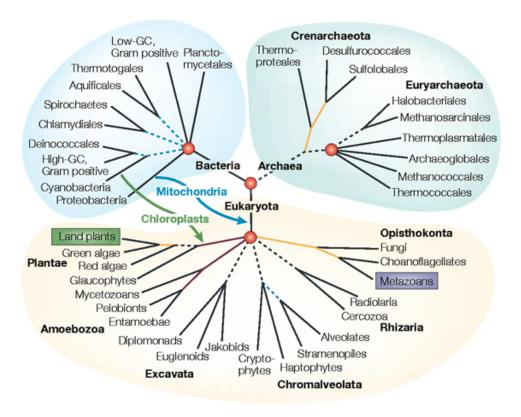
Four techniques for large-scale multiple sequence alignment and phylogenetic estimation

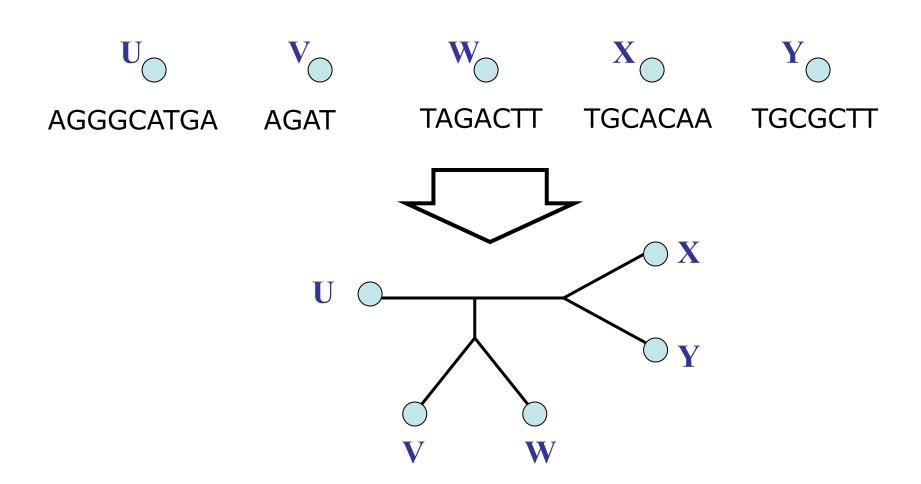
Tandy Warnow

Department of Computer Science

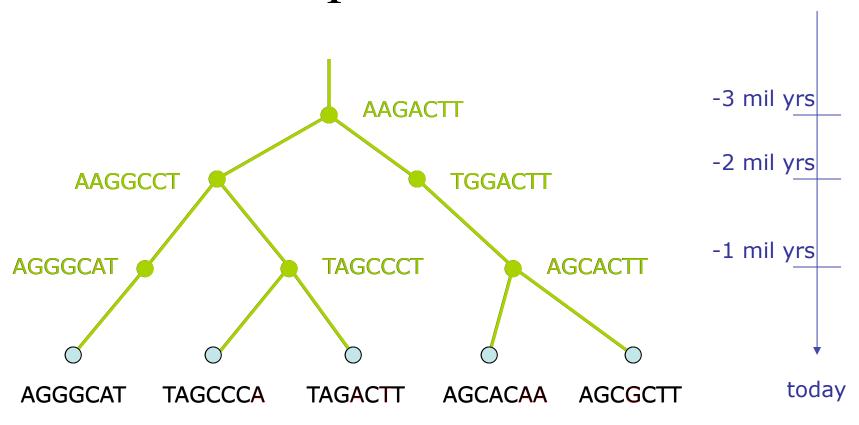
The University of Texas at Austin

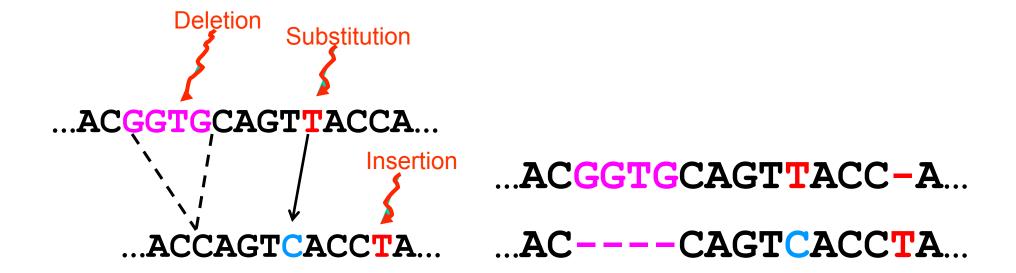
Assembling the Tree of Life





DNA Sequence Evolution





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

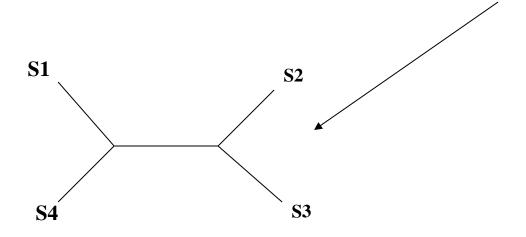
S4 = TCACGACCGACA

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
```



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.

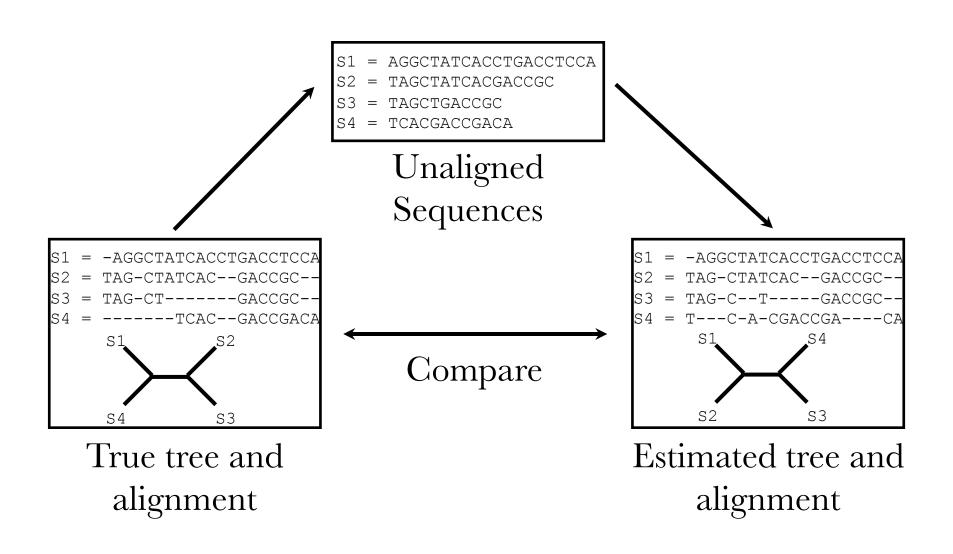
Phylogeny methods

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

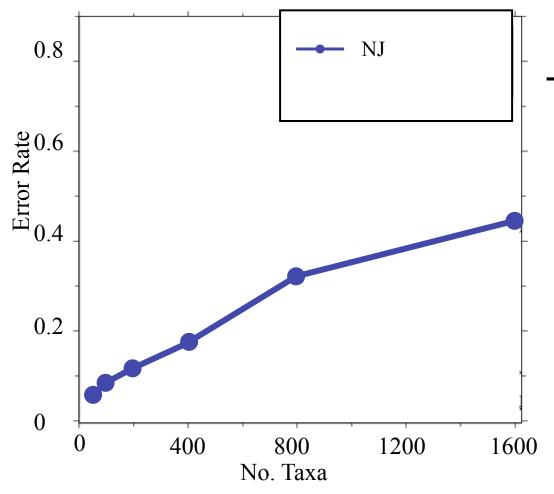
RAXML: heuristic for large-scale ML optimization

Large numbers of taxa

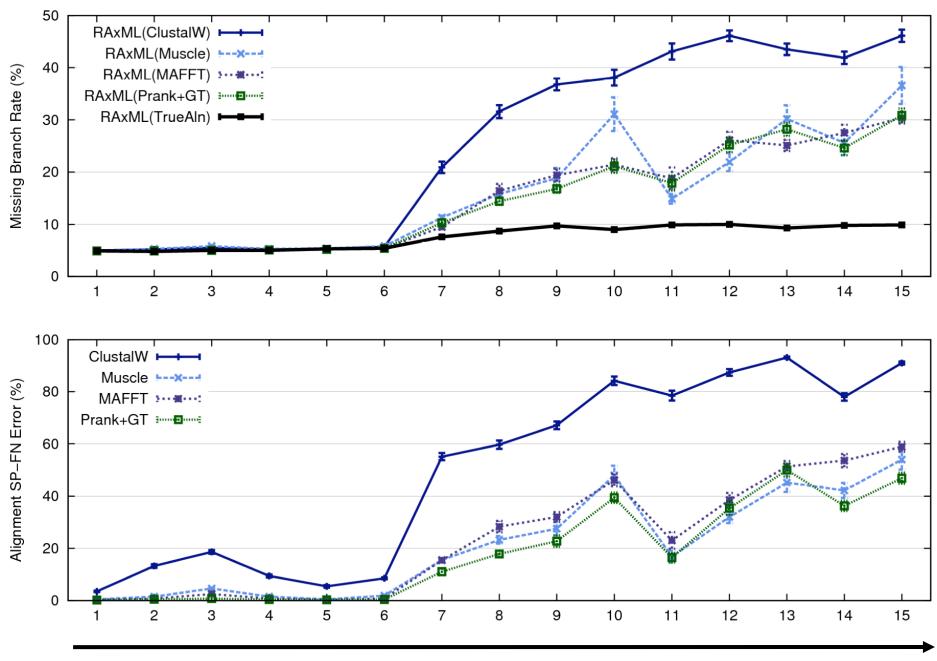
Simulation Studies



Neighbor joining has poor performance on large diameter trees [Nakhleh et al. ISMB 2001]



Theorem (Atteson): **Exponential**sequence length
requirement for
Neighbor Joining!



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

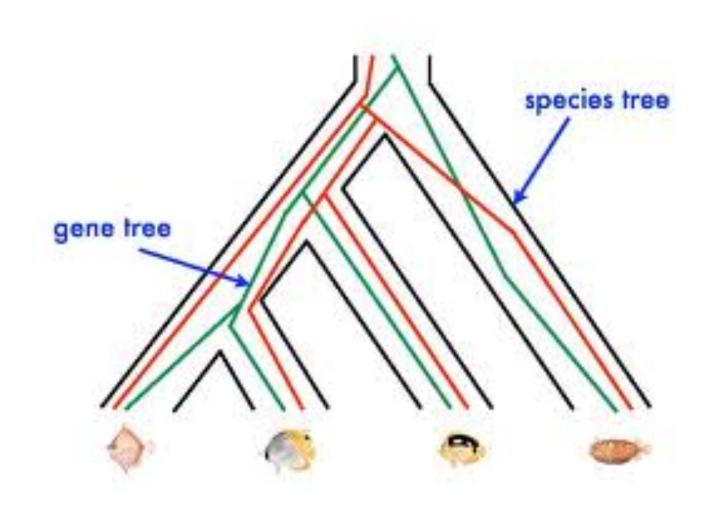
- Large numbers of taxa
 - Accurate multiple sequence alignment is challenging, and has a large impact on phylogeny estimation
 - The best phylogeny estimation methods are heuristics for NP-hard problems (standard polynomial time methods can have poor accuracy, even on the true alignment)

Large numbers of genes

- Large numbers of genes
 - "Concatenation" can become computationally infeasible

- Large numbers of genes
 - "Concatenation" can become computationally infeasible
 - Gene tree incongruence can make accurate species tree estimation challenging

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



1kp (http://www.onekp.com/)

















Gane Ka-Shu Wong U Alberta

Jim Leebens-Mack U Georgia

Norm Wickett Northwestern

Naim Matasci iPlant – U Arizona

Tandy Warnow,

Siavash Mirarab, Nam Nguyen, and UT-Austin

Md. S. Bayzid

- Transcriptomes of approx. 1200 species
- More than 13,000 gene families (most not single copy)
- Multi-institutional project (10+ universities)

Challenges:

Estimating very large gene alignments and trees (100,000+ sequences)

Estimating species trees from incongruent gene trees

Avian Phylogenomics Project

E.Jarvis, HHMI

MTP Gilbert, Copenhagen

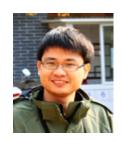
G. Zhang, BGI

S. Mirarab,

T. Warnow, and Md. S.Bayzid, UT-Austin













- Approx. 50 species, whole genomes
- 8000+ genes, UCEs
- Gene trees and sequence alignments computed using SATé
- Species tree estimated using maximum likelihood (RAxML)
- Multi-national team (20+ investigators)

Biggest challenges:

Estimating species tree from incongruent gene trees, Poor phylogenetic signal in most genes

Major Challenges: large datasets, fragmentary sequences

- Multiple sequence alignment: Few methods can run on large datasets, and alignment accuracy is generally poor for large datasets with high rates of evolution.
- Gene Tree Estimation: standard methods have poor accuracy on even moderately large datasets, and the most accurate methods are enormously computationally intensive (weeks or months, high memory requirements).
- Species Tree Estimation: gene tree incongruence makes accurate estimation of species tree challenging.

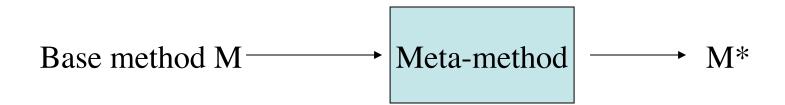
Both phylogenetic estimation and multiple sequence alignment are also impacted by *fragmentary data*.

This Talk

- SATé co-estimating trees and alignments (Science, 2009)
- DACTAL estimating trees (almost) without alignments (ISMB 2012)
- **SEPP** phylogenetic placement of fragmentary sequence data (e.g., short reads) (PSB 2012)
- UPP Ultra-large alignment using SEPP (unpublished)

Meta-Methods

 Meta-methods "boost" the performance of base methods (phylogeny reconstruction, alignment estimation, etc).



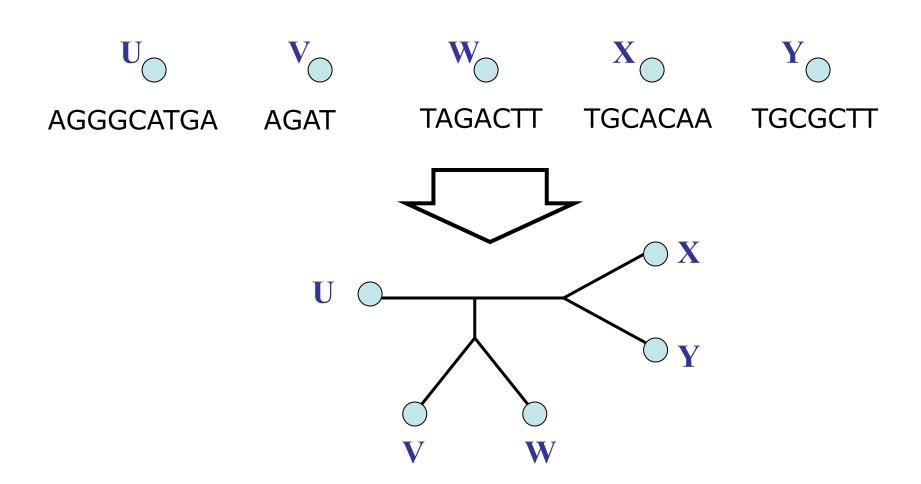
Part I: SATé

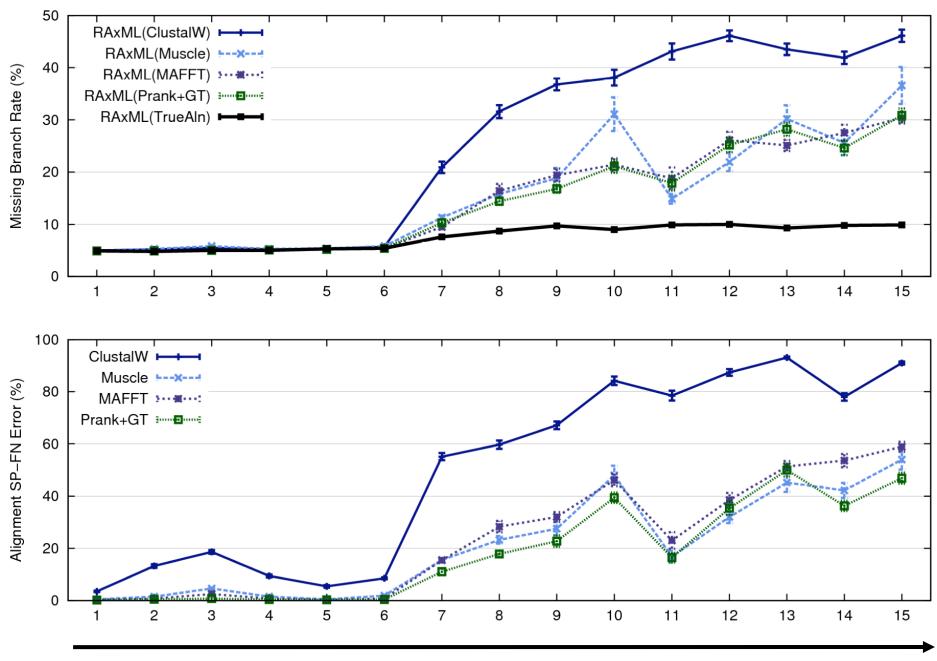
Simultaneous Alignment and Tree Estimation

Liu, Nelesen, Raghavan, Linder, and Warnow, *Science*, 19 June 2009, pp. 1561-1564.

Liu et al., Systematic Biology 2012

Public software distribution (open source) through Mark Holder's group at the University of Kansas





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

SATé Algorithm

Obtain initial alignment and estimated ML tree

Tree

SATé Algorithm

Obtain initial alignment and estimated ML tree

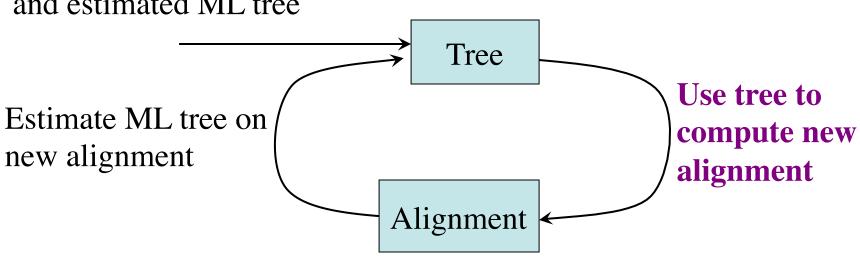
Tree

Use tree to compute new alignment

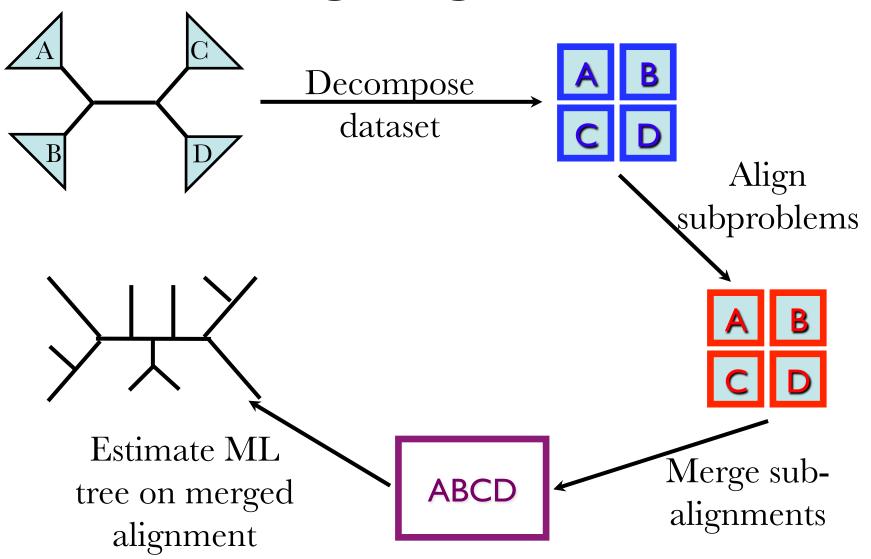
Alignment

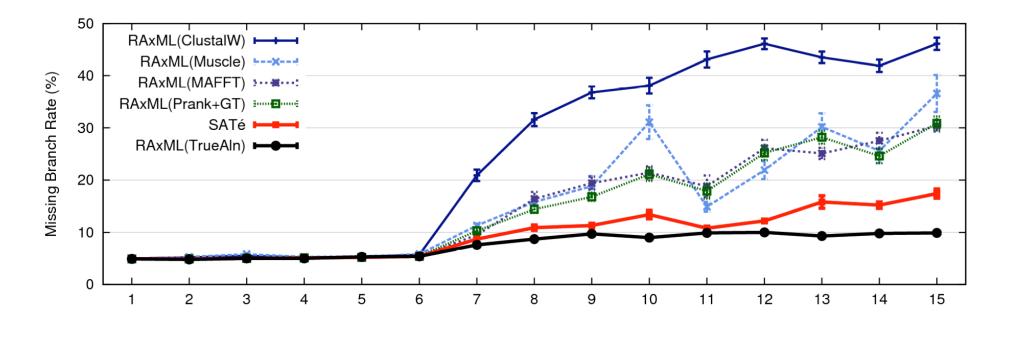
SATé Algorithm

Obtain initial alignment and estimated ML tree



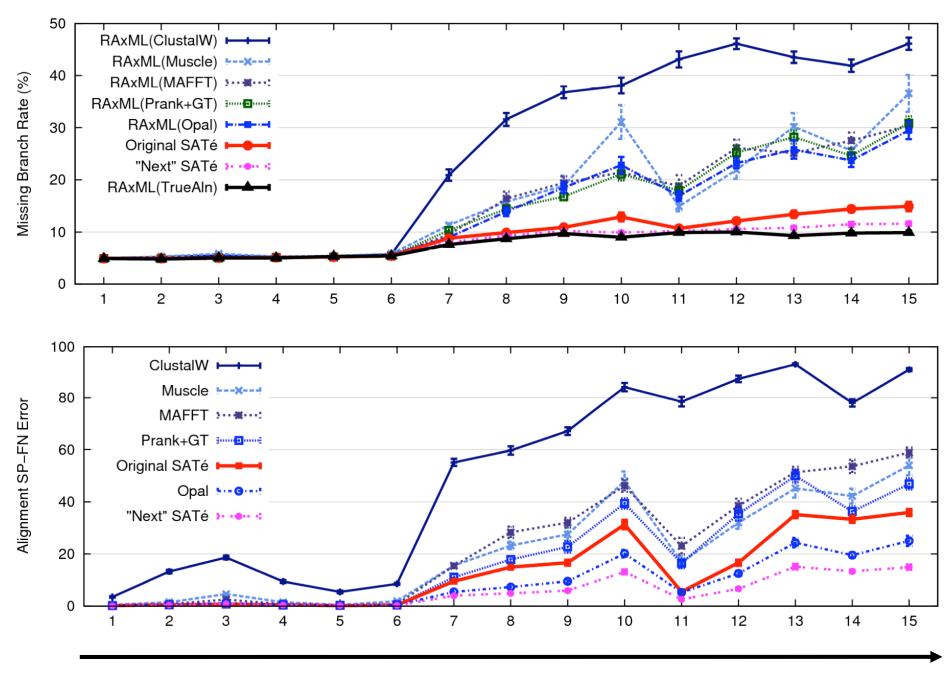
Re-aligning on a tree





1000 taxon models, ordered by difficulty

24 hour SATé analysis, on desktop machines (Similar improvements for biological datasets)

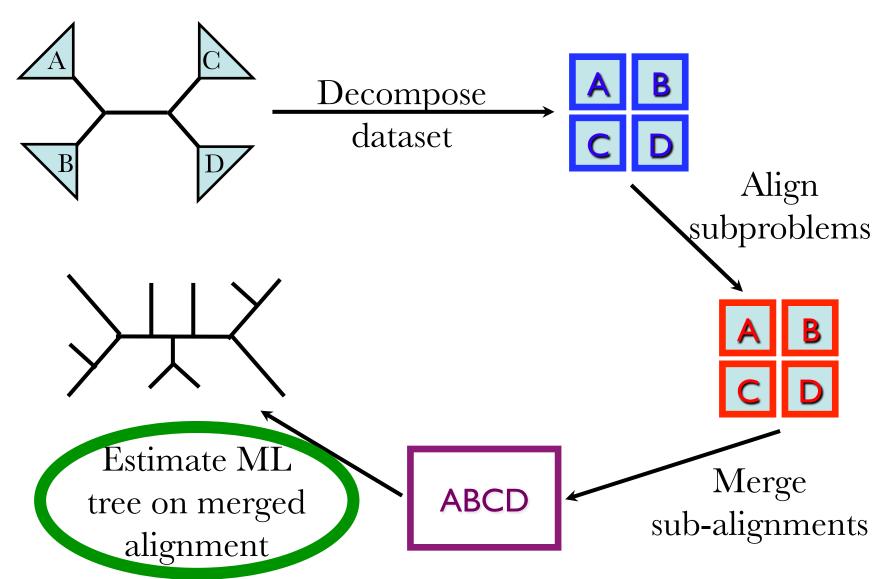


1000 taxon models ranked by difficulty

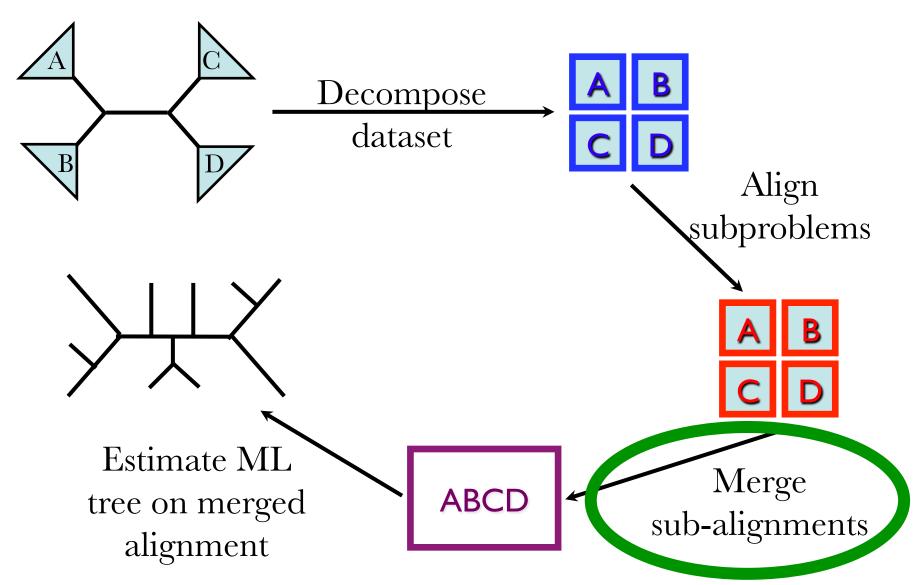
Brief discussion

- SATé "boosts" the base methods. Results shown are for SATé used with MAFFT and Muscle. Similar improvements seen for use with Prank, Opal, Muscle, ClustalW, etc.
- Biological datasets: Similar results on large benchmark datasets (structurally-based rRNA alignments)
- No statistical guarantees!!! In fact, it's all bad news: ML, treating gaps as missing data (even given the true alignment), can be inconsistent!
- Performance in practice results from use of base methods (and ability to use best versions of base methods).
- Alignment of genome-scale sequences is a different problem.
- SATé is designed for full-length sequences, not fragmentary datasets

Limitations



Limitations

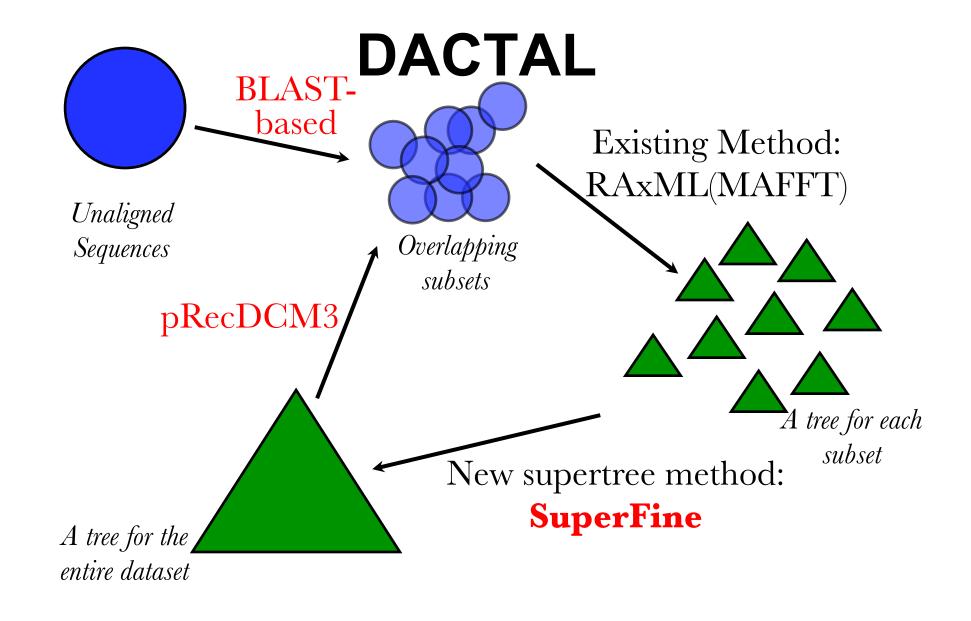


Part II: DACTAL

Divide-And-Conquer Trees (Almost) without alignments

- Input: set S of unaligned sequences
- Output: tree on S (but no alignment)

Nelesen, Liu, Wang, Linder, and Warnow, ISMB 2012 and Bioinformatics 2012



Average of 3 Largest CRW Datasets

CRW: Comparative RNA database,

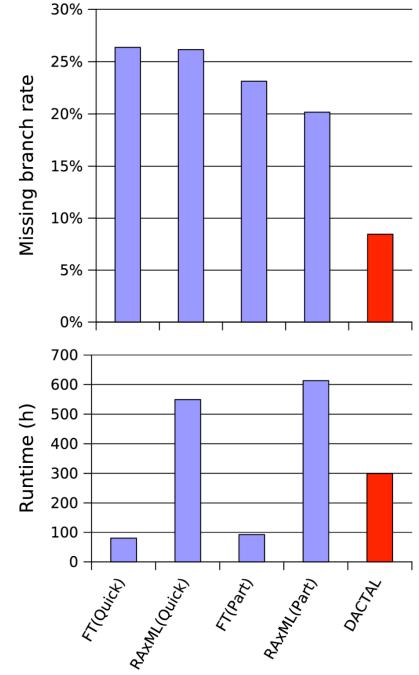
Three 16S datasets with 6,323 to 27,643 sequences

Reference alignments based on secondary structure

Reference trees are 75% RAxML bootstrap trees

DACTAL (shown in red) run for 5 iterations starting from FT(Part)

FastTree (FT) and RAxML are ML methods



DACTAL and SATe have comparable accuracy

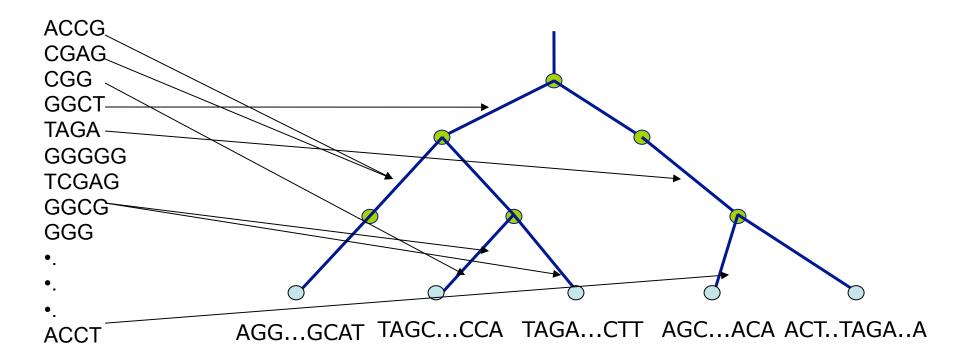
Part III: SEPP

- 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 singlegene datasets with fragmentary sequences

Phylogenetic Placement

Fragmentary sequences from some gene

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



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

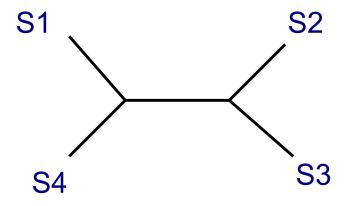
```
S1 = -AGGCTATCACCTGACCTCCA-AA

S2 = TAG-CTATCAC--GACCGC--GCA

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

S4 = TAC---TCAC--GACCGACAGCT

Q1 = TAAAAC
```



Align Sequence

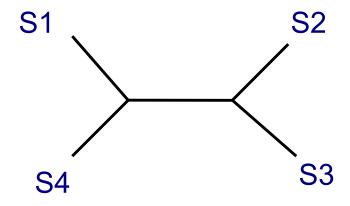
```
S1 = -AGGCTATCACCTGACCTCCA-AA

S2 = TAG-CTATCAC--GACCGC--GCA

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

S4 = TAC---TCAC--GACCGACAGCT

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



Place Sequence

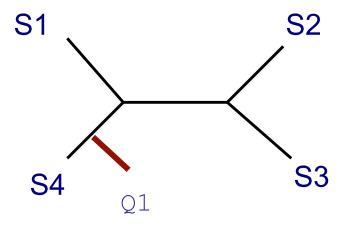
```
S1 = -AGGCTATCACCTGACCTCCA-AA

S2 = TAG-CTATCAC--GACCGC--GCA

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

S4 = TAC---TCAC--GACCGACAGCT

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

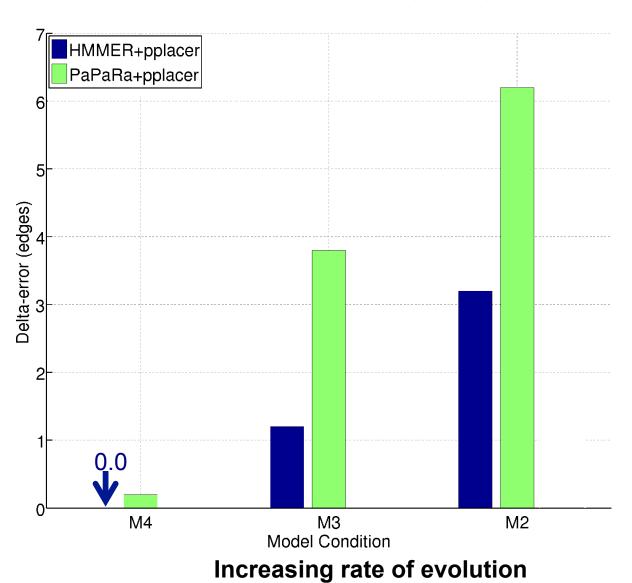


Phylogenetic Placement

- 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)

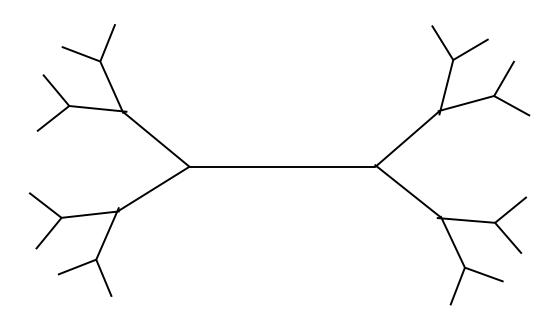
Note: pplacer and EPA use maximum likelihood, and are reported to have the same accuracy.

HMMER vs. PaPaRa

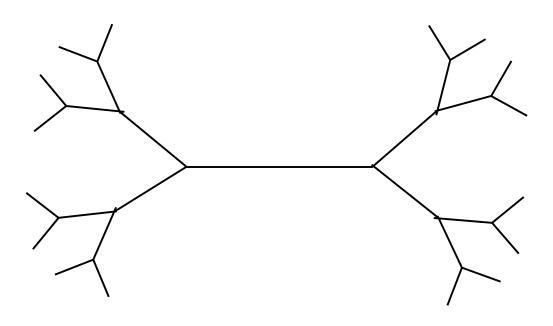


HMMER+pplacer:

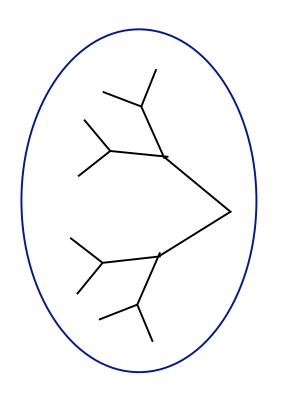
- 1) build one HMM for the entire alignment
- 2) Align fragment to the HMM, and insert into alignment
- 3) Insert fragment into tree to optimize likelihood

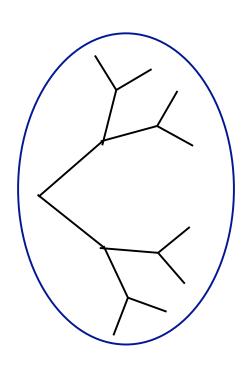


One Hidden Markov Model for the entire alignment?

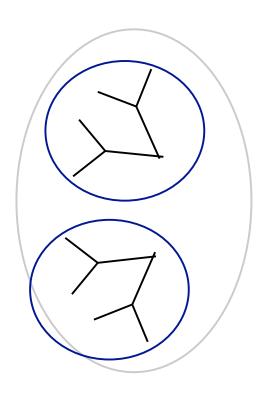


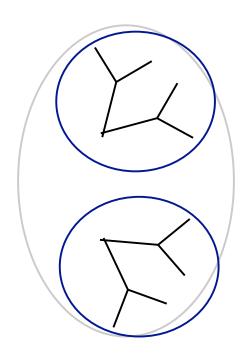
Or 2 HMMs?



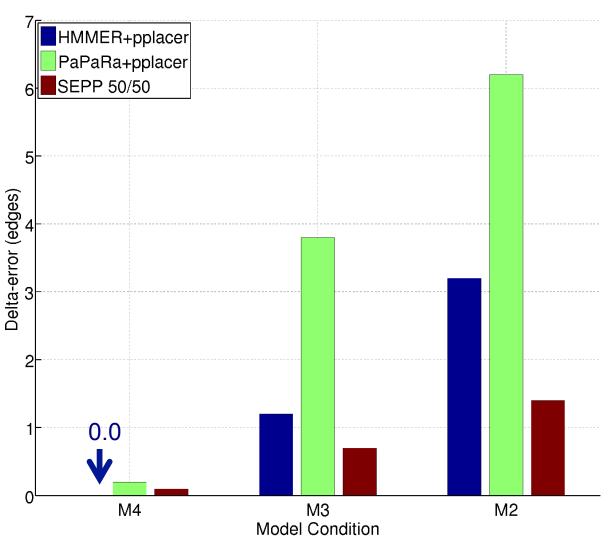


Or 4 HMMs?



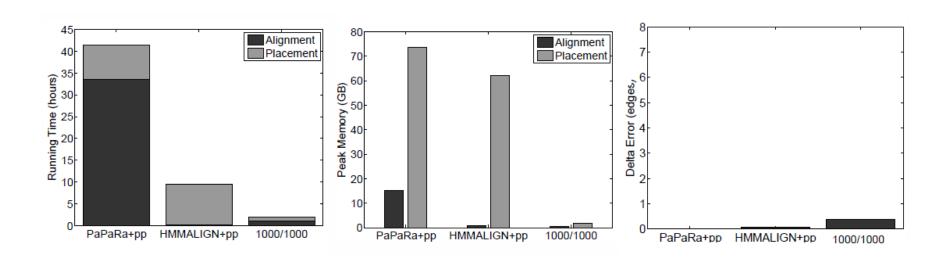


SEPP(10%), based on ~10 HMMs



Increasing rate of evolution

SEPP (10%) on Biological Data



16S.B.ALL dataset, 13k curated backbone tree, 13k total fragments For 1 million fragments:

PaPaRa+pplacer: ~133 days

HMMALIGN+pplacer: ~30 days

SEPP 1000/1000: ~6 days

Applications of SEPP (unpublished)

UPP: Ultra-large alignment using SEPP

 TIPP: taxon identification of fragmentary data (for metagenomic analysis)

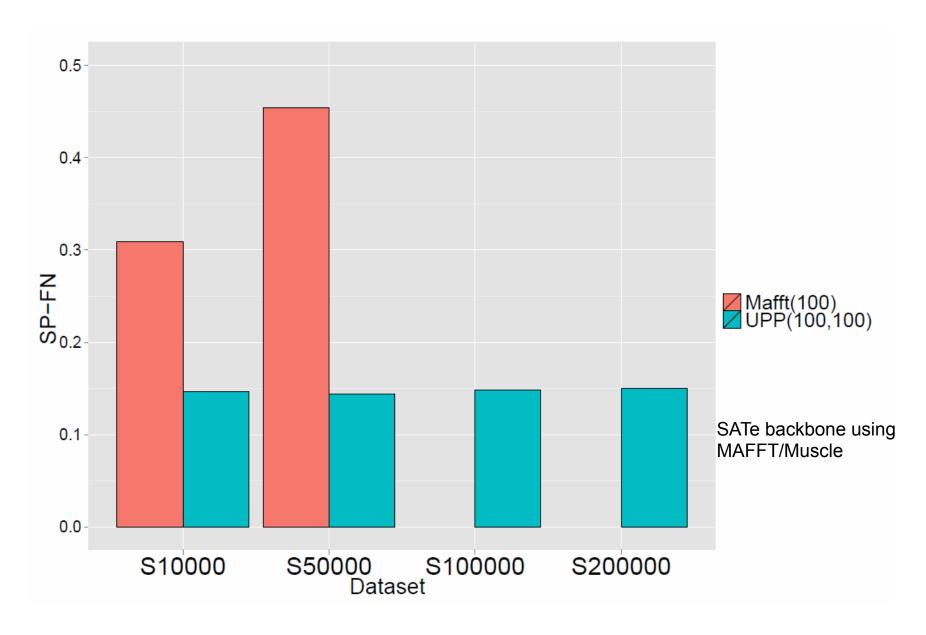
Part IV: UPP: Ultra-large alignment using SEPP

Input: set S of unaligned sequences

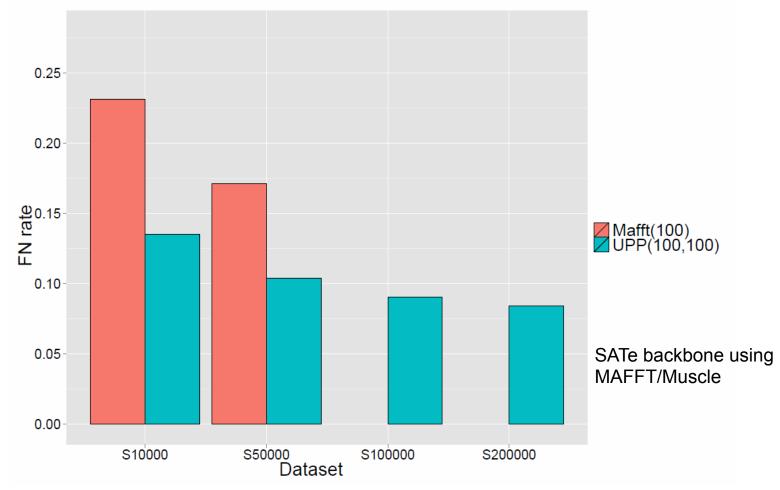
Output: alignment and tree on S

- Select random subset X of sequences
- Estimate alignment and tree on X
- Run SEPP to align remaining sequences
- Run favorite tree estimation method on alignment
- UPP(x,y) refers to UPP using backbones of size y and alignment subsets of size x

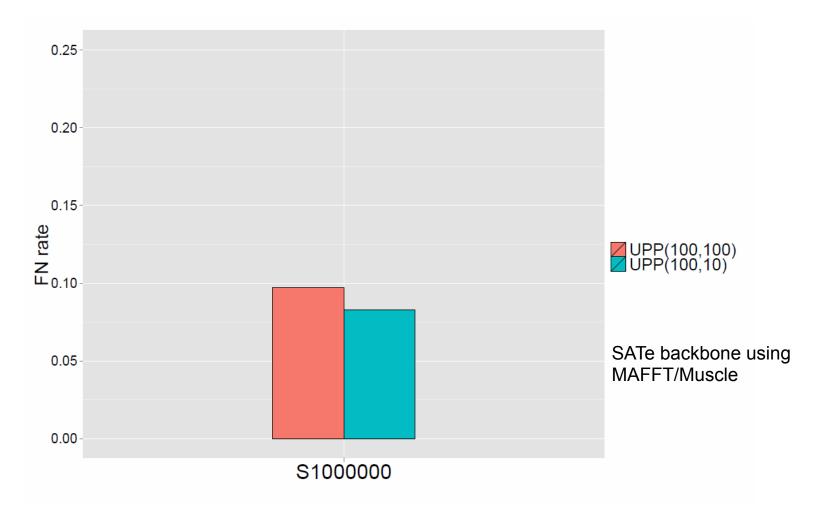
RNASim: SP-FN Score



UPP vs. MAFFT Tree error on 10K-200K sequences



UPP(100,100) vs. UPP(100,10) One Million Taxa: Tree Error



Note improvement obtained by using SEPP decomposition

Four "Boosters"

SATé: co-estimation of alignments and trees

DACTAL: tree estimation (almost) without alignments

SEPP: phylogenetic placement of short reads

UPP: ultra-large multiple sequence alignment

Phylogenetic "boosters" (meta-methods)

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

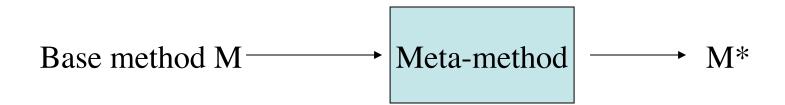
Techniques: divide-and-conquer, iteration, and "bin-and-conquer"

Examples:

- DCM-boosting for distance-based methods (1999)
- DCM-boosting for heuristics for NP-hard problems (1999)
- SATé-boosting for alignment methods (2009)
- SuperFine-boosting for supertree methods (2011)
- DACTAL-boosting: almost alignment-free phylogeny estimation methods (2011)
- SEPP-boosting for phylogenetic placement of short sequences (2012)
- UPP-boosting for alignment methods (unpublished)
- TIPP-boosting for metagenomic taxon identification (unpublished)
- Binning to improve coalescent-based species tree estimation methods (2013)

Meta-Methods

 Meta-methods "boost" the performance of base methods (phylogeny reconstruction, alignment estimation, etc).



Algorithmic Strategies

- Divide-and-conquer
- Iteration
- Multiple HMMs instead of one (for classification problems)
- Bin-and-conquer

Acknowledgments

- Guggenheim Foundation Fellowship
- Packard Fellowship
- NSF: ATOL, ITR, and IGERT grants
- David Bruton Jr. Professorship
- HHMI
- Microsoft Research
- Texas Advanced Computing Center (TACC)

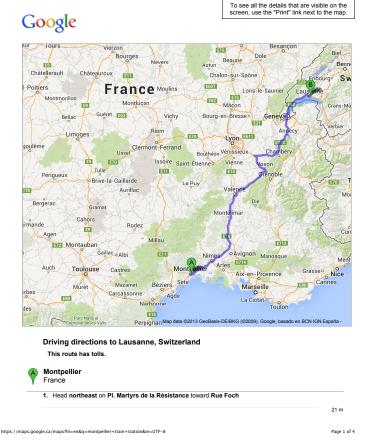
Collaborators:

- SATé: Kevin Liu, Serita Nelesen, Sindhu Raghavan, and Randy Linder (and Mark Holder's lab at Kansas for public distribution)
- DACTAL: Serita Nelesen, Kevin Liu, and Randy Linder
- SEPP and UPP: Siavash Mirarab and Nam Nguyen

2008: David and I deal with a train strike in France











Montpellier, France to Lausanne, Switzerland

Thank you David!

