

Evolution of Proteins 2: Proteins 7350

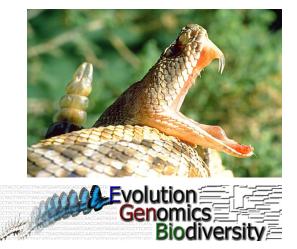
Pollock_ProteinEvol6.ppt

Slides with unpublished data are deleted



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Overview

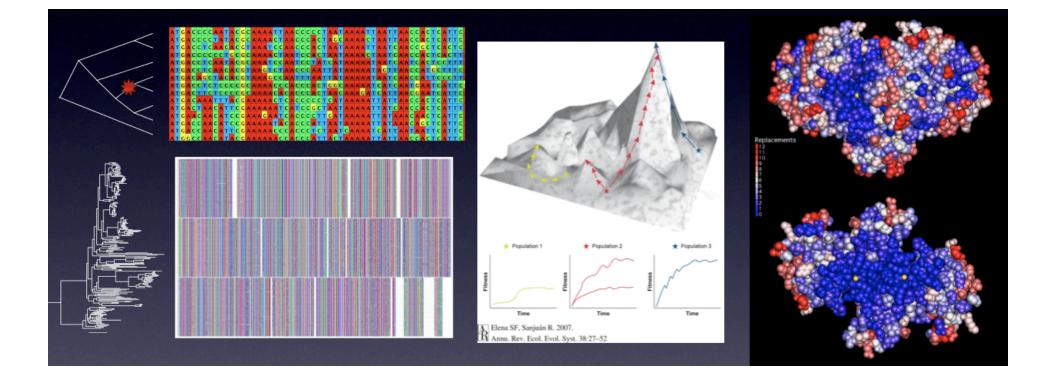
Explanation of what we do

Mitochondrial genomics: a pilot project

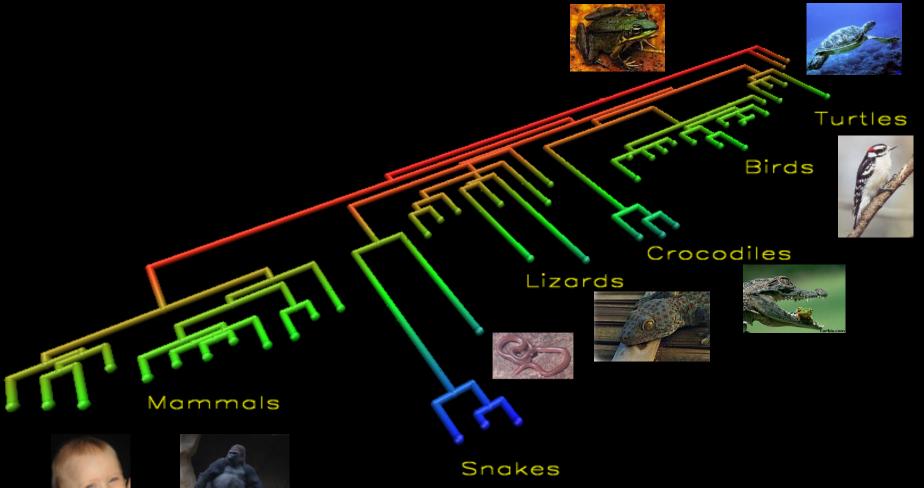
PLEX: Context-dependent evolutionary genomics in a practical time frame

Coevolution of transcription factors and their binding sites: an example

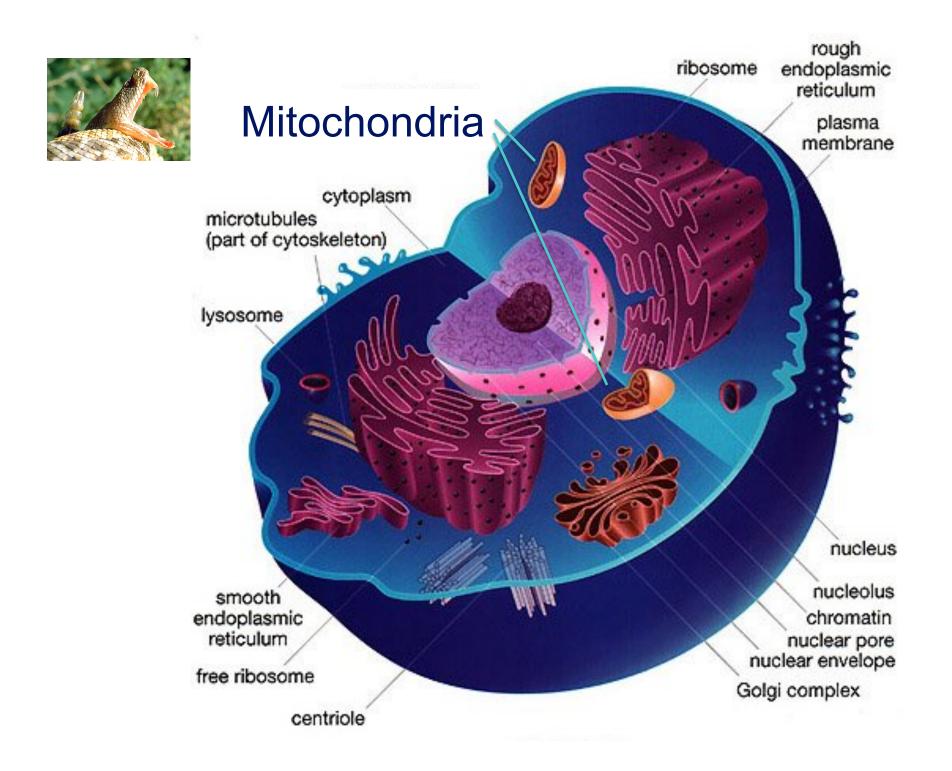
Evolution of Proteins



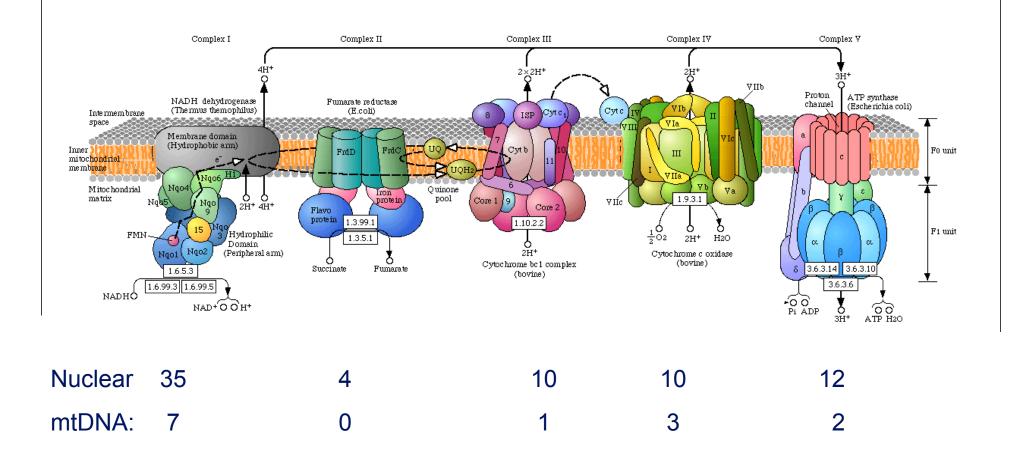
Jason de Koning







Oxidative Phosphorylation



Molecular Evolution Structure, Function & Rates

- Conserved sites correspond to structurally or functionally important residues
- Changes in evolutionary rates correspond to:
 - Loss of function
 - Altered function (functional divergence)
- Synonymous rates (dS) are compared to nonsynonymous rates (dN) as a "neutral" standard.

$$\mu N * \frac{1}{N} = \mu$$

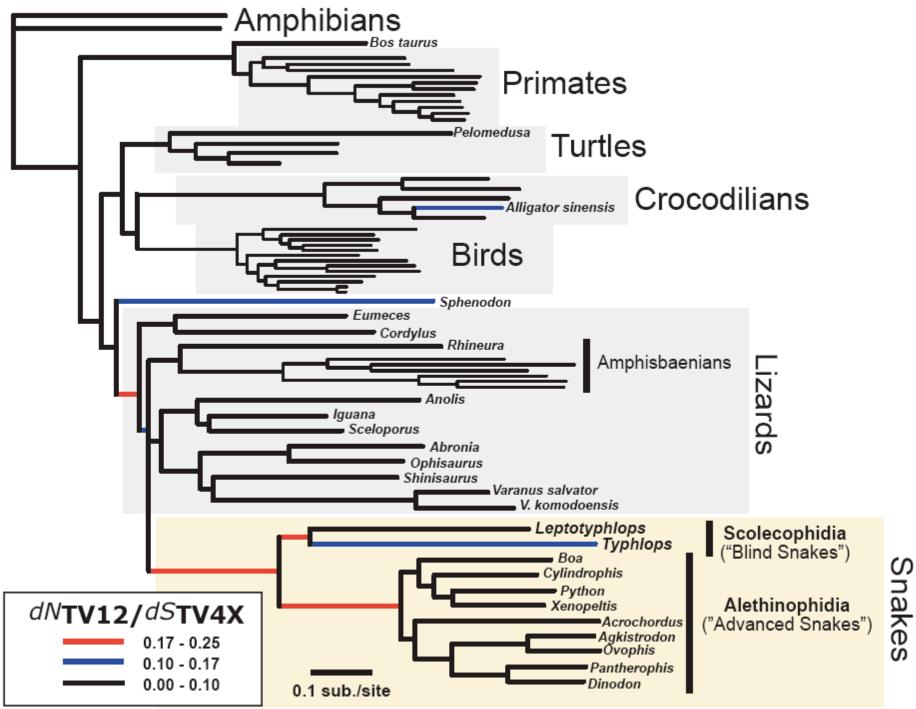
- Convergence is a sign of adaptive importance, and is rare at the molecular level
- Coevolution is usually distributed among many sites, often weak

Positive Selection in Snake Mitochondria

- Standard programs strongly indicate positive selection in proteins throughout the mitochondria
 - Especially cytochrome oxidase subunit I and cytochrome b (the hearts of complex IV and III)

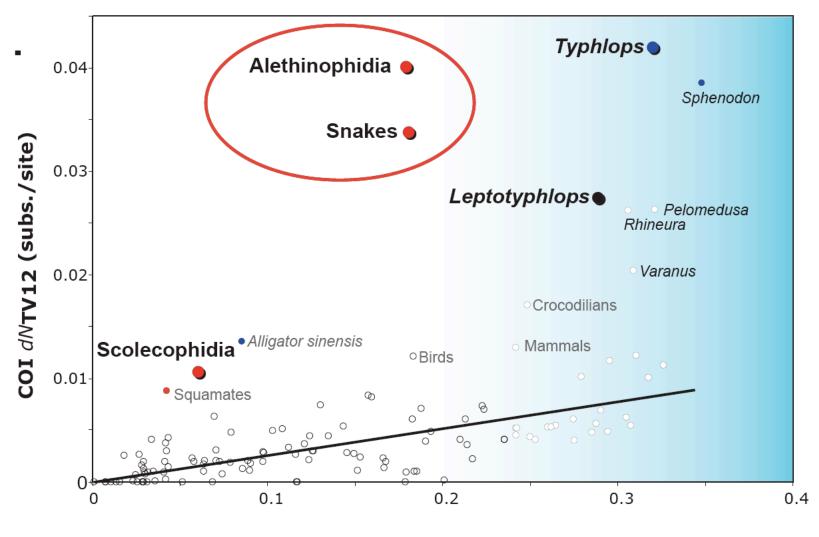
Concern over "saturation", inaccurate models

Focus on transversions



Squamates

Excess Replacements in COI

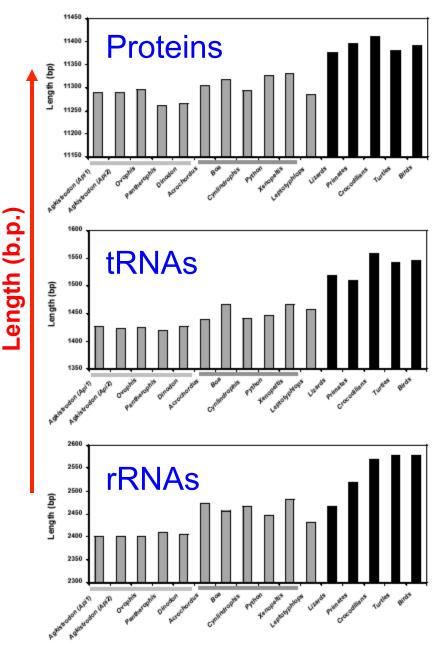


 dS_{TV4X} (substitutions per site)

There is very little in tetrapod mtDNA that is not functional

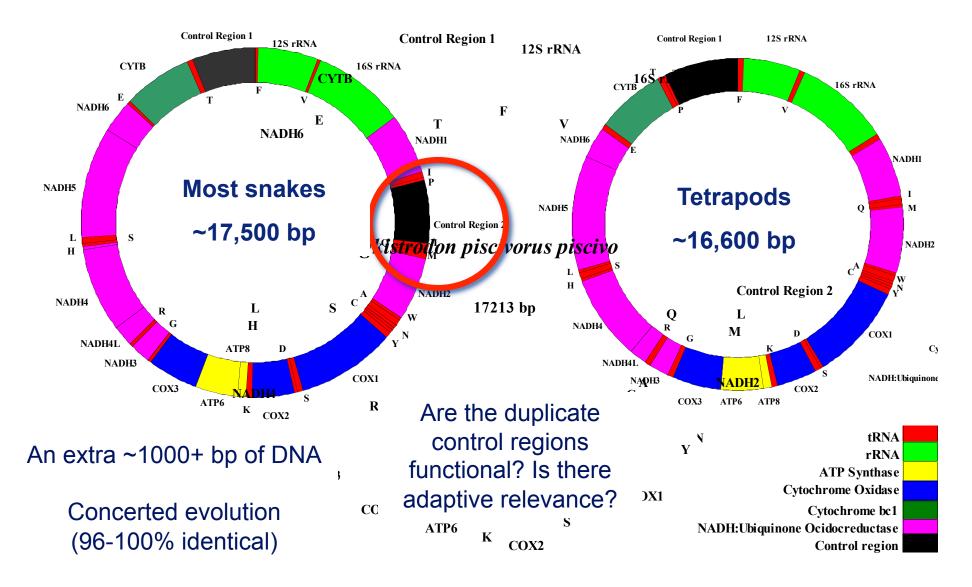
- Implies selection to remove junk
- Most snake mtDNA genes are short
 - Implies even stronger selection to reduce excess nucleotide length

SNAKES / OTHER TETRAPODS

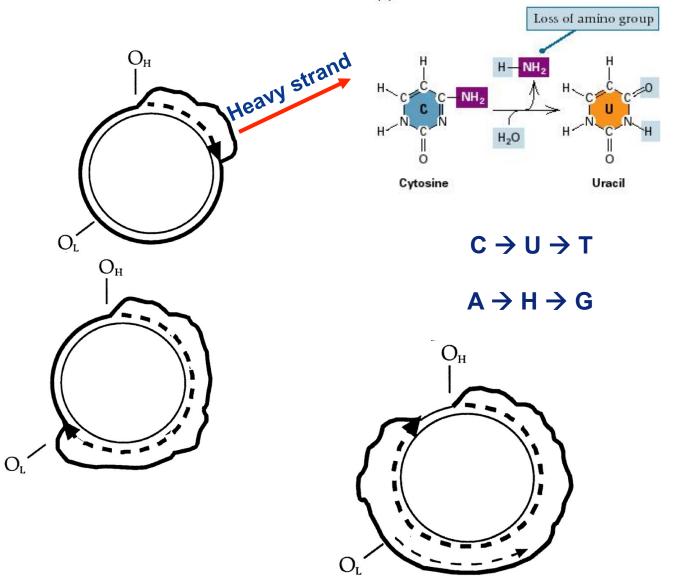


Duplicate Control Regions (CR) in Most Snake mtDNA

Origin of genome replication and bidirectional transcription initiation

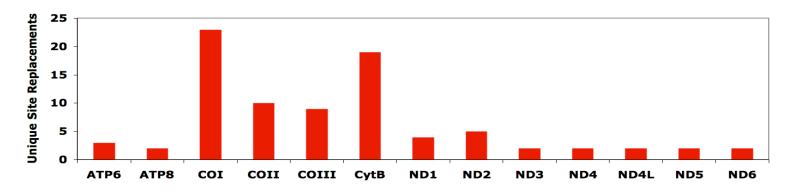


Typical Mitochondrial Genome Replication (single control region)

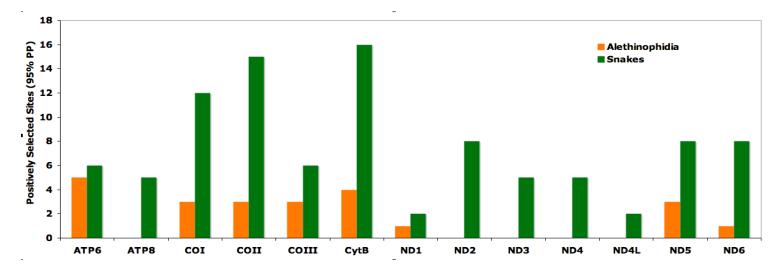


Accelerated Evolution Early in Snake Evolution at Conserved Sites and Genome-wide

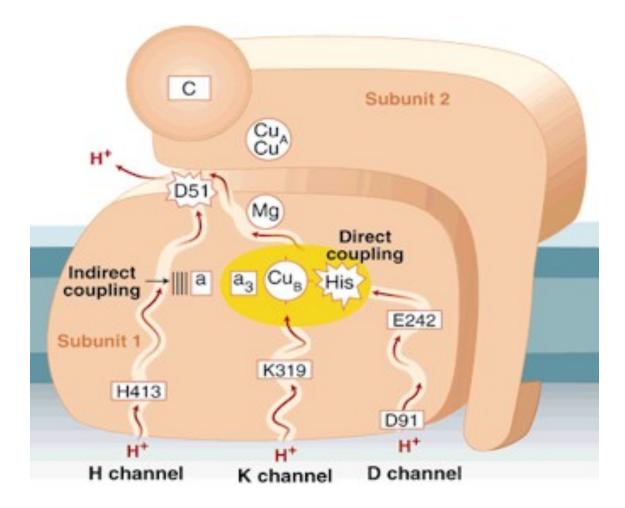
Replacements at sites otherwise conserved across tetrapods



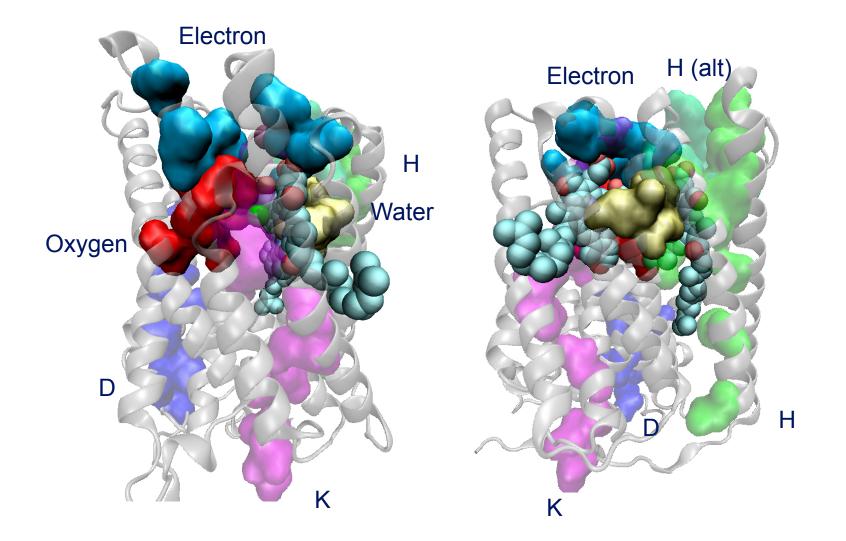
Positively selected sites along Alethinophidian and Ancestral Snake branches



Predicted Function of Channels



COI Functional Regions



Unique Sites

Altered in snakes Otherwise conserved across most tetrapods Focus on sites most likely to be functionally relevant (limit numbers) Associated with coevolving site pairs Coevolution in snake mtDNA is very high 22% of site pairs at p<0.01

Unique Residue Clusters

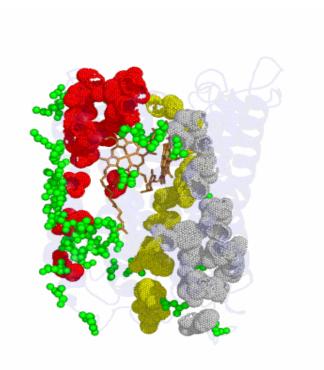
| Cluster Number | Residues | Ca Distance | Location |
|----------------|------------------------|-----------------|-------------------------|
| 1 | 35L – 37I – 54Y | 5.0 Å*, 10.6 Å* | H Channel |
| 2 | 443Y - 447Y | 6.7 Å* | H Channel |
| 3 | 256A – 258V | 5.6 Å* | K Channel |
| 4 | 266E – 267P | 3.8 Å* | K Channel |
| 5 | 26A – 108S | 11.9 Å | D Channel |
| 6 | 205G – 231Y | 6.3 Å* | O ₂ Delivery |
| 7 | 299V – 301T | 5.5 Å* | O ₂ Delivery |
| 8 | 194L – 281G | 6.9 Å* | O ₂ Delivery |

Coevolution is Usually Distributed

Not usually strong pairwise

Oftentimes adjacent to function or binding

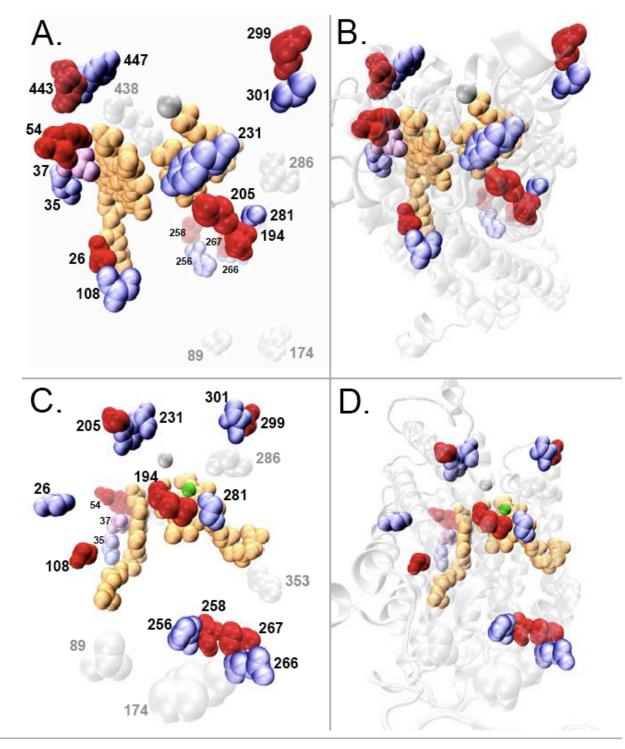
> Pollock and Wang, 2005, 2007
> Yeang and Haussler, 2007



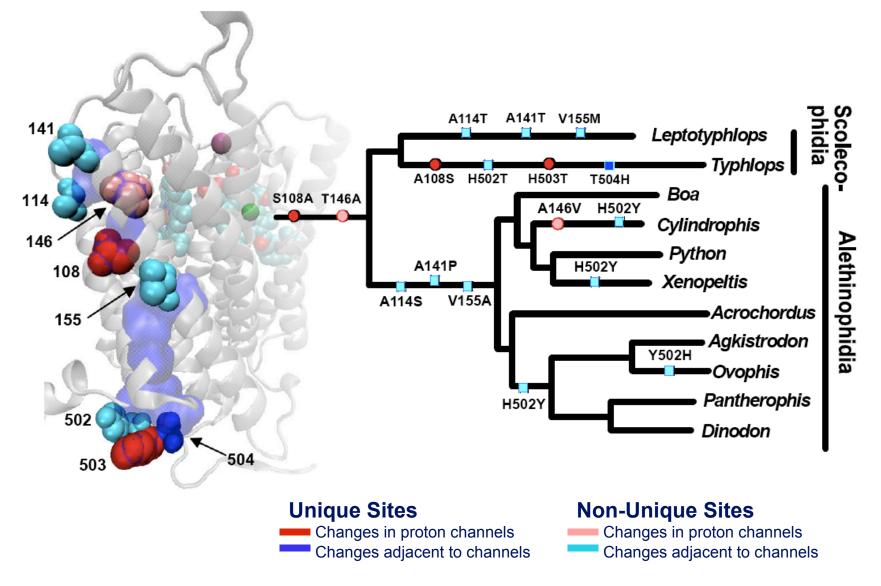


Coevolution in COI

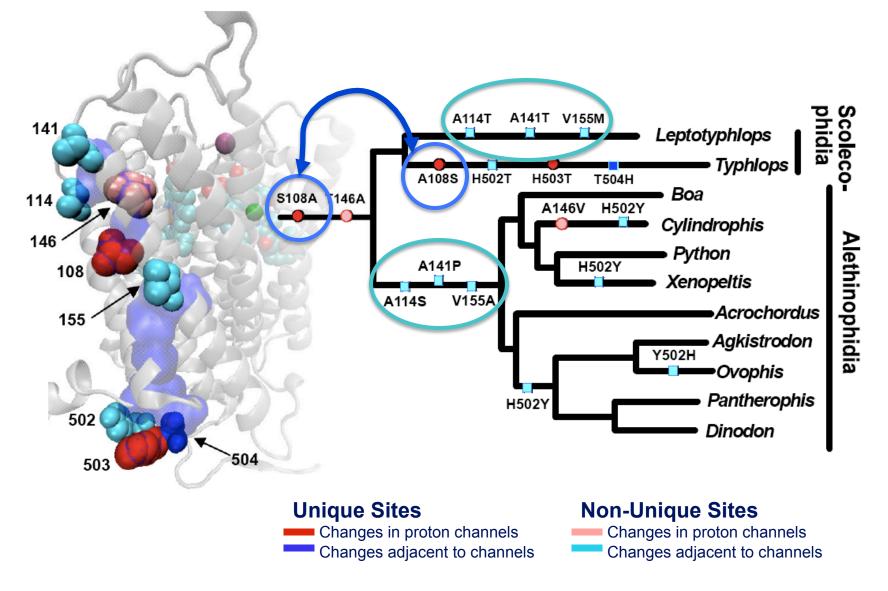
Physically paired unique substitutions



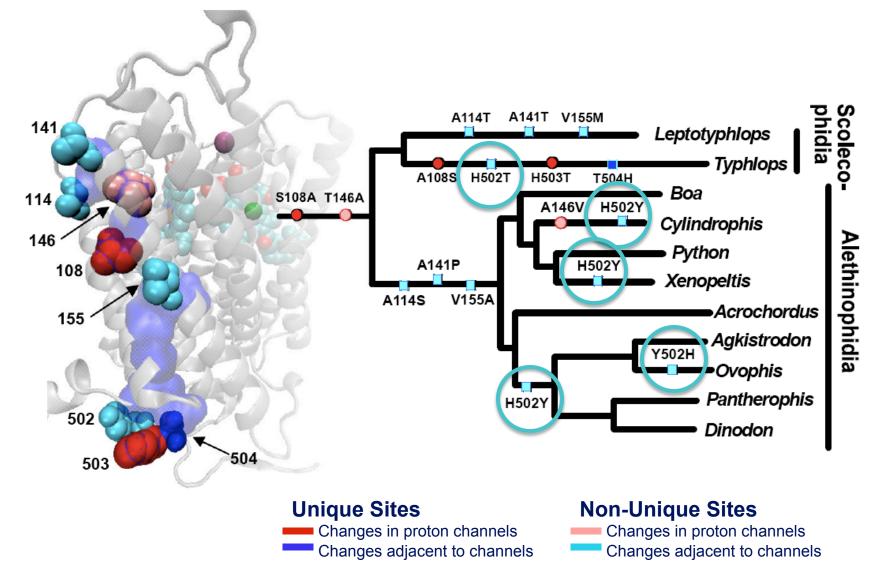
Channel D (Direct Coupling) Loss of Polarity, then Recovery?



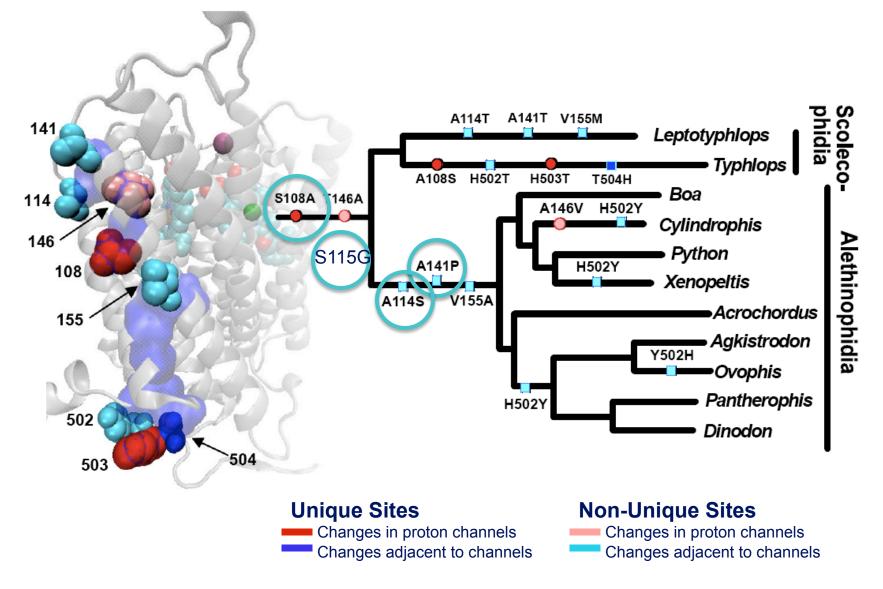
Convergence and Reversion



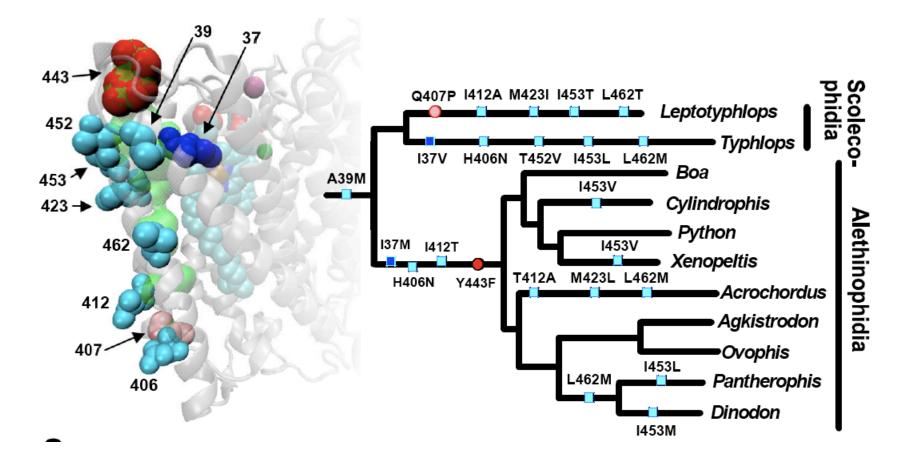
Channel D (Direct Coupling) Repeated Convergence (and a reversion)



Channel D in *Rhineura* a distantly-related legless tubular squamate



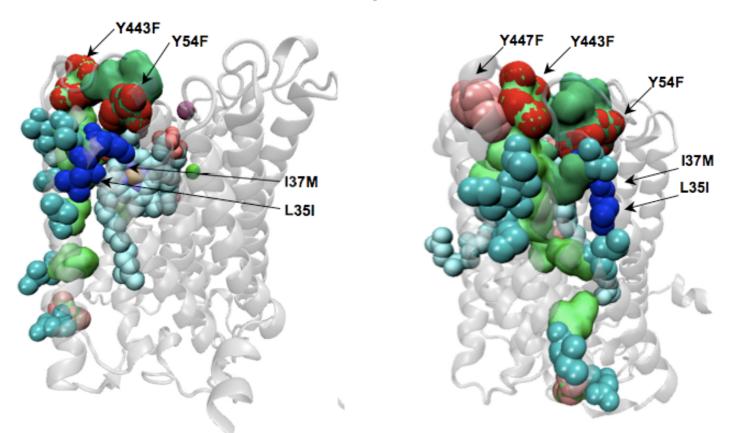
Channel H: Exclusive Pumping, Indirect Coupling Controversial function is shut down



Controversy over channel H

- Is it really used?

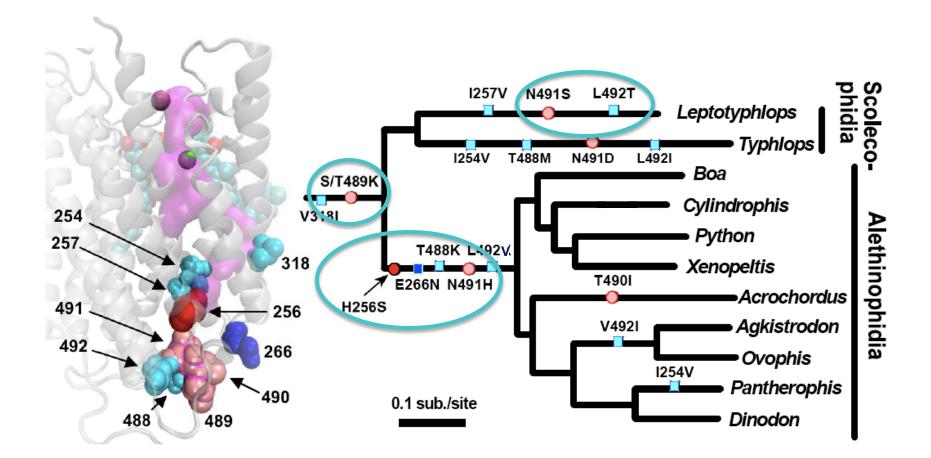
- Which of two different paths is it?



Snakes go out of their way to completely destroy all possible outlets of the alternative Channel H; tyrosine (Y) to Phenylalanine (F) substitutions are usually quite rare

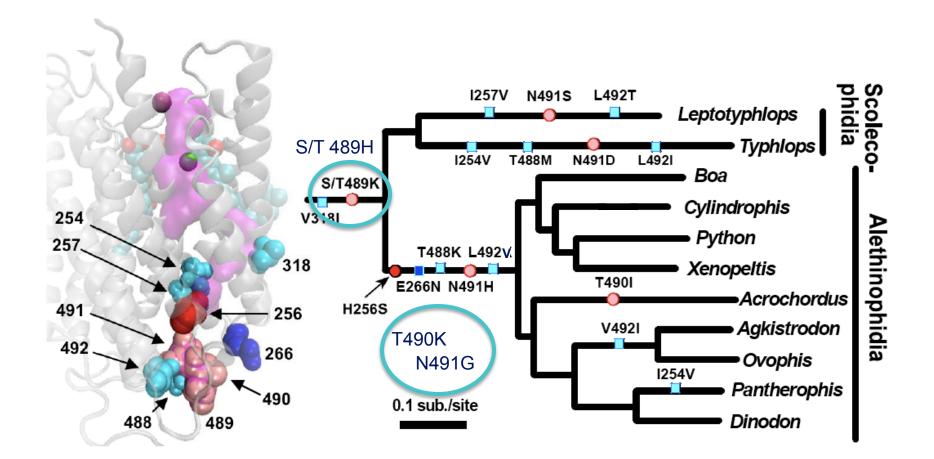
Channel K: Proton Delivery to Reaction Center

Not shut down; increase in positive charge at entrance

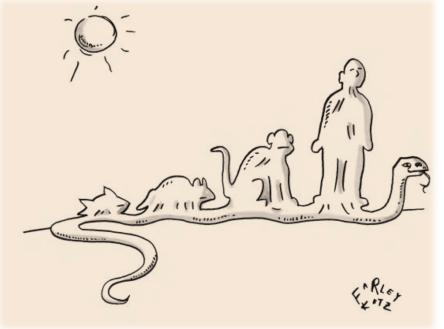


Channel K in Rhineura

Not shut down; increase in positive charge at entrance



Aerobic Metabolism
 One of the lowest basal metabolic rates
 Highest fluctuation between basal and max
 Fluctuations of 40-fold in 48 hours





Aerobic Metabolism
 Physiological Remodeling to Digest Prey
 Heart muscle - may enlarge 50%
 Liver – may enlarge 100%
 Gut - may enlarge 100 - 150%

progress

A vertebrate model of extreme physiological regulation

Stephen M. Secor & Jared Diamond

Secor & Diamond.

Nature, 1998

Department of Physiology, University of California Medical School, Los Angeles, California 90095-1751, USA

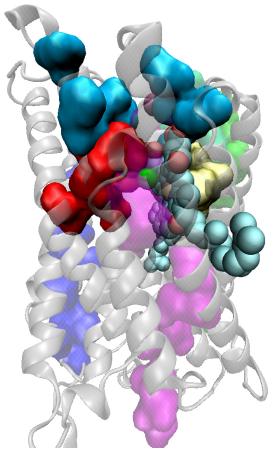
Investigation of vertebrate regulatory biology is restricted by the modest response amplitudes in mammalian model species that derive from a lifestyle of frequent small meals. By contrast, ambush-hunting snakes eat huge meals after long intervals. In juvenile pythons during feeding, there are large and rapid increases in metabolism and secretion, in the activation of enzymes and transporter proteins, and in tissue growth. These responses enable an economic hypothesis concerning the evolution of regulation to be tested. Combined with other experimental advantages, these features recommend juvenile pythons as the equivalent of a squid axon in vertebrate regulatory biology.



Aerobic Metabolism
 Physiological Remodeling
 Venom
 Diverse arsenal of deadly venom proteins
 Widespread adaptive evolution of venom proteins



Massive Multi-Protein Adaptation



Most extreme adaptation known in metabolic proteins
 Molecular coevolution – best example known
 Molecular convergence
 Chift in mitochoodrial function

Shift in mitochondrial function ✓Increase proton flow to the reaction center?

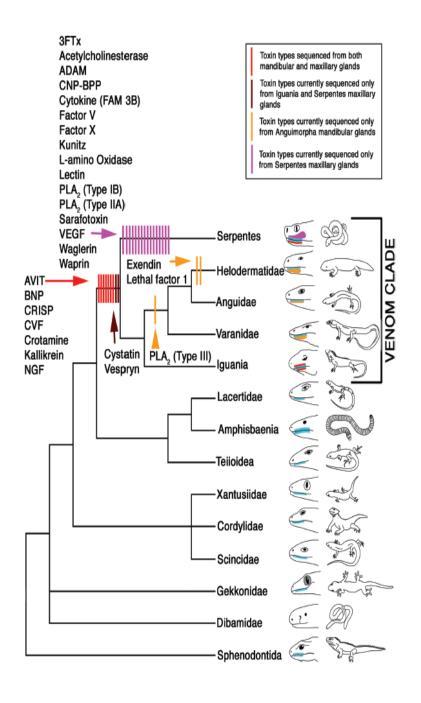
Likely important for metabolic fluctuations in snakes

✓Microevolutinary event \rightarrow macroevolutionary adaptation

Aerobic Metabolism
 Physiological Remodeling
 Venom
 Evolutionary History

 => Fossorial and inactive
 => Terrestrial and capable of switching from inactive to very active





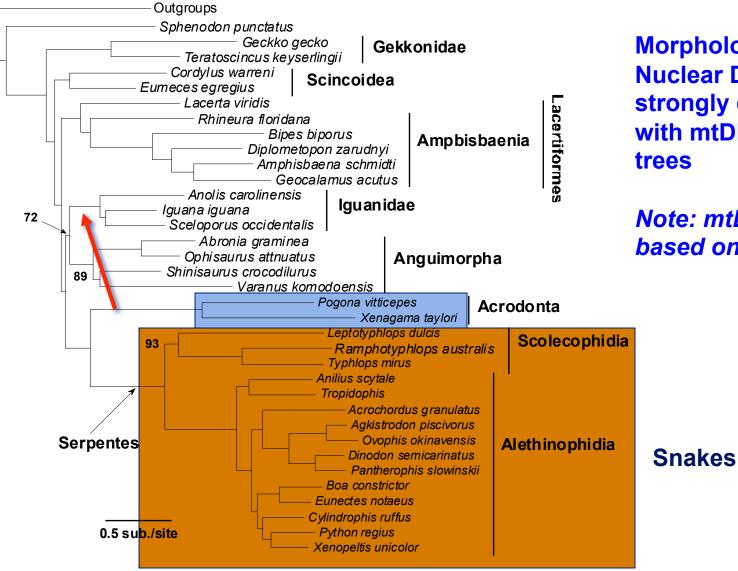
Did Venom Play a Role?

Snake venom genes were present (and expressed in salivary glands) in lizards *PRIOR* to snake evolution

A broad arsenal of amazingly toxic proteins evolved only in some snakes

Venom is also one of the main known causes of accelerated or diversifying evolution

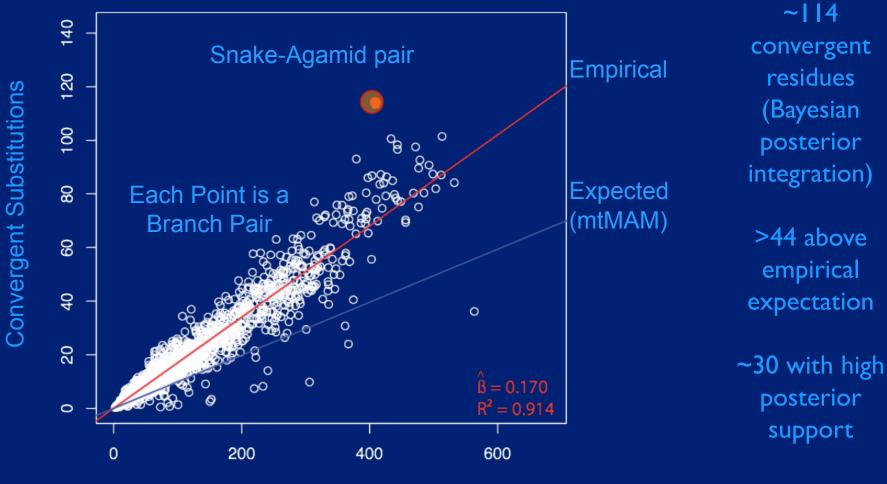
Snake / Lizard Phylogeny



Morphological and Nuclear Data strongly disagree with mtDNA-based trees

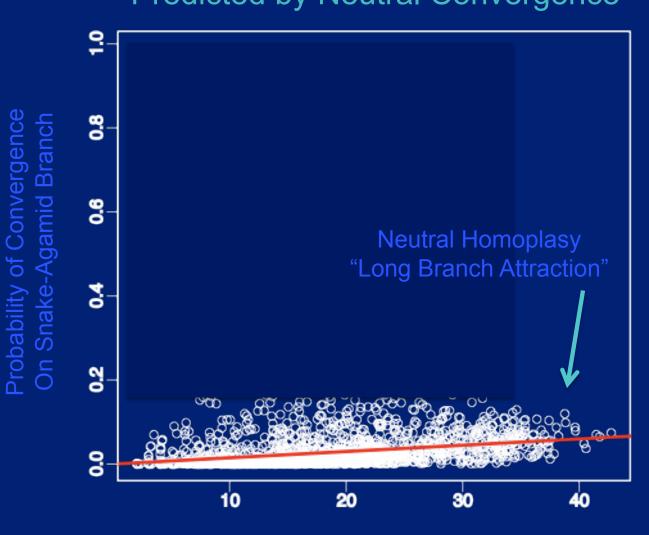
Note: mtDNA tree is based on 12,000 bp!

Excess Convergence in Mitochondrial Proteins



Divergent Substitutions

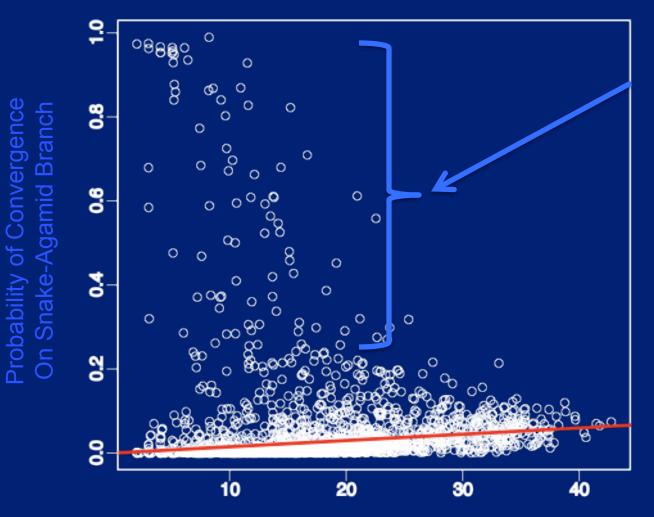
Fast Sites Converge a Little Bit Predicted by Neutral Convergence



Posterior Number of Substitutions per Site

Converged Sites Evolved Slowly

Consistent with Adaptive Convergence

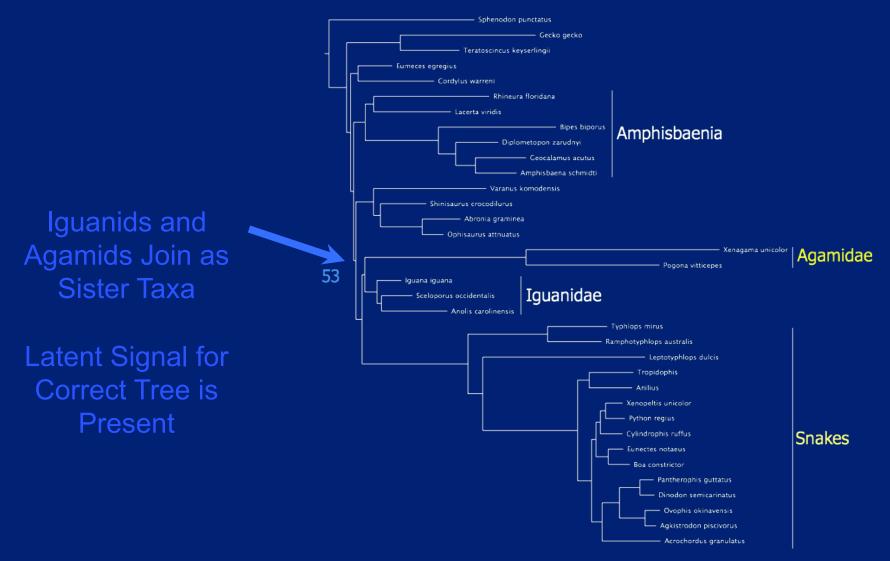


Convergence at Highly Conserved Sites

Posterior Number of Substitutions per Site

Screening Convergent Sites Restores Nuclear Tree

top 5% of convergent sites were screened



Ruggedness, Dimensionality, and Changing Landscapes

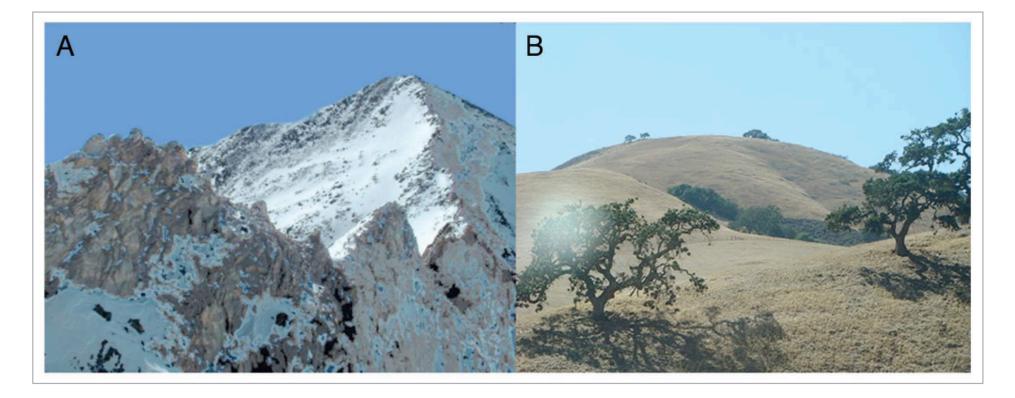


Figure 1. Alternative views of potential protein adaptive landscapes. In (A), the protein adaptive landscape is viewed as being like an arrêt ridge, with only a single narrow path leading from the current adaptive peak in the foreground to a new adaptive peak in the distance. This landscape is conducive to convergence. In (B), the adaptive landscape is viewed as being like rolling hills, with many alternative routes to nearby adaptive hilltops that are not substantially different from one another. With so many alternative paths and alternative similar hilltops, under this scenario sequences would be unlikely to converge (i.e., follow the same path) even under similar adaptive pressure.

PLEX: Context-dependent evolutionary genomics in a practical time frame

Large phylogenomic datasets now common
 Parametric inference with realistic evolutionary models is (was) computationally burdensome

- MCMC + data augmentation of ancestral states and substitution histories can be extremely fast
 - Augmentation step is (was) a major performance bottleneck (>99% of computation)
- Order of magnitude speed improvements and excellent scaling can be achieved
 - partially sampling substitution histories

Mutation-selection models of coding sequence evolution with site-heterogeneous amino acid fitness profiles

Nicolas Rodrigue^{a,1}, Hervé Philippe^b, and Nicolas Lartillot^b

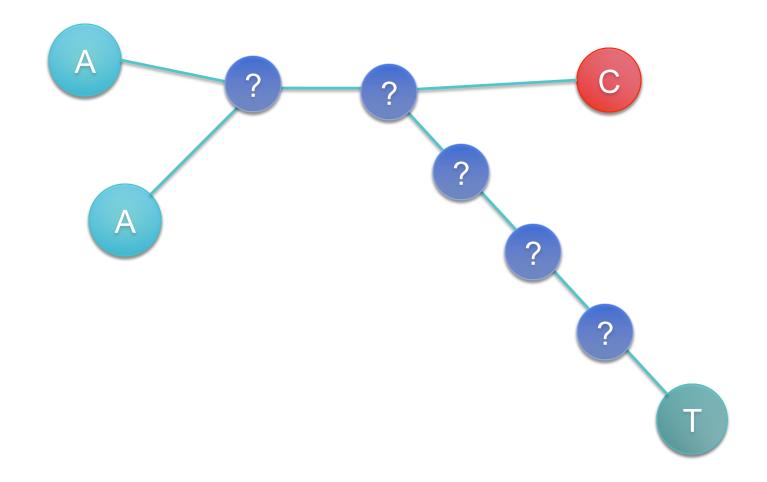
^aDepartment of Biology, University of Ottawa, Ottawa, Ontario, K1N 6N5 Canada; and ^bDepartment of Biochemistry, Centre Robert Cedergren, Université de Montréal, Montréal, Québec, H3C 3J7 Canada

Edited by David M. Hillis, University of Texas at Austin, Austin, TX, and approved January 27, 2010 (received for review September 24, 2009)

in some cases we study here. Although the empirical mixture approaches can provide less taxing models, the Bayes factors reported above (computed using pruning-based sampling) still required over 2 months of CPU time. It is thus of interest to advance further computational methods, both to ameliorate our current data-augmentation-based sampler and to bridge this type of MCMC sampling with our thermodynamic integration methods.



Time Complexity of Integrated Likelihood Calculations on a Phylogeny



Time Complexity of Integrated Likelihood Calculations on a Phylogeny

(N states, b branches between nodes, s sites)

$$O(N^4 + N^3b + N^2bs)$$

Substitution Histories Ancestral States

Calculation gets overwhelming with increased complexity

Spatial Variation (many rate matrices) Gradient Mixture Models Context dependence

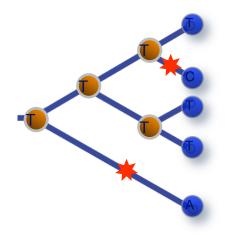
Temporal Variation Markov-modulated codon models Switching selection regimes

$$O(N^4s + N^3bs + N^2bs)$$

$$O(N^4 + N^3b + N^2bs)$$

N is very large (e.g., 183 x 183)

Time Complexity Using Data Augmentation



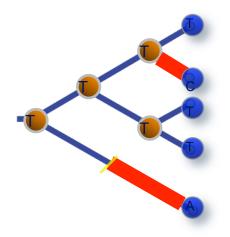
Complete sampling in continuous time (Nielsen, Rodrigue, Lartillot)

"Don't need to use fully integrated likelihood calculations"

 $O(N^2)$

Likelihood:

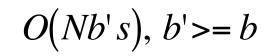
 $O(N^4 + N^3b + N^2bs + Nbs) + more$



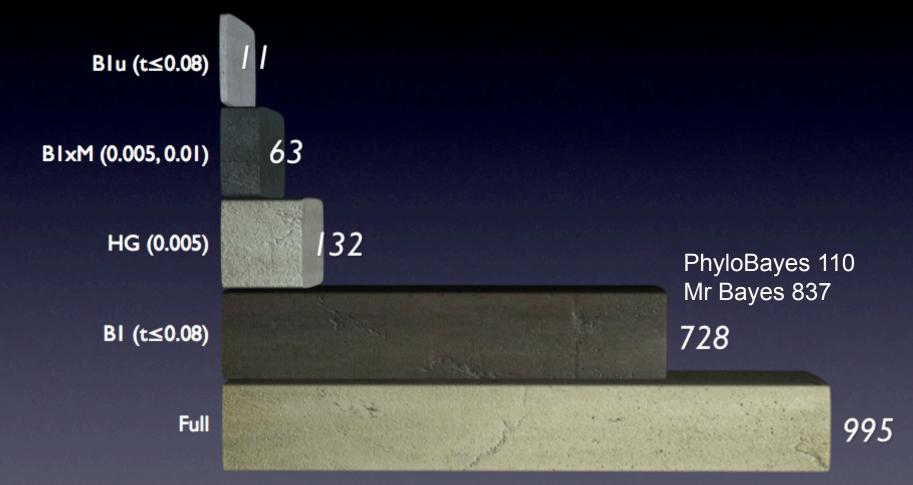
Partial sampling in continuous time (de Koning et al. 2010)

"Don't need to fully sample the timing of substitution either"

 $O(N^2)$



Likelihood Analysis at the Speed of Parsimony



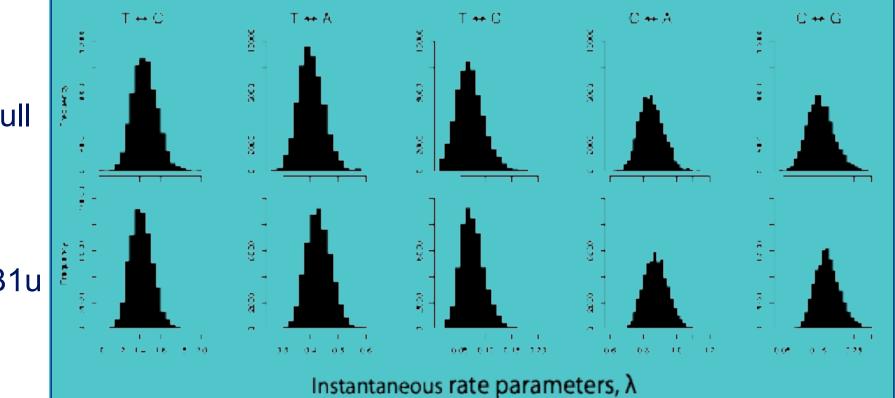
Time to analyse 224 taxon dataset, GTR model(100,000 generations of MCMC)

Dramatically Improved Scaling

| Model | Standard Likelihood | Blu Speedup |
|------------|------------------------|----------------|
| DNA | 1000 sec | II sec (I00x) |
| Amino Acid | 7.5 hrs | 17 sec (1600x) |
| Codon | 2 months | 12 min (7000x) |

Much better than using "exotic" computation strategies: GPU speedup is only ~100x for codon models (Suchard & Rambaut 2009)

Posterior Parameter Distributions GTR, mammalian *cyt-b*



B1u (t≤0.08) *B1u* (t≤0.02) Full -7,586.47 -7,586.90 10 taxa 224 taxa -99,391.12 -99,542.03 -99,446.88

Full

B1u

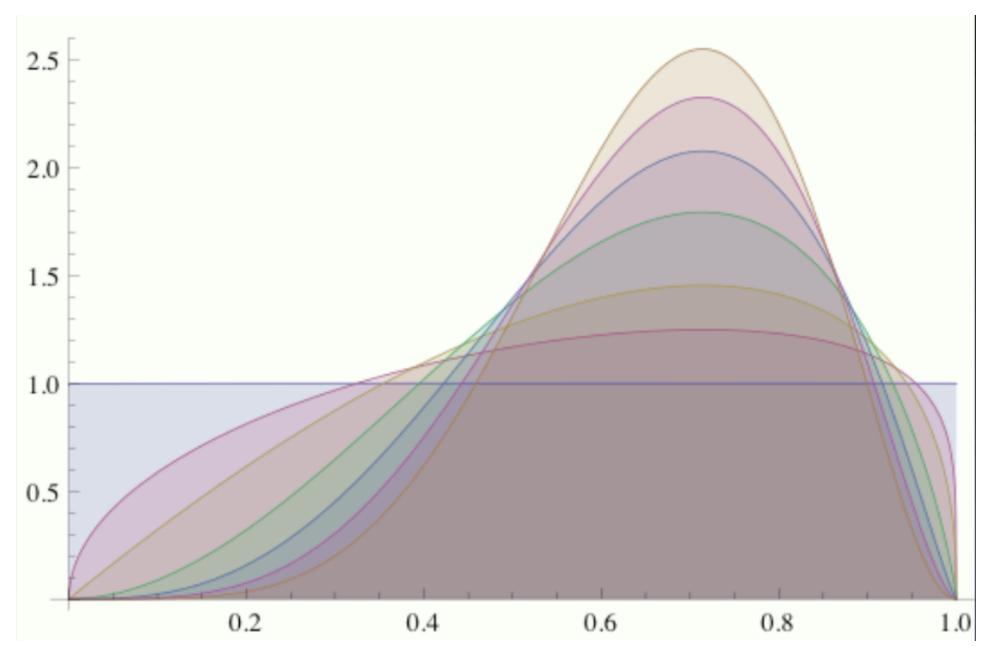
Evaluating Changes in Rate and Branch Length

Classic likelihood calculations $O(N^4 + N^3b + N^2b)$

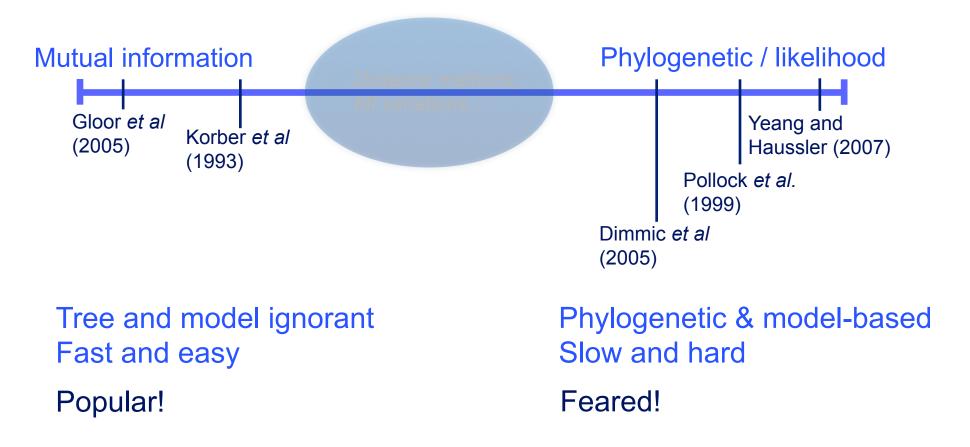
Sampled substitution histories O(1)

Branch lengths and rate parameters can be evaluated separately and have analytically solvable posterior distributions

Thermodynamic Integration



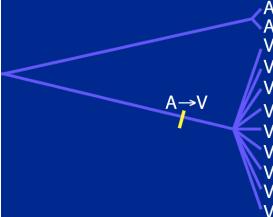
Pairwise Coevolution Approaches



$$MI = \sum_{i} \sum_{j} P(i,j) \log \frac{P(i,j)}{P(i)P(j)}$$

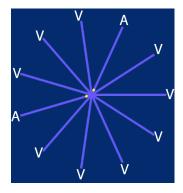
Mutual information methods are misled by:

(1) Phylogeny

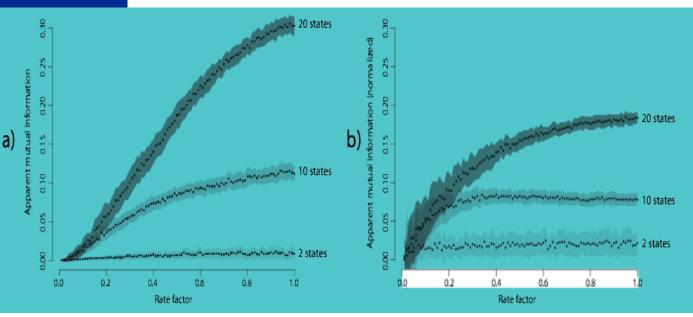


Observed Frequencies V: 9/11 = 82% A: 2/11 = 18%

Actual Time Spent V: 3.3/6.9=<u>48%</u> A: 3.6/6.9=<u>52%</u>



(2) Molecular Evolution



Phylogenetically-integrated MI (*pMI*)

$$\mathsf{P}(Z) = \int_{0}^{t_{b'}} \frac{P(x \to z|s) P(z \to y|(t_{b'} - s))}{P(x \to y|t_{b'})} ds^{\frac{1}{2}} \begin{cases} \frac{1}{\Lambda_x - \Lambda_y} + \frac{e^{\Lambda_y t_{b'}}t_{b'}}{e^{\Lambda_y t_{b'}} - e^{\Lambda_x t_{b'}}} & z = x\\ t_{b'} - \frac{1}{\Lambda_x - \Lambda_y} - \frac{e^{\Lambda_y t_{b'}}t_{b'}}{e^{\Lambda_y t_{b'}} - e^{\Lambda_x t_{b'}}} & z = y\\ t_{b'} & z = x = y\\ 0 & otherwise \end{cases}$$

$$\mathsf{P}(\mathsf{Z}_{i},\mathsf{Z}_{j}) = \int_{0}^{t_{b'}} \frac{P(x_{i} \to z_{i}|s) P(z_{i} \to y_{i}|(t_{b'} - s))}{P(x_{i} \to y_{i}|t_{b'})} \frac{P(x_{j} \to z_{j}|s) P(z_{j} \to y_{j}|(t_{b'} - s))}{P(x_{j} \to y_{j}|t_{b'})} ds$$

=24 cases

- Unrestricted non-reversible amino-acid substitution with gamma-distributed rate variation among sites
- Posterior-predictive null distribution for automated significance testing
- Fast! Roughly 6,200 pairs of sites per second (Yeang and Haussler, 2008: 29.35 seconds per site pair on a slower CPU):
 approximate corrected speed-up about 100,000X

Work in Progress

Context dependent nucleotide substitution
 Amino acid mixture models (dependent rates)
 Overlaid nucleotide and fitness models
 Gradient mixture models
 Whole molecule fitness
 Transcription factors and binding sites
 Protein stability and function

Summary

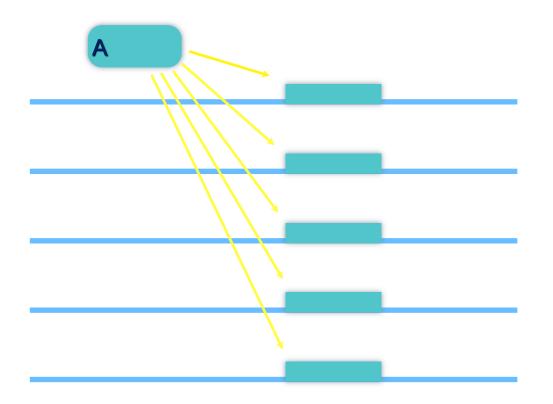
Partial sampling of substitution histories with B1u integration eliminates the most burdensome aspect of MCMC based phylogenomic analysis

Accuracy is high; precision can be tuned by decreasing the threshold of branch bisection

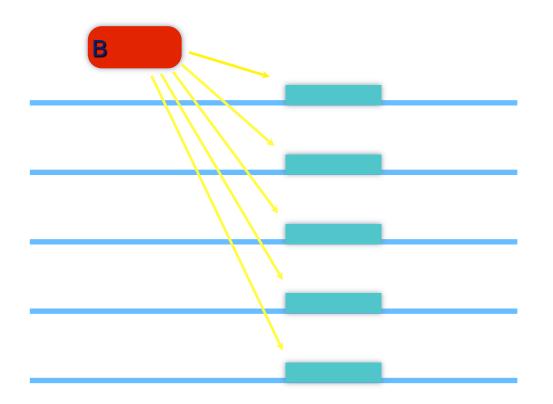
Should largely alleviate the pressure for convenience-motivated simplifications

Acknowledgements: Jason de Koning, Wanjun Gu, Todd Castoe; Richard Goldstein, Nicolas Rodrigue

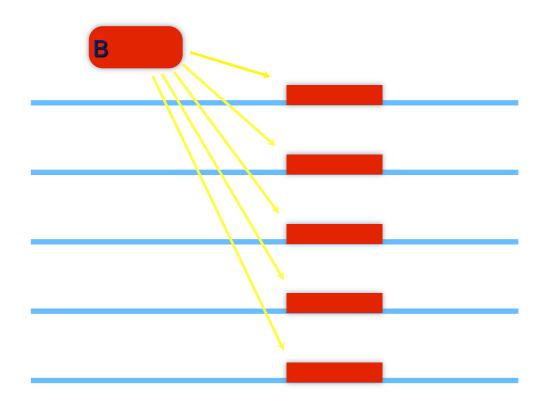
TF binding modifications

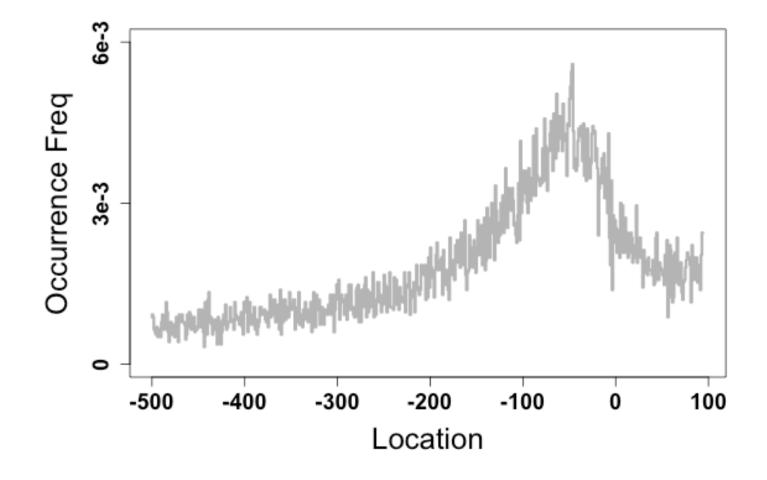


TF binding modifications



TF binding modifications





Adaptation, Coevolution and Convergence

Normal non-Adaptive Evolution

Adaptive evolution drives a different mode of coevolution and convergence