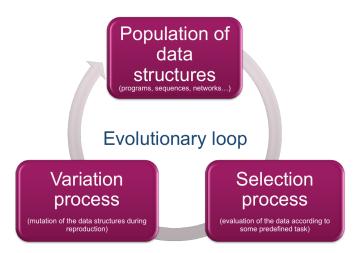
Introducting mini-Aevol: Underlying models, implementation and possible optimizations and parallelization

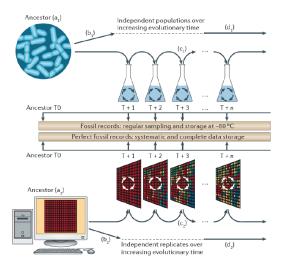
Jonathan Rouzaud-Cornabas

LIRIS / Insa de Lyon – Inria Beagle

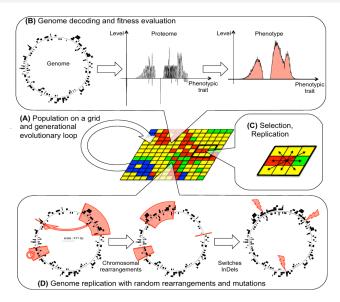
Evolutionary Loop

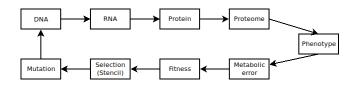


In-silico experimental evolution

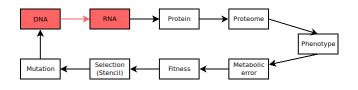


The biological model of Aevol

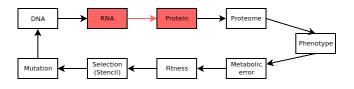




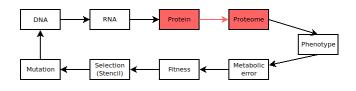
- The workflow is repeat for every cell of the grid
- Each cell contains one and only one organism



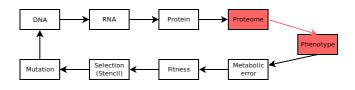
- DNA \rightarrow RNA : **Transcription** : Looking for 2 motifs
 - Promoter (size = 22): Hamming distance (predefined motif, max. error = 4)
 - ② Terminator (size = 4) : Exact match (DNA[pos : pos + 4] == DNA[pos + 10 : pos + 14])



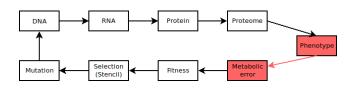
- DNA → RNA : **Transcription** : Looking for 2 motifs
- ullet RNA o Protein : **Translation** : Looking for 2 motifs
 - Shine Dalgarno (size = 6) + space (size = 4) + Start (size = 3) : Exact match
 - 2 Stop (size = 3): Exact match



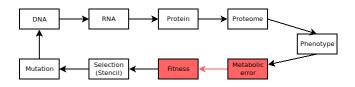
- DNA → RNA : Transcription : Looking for 2 motifs
- RNA → Protein : Translation : Looking for 2 motifs
- Protein → Proteome : Folding :
 - Looking for $N = \frac{Protein_Length}{3}$ motifs
 - Decoding each triple to addition to M, W, H variables
 - Translation *M*, *W*, *H* into triangle and approx. them into 300 double array



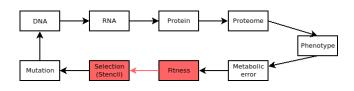
- DNA → RNA : Transcription : Looking for 2 motifs
- RNA → Protein : Translation : Looking for 2 motifs
- Protein → Proteome : Folding :
- Proteome → Phenotype : Summing arrays :
 - Looking for $N = \frac{Protein_Length}{3}$ motifs
 - Decoding each triple to addition to M, W, H variables
 - Translation *M*, *W*, *H* into triangle and approx. them into 300 double array
 - Summing into a global 300 double array



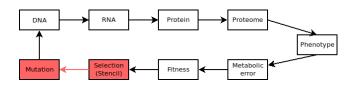
- ullet DNA ightarrow RNA : **Transcription** : Looking for 2 motifs
- RNA → Protein : Translation : Looking for 2 motifs
- Protein → Proteome : Folding :
- Proteome → Phenotype : Summing arrays :
- Phenotype → Metabolic error: Difference betwen 2 arrays
 - Phenotypical Target : Pre-defined array built from a sum of gaussians
 - Compute the difference the Phenotype and the phenotypical target



- Metabolic error → Fitness:
 - exp(-selection_pressure × metabolic_error)



- Metabolic error → Fitness:
- Fitness → Selection : Stencil 9-point 2D
 - Only kernel requiring data coming from other cells of the grid
 - Fetch fitness of the neighboring cells
 - Use the fitness proportionate algorithms to select the organism that will reproduce in the cell
 - Delete the old organism
 - Copy the DNA of the organism into the cell



- Metabolic error → Fitness:
- Fitness → Selection : Stencil 9-point 2D
- Selection → Mutation : Modify the DNA
 - Switch of the value of up to 6 value in the DNA array

Aevol VS mini-Aevol

- Aevol VS micro-Aevol SLOC: 87,000 VS 2,800
- Simplify model
 - Only a strand of DNA (and not too) i.e. the DNA is read one-way and not both-way
 - A reduce mutation sets: no mutation that change the size
 - A lot of advances features are missing: no plasmid, 4bp DNA, ...
- Implementation
 - Lot less robust to input errors
 - Can not change model parameters during an evolution
 - No phylogenetic tree
 - No postprocessing tool

Optimizations

- Working only on diff (where the DNA has changed during the mutation) i.e. do not reparse the whole DNA at each generation (available in the git repository for sequential CPU)
- Remove the modulo that allows to simulate circular DNA (copy a part of the beginning of the DNA at the end to do it)
- Change the coding of the DNA from a char array to a bitset
- Vectorization of the comparison between DNA and motifs
- Search multiple motifs at once through vectorization
- Evaluate classic mutation operator (malloc+memcpy+mutation) and backtrace mutation operator (malloc+copy)
- One large 1D structure (merge all DNAs into a big DNA) to ease load balancing
- Include metadata (promoter, terminator, start, stop) into a bitset

Parallelization

- OpenMP (parallel for and/or tasks with or without dependency)
- CUDA
- OpenACC
- MPI
- Hybrid e.g. OpenMP+OpenACC, OpenMP+MPI, ...

• Implement the stencil (i.e. asynchronous time between grid cells)

Methodology

- Use GIT (or another versioning software)
- For each optimization/parallelization, tests its performance, scalability (weak and strong) and create visualization of them
- You need to verify the reproduceability of your code i.e., same parameters must lead to same results

Simulation parameters:

- Size of the DNA (-g): 500, 5000, 50,000 (you can go up to 500,000)
- Population size (-w and -h) : 32×32, 128×128, 256×256 (you can up to 1024×1024)
- Mutation rate (-m): 0.0001, 0.00001, 0.000001