# Evolutionary Reasoning Does Not Arise in Standard Usage of Protein Language Models

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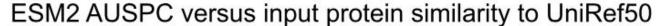


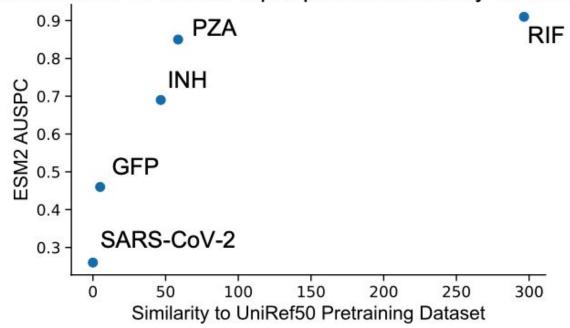






Existing protein language models struggle to generalize to out of distribution sequences





# How can we improve generalizability?

#### Evolutionary Modeling versus Evolutionary Reasoning

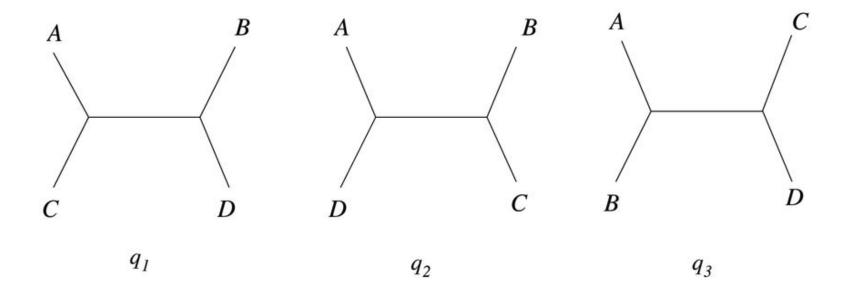
- Evolutionary modeling: learn residue-level distributions on a single-sequence, matching the marginal sequence distribution
- Evolutionary reasoning: inferring relationships among sequences from unaligned sequences

#### Open questions

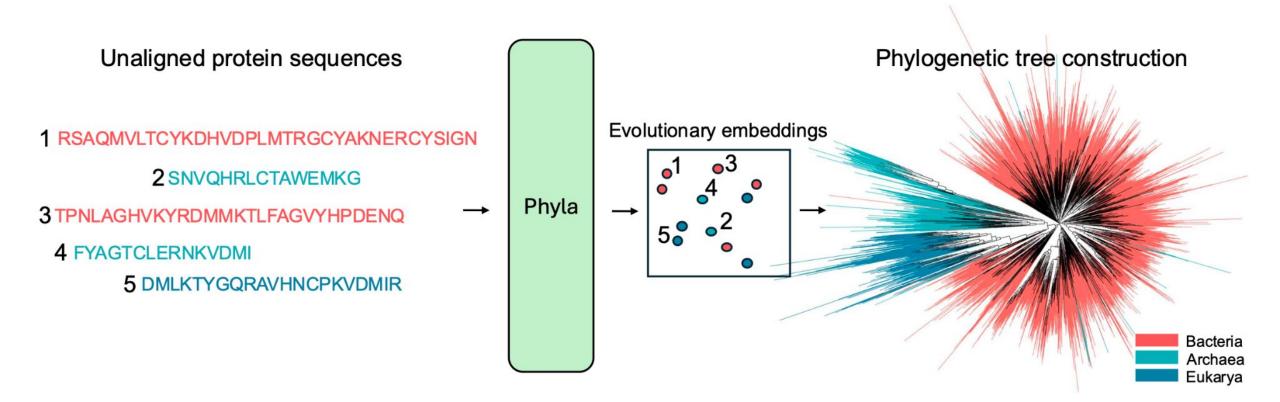
- 1. Are current PLMs capable of evolutionary reasoning?
- 2. And if not, how must we reimagine their architecture and training to enable them to do so?

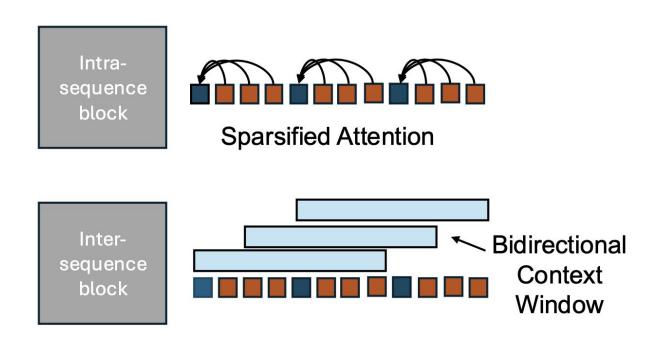
#### Tree Loss

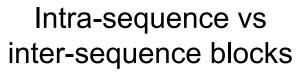
- Quartet loss
  - Samples sets of quartets
  - Considers all possible tree orientations in the quartet
  - Teaches the model to recapitulate ground truth distances

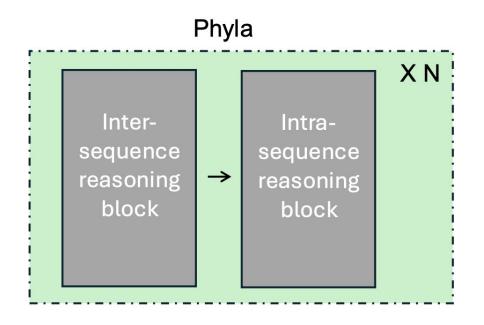


## Phyla Overview



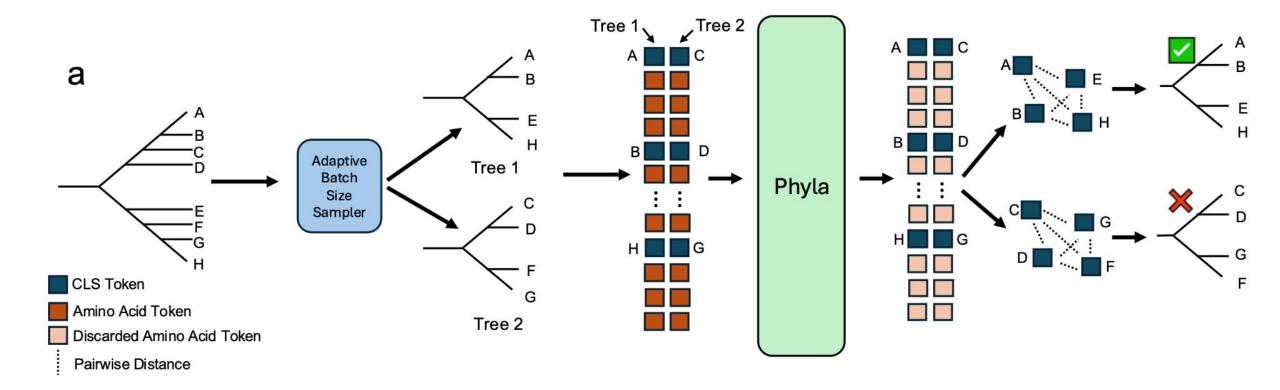






Phyla blocks

# Phyla Model Training

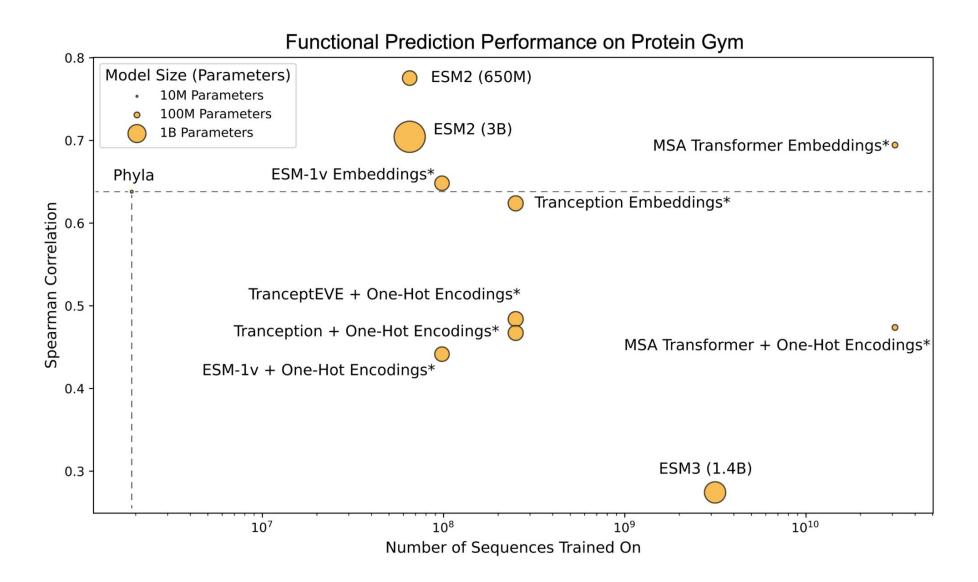


## Evolutionary Reasoning Benchmark

- Tree Reconstruction
  - How well do the models reconstruct phylogenetic trees?
  - TreeBase: 1,533 trees
  - TreeFam: 9,586 trees
- Taxonomic Clustering
  - How well do the models cluster evolutionarily similar sequences?
  - Genome Taxonomy Database (GTDB): bacterial isolates
  - Stratified by Class, Order, Family, Genus, Species

Model	$\textbf{TreeBase} \downarrow$	$\mathbf{TreeFam} \downarrow$	<b>Class</b> ↑	Order $\uparrow$	Family $\uparrow$	<b>Genus</b> ↑	Species ↑
Hamming Distance	0.75	0.75	_	=	_	_	_
MAFFT+FastTree	0.65	0.32	1_	_	_	_	_
ESM2 (650M)	0.78	0.67	0.64	0.66	0.68	0.71	0.75
ESM2 (3B)	0.79	0.67	0.55	0.56	0.57	0.59	0.67
ESM3 (1.4B)	0.81	0.72	0.61	0.63	0.66	0.67	0.72
ESM C (300M)	0.77	0.71	0.57	0.60	0.62	0.67	0.71
ESM C (600M)	0.80	0.73	0.61	0.66	0.66	<u>0.71</u>	0.75
Evo 2 (7B)	0.84	0.84	0.50	0.54	0.55	0.55	0.64
ProGen2-Large (2.7B)	0.77	0.68	0.60	0.65	0.66	<u>0.71</u>	0.75
ProGen2-XLarge (6.4B)	0.86	0.82	0.52	0.55	0.57	0.61	0.68
PHYLA (24M)	0.73	0.58	0.71	0.76	0.87	0.93	0.98

#### **Functional Prediction**



## Generalizability

Model	Low-Overlap	High-Overlap	All Datasets
ESM2	0.59	0.80	0.78
PHYLA-MLM	0.53	0.61	0.55
PHYLA-NoAttention	0.40	0.53	0.44
PHYLA	0.62	0.68	0.64

Also minimal train-test overlap <0.6% on Treebase and <3.4% on TreeFam

#### Applications of Phyla

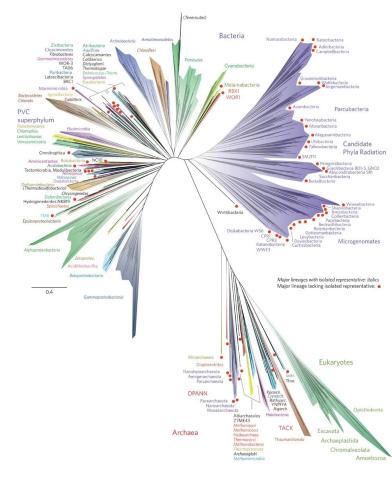
A new view of the tree of life. Laura A. Hug, Brett J. Baker, et al. Nature

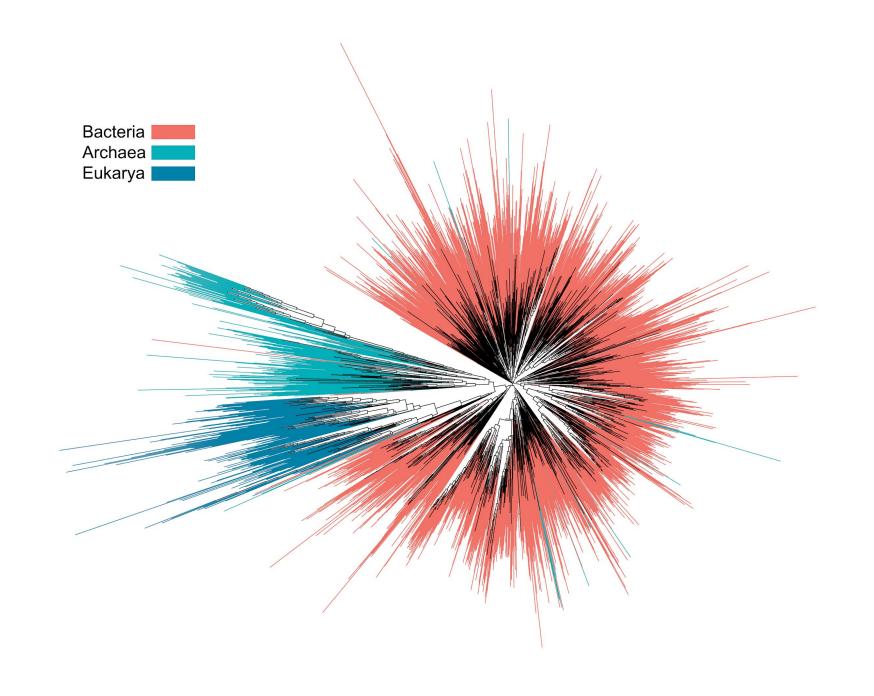
Microbiology. 2016

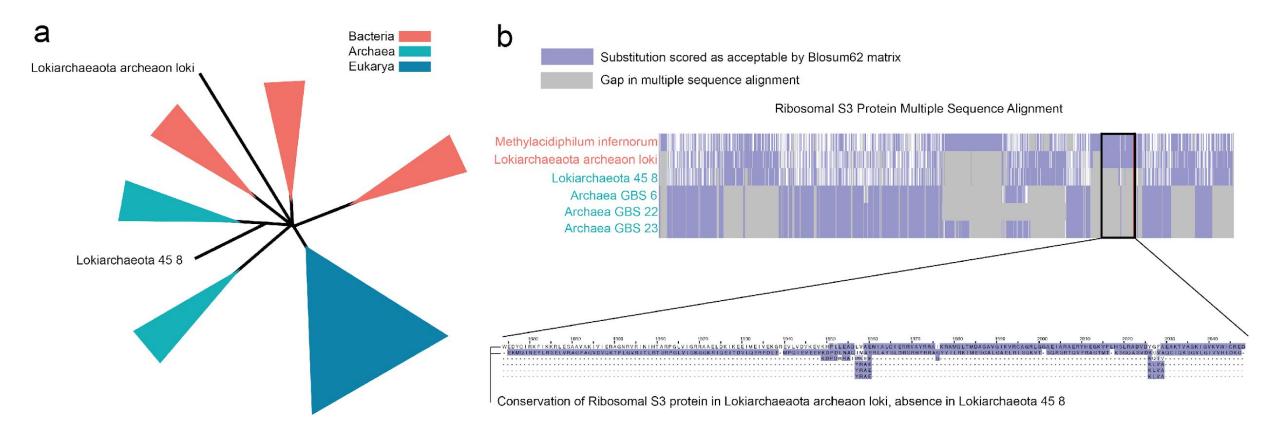
3083 Ribosomal Protein Sequences

3,840 computational hours

- Feed this to Phyla
  - Constructed tree in 16 hours
- How does this tree differ from tree of life?
  - Are these differences meaningful?







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#### Co-authors

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