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# Artificial Regulatory Network Evolution

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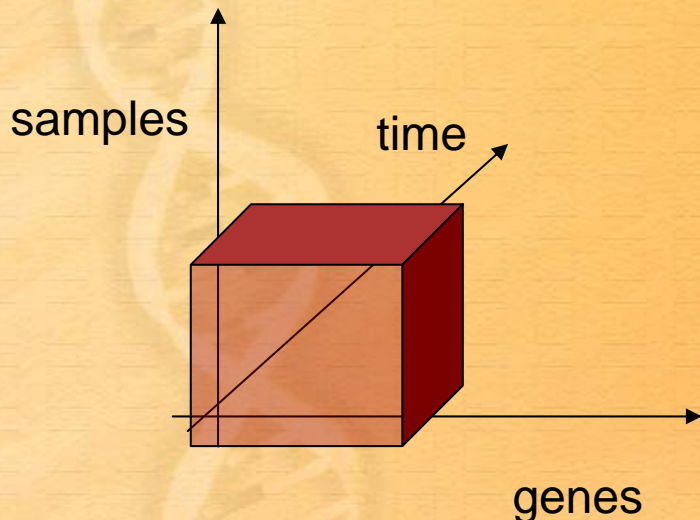


# Context: from data to knowledge

- Large kinetic transcriptome data sets are

announced

- Genetic Network (GN) inference



- We need to design NOW the related data mining algorithms

# Problems

- Just a few real data sets are available
- Today, benchmarking is performed on:
  - Randomly generated data
  - Synthetic data w.r.t. models from other fields
  - Data from GN generators biased by topology

# Approach

- Can we use simulation to build biologically plausible GNs and thus more relevant kinetic data sets?
- GNs are built by an evolutionary process
- We propose to use artificial evolution to generate plausible GNs

# Biologically plausible GN

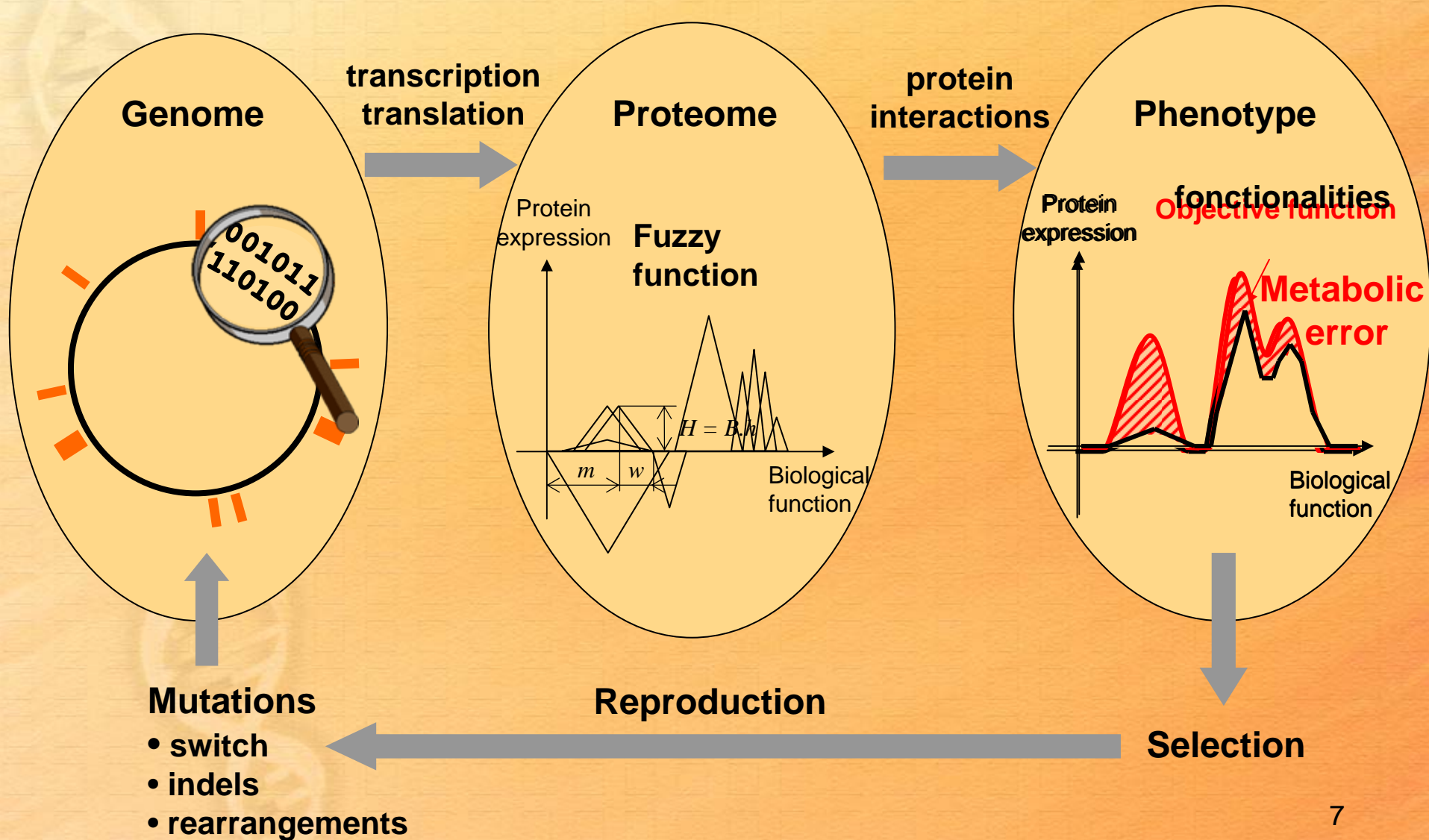
- To obtain plausible GNs we must respect biological bases of network evolution:
  - GNs are derived from a genome sequence and a proteome component
  - Mutation of the genetic sequence
  - Selection on the phenotype

We have developed the **RAevol Model**

# Based on the Aevol\* Model

- Studying robustness and evolvability in artificial organisms:
  - Artificial genome, non-coding sequences, variable number of genes
  - Genome: circular double-strand binary string
  - Mutation/selection process

# Ævol : Artificial Evolution



# From Aevol to RAevol

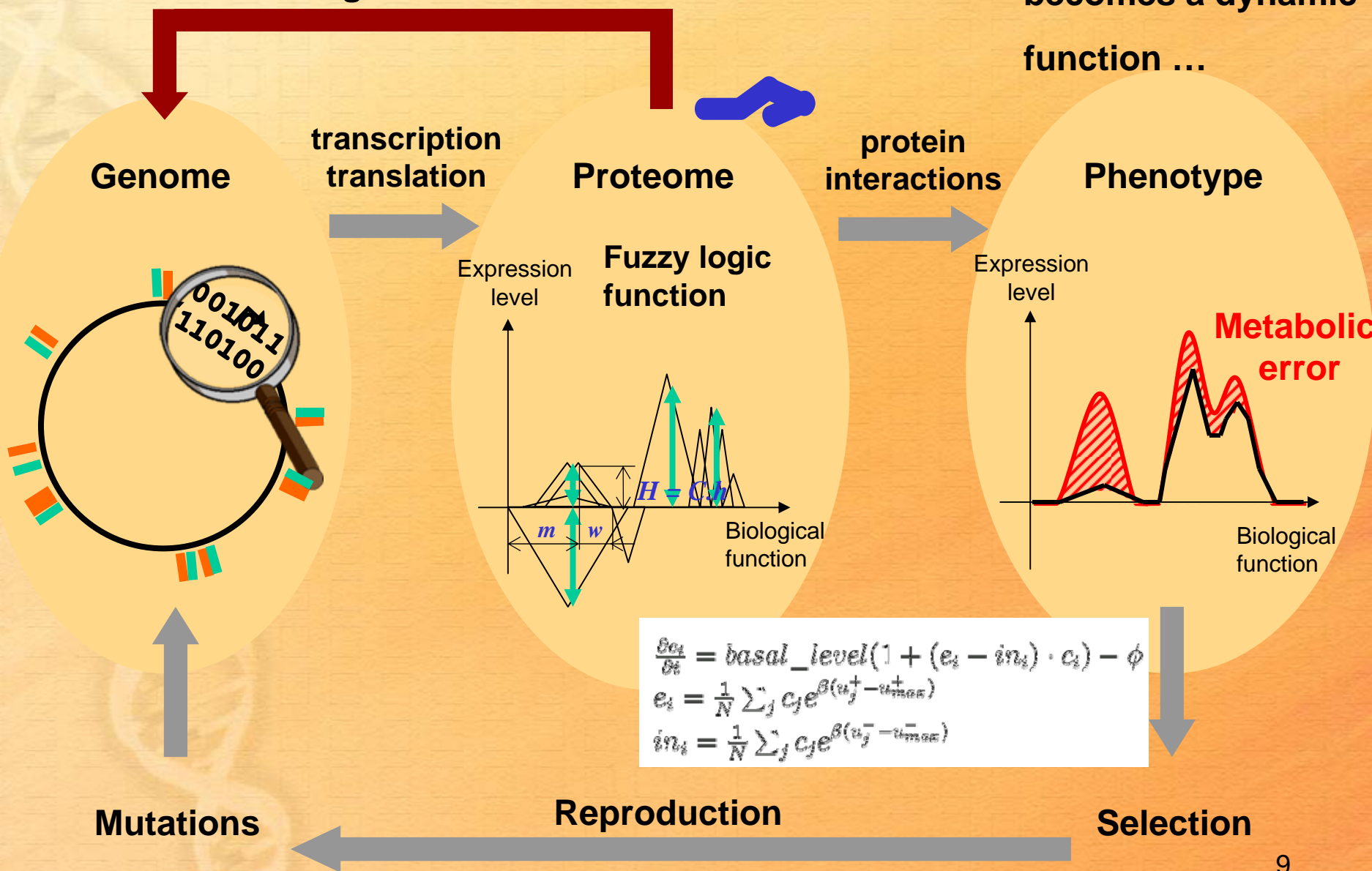
- Interesting properties of Aevol to understand genome evolution:
  - See **C. Knibbe**, A long-term evolutionary pressure on the amount of non-coding DNA (2007). *Molecular Biology and Evolution*, in press. doi: 10.1093/molbev/msm165
- We need to add a regulatory process  
→ RAevol



# RÆvol : Artificial Evolution

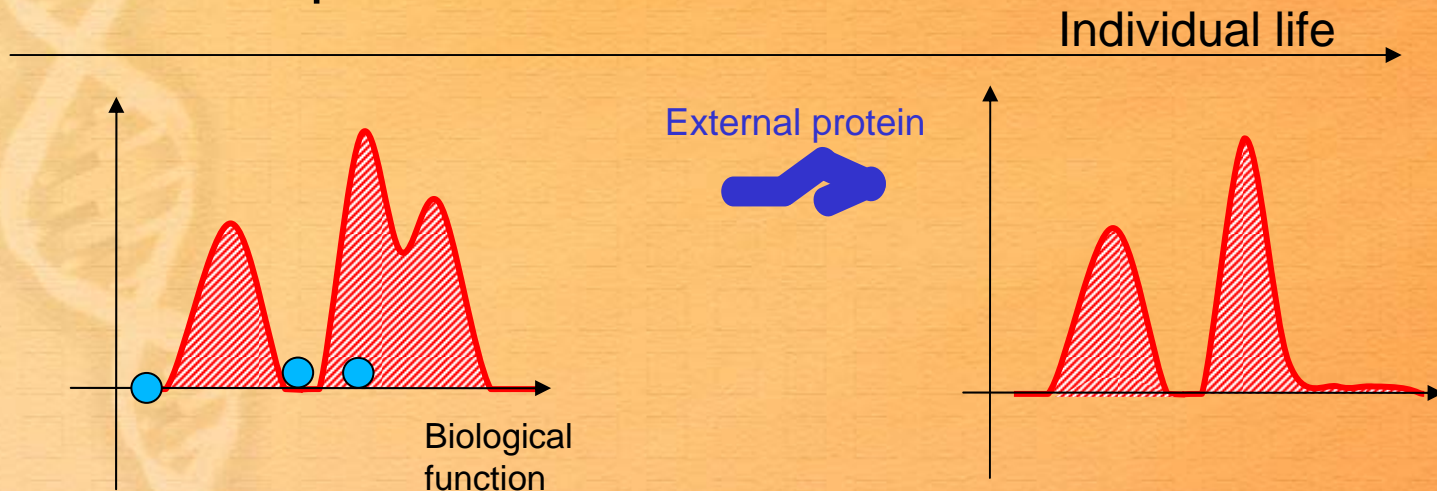
## Regulation

The phenotype becomes a dynamic function ...



# Experimental setup

- **Simulations:** 1000 individuals, mutation rate  $1 \cdot 10^{-5}$ , 15000 generations
- Organisms must perform 3 metabolic functions
- The incoming of an external signal (protein) triggers an inhibition process



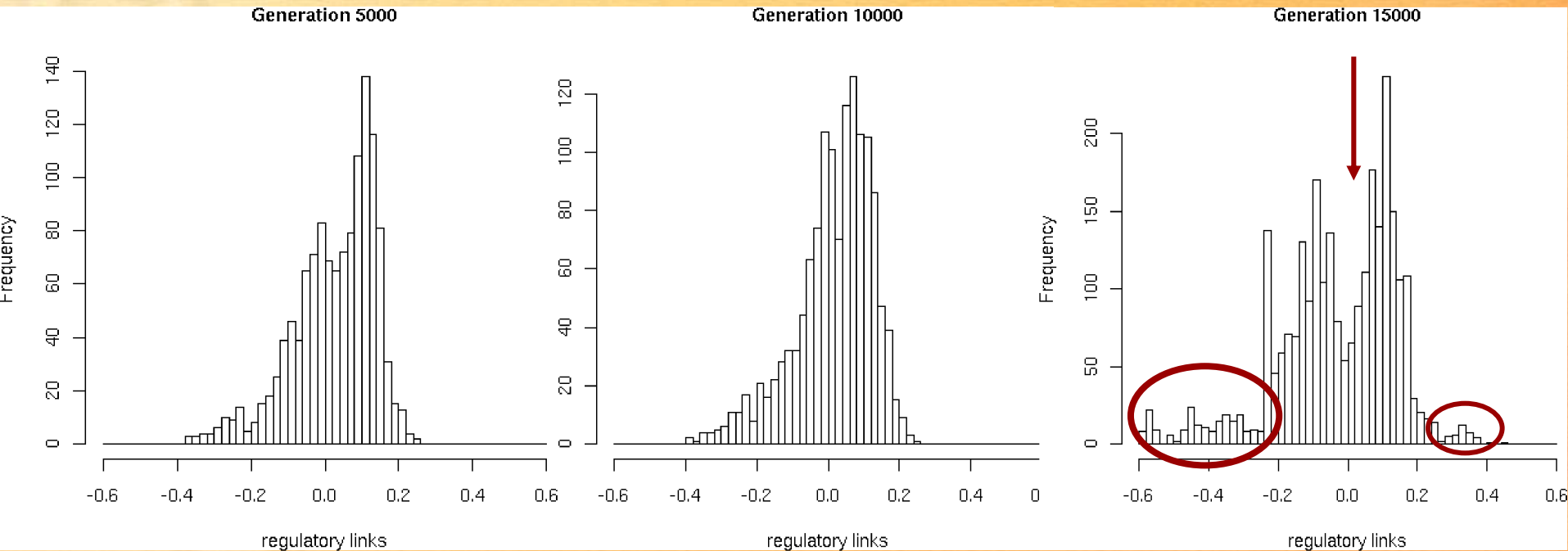
# First results

Generation	0	1000	5000	10000	15000
Metabolic error	0.1377	0.0616	0.0206	0.0171	0.0161
Genome size (in kilobase pairs)	5.0	21.3	15.3	12.3	12.8
Nb. of nodes in the metabolic network	1	24	34	36	37
Nb. of nodes in the genetic network	3	24	35	37	45
Nb. of transcription factors	2	0	1	1	8
Nb. of links	3	576	1223	1332	2601
positive links	2	508	771	827	1276
negative links	1	68	452	505	1325

- The metabolic network mainly grows during the 5000 first generations → GN grows likely
- Transcription factors appear after 10000 generations  
→ GN grows independently from metabolism

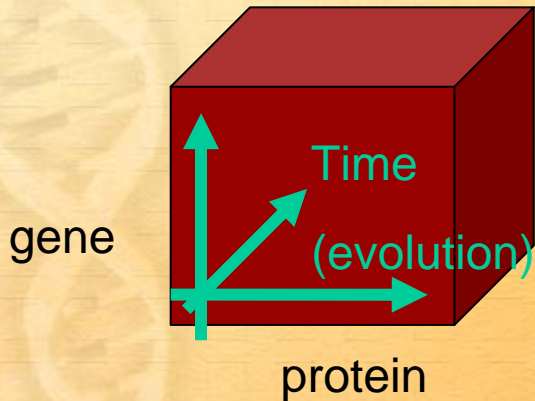
# First results

## Regulatory Links Values

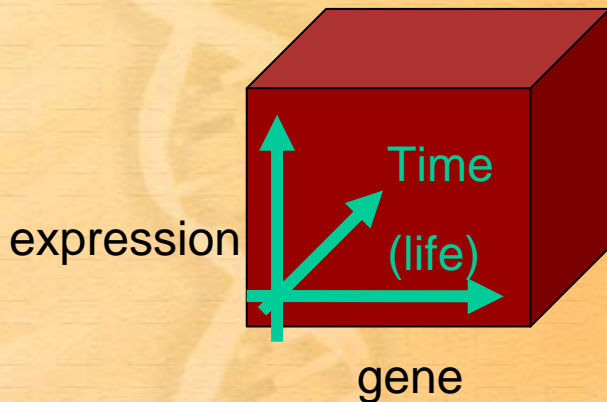


- First phase: quasi-normal distribution
- Second phase: multimodal distribution, strong links (mainly inhibitory) ...

# Conclusion and perspectives



- RAevol generates plausible GNs (protein-gene expression levels) along evolution



- Studying the generation of kinetic transcriptome data sets is ongoing

- Towards more realistic benchmarks for data mining algorithms

# Open issues

- Systematic experiments
  - effect of mutation rates
  - effect of environment stability
- Study the network topology
  - Compare the network topology with real organisms...
  - Do frequent motifs/modules appear in the network ?

# The Aevol Model

- **Interesting properties** of the Aevol Model:
  - Transcription/translation process → Different RNA production levels
  - Explicit (abstract) proteome → interactions between proteins and genetic sequence
  - Variable gene number → Variable network size
  - Complex mutational process (mutations, InDel, rearrangements, ...) → Different topology emergence