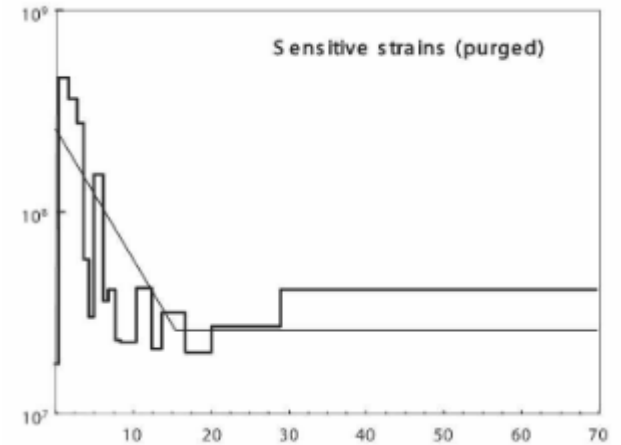
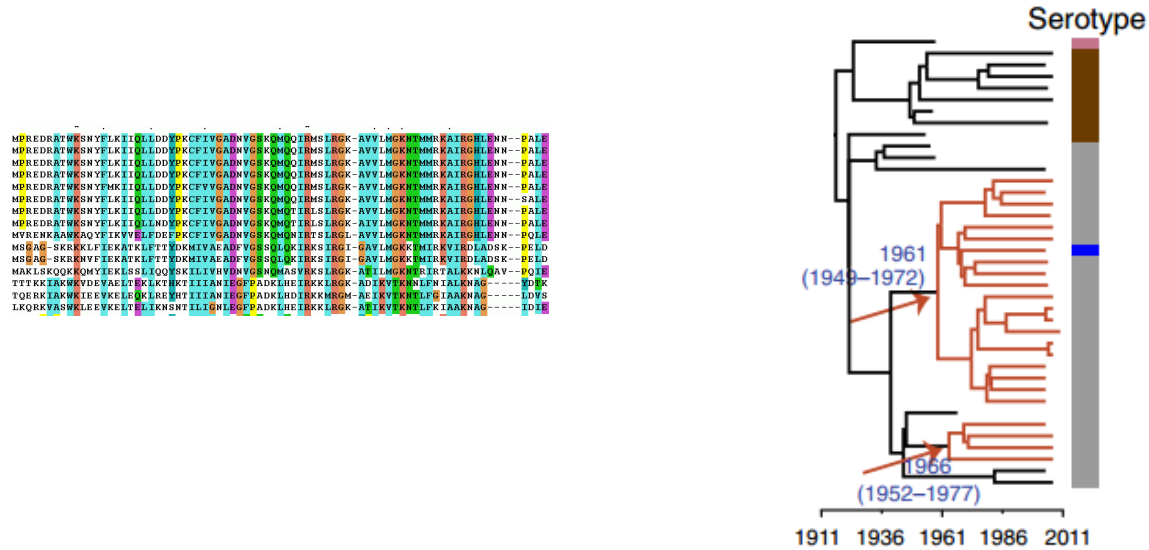


Inferring population sizes of bacterial populations a deep learning approach

MLMicrobial Genomics - ECML -2022

Jean Cury, Théophile Sanchez, Erik Bray, Jazeps Medina-Tretmanis,
Maria Avila-Arcos, Emilia Huerta-Sanchez, Guillaume Charpiat, and Flora
Jay

Bacterial population genetics

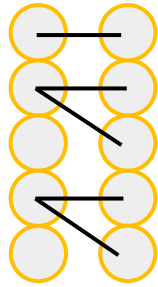


⇒ Focus on population size inference

Intuition



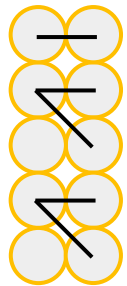
Intuition



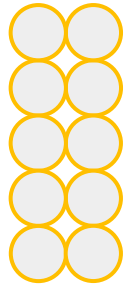
Parental

Daughter cells

Intuition

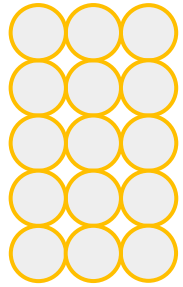


Intuition



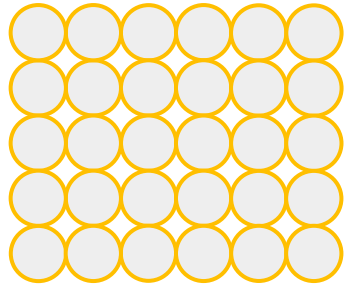
→
Generations

Intuition



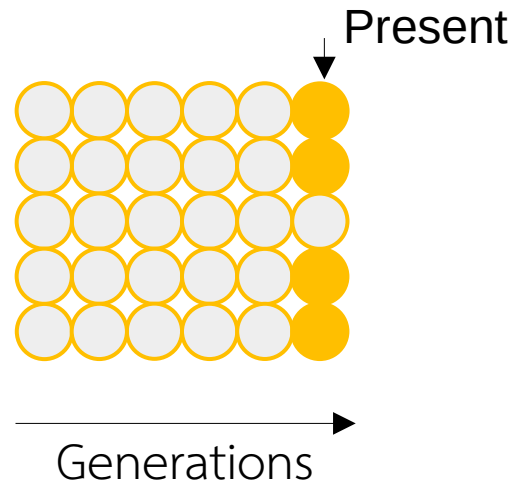
→
Generations

Intuition

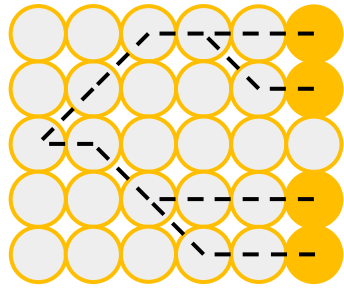


→
Generations

Intuition

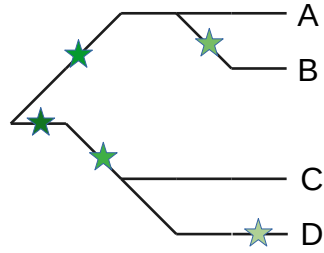
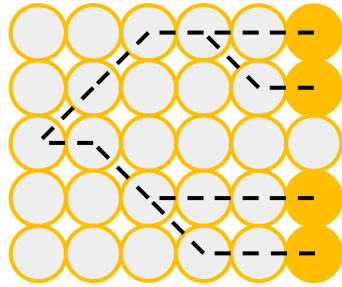


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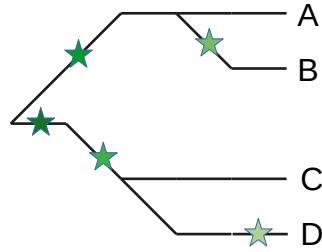
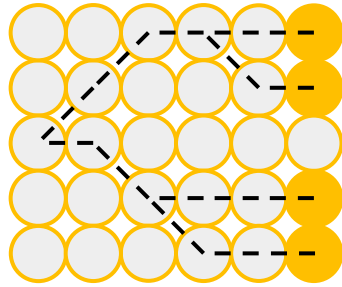


→
Generations

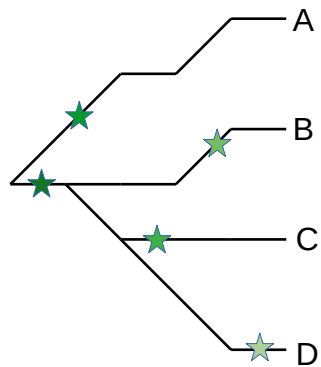
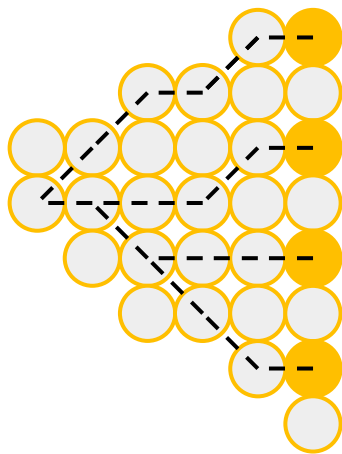
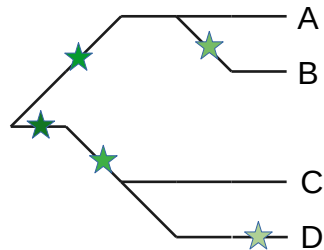
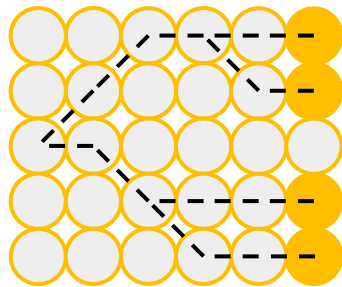
Intuition



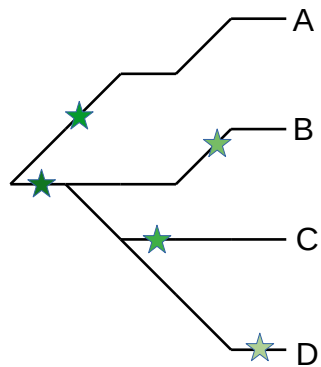
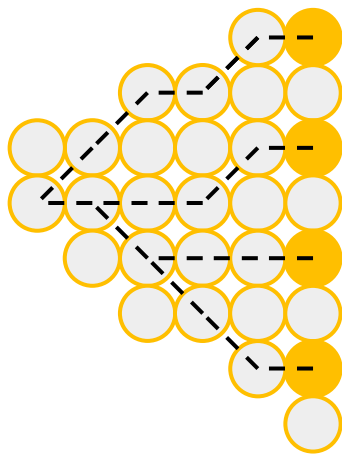
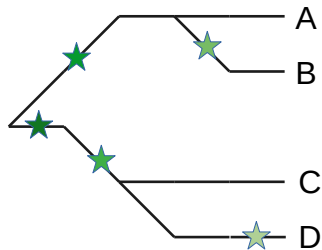
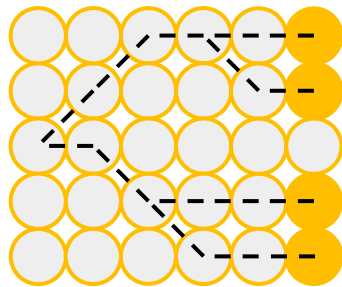
Intuition



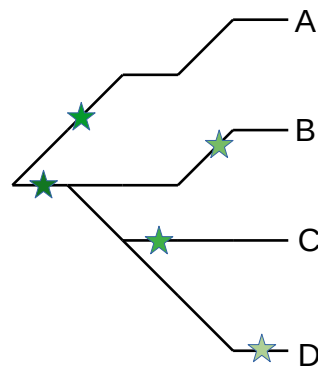
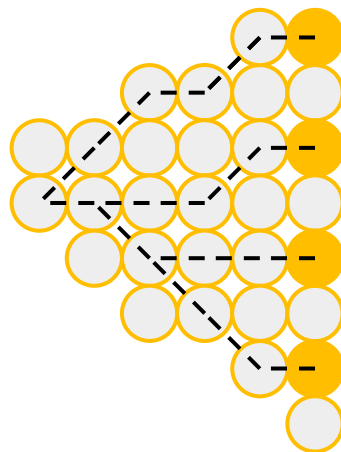
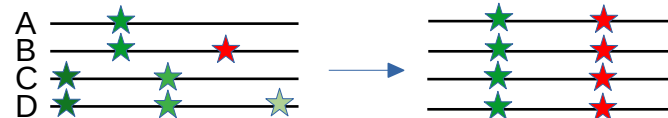
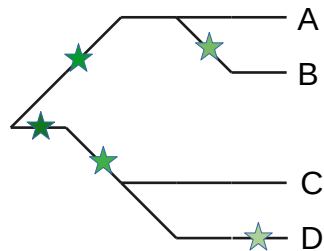
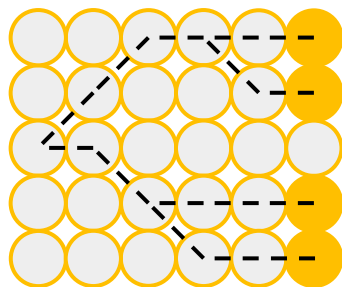
Intuition



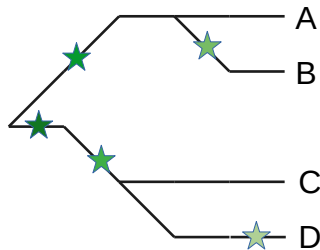
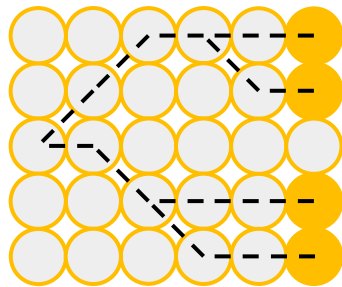
Intuition



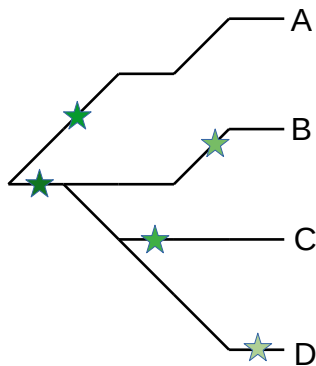
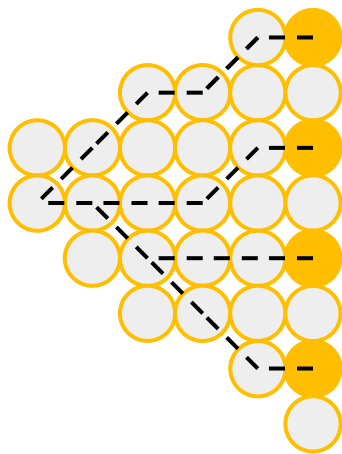
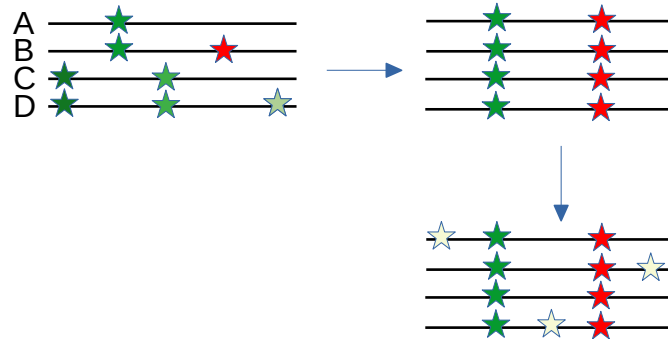
Intuition



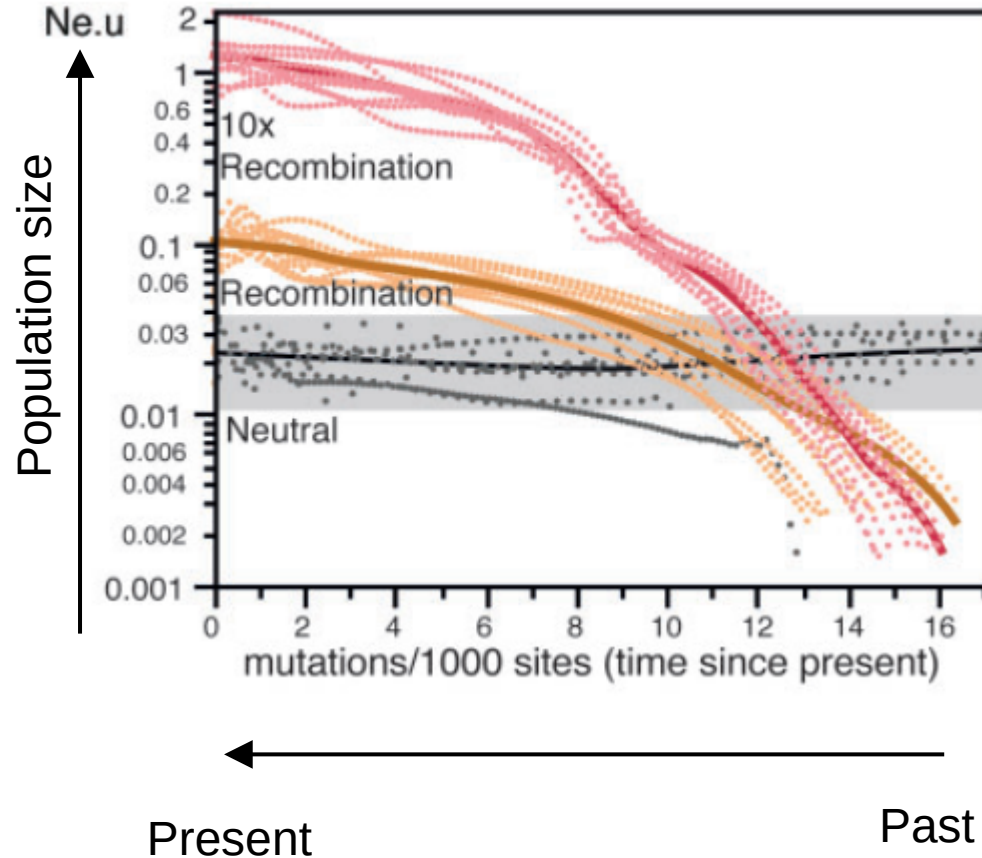
Intuition



Add selection sign



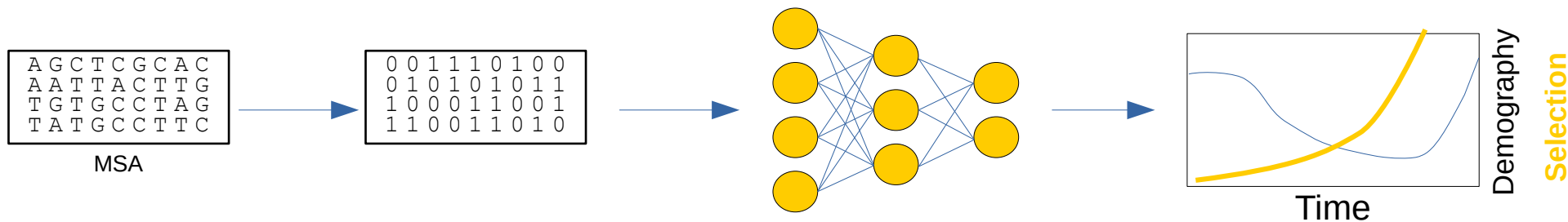
Inference : Skyline Plot



- Using estimated coalescent time, it infers population size
- Non-parametric
- Does not require simulation
- Problem : does not work in bacteria

Project

- End to end deep learning approach for bacterial popgen



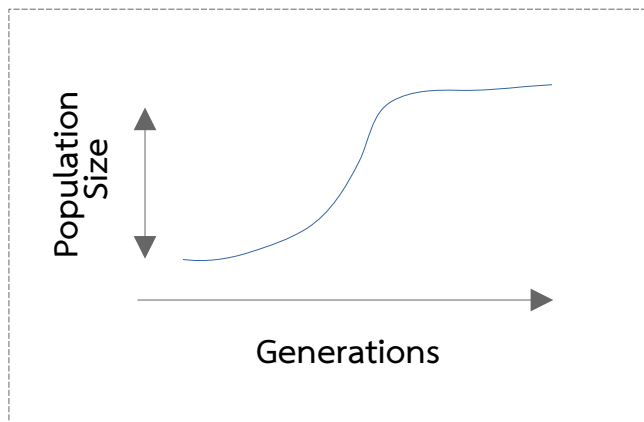
- **Problem:** No ground truth data
- We need a population genetic **simulator** that is :
 - fast
 - Implement bacterial recombination (homologous HGT)
 - Demography, selection, etc..

Input Data

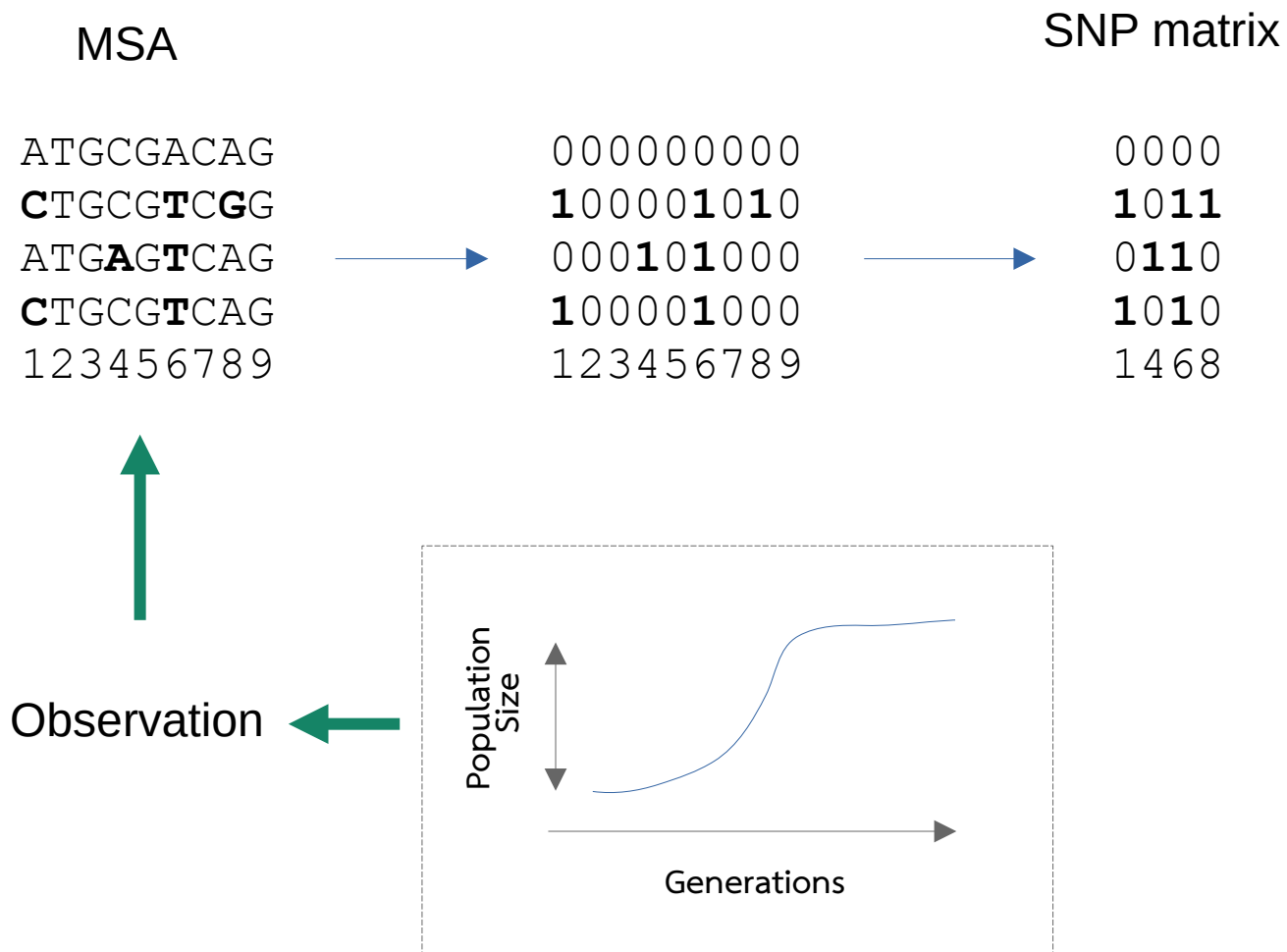
MSA

ATGCGACAG
CTGCG**T**CGG
ATG**A**G**T**CAG
CTGCG**T**CAG
123456789

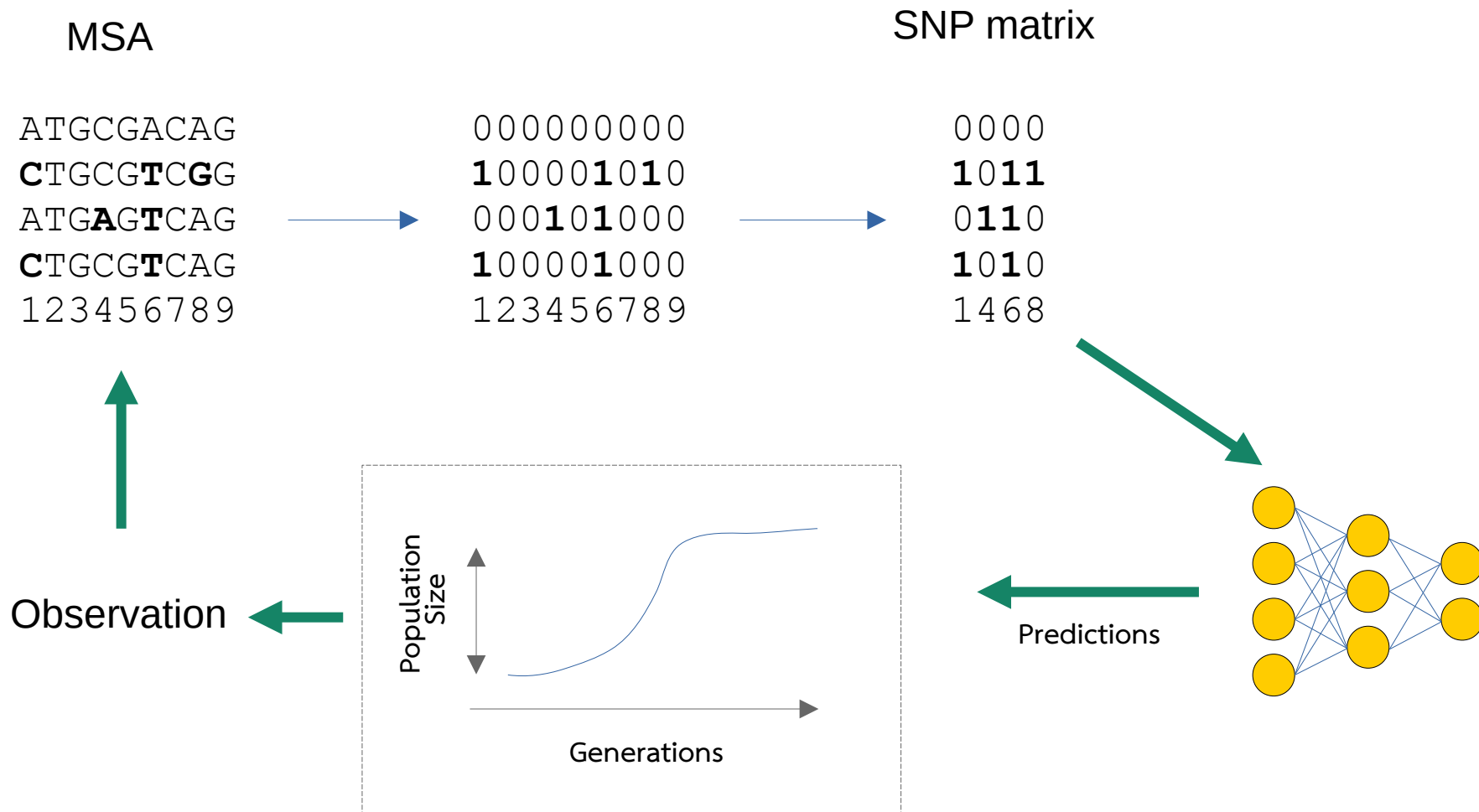
↑
Observation ←



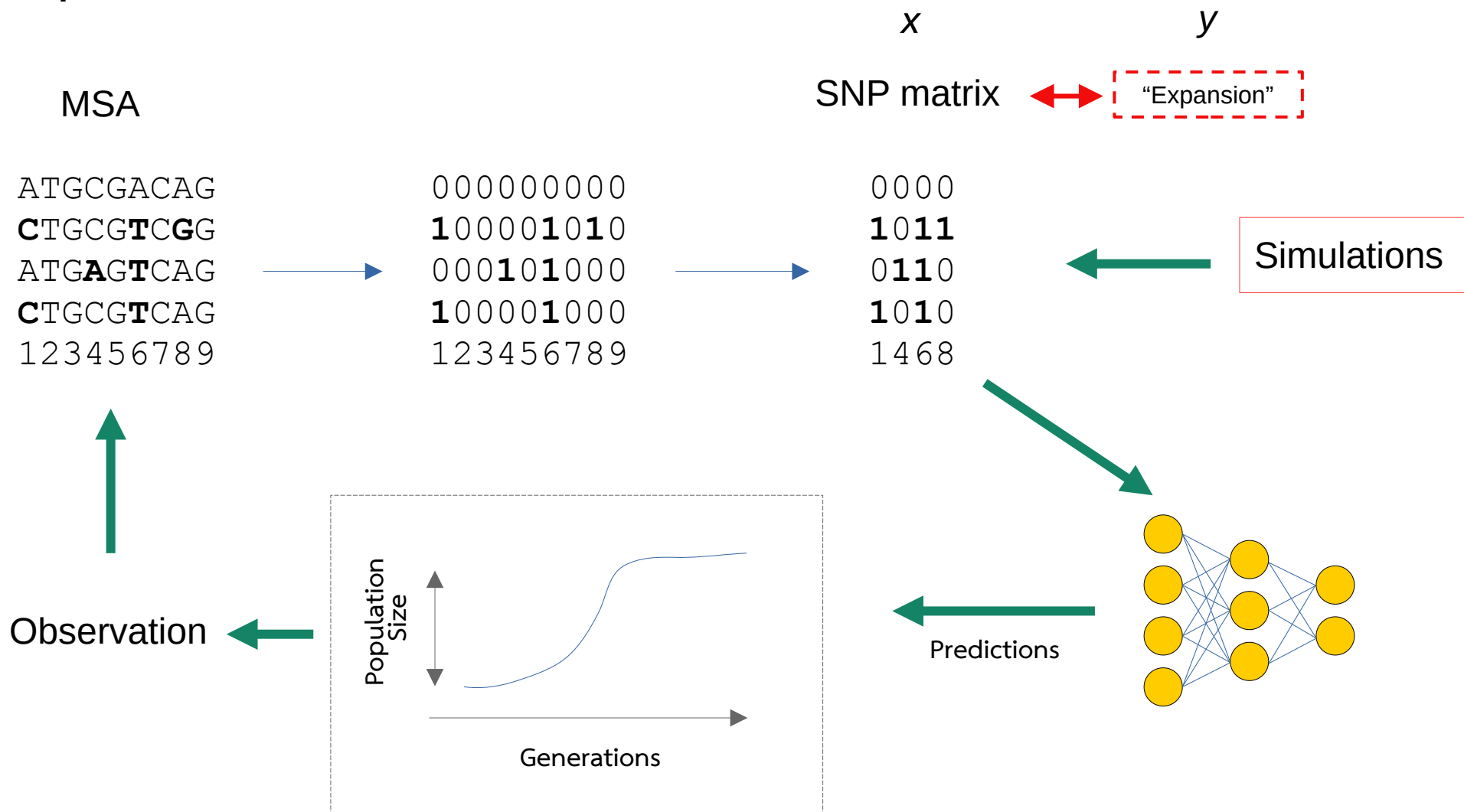
Input Data



Input Data

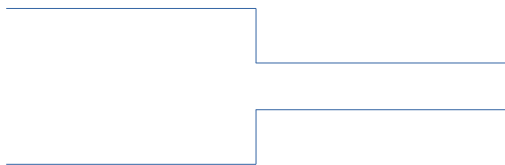


Input Data



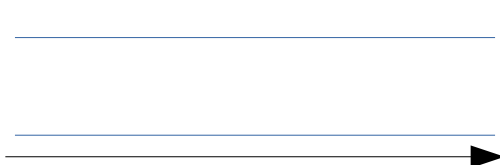
What we simulate

Decline



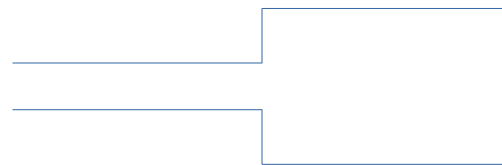
5000x100 simulations

Constant



1000x100 simulations

Expansion



5000x100 simulations

N_e

→ 50% with and 50% without selection

→ Variable parameters:

- **initial population size** ($\sim N_e$)
- mutation rate
- recombination rate (ratio r/m)
- coefficient of selection
- time of selection
- **time of demographical change**
- **strength of bottleneck/expansion**

→ Generated with a generalized Halton sequence

→ Fixed parameters:

- chromosome size
- mean size of gene conversion tracts
- Number of generations

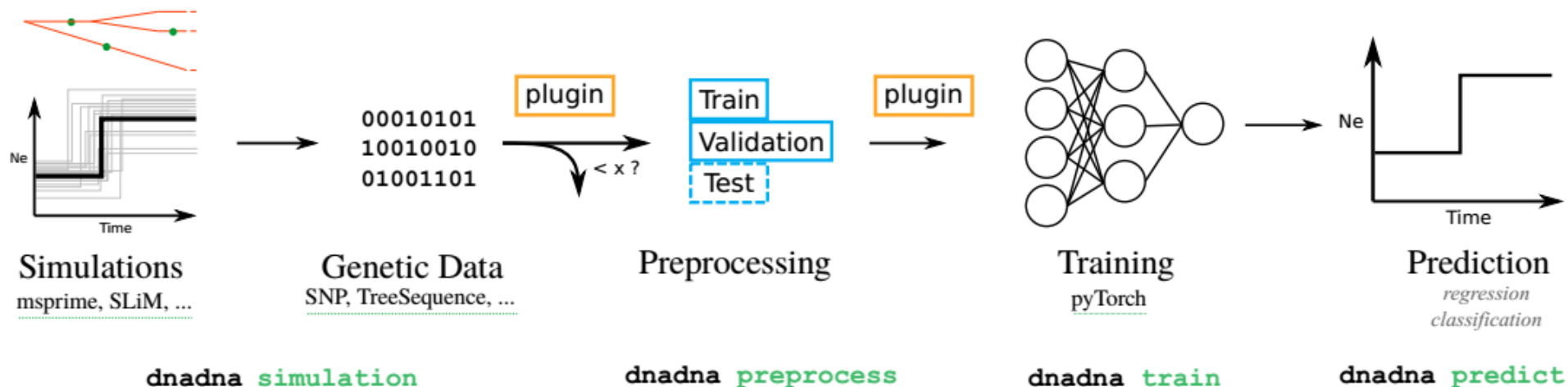
Using SLiM, adapted for
Bacterial population
(Cury et al., 2021)

Approach

- Use of **dnadna**, a package that help to reproduce, share and develop DL methods for population genetics
- Use of **SPIDNA** architecture
 - Invariant to permutation of individuals
 - Adaptive to input dimension
 - Good performance on human populations
- Add **uncertainty** estimation



dnadna : Package for DL in population genetics



Package that allow:

- Development of network
- Reuse of someone else's network
- Reproduce training/prediction

→ Without coding skill (YAML)

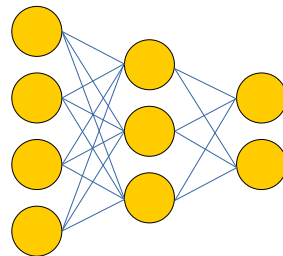


<https://gitlab.com/mlgenetics/dnadna>

Inference of demography

0000
1011
0**11**0
1010
1468

400 first SNPs



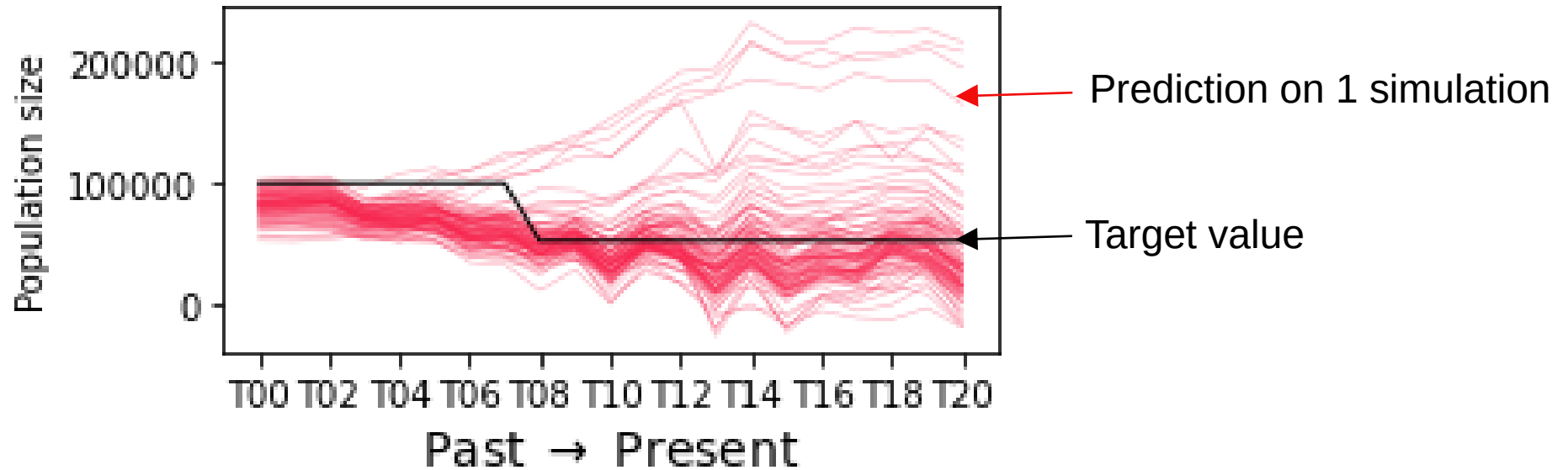
SPIDNA



Population sizes at
21 time steps

Inference of demography

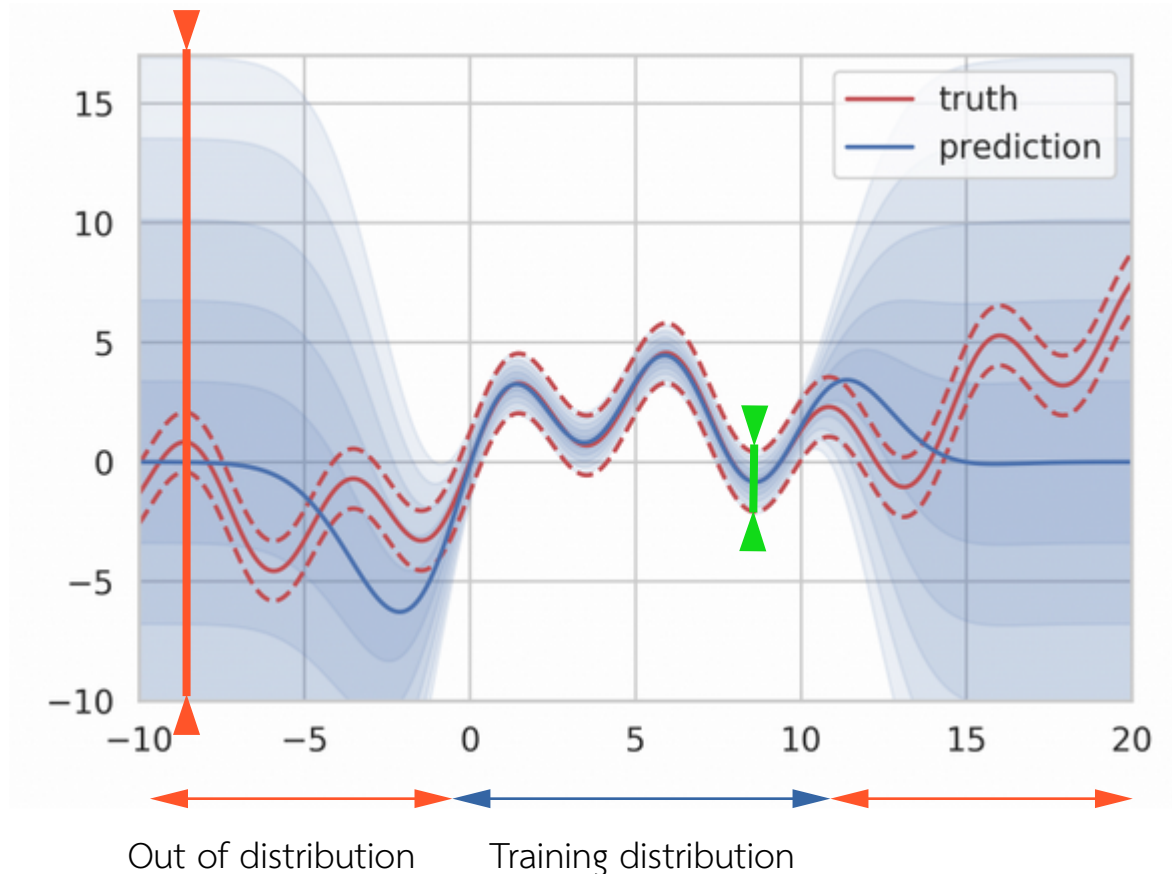
Example with 100 simulations with the same set of parameters



What about uncertainty ?

- DNN output a single value without notion of uncertainty :

- **Aleatoric** : due to the underlying process that is intrinsically stochastic
- **Epistemic** : Your sample is out of the distribution of the simulations



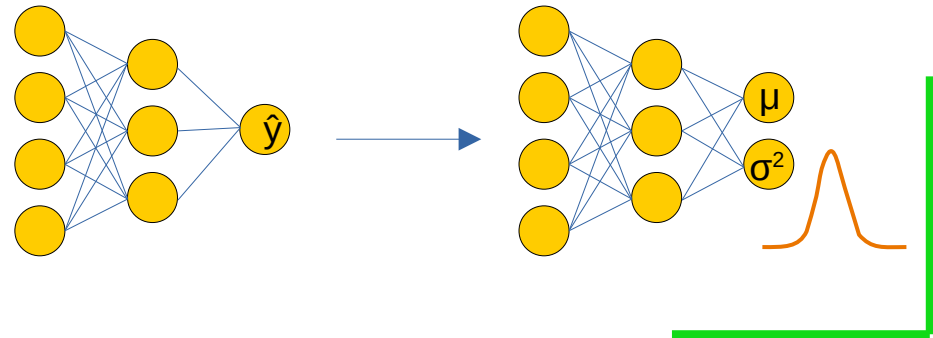
What about uncertainty ?

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- Aleatoric : due to the underlying process that is intrinsically stochastic

- Epistemic : Your sample is out of the distribution of the simulations

Use of Gaussian Negative Log Likelihood Loss to learn a gaussian with parameters μ and σ^2

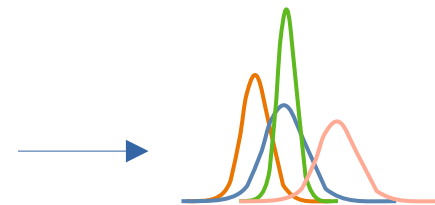
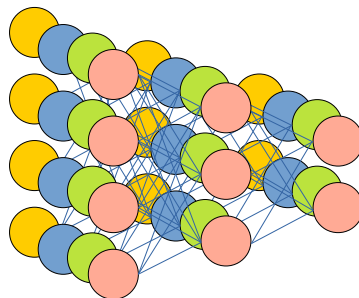


What about uncertainty ?

- DNN output a single value without notion of uncertainty :

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- Epistemic : Your sample is out of the distribution of the simulations

Ensemble of Networks



Weighted mixture of Gaussian distribution

$$\text{weights} \propto 1/\sigma^2$$

Uncertainty estimation

Population size

Without selection

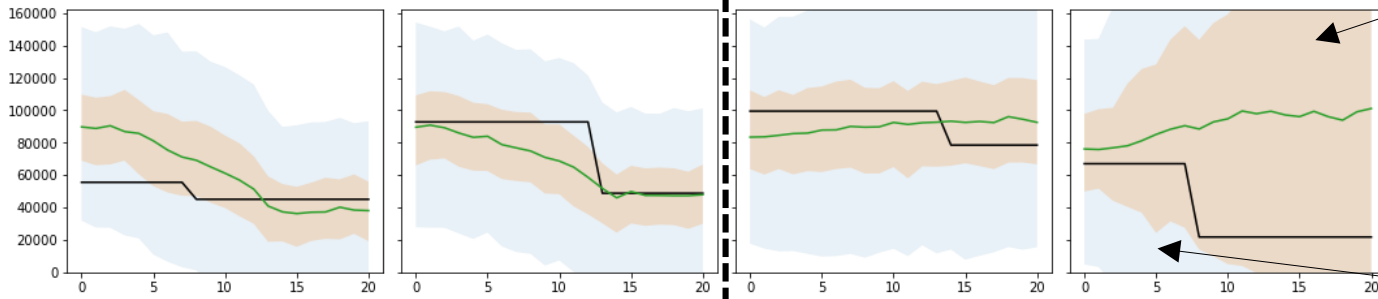
With selection

50% High density Interval

Decline

90% High density Interval

Time



Uncertainty estimation

Population size

Without selection

With selection

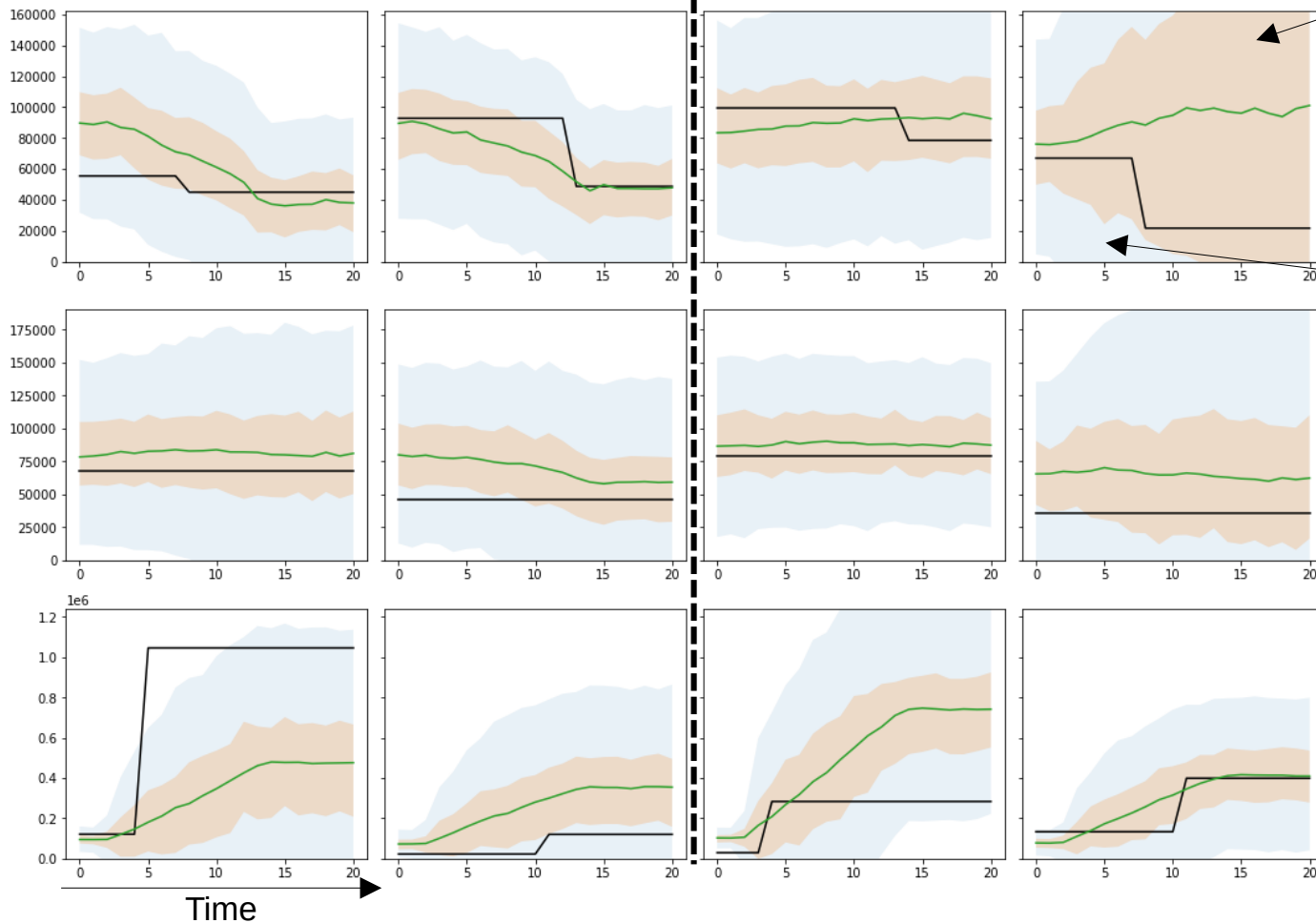
50% High density Interval

Decline

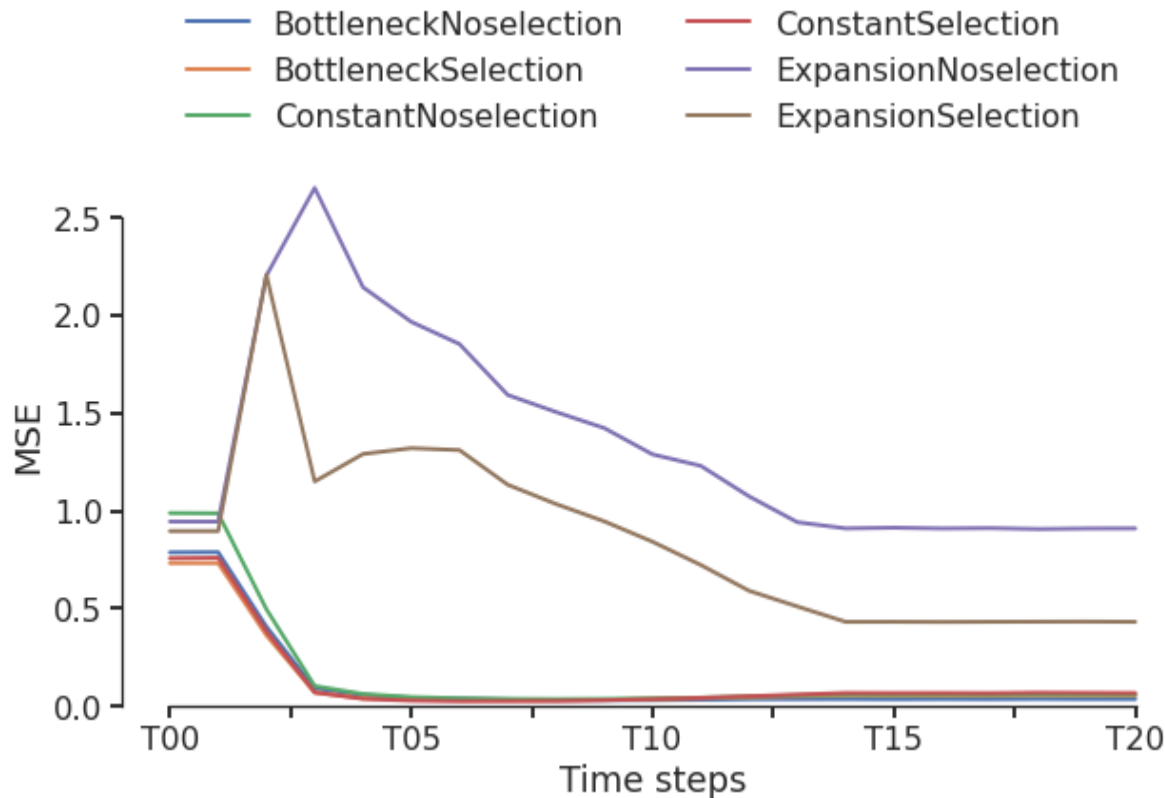
90% High density Interval

Constant

Expansion



Error on Test set (the lower the better)

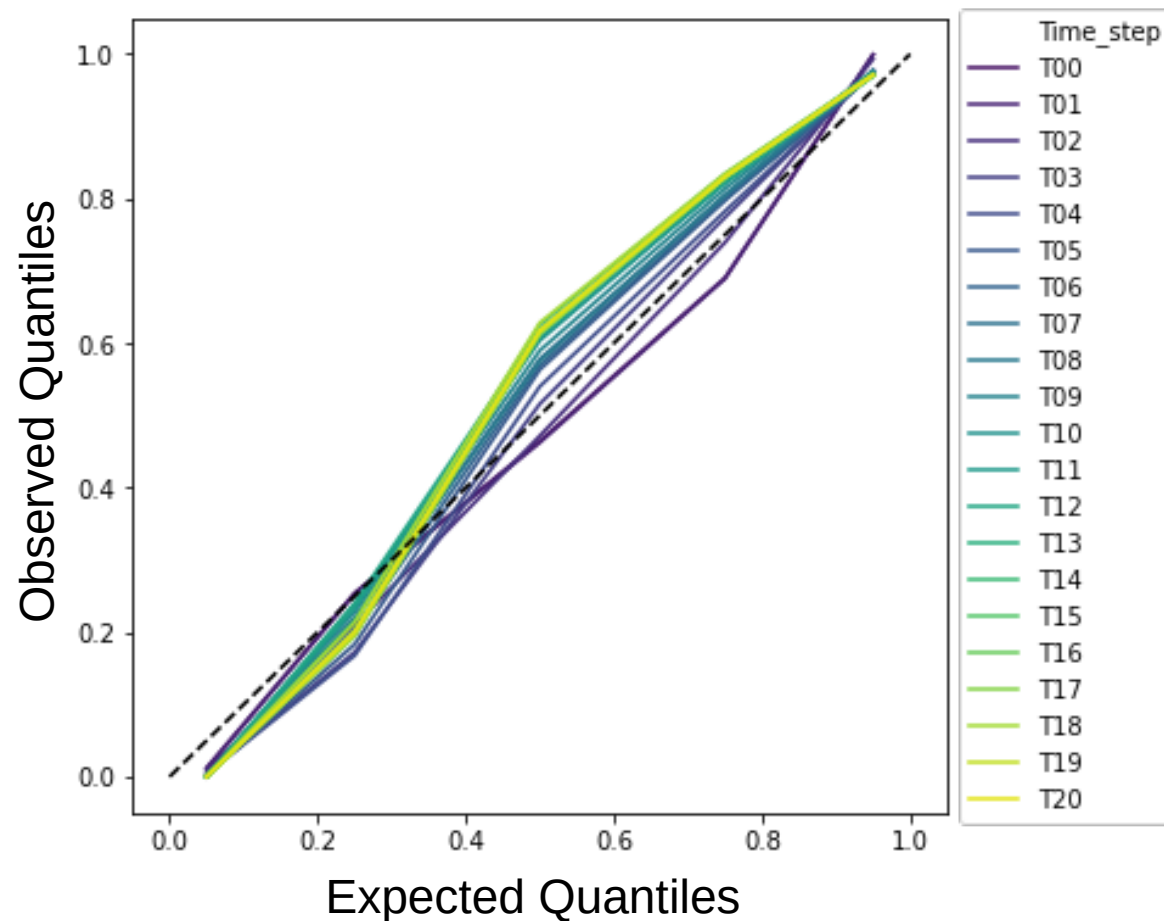


Bad predictions for
Expansion

Good prediction otherwise

Except for ancient times
where predictions follows
the prior of the training set.

Calibration of the Gaussian mixture

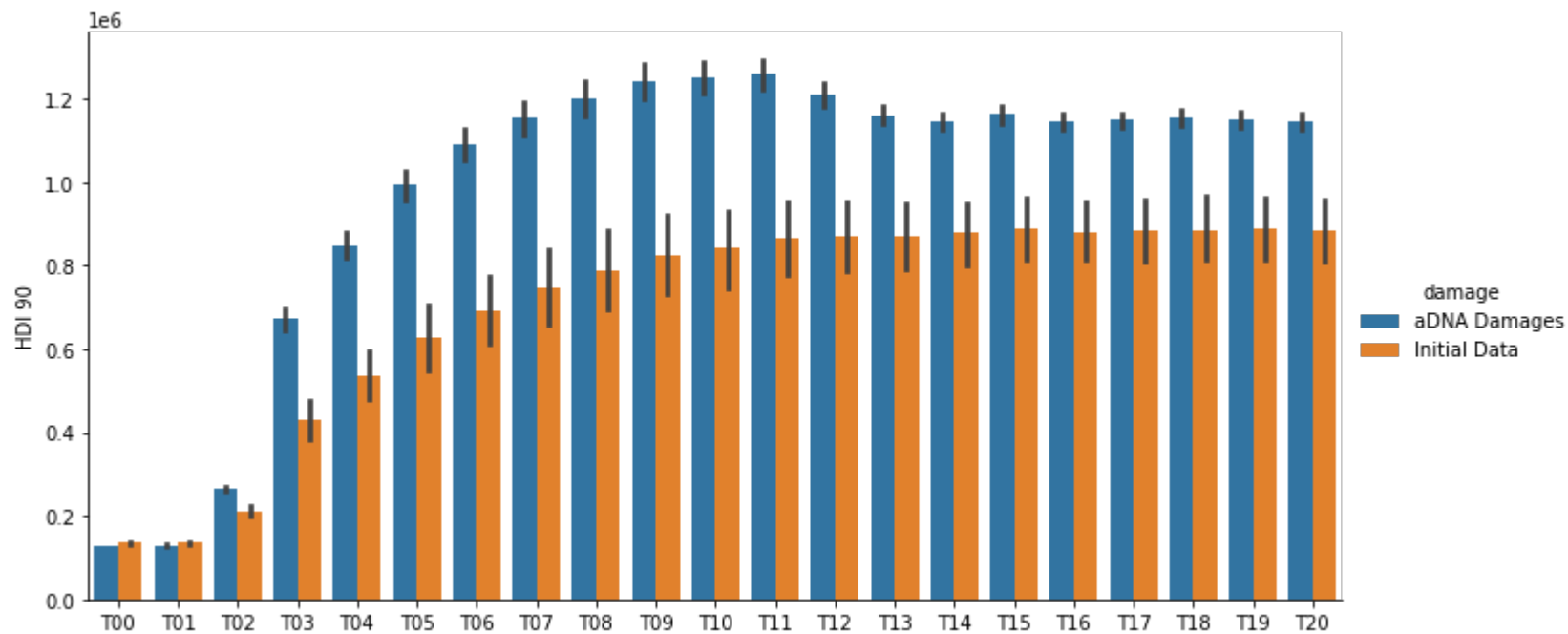


ancient DNA

- Increasing amount of aDNA sequenced as technology improves
- Can help palaeontologist / historian understand distant past
- Problem : low quality of sequences
 - Due to degradation of DNA
 - Higher rate of sequencing error
 - Poor coverage (small amount of DNA)



Uncertainty on ancient DNA



Summary

- Prediction of bacterial population size through time
 - Irrespective of the underlying selection regime and other parameters
- Using **dnadna** package → easy to reuse / reproduce
- Estimation of the aleatoric and epistemic uncertainties

- Transfer learning with aDNA
- Assess interest of transfer learning from other net trained on similar task
- Improve training procedure with SPIDNA (something else than 400 SNP)
- Test on real data

Thanks

- Flora Jay
- Theophile Sanchez
- Guillaume Charpiat
- Erik M. Bray
- Ben Haller

- Jazeps Medina-Tretmanis
- Maria Avila-Arcos
- Emilia Huerta-Sanchez
- Mathieu Michel

