

# Large-scale selection analyses

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# Two decades of large-scale selection scans

#### Inferring Nonneutral Evolution from Human-Chimp-Mouse **Orthologous Gene Trios**

Andrew G. Clark, 1 Stephen Glanowski, 3 Rasmus Paul D. Thomas, Anish Keiariwal, Melissa A. David M. Tanenbaum, 5 Daniel Civello, 6 Fu Lu, 5 Bria Steve Ferriera, Gary Wang, Xianqgun Zhe Thomas J. White, John J. Sninsky, Mark D. A Michele Cargill6†

#### 2005 A Scan for Positively Selected Genes in the Genomes of Humans and Chimpanzees

Rasmus Nielsen 1,2+, Carlos Bustamante 1, Andrew G. Clark 3, Stephen Glanowski 4, Timothy B. Sackton 3, Melissa J. Hubisz<sup>1</sup>, Adi Fledel-Alon<sup>1</sup>, David M. Tanenbaum<sup>5</sup>, Daniel Civello<sup>6</sup>, Thomas J. White<sup>6</sup>, John J. Sninsky<sup>6</sup>, Mark D. Adams<sup>5</sup>, Michele Cargill<sup>6</sup>

2003, Science

PLOS GENETICS

2008

Open access, freely available online

Patterns of Positive Selection in Six Mammalian Carolin Kosiol<sup>1</sup>, Tomáš Vinař<sup>1</sup>, Rute R. da Fonseca<sup>2</sup>, Melissa J. Hubisz<sup>3</sup>, Carlos D. Bustamante<sup>1</sup>, Rasmus

Nielsen<sup>2</sup>, Adam Siepel<sup>1</sup>\*

Research article

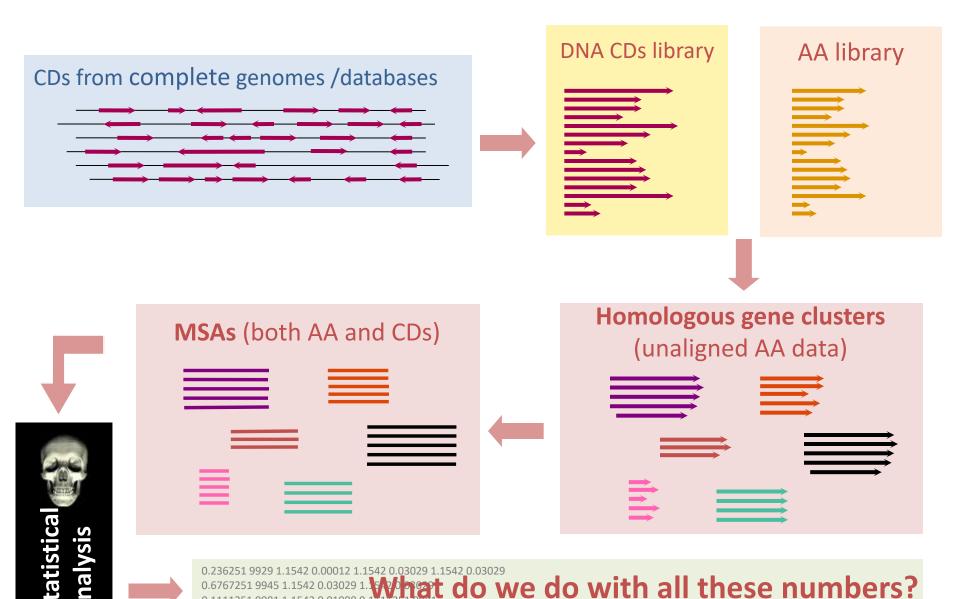
A systematic search for positive selection in higher plants (Embryophytes)

Christian Roth 1,2,3 and David A Liberles \* 1,3

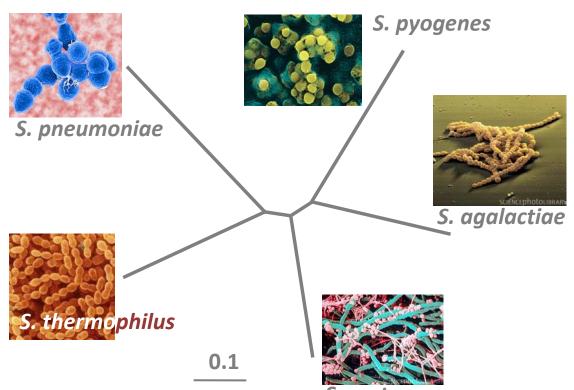
2006

PLOS BIOLOGY

# Large-scale selection scans step-by-step



### **Natural selection in Streptococcus**



Anisimova et al 2007 BMC Evol Biol

12 complete genomes

Positive selection in 136 genes:

29% connected to virulence
10% no ascribable function
7% essential to *S. pneumoniae*19% with body-site specific
patterns of gene expression
during invasive disease in *S. pyogenes* (infected blood,
cerebrospinal fluid,
epithelial cell contact)

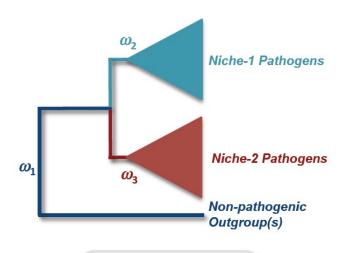
Positive selection affects both core and accessory genes, most likely due to the antagonistic interaction between host and parasite.

Products of both core and auxiliary genes participate in complex networks that comprise the molecular basis of virulence.

# Listeria phylogenomics Mapping selection to phenotype

A: Gene-level data analysis

B: Phenotype-level data analysis



#### Null model (1 parameter):

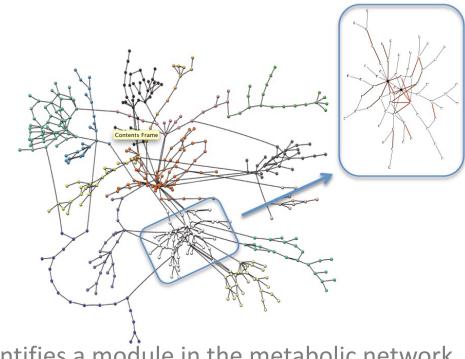
 $H_0: \omega_1 = \omega_2 = \omega_3$ 

#### Alternatives (2 parameters):

 $H_1: \omega_1 \neq \omega_2 = \omega_3$ 

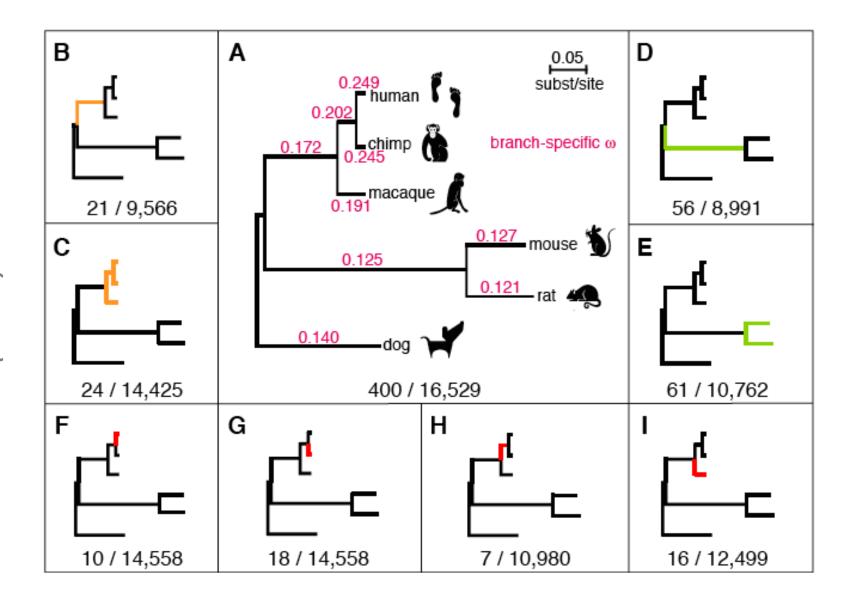
 $H_2: \omega_1 = \omega_2 \neq \omega_3$ 

 $H_3: \omega_1 = \omega_3 \neq \omega_2$ 

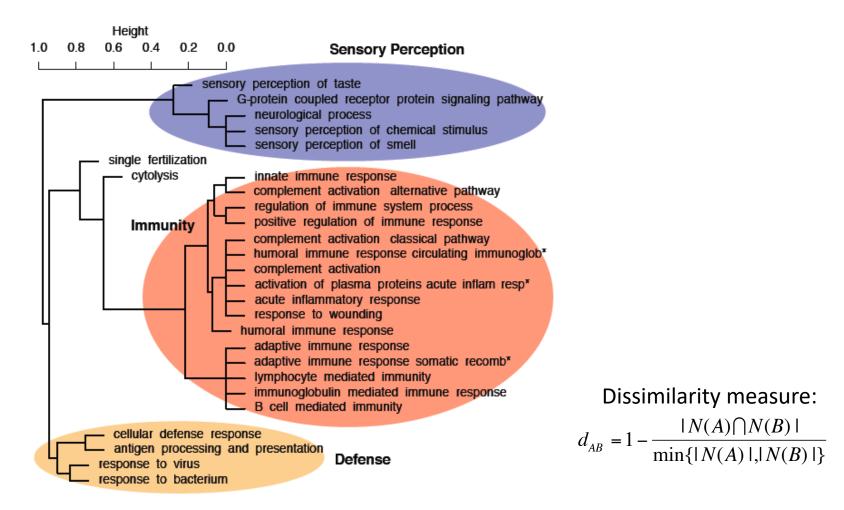


Blue box identifies a module in the metabolic network. Red links in the expanded view of this module indicate a significant cluster of genes subject to niche specific selection in "lineage I" of L. monocytogenes.

# Multiple LRTs: scan of mammalian genomes



# Multiple LRTs: scan of mammalian genomes



Hierarchical clustering of GO categories (biological process) over-represented with genes under positive selection

# Which proteins are under positive selection?

- Host proteins involved in defence or immunity against viral, bacterial, fungal or parasite attacks (MHC, immunoglobulin VH, class 1 chitinas).
- Viral or pathogen proteins involved in evading host defence (HIV env, nef, gap, pol, etc., capsid in FMD virus, flu virus hemagglutinin gene).
- Proteins or pheromones involved in reproduction (abalone sperm lysin, sea urchin bindin, proteins in mammals)
- Proteins that acquired new functions after gene duplication.
- Miscellaneous (diet, globins, etc.)



### **Detecting positive selection**

- Pairwise methods very low power
- Branch models allow variation over time but assume one  $\omega$  for all sites low power
- Site models allow variation among sites but assume selection pressure does not change over time – have higher power if positive selection is long term
- Branch-site models may be more successful at detecting episodic selection but are more difficult to fit, require more data and often have multiple suboptimal peaks (caution with genome scans!)

### **Testing for positive selection**

- LRT is accurate even for small datasets
- Power of LRT is better for larger datasets
- Watch out for recombination
- Accurate parameter estimation is more difficult, depends on model assumptions
- Bayesian site prediction is even more difficult than LRTs and parameter estimation
- There is an optimal window of sequence divergence (sequences should be not too similar and not saturated)
- Robustness of results: Use several models & tests
- Check for local optima, especially for complex models

#### Weaknesses of methods based on codon models

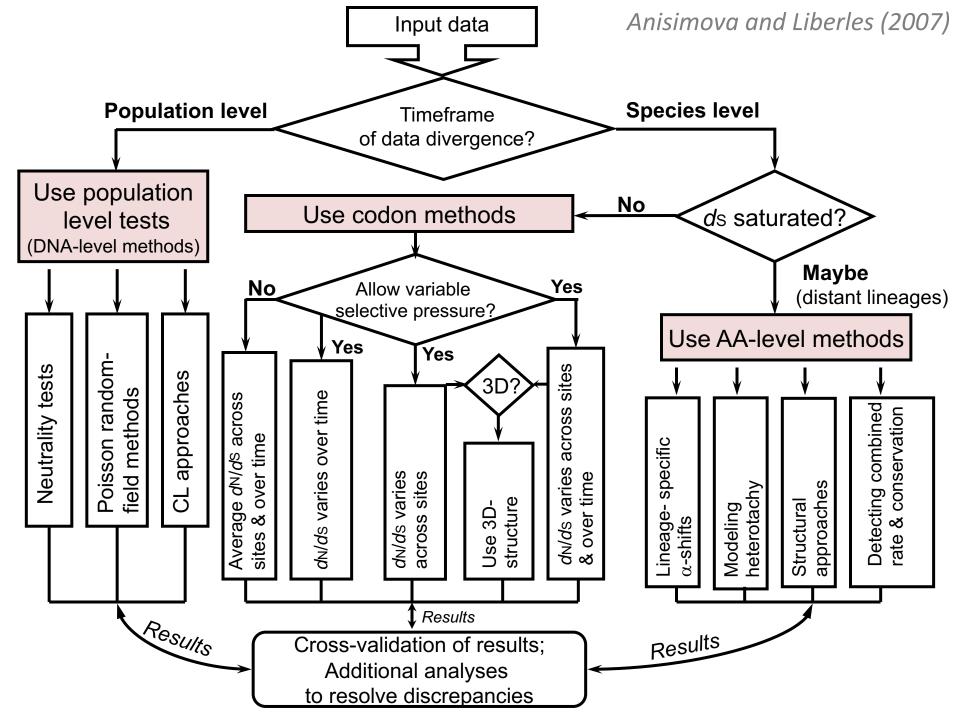
- Model assumptions may be unrealistic (but some assumptions matter more than others)
- The method detects positive selection only if it generates excessive nonsynonymous substitutions. It may lack power in detecting one-off directional selection or when the sequences are highly similar or highly divergent. Little power with population data.
- Do not work for noncoding DNA (but see Wong & Nielsen 2003 Genetics)
- Sensitive to sequence and alignment errors (Fletcher & Yang 2010 Mol Biol Evol 27; Privman et al. 2011 Mol Biol Evol 29; Jordan & Goldman 2012 Mol Biol Evol 29)

#### Criticisms on codon models

#### by M. Nei, Y. Suzuki, & A.L. Hughes

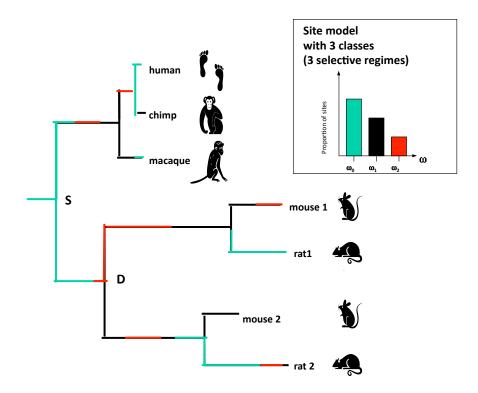
Hughes AL. 2007. Looking for Darwin in all the wrong places: the misguided quest for positive selection at the nucleotide sequence level. *Heredity* 99 Nozawa, Suzuki & Nei. 2009. *PNAS* 106

- Yang Z, dos Reis M. 2011. Statistical properties of the branch-site test of positive selection. *Mol Biol Evol* 28
- Zhai W, Nielsen R, Goldman N, Yang Z. 2012. Looking for Darwin in genomic sequences validity and success of statistical methods. *Mol Biol Evol* 29
- MacCallum, C. & Hill, E. 2006 Being positive about selection. *PLoS Biol* **4**, e87



# The many faces of codon models

- Detecting selection
- Studying codon bias
- Inferring phylogenies
- Dating speciation events
- Ancestral reconstruction
- Changes in time & space
- Predicting coding regions
- Improved alignment
- Inferring gene features (phyloHMM, netHMM)
- Simulation of data



Markov modulated model: Guindon et al. 2004

Reviews of codon models: Kosiol and Anisimova 2012 Anisimova and Kosiol 2009