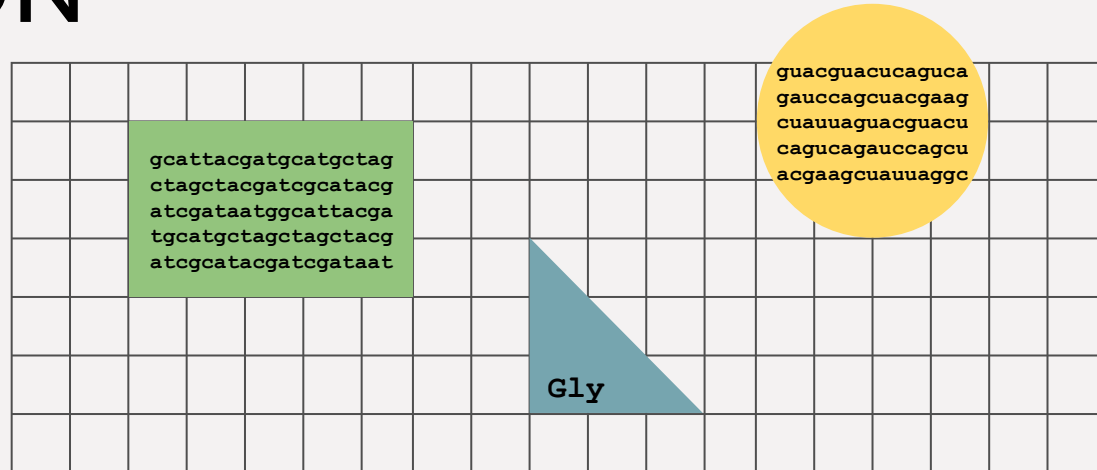


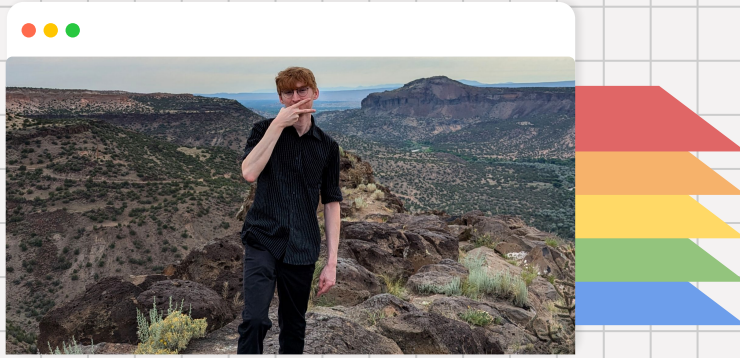
EVOLUTIONARY ALGORITHMS SIMULATING MOLECULAR EVOLUTION

(Yeah, it's kind of a mouthful.)

(And it's fair game for the exams!)



About Me



- Bachelor's in CS from **UAB** in 2020
 - Minor in neurobiology
- Master's in CSSE from **Auburn** in 2022
- Doctorate in CSSE now ongoing
- Research focus on evolutionary algorithms (EAs) and their application to biology

Fun Facts

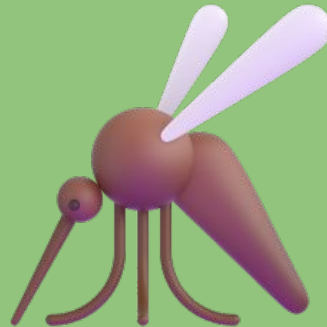
- I have two lizards, a ball python, and a tortoise.
- Insufferable Letterboxd guy.

What's on the agenda?



Bio 101: A Quick Refresh

I'm sure it's been a while.



Where it All Began

The very first EASME project.



What We're Up to Now

I promise we're not creating bioweapons.

1. Biology 101

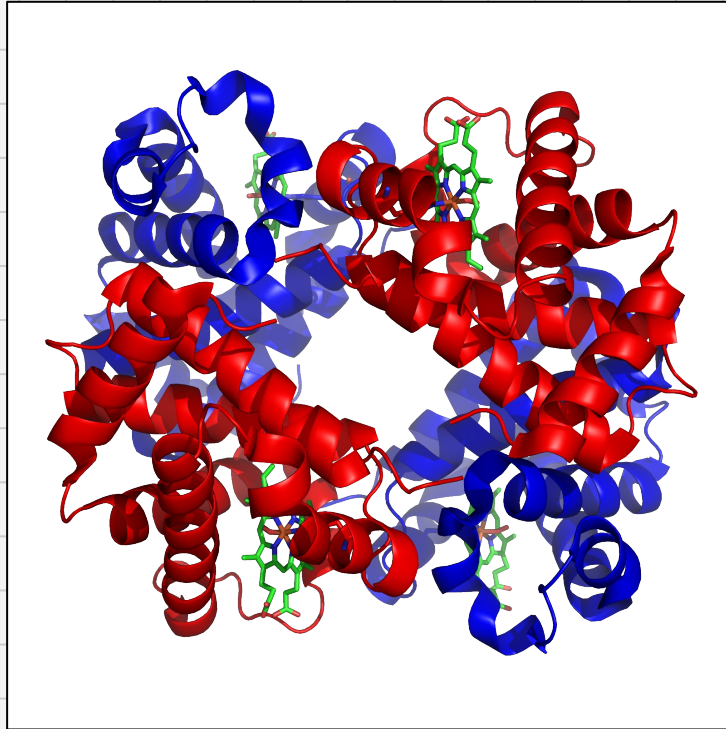
(Just the Important Stuff)

The idea that information flows from DNA to RNA to proteins is the “central dogma” of biology.

- Most of your cells contain deoxyribonucleic acid, or **DNA**, made up of four base pairs: adenine (**A**), thymine (**T**), guanine (**G**), and cytosine (**C**).
- **Transcription** converts **double-helix** DNA into **single-helix RNA**, which replaces thymine (T) with uracil (**U**).
- Ribosomes in your cells **translate** RNA into a string of **amino acids**, which then folds into a unique shape — this folded structure is a **protein**.

Three base pairs of RNA are translated into **one** amino acid. There are 64 possible RNA “codons,” which map to 20 amino acids — so it’s a hash table!

Proteins in More Detail



Hemoglobin

Evolution

(The Old-Fashioned Way)

Trillions of mutations happen inside your body every single day. Luckily, your cells automatically fix most of them — just hope “most” is enough!

- Evolutionary computing was inspired by the natural process of evolution.
- In nature, factors like radiation can cause one base pair in a DNA molecule to **mutate** into another.
 - Transition (A↔G, C↔T) is more common than transversion.
- As in evolutionary computing, some mutations are harmful, some are beneficial, and some are completely neutral.
 - As multiple, often similar codons map to the same amino acid, *most* mutations in biology are neutral.

The codons CGA, CGC, CGG, *and* CGU all map to the amino acid arginine (**R**).

AUG is the **start** codon, and UAA, UAG, and UGA are the **stop** codons.

So what would happen if a CGA mutated to UGA...?

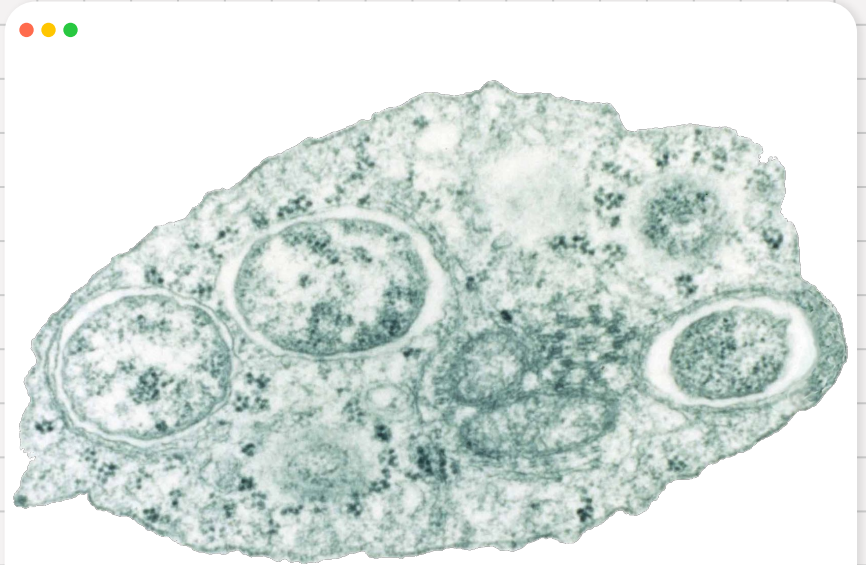
2. The First Project

feat. *Wolbachia pipientis*



The first EASME project, developed by Auburn faculty member John Beckmann and Dr. T, used an EA to simulate the evolution of cytoplasmic incompatibility in *Wolbachia*-infected mosquitoes.

That's a lot of words! But what do they *mean*?



Wolbachia in an infected host cell

Cytoplasmic Incompatibility Explained

	Uninfected Female	Infected Female
Uninfected Male	Uninfected Offspring	Infected Offspring
Infected Male	No Offspring ☠	Infected Offspring

- Inside the sperm of an infected male, the *Wolbachia* bacteria will produce a toxin that will kill the offspring if left unattended.
- Inside the egg of an infected *female*, however, that very same bacteria will produce an antidote to that toxin!
- Through this simple toxin-antidote system, *Wolbachia* has become the single most common reproductive parasite in Earth's biosphere.

Which came first, the chicken or the egg?

Which evolved first, the toxin or the antidote?

Modeling the Emergence of TA Protein Functions

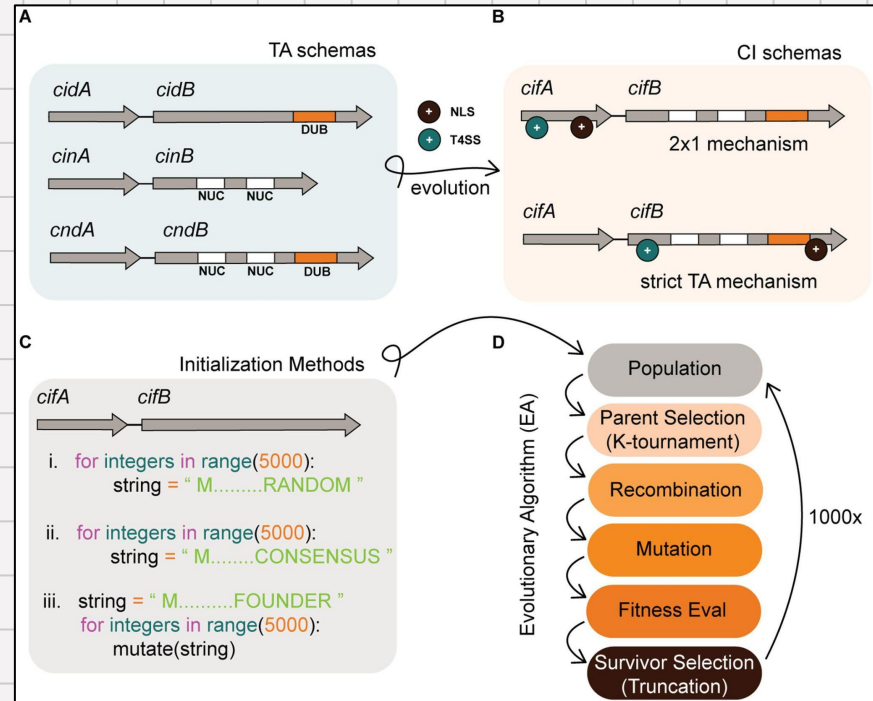
Rather than evolve some abstract representation, this project directly encoded, evolved, and evaluated DNA strings (using the A, T, G, C primitives).

The toxin-antidote system is controlled by a “lock and key” mechanism between the two proteins, which is easy to model.

Algorithm initialized a population of random sequences, then attempted to evolve them towards the modern wild-type sequences.

Fitness was determined by the presence of certain enzymatic domains, a “sliding window” search for binding residues, and the presence of NLS/T4SS.

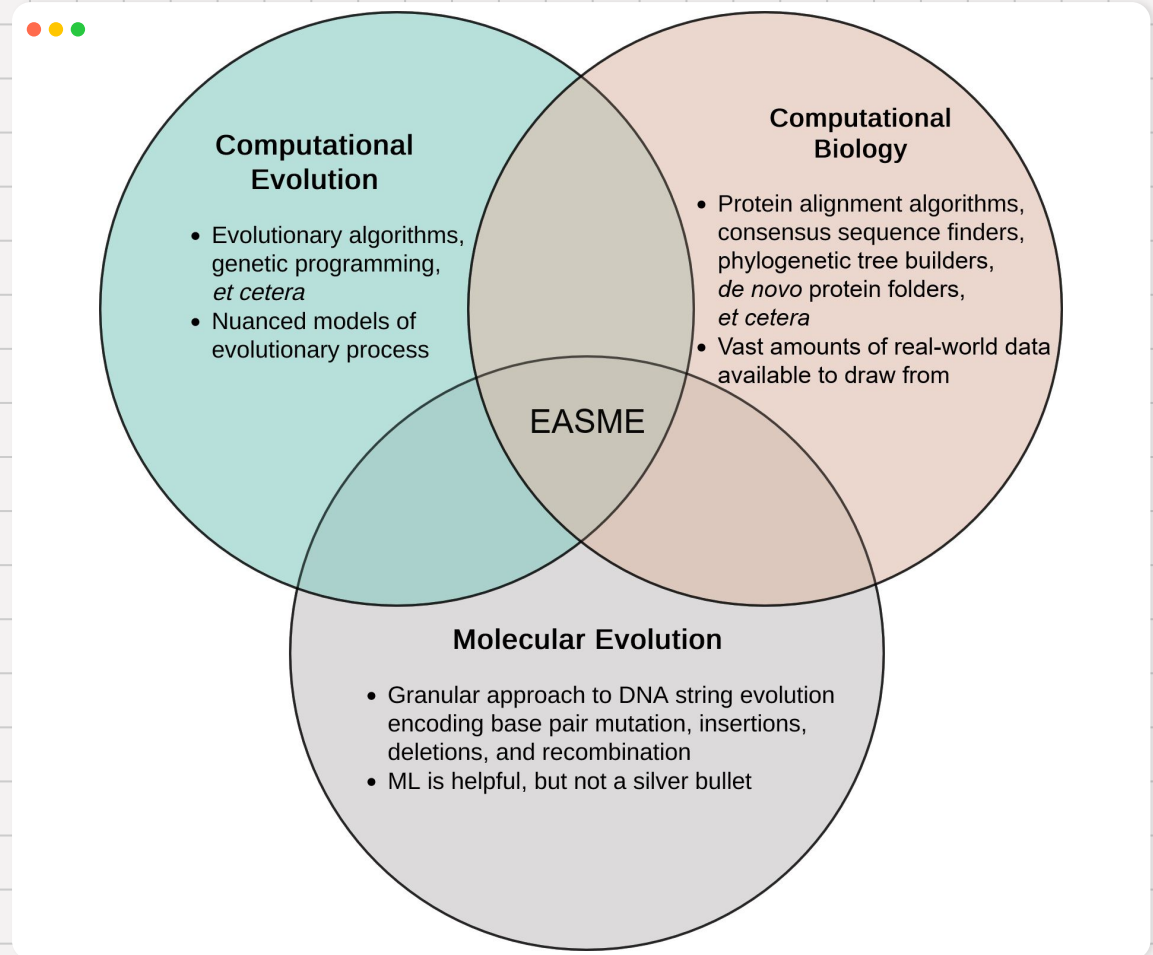
Both wild-type mechanisms can be reached from the random initializations, but starting conditions will determine which one is reached first, and one is easier than the other.



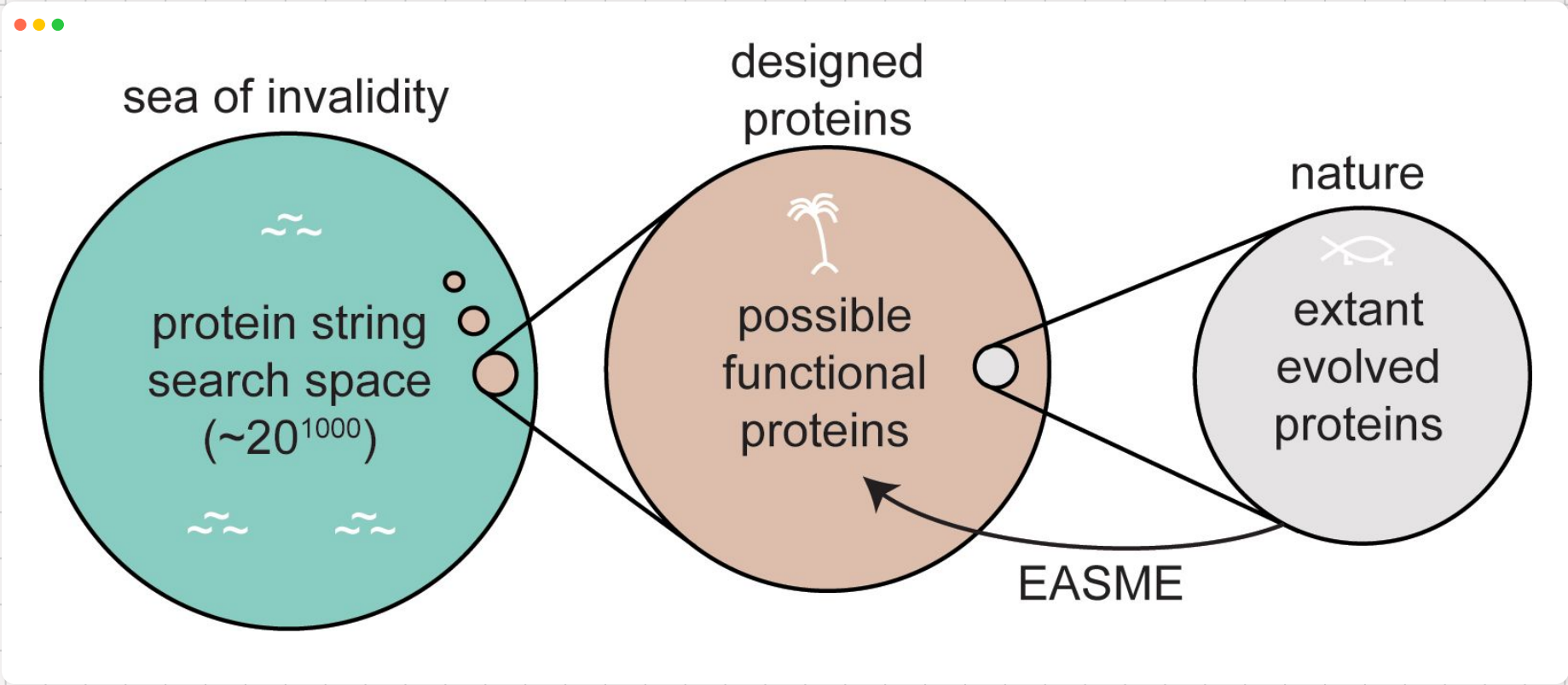
3. EASME

EASME only became possible recently, due to the computational costs associated w/ many biological problems.

Wet lab experiments are a rarity in the CS field, and somewhat unique to EASME.



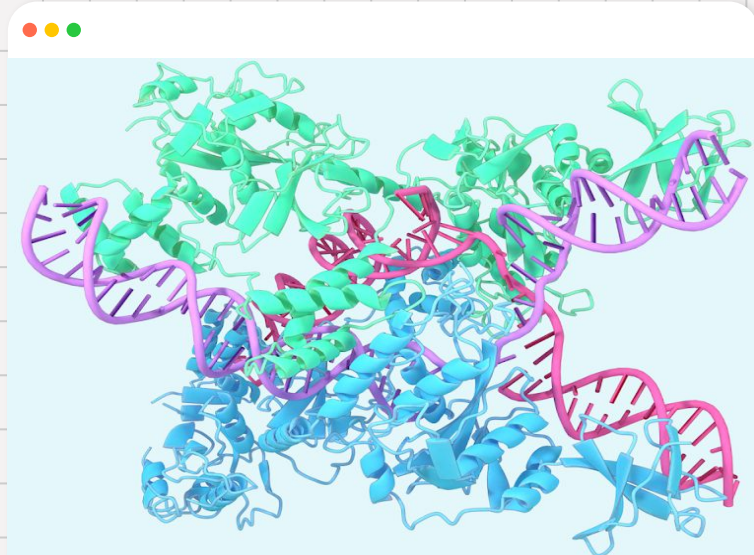
Search Space



Why Not ML?

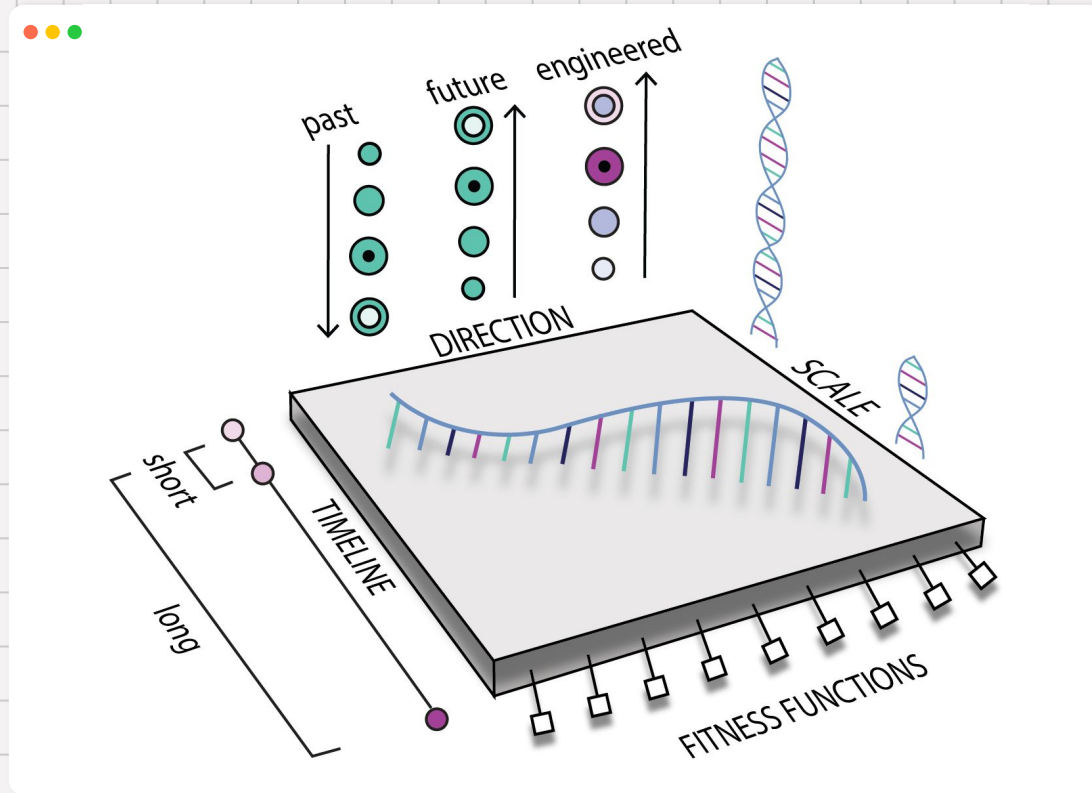
“Well, probably because this is a class on EC, not a class on ML...”

- Machine learning has been making a lot of waves recently, especially with the release of Google’s AlphaFold 3.
- However, EC is naturally a better fit for *evolving* new proteins, as it models the natural process that created existing proteins.
- ML models can be useful components of fitness functions, but would be somewhat limited by their own training data if tasked with designing entirely *new* proteins.



A protein folded by AlphaFold 3

Categorizing Projects



The Big Question: Fitness



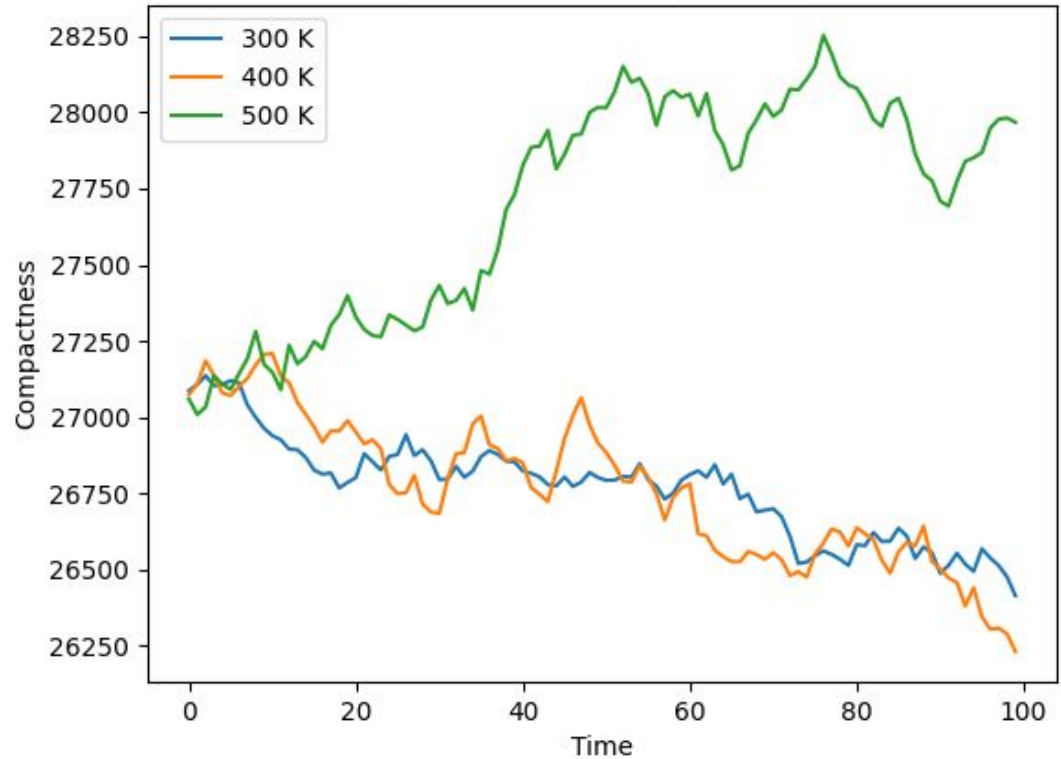
How does one calculate the fitness of a protein?

Protein folding is one of the biggest unsolved problems in science, but machine learning models can help us approximate structures.

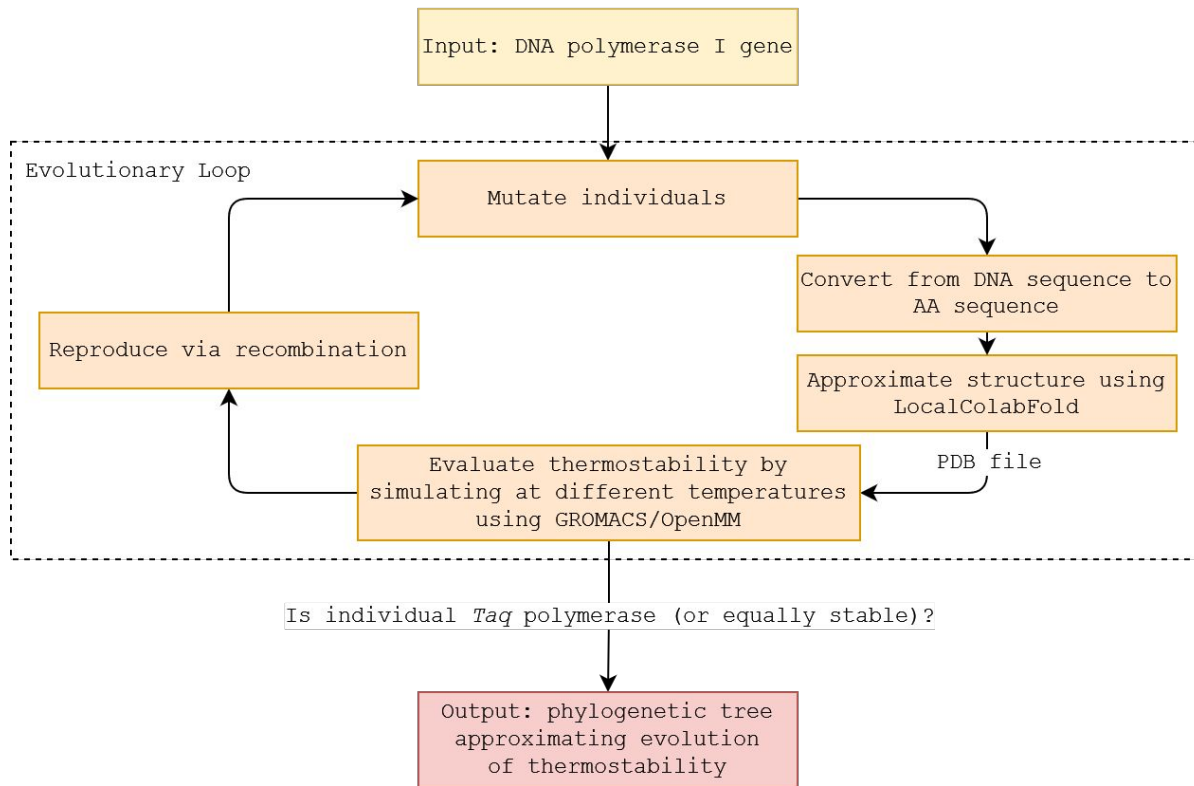
Approximated foldings and simple simulations can serve as rudimentary fitness functions.



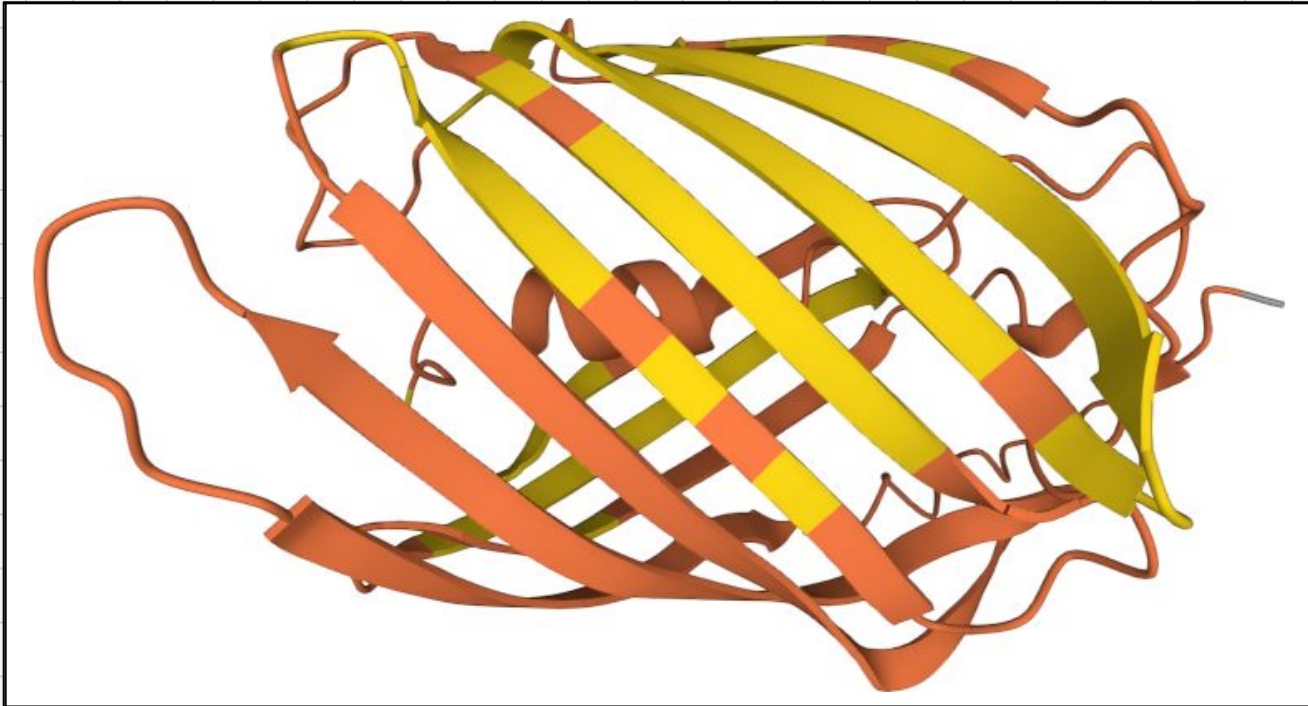
Thermostability of Protein



HEAT: Hybridized Evolution Approximating *Taq*



Next Step: Lab Testing

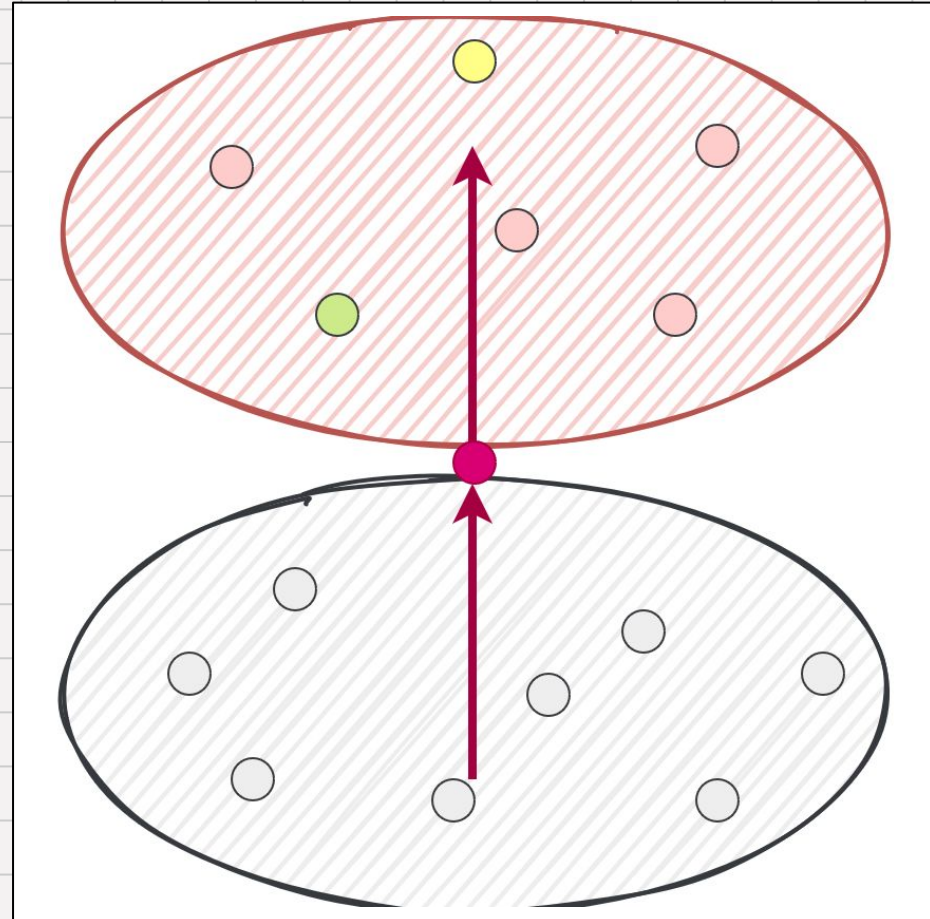


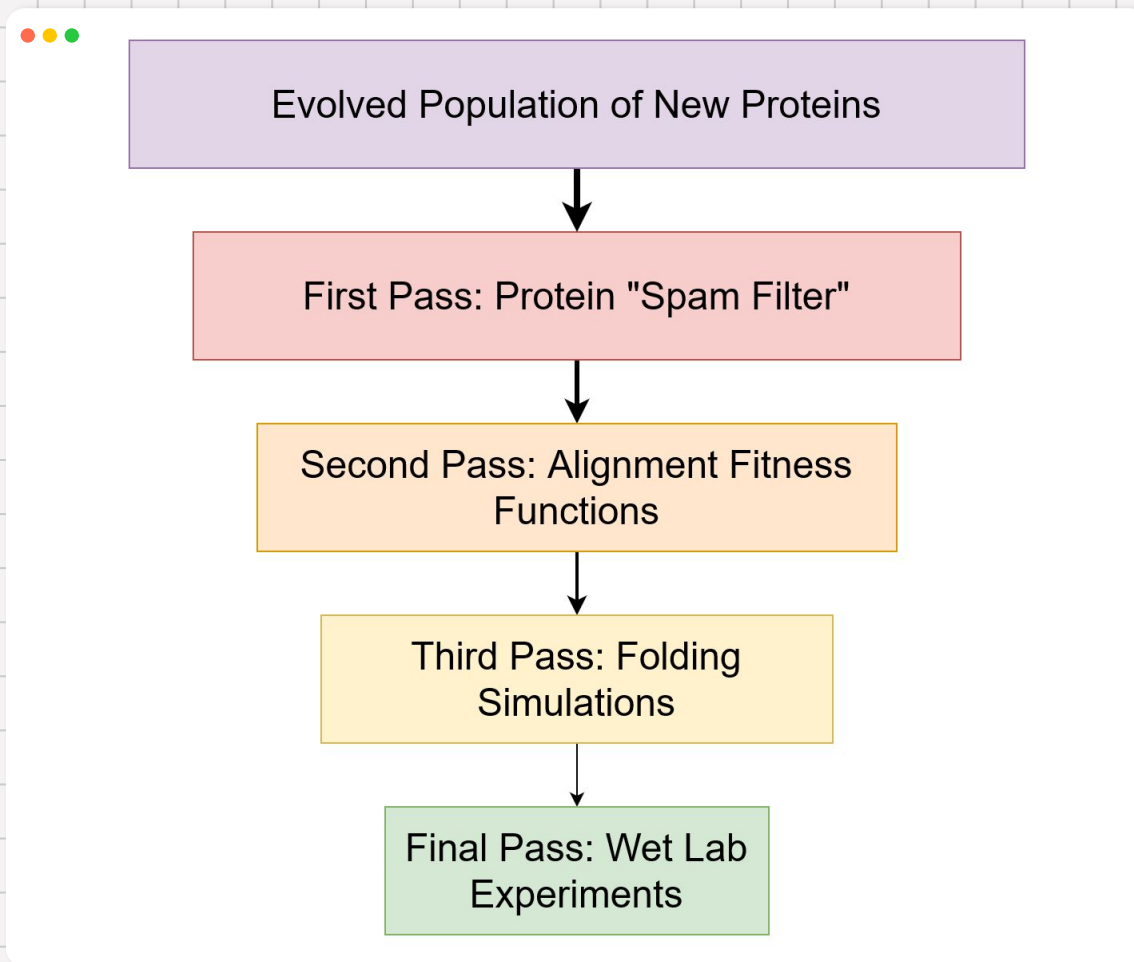
pMagenta

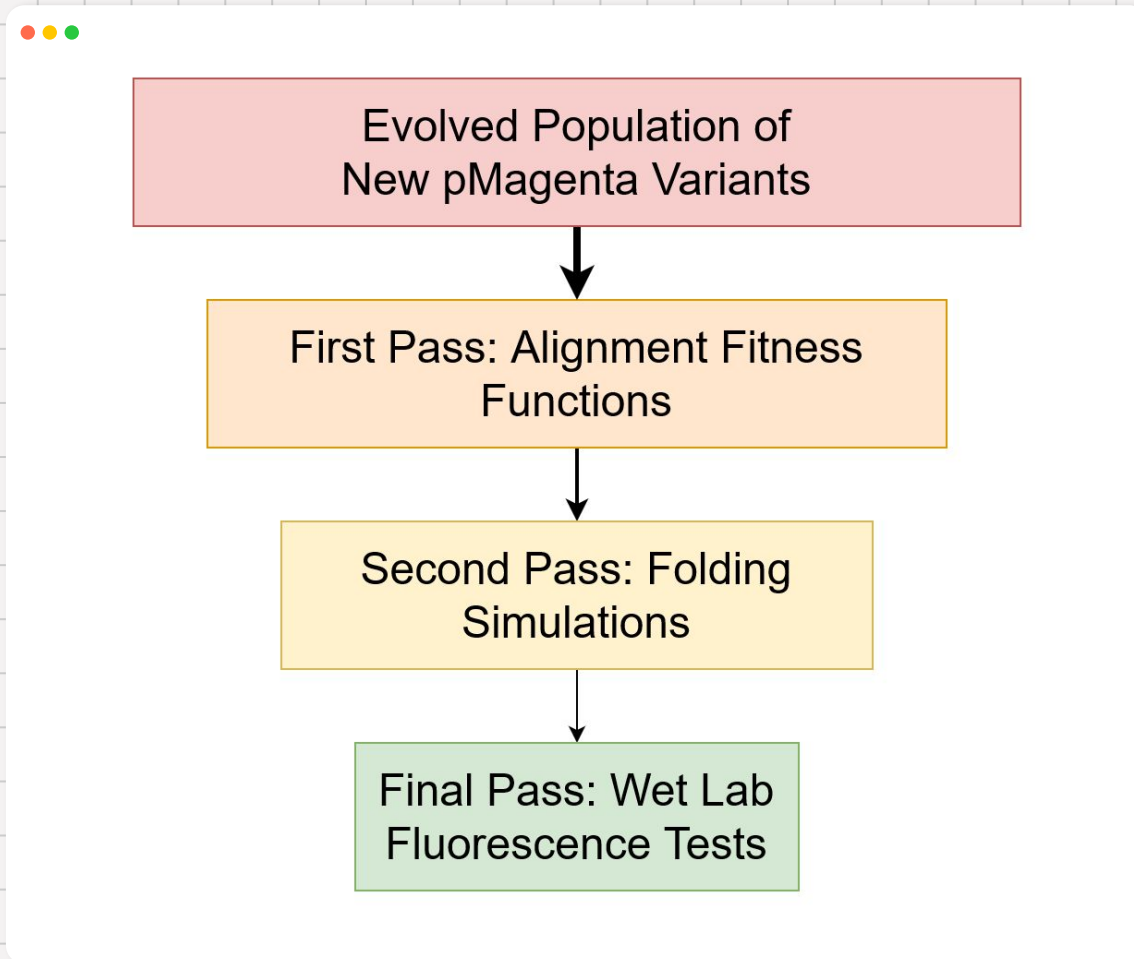
Fluorescent Space

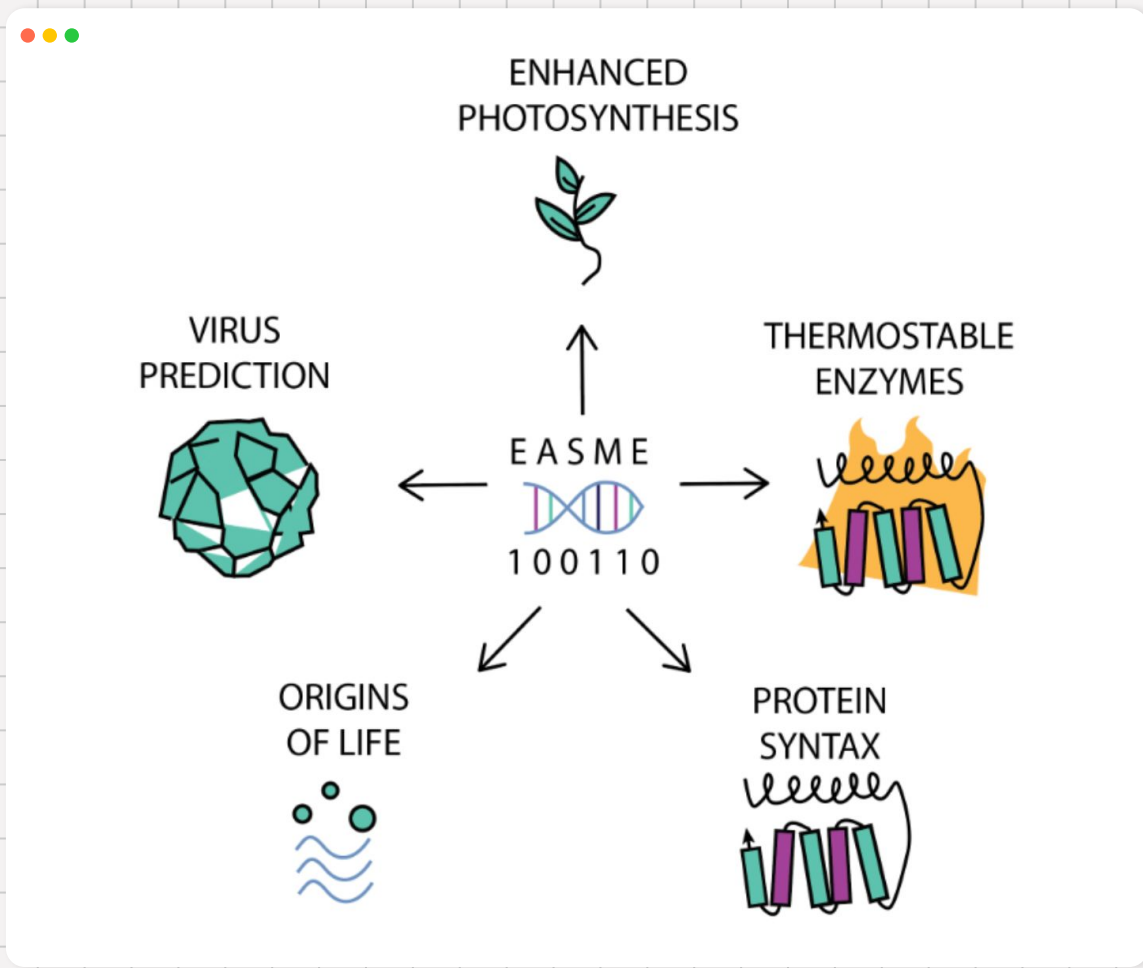


1. Download proteins similar to pMagenta from UniProt, some fluorescent, some not.
2. Group based on fluorescence.
3. Judge fitness of new variants by alignment to all specimens — homology to fluorescents is positive, homology to non-fluorescents is negative.
4. Ideally, new variants will move towards higher fluorescence in this high-dimensional space.











aub.ie/easme

Sources

- Hemoglobin render: https://commons.wikimedia.org/wiki/File:1GZX_Haemoglobin.png
- *Wolbachia* image: <https://commons.wikimedia.org/wiki/File:Wolbachia.png>
- “Modeling emergence of *Wolbachia* toxin-antidote protein functions with an evolutionary algorithm,” Beckmann *et al.* (2023). *Frontiers in Microbiology*, Vol. 14.
- Abramson, J., Adler, J., Dunger, J. et al. Accurate structure prediction of biomolecular interactions with AlphaFold 3. *Nature* **630**, 493–500 (2024).
<https://doi.org/10.1038/s41586-024-07487-w>