% species B

15% species C

14% species D

1% species E

Abundance Profiling

Leading techniques:

PhymmBL (Brady & Salzberg, Nature Methods 2009)

NBC (Rosen, Reichenberger, and Rosenfeld, Bioinformatics 2011)

MetaPhyler (Liu et al., BMC Genomics 2011), from the Pop lab at the University of Maryland

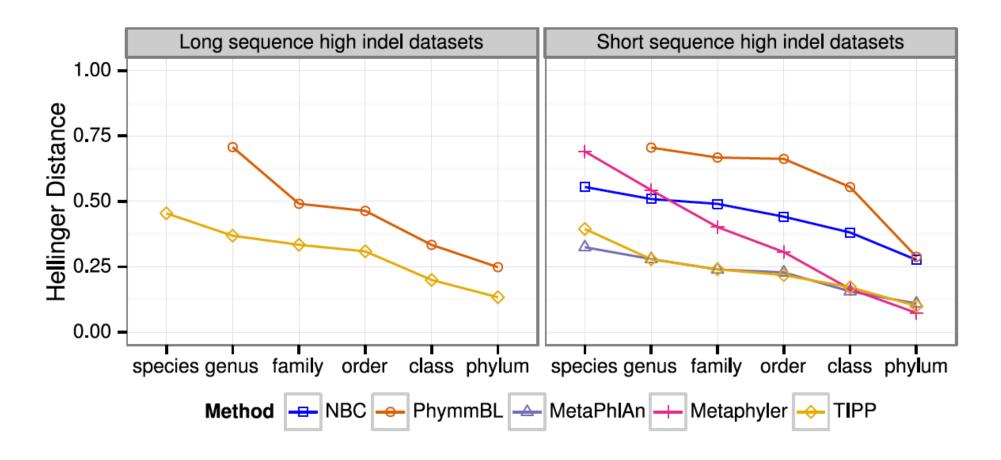
MetaPhlAn (Segata et al., Nature Methods 2012), from the Huttenhower Lab at Harvard

mOTU (Bork et al., Nature Methods 2013)

MetaPhyler, MetaPhlAn, and mOTU are marker-based techniques (but use different marker genes).

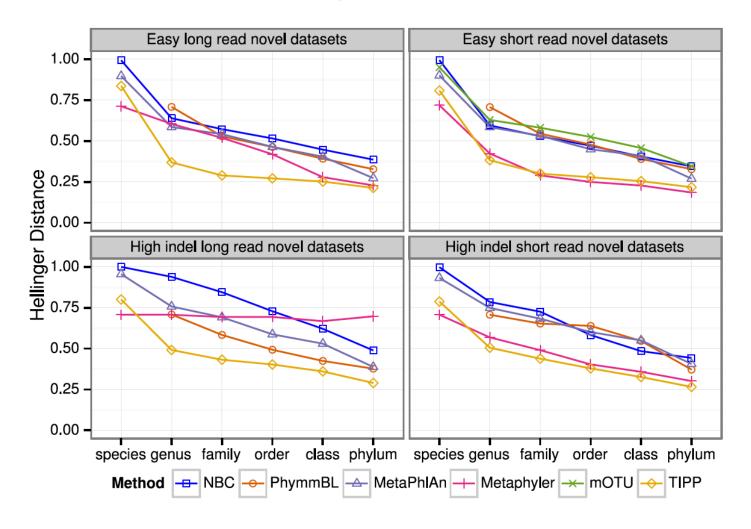
Marker gene are typically single-copy, universal, and resistant to horizontal transmission.

High indel datasets containing known genomes



Note: NBC, MetaPhlAn, and MetaPhyler cannot classify any sequences from at least one of the high indel long sequence datasets, and mOTU terminates with an error message on all the high indel datasets.

"Novel" genome datasets



Note: mOTU terminates with an error message on the long fragment datasets and high indel datasets.

TIPP vs. other abundance profilers

- TIPP is highly accurate, even in the presence of high indel rates and novel genomes, and for both short and long reads.
- All other methods have some vulnerability (e.g., mOTU is only accurate for short reads and is impacted by high indel rates).

Phylogenetic "boosters"

Goal: improve accuracy, speed, robustness, or theoretical guarantees of base methods

Techniques: divide-and-conquer, iteration, chordal graph algorithms, and "bin-and-conquer"

Examples:

- DCM-boosting for distance-based methods (1999)
- DCM-boosting for heuristics for NP-hard problems (1999)
- SATé- and PASTA-boosting for alignment methods (2009, 2012, and 2014)
- SuperFine-boosting for supertree methods (2012)
- DACTAL: almost alignment-free phylogeny estimation methods (2012)
- SEPP-boosting for phylogenetic placement of short sequences (2012)
- TIPP-boosting for metagenomic taxon identification (submitted)
- UPP-boosting for alignment methods (in preparation)
- Bin-and-conquer for coalescent-based species tree estimation (2013 and 2014)

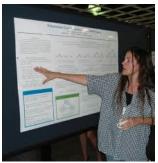
Warnow Laboratory











U Texas PhD students: Siavash Mirarab, Nam Nguyen (former), and Md. S. Bayzid

Postdoctoral Fellow: Ruth Davidson (at UIUC)

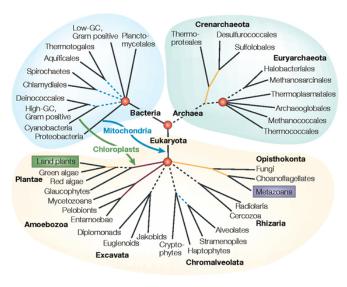
Lab Website: http://www.cs.utexas.edu/users/phylo

Current and Past Funding: Guggenheim Foundation, Packard Foundation, NSF, Microsoft Research New England, and TACC (Texas Advanced Computing Center). HHMI graduate fellowship to Siavash Mirarab and Fulbright graduate fellowship to Md. S. Bayzid.

http://tandy.cs.illiinois.edu Tandy's website

Phylogenomics

(Phylogenetic estimation from whole genomes)

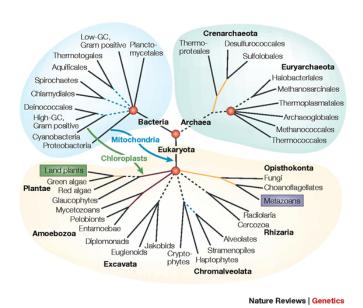






Phylogenomics

(Phylogenetic estimation from whole genomes)





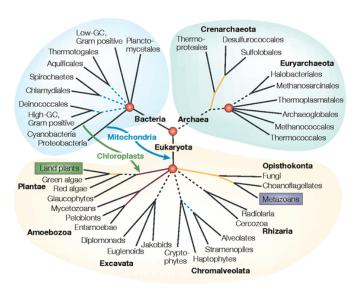
Challenges: Gene tree conflict due to incomplete lineage sorting, duplication and loss, recombination, etc.

Reticulate evolution (horizontal gene transfer, hybridization) requiring phylogenetic networks.

The Tree of Life: Multiple Challenges

Scientific challenges:

- Ultra-large multiple-sequence alignment
- Alignment-free phylogeny estimation
- Supertree estimation
- Estimating species trees from many gene trees
- Genome rearrangement phylogeny
- Reticulate evolution
- Visualization of large trees and alignments
- Data mining techniques to explore multiple optima
- Theoretical guarantees under Markov models of evolution



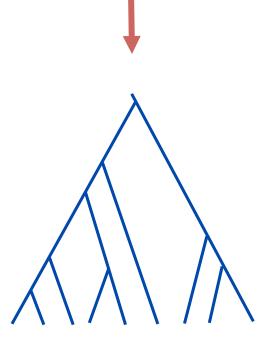
Nature Reviews | Genetics

Techniques:

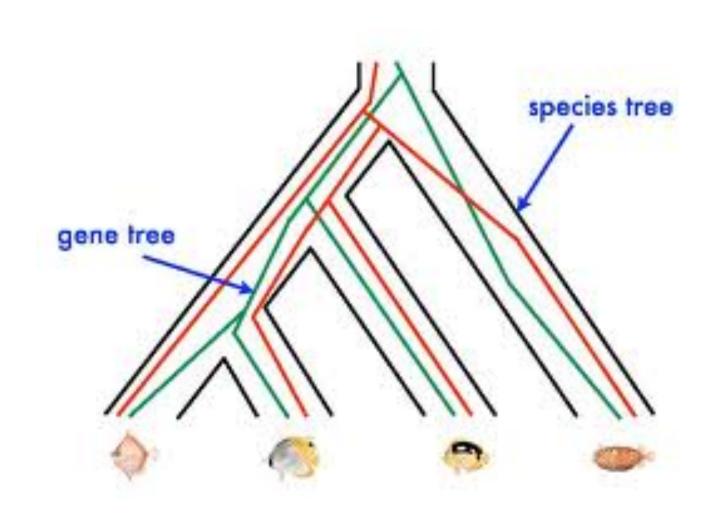
machine learning, applied probability theory, graph theory, combinatorial optimization, supercomputing, and heuristics

Combined analysis

gene 1 gene 2 gene 3

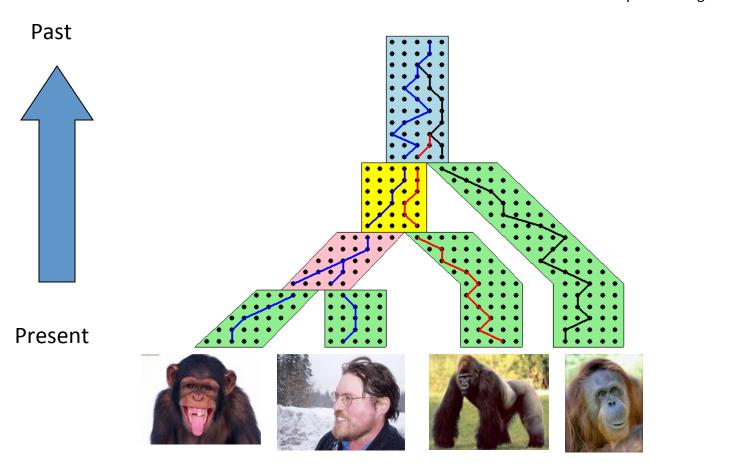


Red gene tree ≠ species tree (green gene tree okay)



The Coalescent

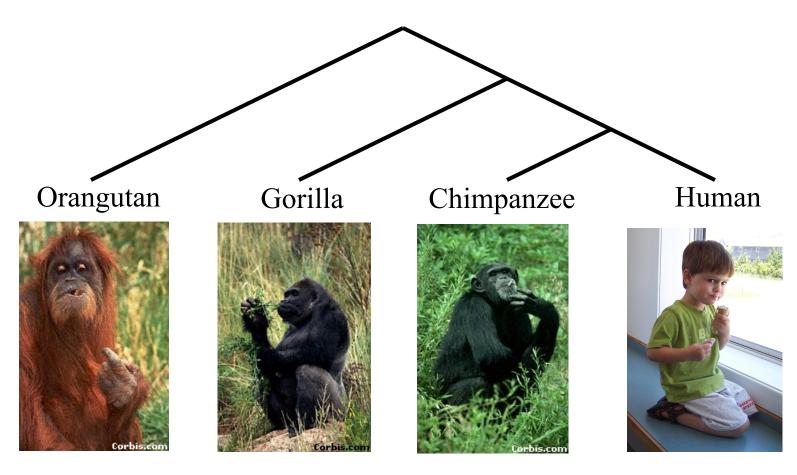
Courtesy James Degnan



Incomplete Lineage Sorting (ILS)

- 1000+ papers in 2013 alone
- Confounds phylogenetic analysis for many groups:
 - Hominids
 - Birds
 - Yeast
 - Animals
 - Toads
 - Fish
 - Fungi
- There is substantial debate about how to analyze phylogenomic datasets in the presence of ILS.

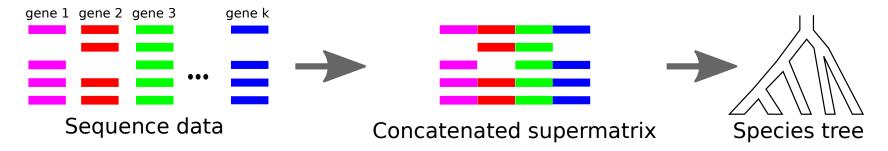
Species tree estimation: difficult, even for small datasets

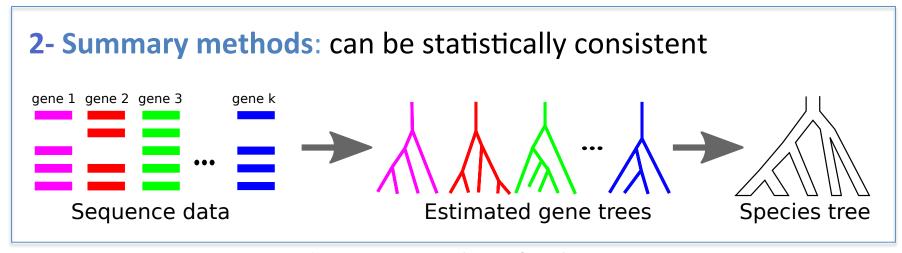


From the Tree of the Life Website, University of Arizona

Species tree estimation

1- Concatenation: statistically inconsistent (Roch & Steel 2014)





3- Co-estimation methods: too slow for large datasets

Is Concatenation Evil?

- Joseph Heled:
 - YES

- John Gatesy
 - No

- Data needed to held understand existing methods and their limitations
- Better methods are needed

Avian Phylogenomics Project (100+ people)

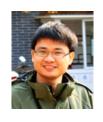
Erich Jarvis, HHMI



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- Approx. 50 species, whole genomes
- 8000+ genes, UCEs
- Gene sequence alignments and trees computed using SATé (Science 2009, Systematic Biology 2012)
- Concatenation analysis (multi-million site) using ExaML (new version of RAxML for very long alignments)
- Massive gene tree incongruence suggestive of incomplete lineage sorting -coalescent-based species tree estimation computed using "Statistical Binning" (Science, in press)

1kp: 1000 Plant Transcriptomes

G. Ka-Shu Wong U Alberta

J. Leebens-Mack U Georgia

N. Wickett Northwestern

N. Matasci iPlant

T. Warnow, UT-Austin

S. Mirarab, UT-Austin

N. Nguyen, UT-Austin















Plus many many other people...

- Whole Transcriptomes of 103 plant species and 850 single copy loci (1200 taxa in next phase)
 - Most accurate summary methods cannot handle this size
- Common ancestor about 1 billion years ago and so gene trees are hard to root
 - Most summary methods need rooted gene trees
- Pre-existing summary methods do not provide reasonable results on this dataset

[Wickett et al. (under review), 2014.]

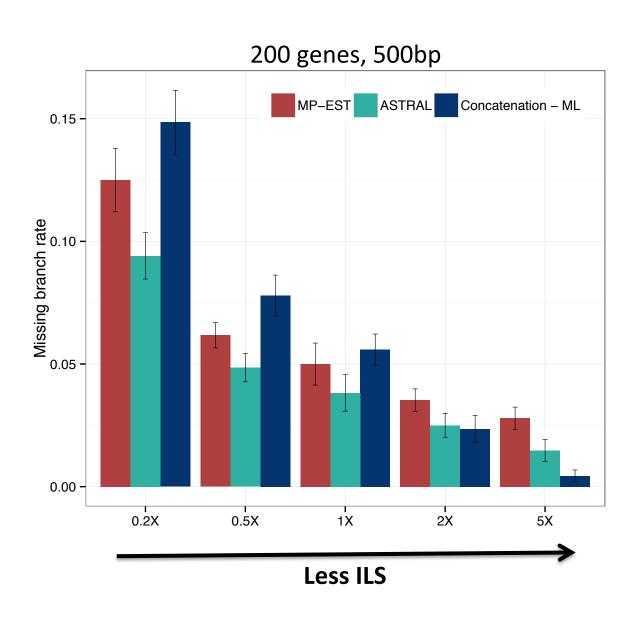
Our methods

- "Statistical Binning" (accepted): uses a statistical method to determine sets of "combinable" gene sequence alignments, improves coalescent-based species tree estimation accuracy when gene trees have poor resolution (used for Avian Phylogenomics Project).
- ASTRAL (Bioinformatics 2014): polynomial time statistically consistent method, can run on very large datasets of unrooted gene trees.

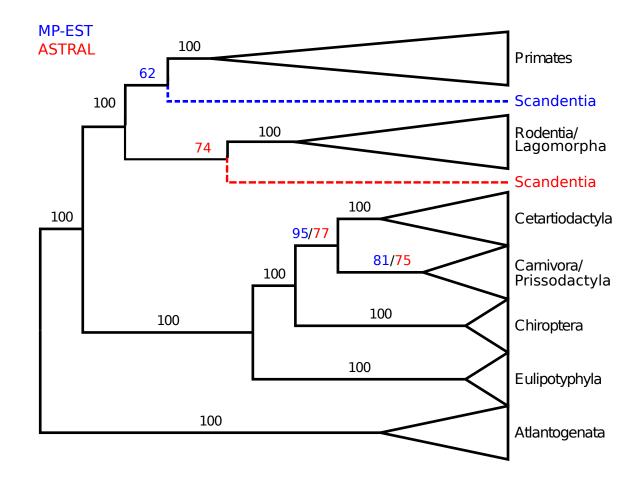
Mammalian simulations

- Based on a biological mammalian dataset of 37 taxa and 442 genes, published by Song et al., PNAS, 2012.
- In simulations, we vary
 - Levels of ILS
 - Number of genes
 - Alignment length to control gene tree estimation error
- Compare ASTRAL to
 - MP-EST, BUCKy
 - Concatenation
 - MRP, Greedy
- Measure species tree error compared to the known true tree

ASTRAL vs. Concatenation



Analyses of the Song et al. Mammalian dataset



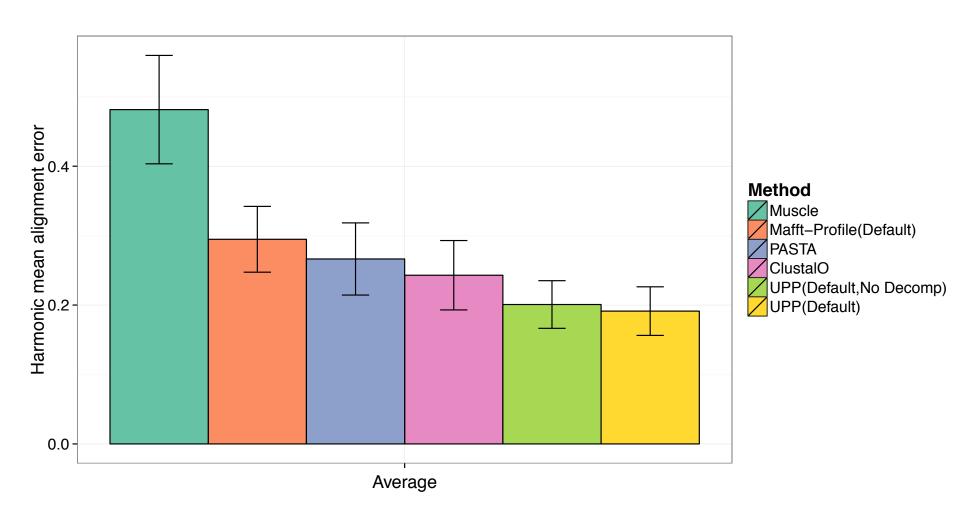
The placement of Scandentia (Tree Shrew) is controversial.

The ASTRAL analysis agrees with maximum likelihood concatenation analysis of this dataset.

Summary

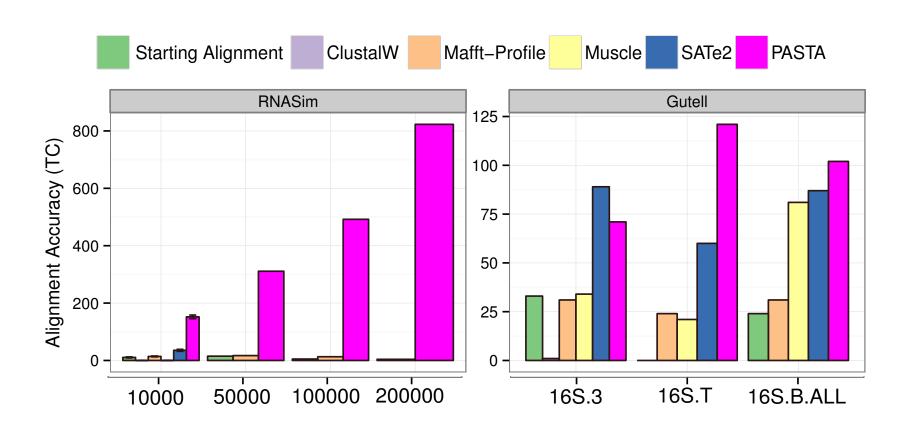
- New multiple sequence alignments with improved accuracy and scalability, as well as high robustness to fragmentary data – SATé used in many studies.
- New coalescent-based species tree estimation methods that have better accuracy than current methods, and can analyze large datasets (used in 1KP and Avian Phylogenomics project analyses)
- New methods for metagenomic taxon identification and abundance profiling with improved accuracy (will be used in analyses here at Illinois)
- Method development inspired by collaboration with biologists, and improves biological analyses.

AA Sequence Alignment Error (13 HomFam datasets)



Homfam: Sievers et al., MSB 2011 10,000 to 46,000 sequences per dataset; 5-14 sequences in each seed alignment Sequence lengths from 46-364 AA

Alignment Accuracy – Correct columns



Other co-estimation methods

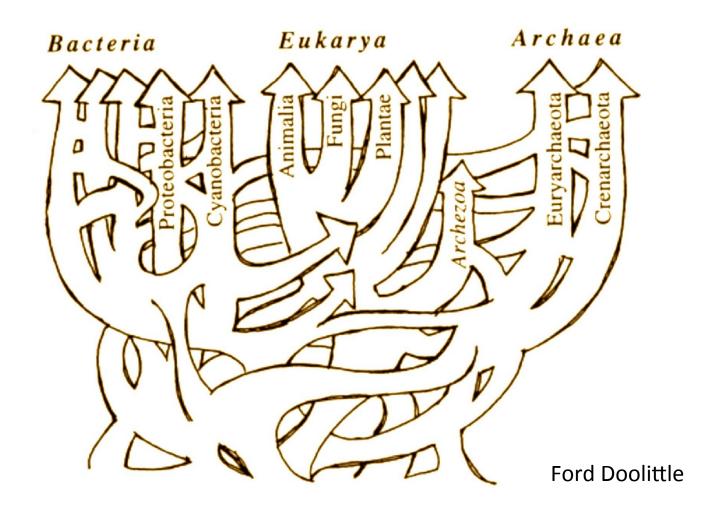
Statistical methods:

- BAli-Phy (Redelings and Suchard): Bayesian software to coestimate alignments and trees under a statistical model of evolution that includes indels. Can scale to about 100 sequences, but takes a very long time.
 - http://www.bali-phy.org/
- StatAlign: http://statalign.github.io/

Extensions of Parsimony

- POY (most well known software)
 - http://www.amnh.org/our-research/computational-sciences/ research/projects/systematic-biology/poy
- BeeTLe (Liu and Warnow, PLoS One 2012)

Horizontal Gene Transfer – Phylogenetic Networks



Warnow Laboratory











PhD students: Siavash Mirarab*, Nam Nguyen, and Md. S. Bayzid**

Undergrad: Keerthana Kumar

Lab Website: http://www.cs.utexas.edu/users/phylo

Funding: Guggenheim Foundation, Packard, NSF, Microsoft Research New England, David Bruton Jr. Centennial Professorship, TACC (Texas Advanced Computing Center), and the University of Alberta (Canada)

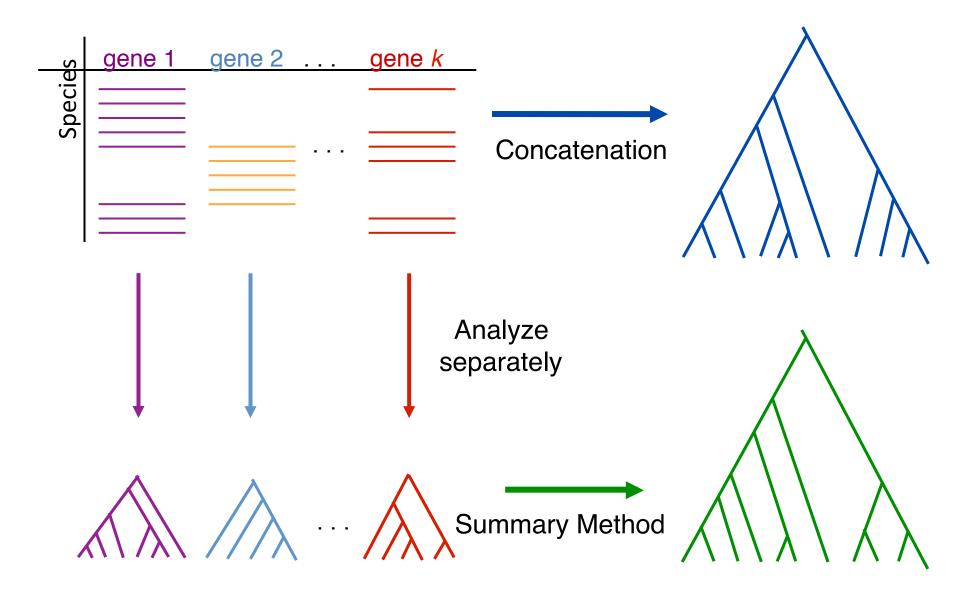
TACC and UTCS computational resources

- * Supported by HHMI Predoctoral Fellowship
- ** Supported by Fulbright Foundation Predoctoral Fellowship

But...

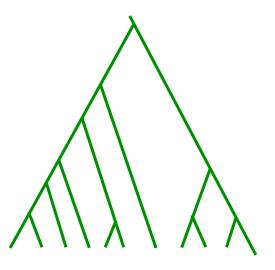
- Gene trees may not be identical to species trees:
 - Incomplete Lineage Sorting (deep coalescence)
 - Gene duplication and loss
 - Horizontal gene transfer
- This makes combined analysis and standard supertree analyses inappropriate

Two competing approaches



How to compute a species tree?





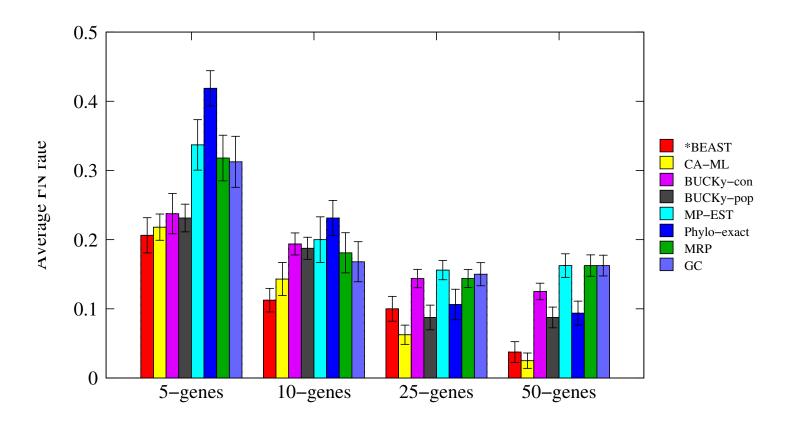
Statistically consistent under ILS?

- MP-EST (Liu et al. 2010): maximum likelihood estimation of rooted species tree – YES
- BUCKy-pop (Ané and Larget 2010): quartet-based Bayesian species tree estimation –YES
- MDC NO
- Greedy NO
- Concatenation under maximum likelihood NO
- MRP (supertree method) open

The Debate: Concatenation vs. Coalescent Estimation

- In favor of coalescent-based estimation
 - Statistical consistency guarantees
 - Addresses gene tree incongruence resulting from ILS
 - Some evidence that concatenation can be positively misleading
- In favor of concatenation
 - Reasonable results on data
 - High bootstrap support
 - Summary methods (that combine gene trees) can have poor support or miss well-established clades entirely
 - Some methods (such as *BEAST) are computationally too intensive to use

Results on 11-taxon datasets with strongILS



*BEAST more accurate than summary methods (MP-EST, BUCKy, etc)
CA-ML: (concatenated analysis) also very accurate

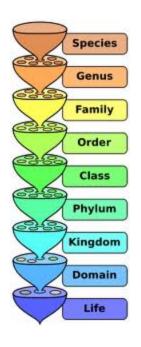
Datasets from Chung and Ané, 2011 Bayzid & Warnow, Bioinformatics 2013

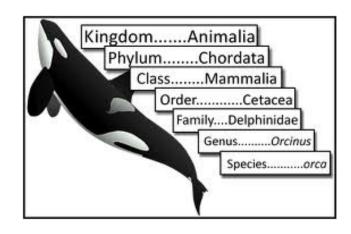
Species tree/network estimation

- Methods have been developed to estimate species phylogenies (trees or networks!) from gene trees, when gene trees can conflict from each other (e.g., due to ILS, gene duplication and loss, and horizontal gene transfer).
- Phylonet (software suite), has effective methods for many optimization problems – including MDC and maximum likelihood.
- Tutorial on Wednesday.
- Software available at http://bioinfo.cs.rice.edu/phylonet?destination=node/3

Metagenomic Taxon Identification

Objective: classify short reads in a metagenomic sample





Two Basic Questions

1. What is this fragment? (Classify each fragment as well as possible.)

2. What is the taxonomic distribution in the dataset? (Note: helpful to use marker genes.)

SEPP

- SEPP: SATé-enabled Phylogenetic
 Placement, by Mirarab, Nguyen, and Warnow
- Pacific Symposium on Biocomputing, 2012 (special session on the Human Microbiome)
- Tutorial on Thursday.

Other problems

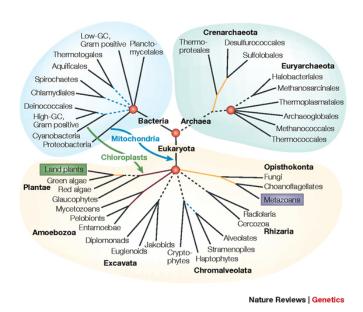
- Genomic MSA estimation:
 - Multiple sequence alignment of very long sequences
 - Multiple sequence alignment of sequences that evolve with rearrangement events
- Phylogeny estimation under more complex models
 - Heterotachy
 - Violation of the rates-across-sites assumption
 - Rearrangements
- Estimating branch support on very large datasets

New methods for estimating sequence alignments, gene trees, and species trees, with applications to metagenomics

Tandy Warnow
Departments of Bioengineering and
Computer Science
The University of Illinois at Urbana-Champaign
http://tandy.cs.illinois.edu

Phylogenomics

(Phylogenetic estimation from whole genomes)



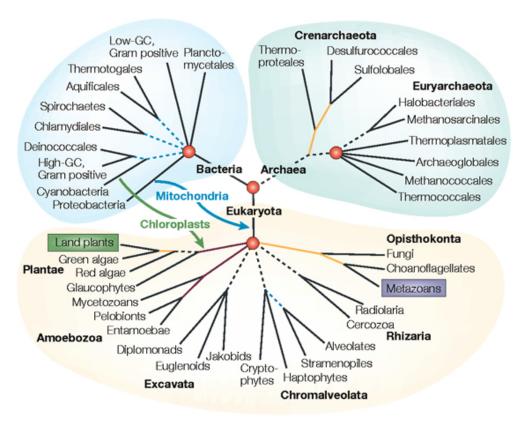


Challenges: gene tree conflict due to incomplete lineage sorting, horizontal gene transfer, or gene duplication and loss.

Phylogenomic pipeline

- Select taxon set and markers
- Gather and screen sequence data, possibly identify orthologs
- Compute multiple sequence alignments for each locus
- Compute species tree or network:
 - Compute gene trees on the alignments and combine the estimated gene trees, OR
 - Perform "concatenation analysis" (aka "combined analysis")
- Get statistical support on each branch (e.g., bootstrapping)
- Estimate dates on the nodes of the phylogeny
- Use species tree with branch support and dates to understand biology

Phylogenetic Estimation: Big Data Challenges



Nature Reviews | Genetics

NP-hard problems

Large datasets:

100,000+ sequences 10,000+ genes

"BigData" complexity
missing data
mixture models
errors in input data
model misspecification
streaming data

Basic Questions

1. What is this fragment? (Classify each fragment as well as possible.)

2. What is the taxonomic distribution in the dataset? (Note: helpful to use marker genes.)

3. What are the organisms in this metagenomic sample doing together?

What I do

• I develop new methods for difficult computational problems, to improve the accuracy of the biological analyses.

Collaborations drive method development.

Avian Phylogenomics Project (100+ people)

Erich Jarvis, HHMI



MTP Gilbert, Copenhagen



G Zhang, BGI



T. Warnow UT-Austin



S. Mirarab Md. S. Bayzid UT-Austin UT-Austin





- Approx. 50 species, whole genomes
- 8000+ genes, UCEs
- Gene sequence alignments and trees computed using SATé (Science 2009, Systematic Biology 2012)
- Concatenation analysis (multi-million site) using ExaML (new version of RAxML for very long alignments)

CHALLENGE:

Massive gene tree incongruence suggestive of incomplete lineage sorting, but gene trees had poor resolution due to low phylogenetic signal

Coalescent-based estimation methods had poor accuracy

1kp: Thousand Transcriptome Project (~40 people)

G. Ka-Shu Wong U Alberta J. Leebens-Mack U Georgia N. Wickett Northwestern N. Matasci iPlant T. Warnow, UT-Austin

S. Mirarab, UT-Austin N. Nguyen, UT-Austin Md. S.Bayzid UT-Austin

















- Plant Tree of Life based on transcriptomes of ~1200 species
- More than 13,000 gene families (most not single copy)

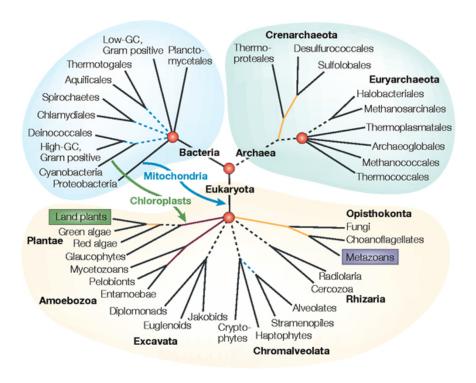
Challenges:

- 1. Alignment of datasets with > 100,000 sequences, with many fragmentary sequences
- 2. Gene tree incongruence due to ILS, too many species for standard methods

This talk

- Multiple sequence alignment methods, <u>SATé</u>
 (2009), <u>PASTA</u> (2014), and <u>UPP</u> (in preparation),
 and applications to metagenomic taxon
 identification (TIPP, submitted)
- New coalescent-based species tree estimation methods for:
 - Avian Phylogenomics Project (submitted) and
 - 1KP, the Thousand Plant Transcriptome Project (submitted)

Phylogenies and Applications



Basic Biology:
How did life evolve?

Applications of phylogenies to:
 protein structure and function
 population genetics
 human migrations

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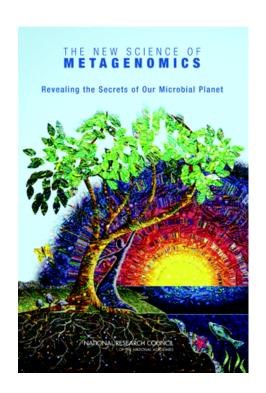
The NIH Human Microbiome Project

25,000 human genes, 1,000,000 bacterial genes



Computational Phylogenetics and Metagenomics





Courtesy of the Tree of Life project

Multiple Sequence Alignment (MSA): another grand challenge¹

```
S1 = AGGCTATCACCTGACCTCCA S1 = -AGGCTATCACCTGACCTCCA S2 = TAGCTATCACGACCGC S2 = TAG-CTATCAC--GACCGC-- S3 = TAGCTGACCGC S3 = TAG-CT-----GACCGC-- ... Sn = TCACGACCGACA Sn = -----TCAC--GACCGACA
```

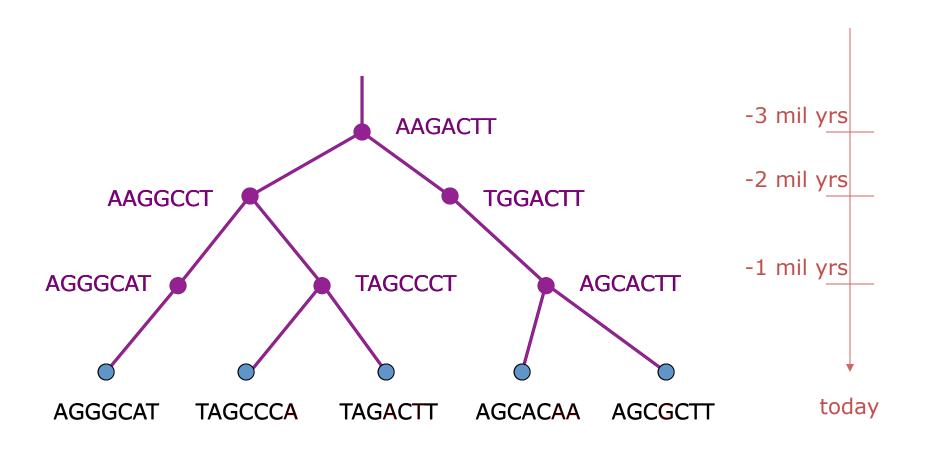
Novel techniques needed for scalability and accuracy

NP-hard problems and large datasets
Current methods do not provide good accuracy
Few methods can analyze even moderately large datasets

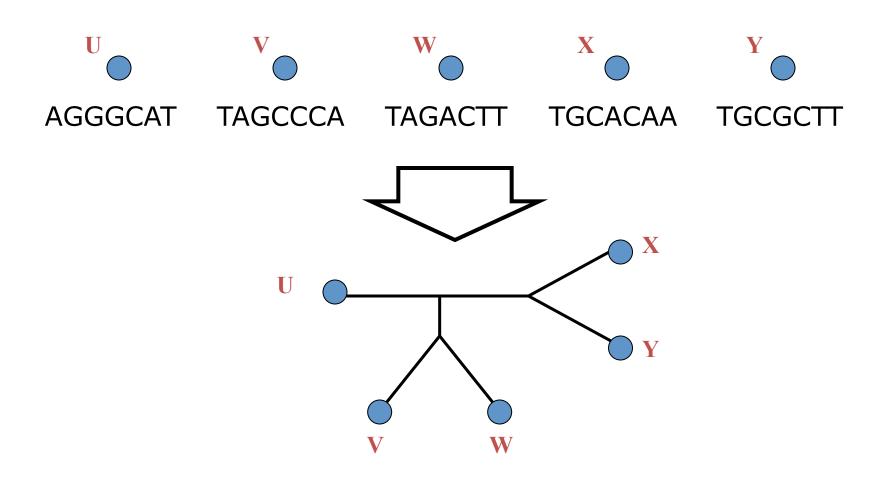
Many important applications besides phylogenetic estimation

¹ Frontiers in Massive Data Analysis, National Academies Press, 2013

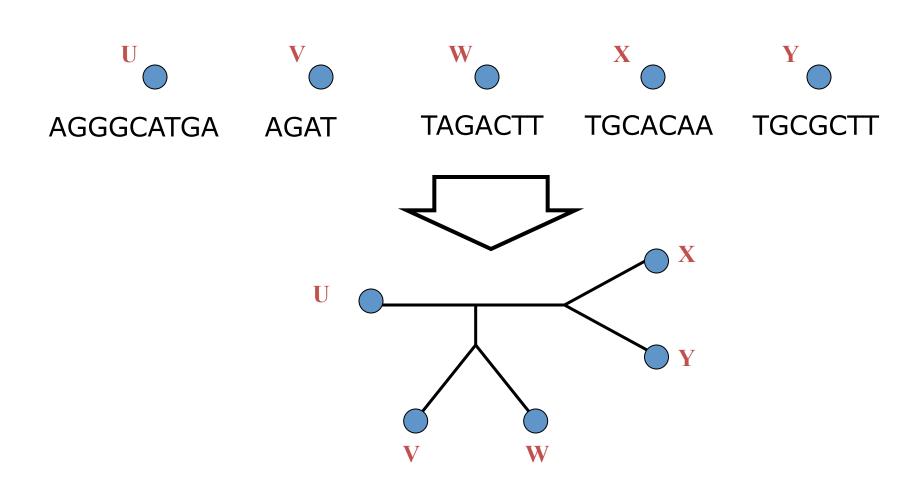
DNA Sequence Evolution



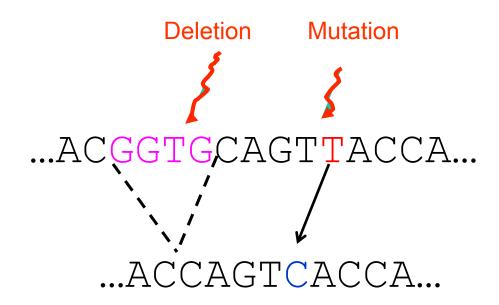
Phylogeny Problem

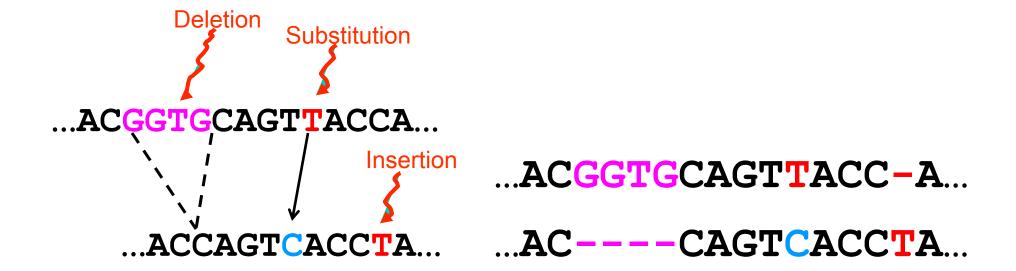


The "real" problem



Indels (insertions and deletions)





The true multiple alignment

- Reflects historical substitution, insertion, and deletion events
- Defined using transitive closure of pairwise alignments computed on edges of the true tree

Input: unaligned sequences

```
S1 = AGGCTATCACCTGACCTCCA
```

S2 = TAGCTATCACGACCGC

S3 = TAGCTGACCGC

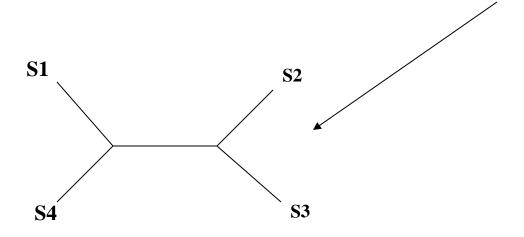
S4 = TCACGACCGACA

Phase 1: Alignment

```
S1 = AGGCTATCACCTGACCTCCA
S2 = TAGCTATCACGACCGC
S3 = TAGCTGACCGC
S4 = TCACGACCGACA
S1 = -AGGCTATCACCTGACCTCCA
S2 = TAG-CTATCAC--GACCGC--
S3 = TAG-CT-----GACCGC--
S4 = -----TCAC--GACCGACA
```

Phase 2: Construct tree

```
S1 = AGGCTATCACCTGACCTCCA
S2 = TAGCTATCACGACCGC
S3 = TAGCTGACCGC
S4 = TCACGACCGACA
S1 = -AGGCTATCACCTGACCTCCA
S2 = TAG-CTATCAC--GACCGC--
S3 = TAG-CT------GACCGC--
S4 = -----TCAC--GACCGACA
```



Phylogenomic pipeline

- Select taxon set and markers
- Gather and screen sequence data, possibly identify orthologs
- Compute multiple sequence alignments for each locus
- Compute species tree or network:
 - Compute gene trees on the alignments and combine the estimated gene trees, OR
 - Estimate a tree from a concatenation of the multiple sequence alignments
- Get statistical support on each branch (e.g., bootstrapping)
- Estimate dates on the nodes of the phylogeny
- Use species tree with branch support and dates to understand biology

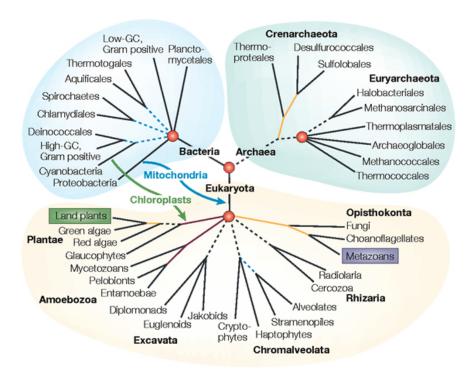
Phylogenomic pipeline

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- Estimate dates on the nodes of the phylogeny
- Use species tree with branch support and dates to understand biology

Large-scale Alignment Estimation

- Many genes are considered unalignable due to high rates of evolution
- Only a few methods can analyze large datasets
- iPlant (NSF Plant Biology Collaborative) and other projects planning to construct phylogenies with 500,000 taxa

Hard Computational Problems



Nature Reviews | Genetics

NP-hard problems

Large datasets
100,000+ sequences
thousands of genes

"Big data" complexity:
model misspecification
fragmentary sequences
errors in input data
streaming data

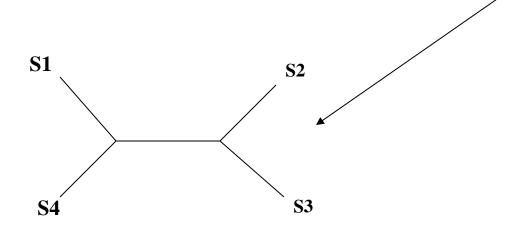
This talk

- "Big data" multiple sequence alignment
- SATé (2009, 2012) and PASTA (2014), methods for co-estimation of alignments and trees
- The HMM Family technique, and applications to
 - phylogenetic placement (SEPP, PSB 2012),
 - multiple sequence alignment (<u>UPP</u>, in preparation), and
 - metagenomic taxon identification (TIPP, submitted).

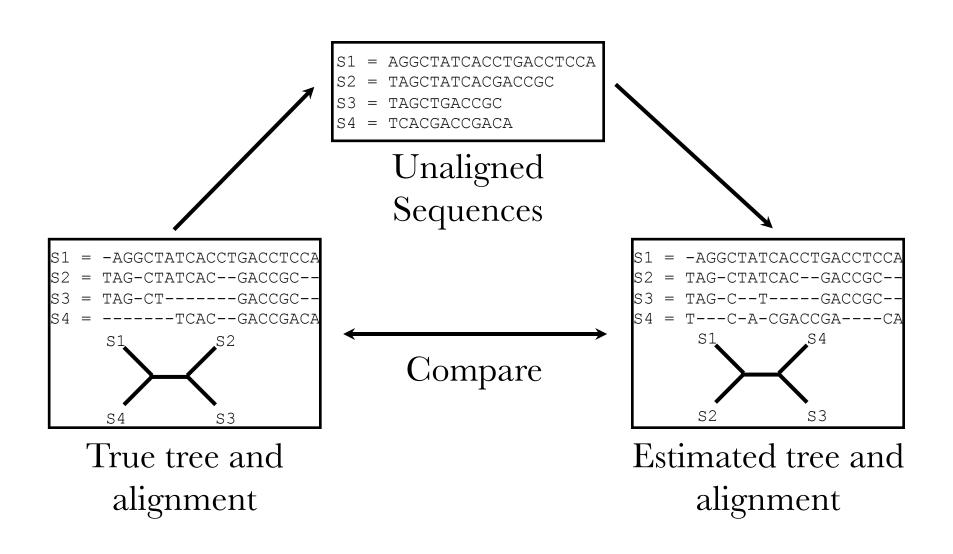
Multiple Sequence Alignment

First Align, then Compute the Tree

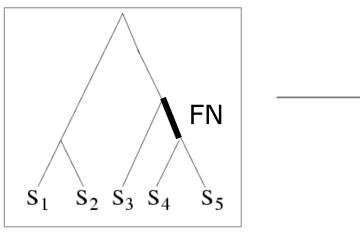
```
S1 = AGGCTATCACCTGACCTCCA
S2 = TAGCTATCACGACCGC
S3 = TAGCTGACCGC
S4 = TCACGACCGACA
S1 = -AGGCTATCACCTGACCTCCA
S2 = TAG-CTATCAC--GACCGC--
S3 = TAG-CT------GACCGC--
S4 = -----TCAC--GACCGACA
```



Simulation Studies



Quantifying Error



TRUE TREE

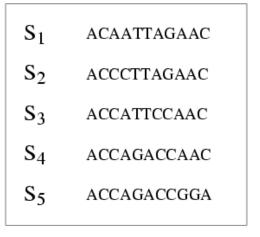
FN: false negative

(missing edge)

FP: false positive

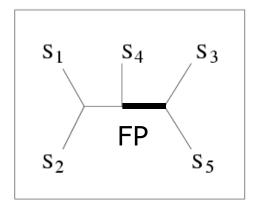
(incorrect edge)

50% error rate



DNA SEQUENCES





INFERRED TREE

Two-phase estimation

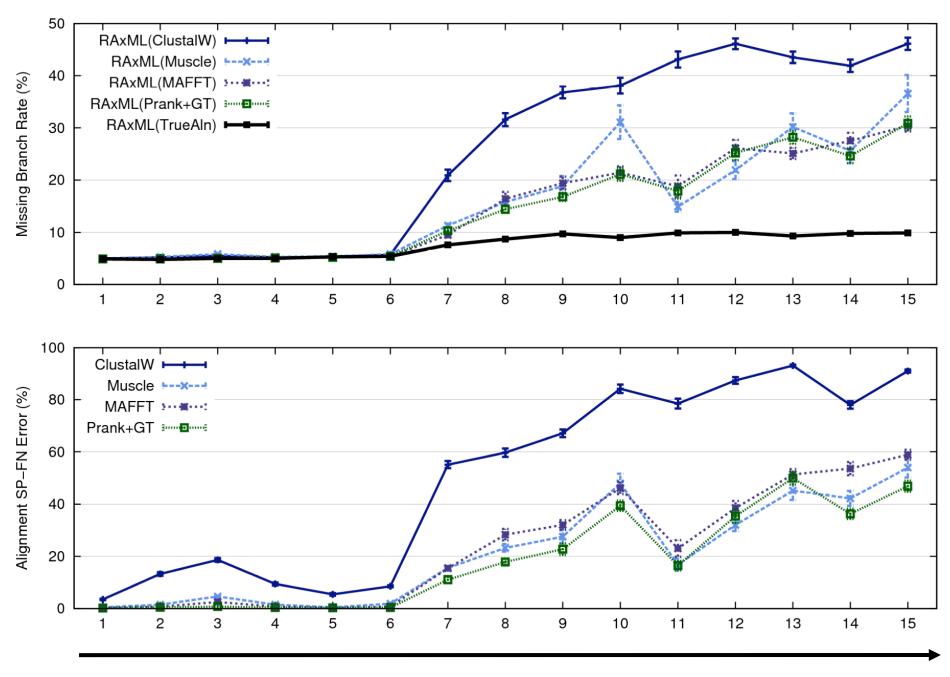
Alignment methods

- Clustal
- POY (and POY*)
- Probcons (and Probtree)
- Probalign
- MAFFT
- Muscle
- Di-align
- T-Coffee
- Prank (PNAS 2005, Science 2008)
- Opal (ISMB and Bioinf. 2007)
- FSA (PLoS Comp. Bio. 2009)
- Infernal (Bioinf. 2009)
- Etc.

Phylogeny methods

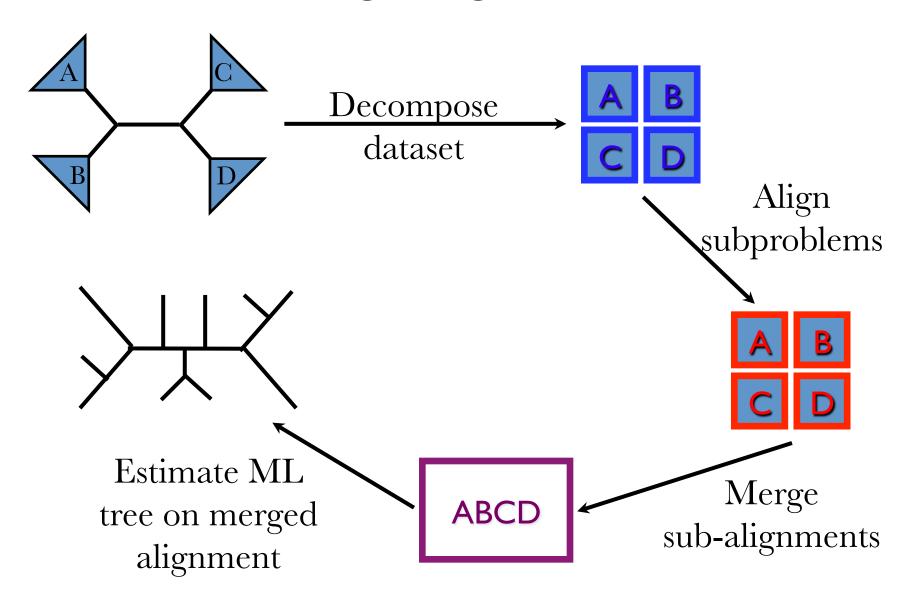
- Bayesian MCMC
- Maximum parsimony
- Maximum likelihood
- Neighbor joining
- FastME
- UPGMA
- Quartet puzzling
- Etc.

RAXML: heuristic for large-scale ML optimization



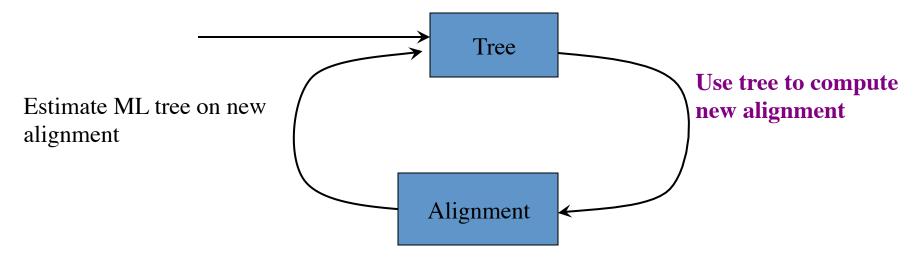
1000-taxon models, ordered by difficulty (Liu et al., 2009)

Re-aligning on a tree



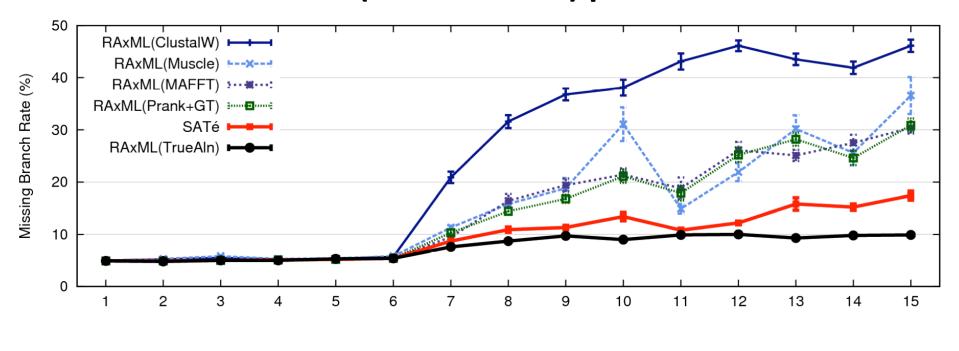
SATé and PASTA Algorithms

Obtain initial alignment and estimated ML tree



If new alignment/tree pair has worse ML score, realign using a different decomposition Repeat until termination condition (typically, 24 hours)

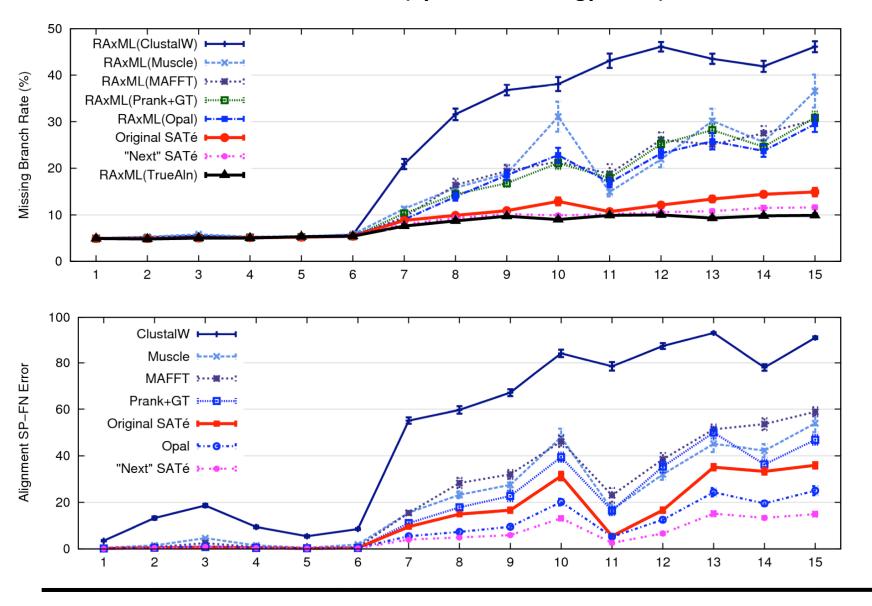
SATé-1 (Science 2009) performance



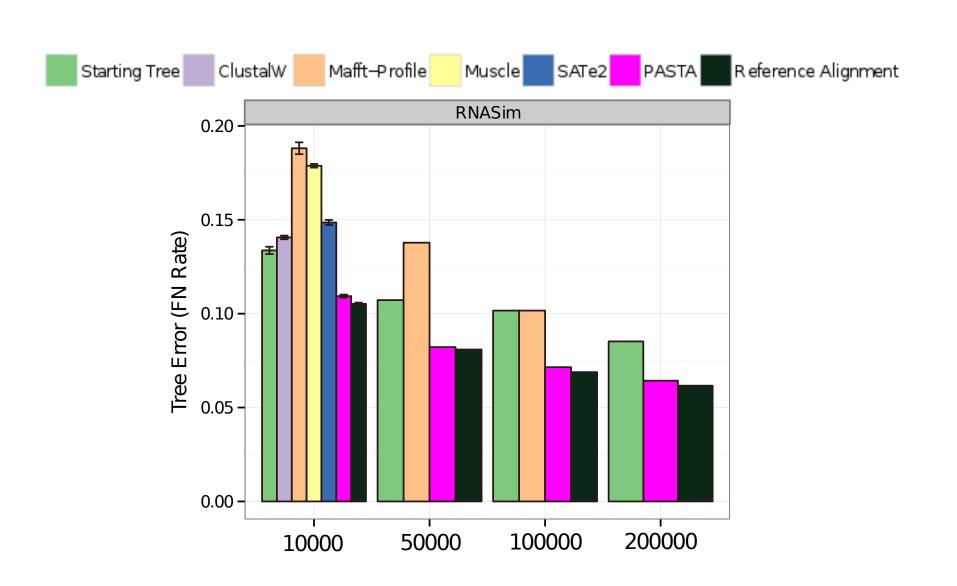
1000 taxon models, ordered by difficulty

SATé-1 24 hour analysis, on desktop machines (Similar improvements for biological datasets)
SATé-1 can analyze up to about 30,000 sequences.

SATé-1 and SATé-2 (Systematic Biology, 2012)



PASTA (2014): even better than SATé-2



1kp: Thousand Transcriptome Project

G. Ka-Shu Wong U Alberta J. Leebens-Mack U Georgia N. Wickett Northwestern N. Matasci iPlant

T. Warnow, UIUC

S. Mirarab, UT-Austin N. Nguyen, UT-Austin Md. S.Bayzid UT-Austin















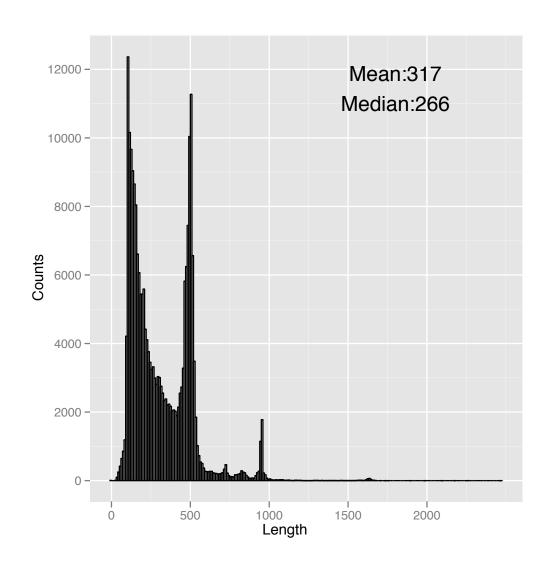


Plus many many other people...

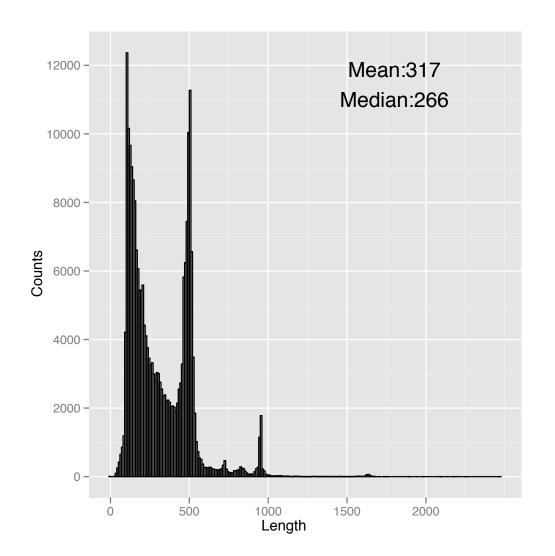
- Plant Tree of Life based on transcriptomes of ~1200 species
- More than 13,000 gene families (most not single copy)

Challenge:

Alignment of datasets with > 100,000 sequences with many fragmentary sequences



1KP dataset: more than 100,000 p450 amino-acid sequences, many fragmentary



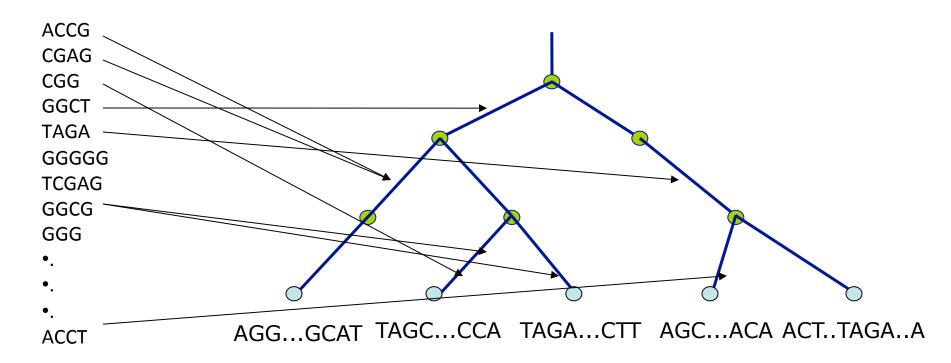
1KP dataset: more than 100,000 p450 amino-acid sequences, many fragmentary

All standard multiple sequence alignment methods we tested performed poorly on datasets with fragments.

Phylogenetic Placement

Fragmentary sequences from some gene

Full-length sequences for same gene, and an alignment and a tree



SEPP

 SEPP: SATé-enabled Phylogenetic Placement, by Mirarab, Nguyen, and Warnow. Pacific Symposium on Biocomputing, 2012, special session on the Human Microbiome

- Objective:
 - phylogenetic analysis of single-gene datasets with fragmentary sequences
- Introduces "HMM Family" technique

Phylogenetic Placement

Step 1: Align each query sequence to backbone alignment

Step 2: Place each query sequence into backbone tree, using extended alignment

Align Sequence

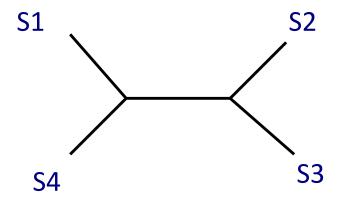
```
S1 = -AGGCTATCACCTGACCTCCA-AA

S2 = TAG-CTATCAC--GACCGC--GCA

S3 = TAG-CT----GACCGC--GCT

S4 = TAC---TCAC--GACCGACAGCT

Q1 = TAAAAC
```



Align Sequence

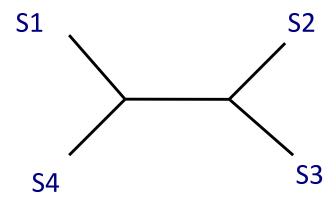
```
S1 = -AGGCTATCACCTGACCTCCA-AA

S2 = TAG-CTATCAC--GACCGC--GCA

S3 = TAG-CT----GACCGC--GCT

S4 = TAC---TCAC--GACCGACAGCT

Q1 = -----T-A--AAAC-----
```



Place Sequence

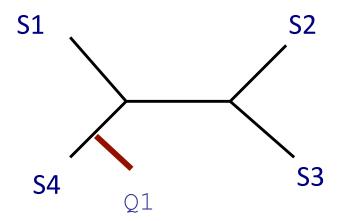
```
S1 = -AGGCTATCACCTGACCTCCA-AA

S2 = TAG-CTATCAC--GACCGC--GCA

S3 = TAG-CT-----GACCGC--GCT

S4 = TAC---TCAC--GACCGACAGCT

Q1 = -----T-A--AAAC-----
```

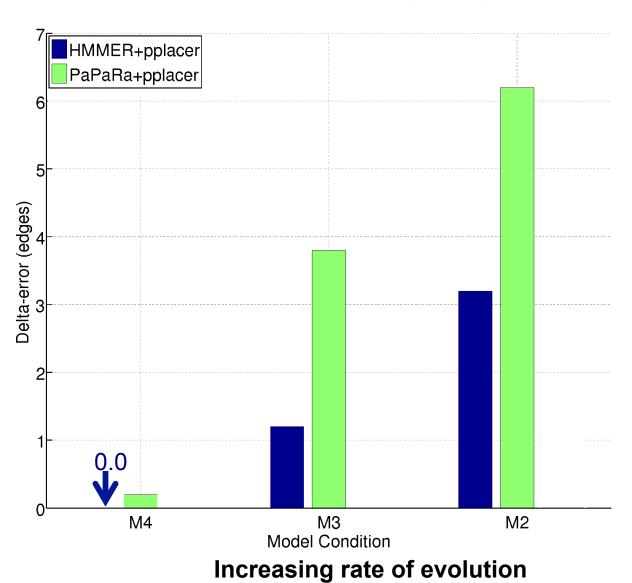


Phylogenetic Placement

- Align each query sequence to backbone alignment
 - HMMALIGN (Eddy, Bioinformatics 1998)
 - PaPaRa (Berger and Stamatakis, Bioinformatics 2011)
- Place each query sequence into backbone tree
 - Pplacer (Matsen et al., BMC Bioinformatics, 2011)
 - EPA (Berger and Stamatakis, Systematic Biology 2011)

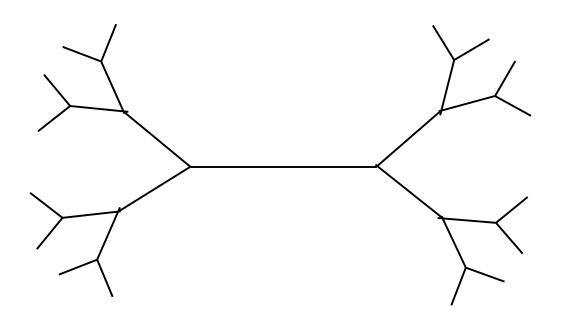
Note: pplacer and EPA use maximum likelihood, and are reported to have the same accuracy.

HMMER vs. PaPaRa

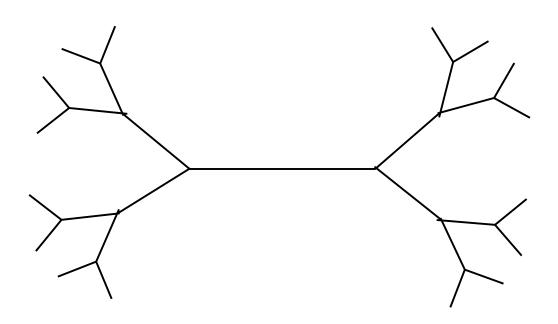


HMMER+pplacer:

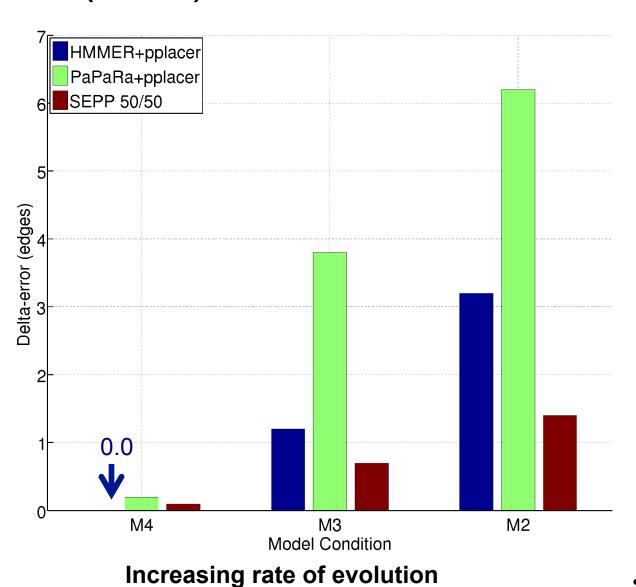
- 1) build one HMM for the entire alignment
- 2) Align fragment to the HMM, and insert into alignment
- 3) Insert fragment into tree to optimize likelihood



One Hidden Markov Model for the entire alignment?



SEPP(10%), based on ~10 HMMs



SEPP vs. HMMER+pplacer

SEPP produced more accurate phylogenetic placements than HMMER+pplacer.

The only difference is the use of a Family of HMMs instead of one HMM.

 The biggest differences are for datasets with high rates of evolution.

UPP: large-scale MSA estimation

UPP = "Ultra-large multiple sequence alignment using Phylogeny-aware Profiles"

Nguyen, Mirarab, and Warnow. In preparation.

Objective: highly accurate large-scale multiple sequence alignments, even in the presence of fragmentary sequences.

Uses a variant of the HMM Family technique in SEPP

UPP Algorithmic Approach

- Select random subset of sequences, and build "backbone alignment"
- Construct a "Family of Hidden Markov Models" on the backbone alignment (the family has HMMs on many subsets of different sizes, not disjoint)
- Add all remaining sequences to the backbone alignment using the Family of HMMs

Evaluation

- Simulated datasets (some have fragmentary sequences):
 - 10K to 1,000,000 sequences in RNASim (Guo, Wang, and Kim, arxiv)
 - 1000-sequence nucleotide datasets from SATé papers
 - 5000-sequence AA datasets (from FastTree paper)
 - 10,000-sequence Indelible nucleotide simulation
- Biological datasets:
 - Proteins: largest BaliBASE and HomFam
 - RNA: 3 CRW datasets up to 28,000 sequences

Summary

- SATé-1 (Science 2009), SATé-2 (Systematic Biology 2012), and PASTA (RECOMB 2014): methods for co-estimating gene trees and multiple sequence alignments. PASTA can analyze up to 200,000 sequences, and is highly accurate for full-length sequences.
 - But none of these methods are robust to fragmentary sequences.
- HMM Family technique: uses an ensemble of HMMs to represent a "backbone alignment". HMM families improve accuracy, especially in the presence of high rates of evolution.
- Applications of HMM Families in:
 - SEPP (phylogenetic placement)
 - UPP (ultra-large multiple sequence alignment) up to 1,000,000 sequences
 - TIPP (metagenomic taxon identification and abundance profiling)

Acknowledgments









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Undergrad: Keerthana Kumar

Lab Website: http://www.cs.utexas.edu/users/phylo

Personal Website: http://tandy.cs.illinois.edu

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TACC and UTCS computational resources

* Supported by HHMI Predoctoral Fellowship

** Supported by Fulbright Foundation Predoctoral Fellowship

The NIH Human Microbiome Project

25,000 human genes, 1,000,000 bacterial genes



Basic Questions

1. What is this fragment? (Classify each fragment as well as possible.)

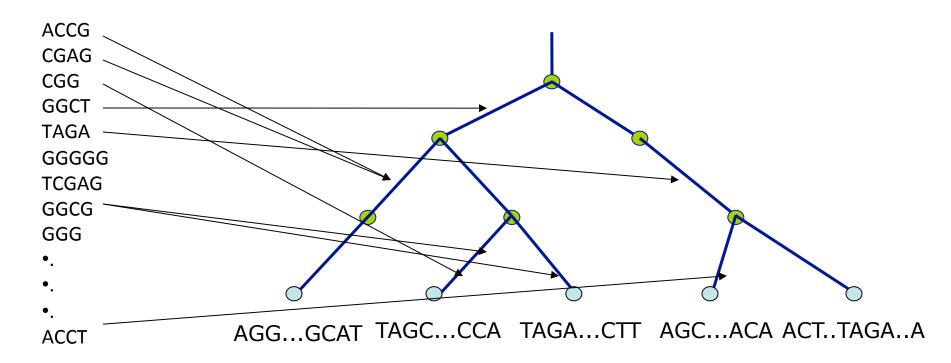
2. What is the taxonomic distribution in the dataset? (Note: helpful to use marker genes.)

3. What are the organisms in this metagenomic sample doing together?

Phylogenetic Placement

Fragmentary sequences from some gene

Full-length sequences for same gene, and an alignment and a tree



TIPP: SEPP + statistics

SEPP has high recall but low precision (classifies almost everything)

TIPP: dramatically reduces false positive rate with small reduction in true positive rate, by considering uncertainty in alignment (HMMER) and placement (pplacer)

TIPP: Taxon Identification and Phylogenetic Profiling. N. Nguyen, S. Mirarab, and T. Warnow. Under review.

Algorithmic Strategies

- Divide-and-conquer
- Chordal graph decompositions
- Iteration
- Multiple HMMs
- Bin-and-conquer (technique used for improving species tree estimation from multiple gene trees, Bayzid and Warnow, Bioinformatics 2013)

1kp: 1000 Plant Transcriptomes

G. Ka-Shu Wong U Alberta

J. Leebens-Mack U Georgia

N. Wickett Northwestern

N. Matasci iPlant

T. Warnow, UT-Austin

S. Mirarab, UT-Austin

N. Nguyen, UT-Austin















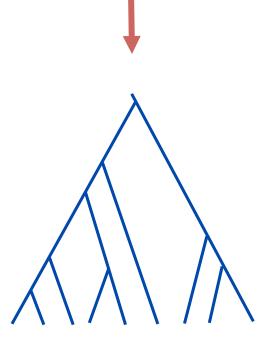
Plus many many other people...

- Whole Transcriptomes of 103 plant species and 850 single copy loci (1200 taxa in next phase)
 - Most accurate summary methods cannot handle this size
- Common ancestor about 1 billion years ago and so gene trees are hard to root
 - Most summary methods need rooted gene trees
- Pre-existing summary methods do not provide reasonable results on this dataset

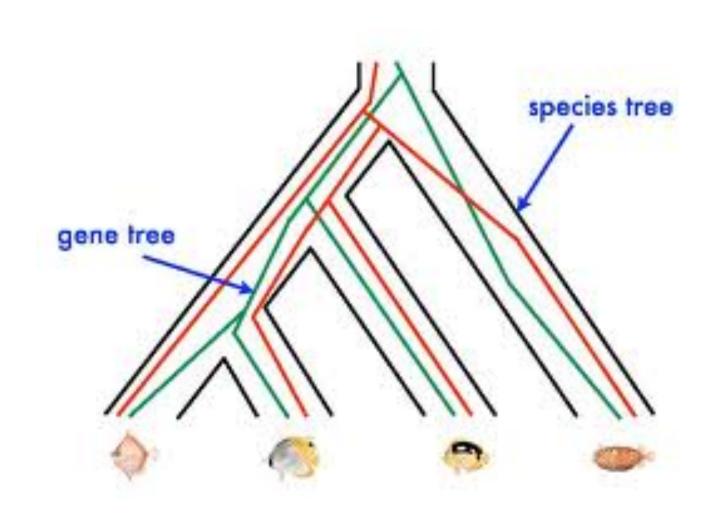
[Wickett et al. (under review), 2014.]

Combined analysis

gene 1 gene 2 gene 3

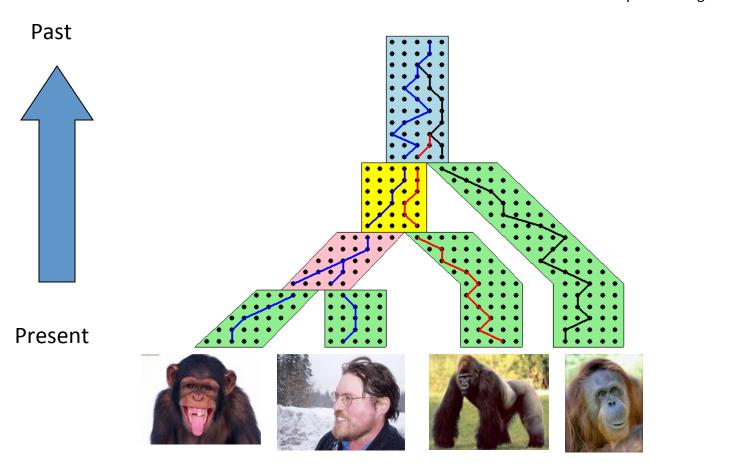


Red gene tree ≠ species tree (green gene tree okay)



The Coalescent

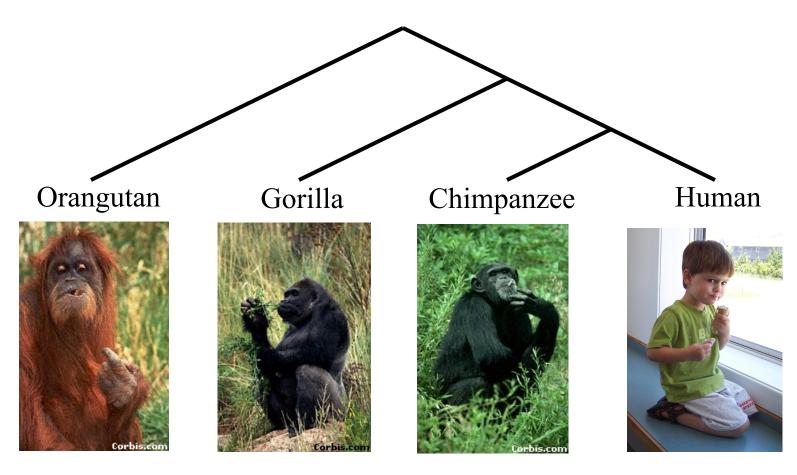
Courtesy James Degnan



Incomplete Lineage Sorting (ILS)

- 1000+ papers in 2013 alone
- Confounds phylogenetic analysis for many groups:
 - Hominids
 - Birds
 - Yeast
 - Animals
 - Toads
 - Fish
 - Fungi
- There is substantial debate about how to analyze phylogenomic datasets in the presence of ILS.

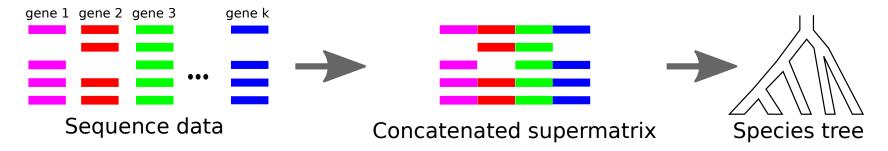
Species tree estimation: difficult, even for small datasets

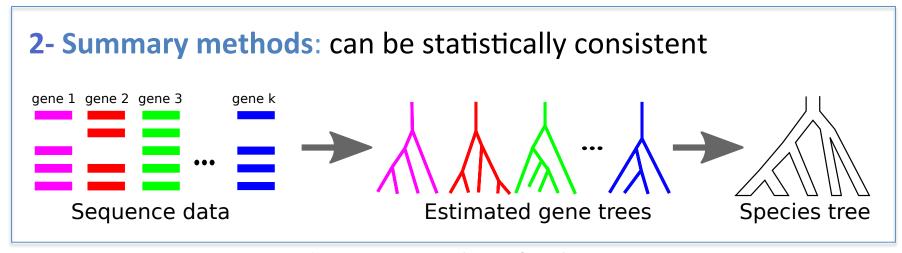


From the Tree of the Life Website, University of Arizona

Species tree estimation

1- Concatenation: statistically inconsistent (Roch & Steel 2014)





3- Co-estimation methods: too slow for large datasets

Is Concatenation Evil?

- Joseph Heled:
 - YES

- John Gatesy
 - No

- Data needed to held understand existing methods and their limitations
- Better methods are needed

Avian Phylogenomics Project (100+ people)

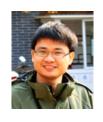
Erich Jarvis, HHMI



MTP Gilbert, Copenhagen



G Zhang, BGI



T. Warnow UT-Austin



S. Mirarab Md. S. Bayzid UT-Austin UT-Austin





- Approx. 50 species, whole genomes
- 8000+ genes, UCEs
- Gene sequence alignments and trees computed using SATé (Science 2009, Systematic Biology 2012)
- Concatenation analysis (multi-million site) using ExaML (new version of RAxML for very long alignments)
- Massive gene tree incongruence suggestive of incomplete lineage sorting -coalescent-based species tree estimation computed using "Statistical Binning" (Science, in press)

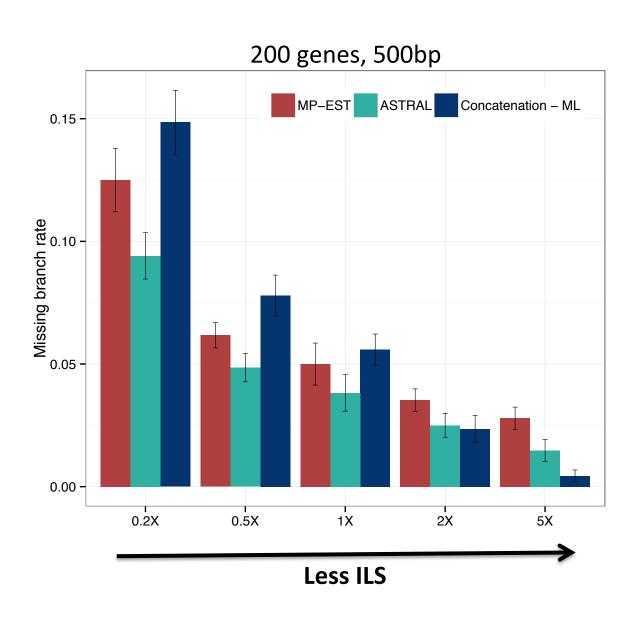
Our methods

- "Statistical Binning" (accepted): uses a statistical method to determine sets of "combinable" gene sequence alignments, improves coalescent-based species tree estimation accuracy when gene trees have poor resolution (used for Avian Phylogenomics Project).
- ASTRAL (Bioinformatics 2014): polynomial time statistically consistent method, can run on very large datasets of unrooted gene trees.

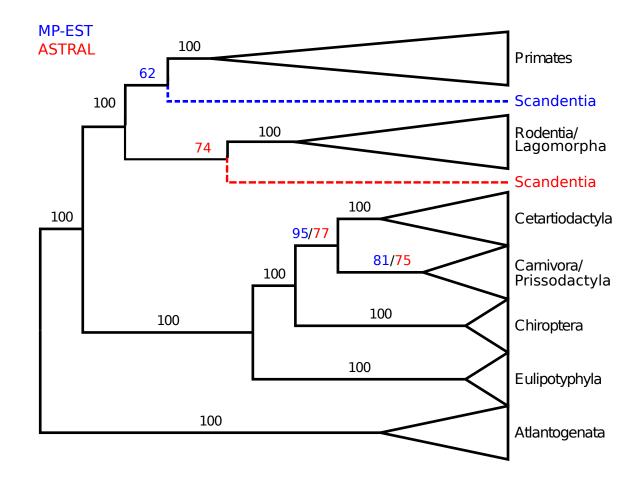
Mammalian simulations

- Based on a biological mammalian dataset of 37 taxa and 442 genes, published by Song et al., PNAS, 2012.
- In simulations, we vary
 - Levels of ILS
 - Number of genes
 - Alignment length to control gene tree estimation error
- Compare ASTRAL to
 - MP-EST, BUCKy
 - Concatenation
 - MRP, Greedy
- Measure species tree error compared to the known true tree

ASTRAL vs. Concatenation



Analyses of the Song et al. Mammalian dataset



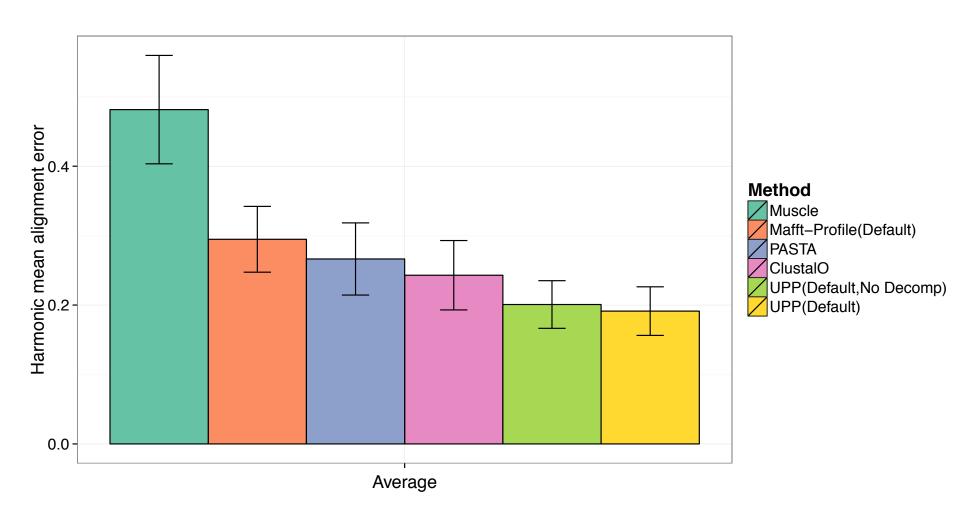
The placement of Scandentia (Tree Shrew) is controversial.

The ASTRAL analysis agrees with maximum likelihood concatenation analysis of this dataset.

Summary

- New multiple sequence alignments with improved accuracy and scalability, as well as high robustness to fragmentary data – SATé used in many studies.
- New coalescent-based species tree estimation methods that have better accuracy than current methods, and can analyze large datasets (used in 1KP and Avian Phylogenomics project analyses)
- New methods for metagenomic taxon identification and abundance profiling with improved accuracy (will be used in analyses here at Illinois)
- Method development inspired by collaboration with biologists, and improves biological analyses.

AA Sequence Alignment Error (13 HomFam datasets)



Homfam: Sievers et al., MSB 2011 10,000 to 46,000 sequences per dataset; 5-14 sequences in each seed alignment Sequence lengths from 46-364 AA

Other co-estimation methods

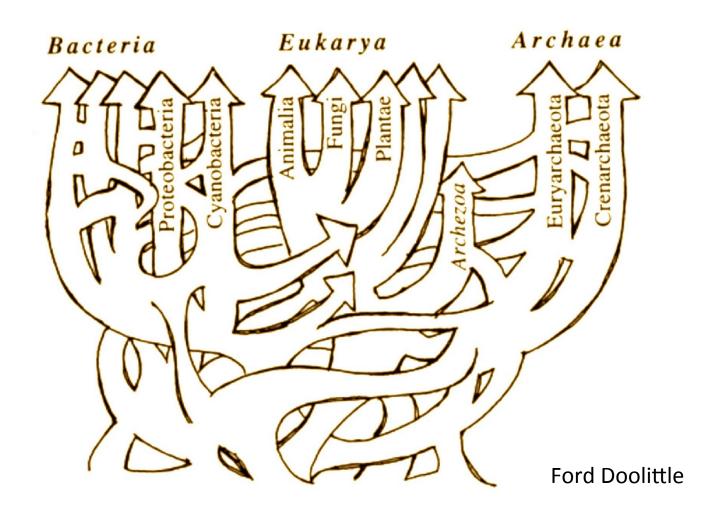
Statistical methods:

- BAli-Phy (Redelings and Suchard): Bayesian software to coestimate alignments and trees under a statistical model of evolution that includes indels. Can scale to about 100 sequences, but takes a very long time.
 - http://www.bali-phy.org/
- StatAlign: http://statalign.github.io/

Extensions of Parsimony

- POY (most well known software)
 - http://www.amnh.org/our-research/computational-sciences/ research/projects/systematic-biology/poy
- BeeTLe (Liu and Warnow, PLoS One 2012)

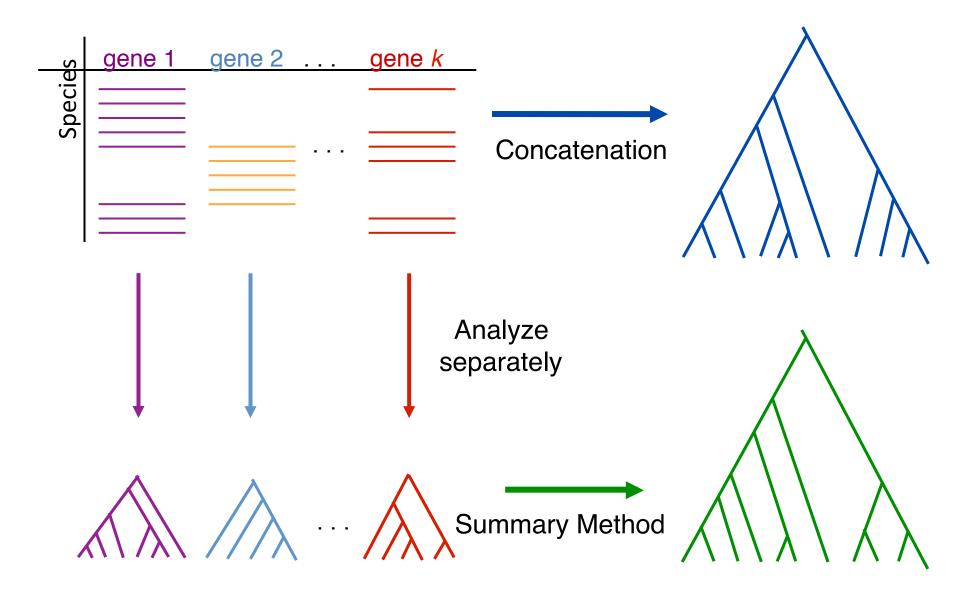
Horizontal Gene Transfer – Phylogenetic Networks



But...

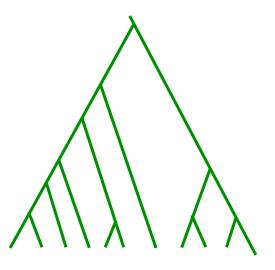
- Gene trees may not be identical to species trees:
 - Incomplete Lineage Sorting (deep coalescence)
 - Gene duplication and loss
 - Horizontal gene transfer
- This makes combined analysis and standard supertree analyses inappropriate

Two competing approaches



How to compute a species tree?





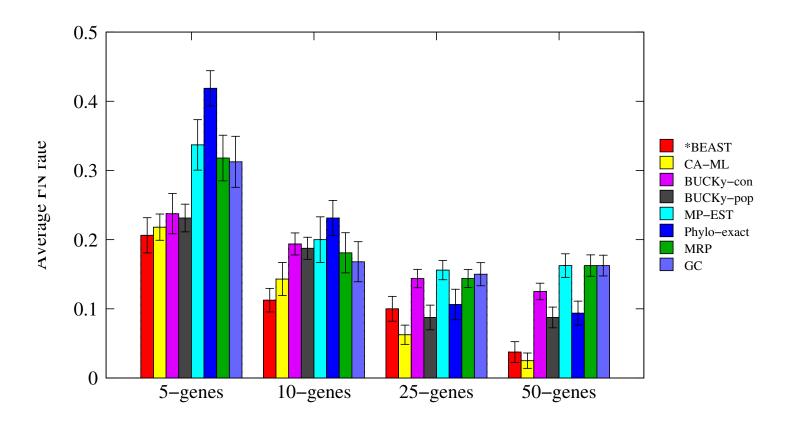
Statistically consistent under ILS?

- MP-EST (Liu et al. 2010): maximum likelihood estimation of rooted species tree – YES
- BUCKy-pop (Ané and Larget 2010): quartet-based Bayesian species tree estimation –YES
- MDC NO
- Greedy NO
- Concatenation under maximum likelihood NO
- MRP (supertree method) open

The Debate: Concatenation vs. Coalescent Estimation

- In favor of coalescent-based estimation
 - Statistical consistency guarantees
 - Addresses gene tree incongruence resulting from ILS
 - Some evidence that concatenation can be positively misleading
- In favor of concatenation
 - Reasonable results on data
 - High bootstrap support
 - Summary methods (that combine gene trees) can have poor support or miss well-established clades entirely
 - Some methods (such as *BEAST) are computationally too intensive to use

Results on 11-taxon datasets with strongILS



*BEAST more accurate than summary methods (MP-EST, BUCKy, etc)
CA-ML: (concatenated analysis) also very accurate

Datasets from Chung and Ané, 2011 Bayzid & Warnow, Bioinformatics 2013

Species tree/network estimation

- Methods have been developed to estimate species phylogenies (trees or networks!) from gene trees, when gene trees can conflict from each other (e.g., due to ILS, gene duplication and loss, and horizontal gene transfer).
- Phylonet (software suite), has effective methods for many optimization problems – including MDC and maximum likelihood.
- Tutorial on Wednesday.
- Software available at http://bioinfo.cs.rice.edu/phylonet?destination=node/3

Two Basic Questions

1. What is this fragment? (Classify each fragment as well as possible.)

2. What is the taxonomic distribution in the dataset? (Note: helpful to use marker genes.)

SEPP

- SEPP: SATé-enabled Phylogenetic
 Placement, by Mirarab, Nguyen, and Warnow
- Pacific Symposium on Biocomputing, 2012 (special session on the Human Microbiome)
- Tutorial on Thursday.

Other problems

- Genomic MSA estimation:
 - Multiple sequence alignment of very long sequences
 - Multiple sequence alignment of sequences that evolve with rearrangement events
- Phylogeny estimation under more complex models
 - Heterotachy
 - Violation of the rates-across-sites assumption
 - Rearrangements
- Estimating branch support on very large datasets