



NETADIS
Statistical Physics Approaches
to
Networks Across Disciplines



MID-TERM REVIEW

Institut Henri Poincaré, Paris

23-24 January 2014

Approximate inference for stochastic dynamics in large biological networks

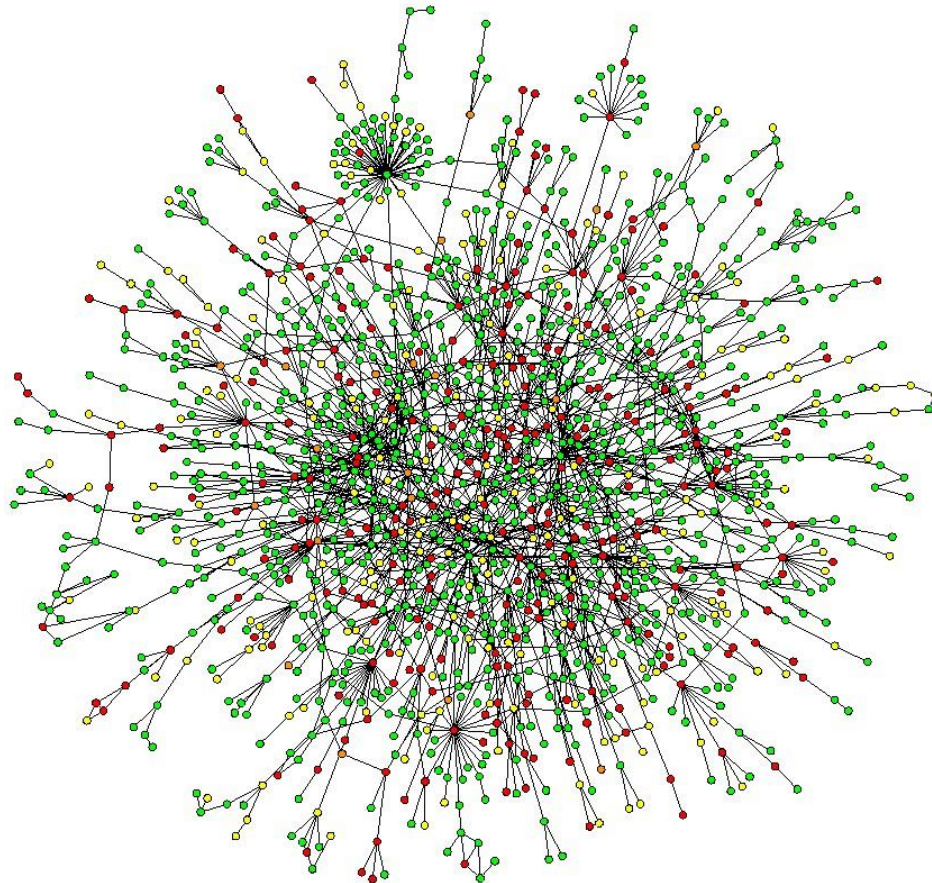
Ludovica Bachschmid Romano

Supervisor: Prof. Manfred Opper

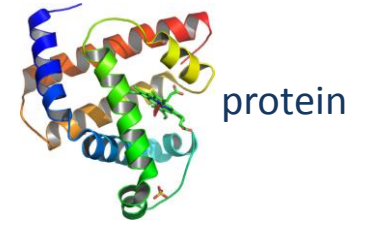
Artificial Intelligence group, TU Berlin



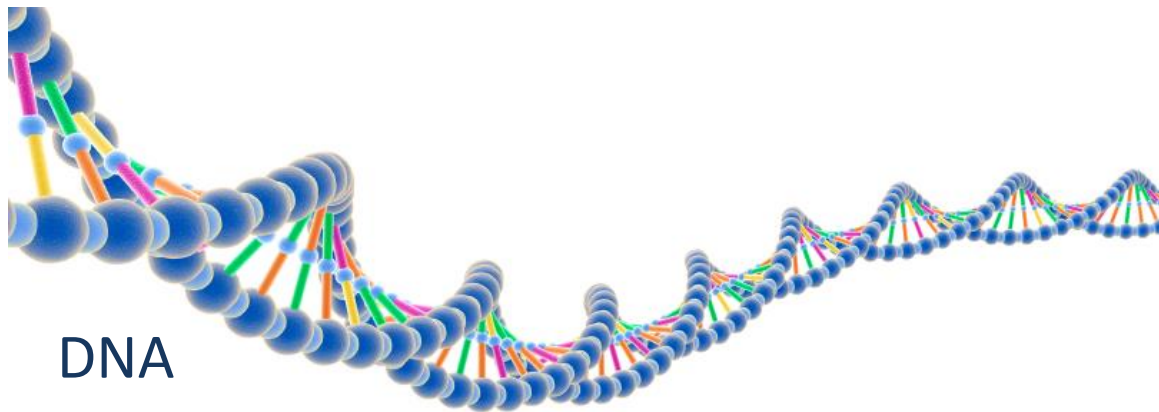
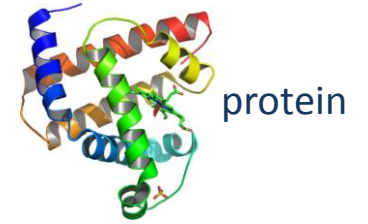
The functions of most biological systems arise from complex interactions between the numerous system's components



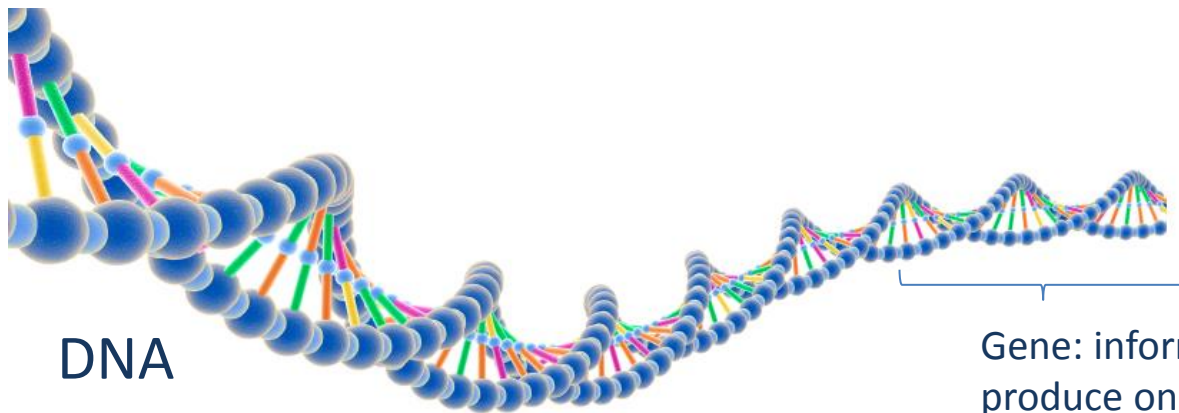
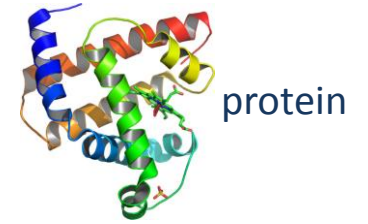
Gene regulatory network



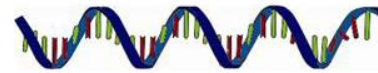
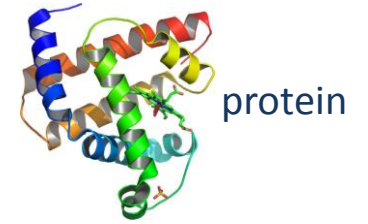
Gene regulatory network



Gene regulatory network



Gene regulatory network



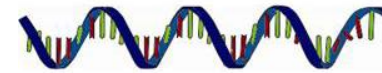
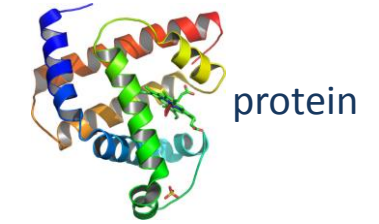
RNA: copy of the genetic information

TRANSCRIPTION

DNA

Gene: information needed to produce one kind of proteins

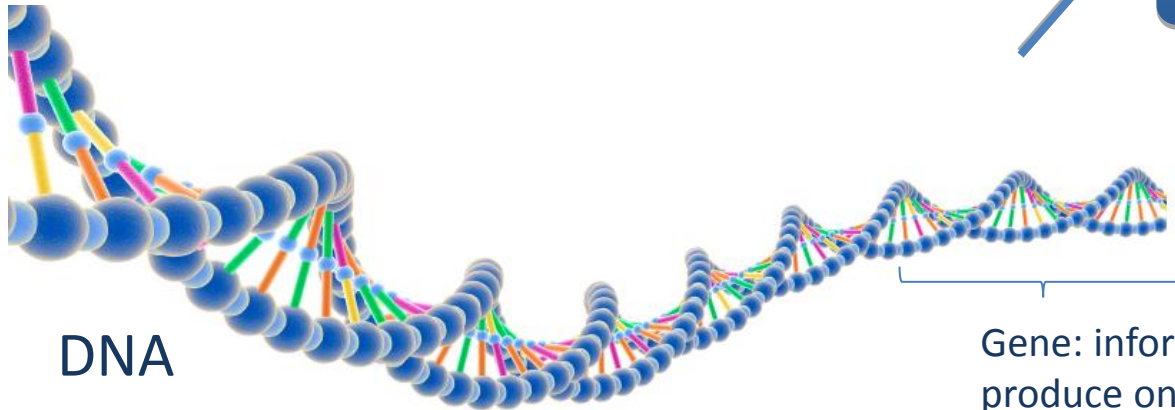
Gene regulatory network



RNA: copy of the genetic information



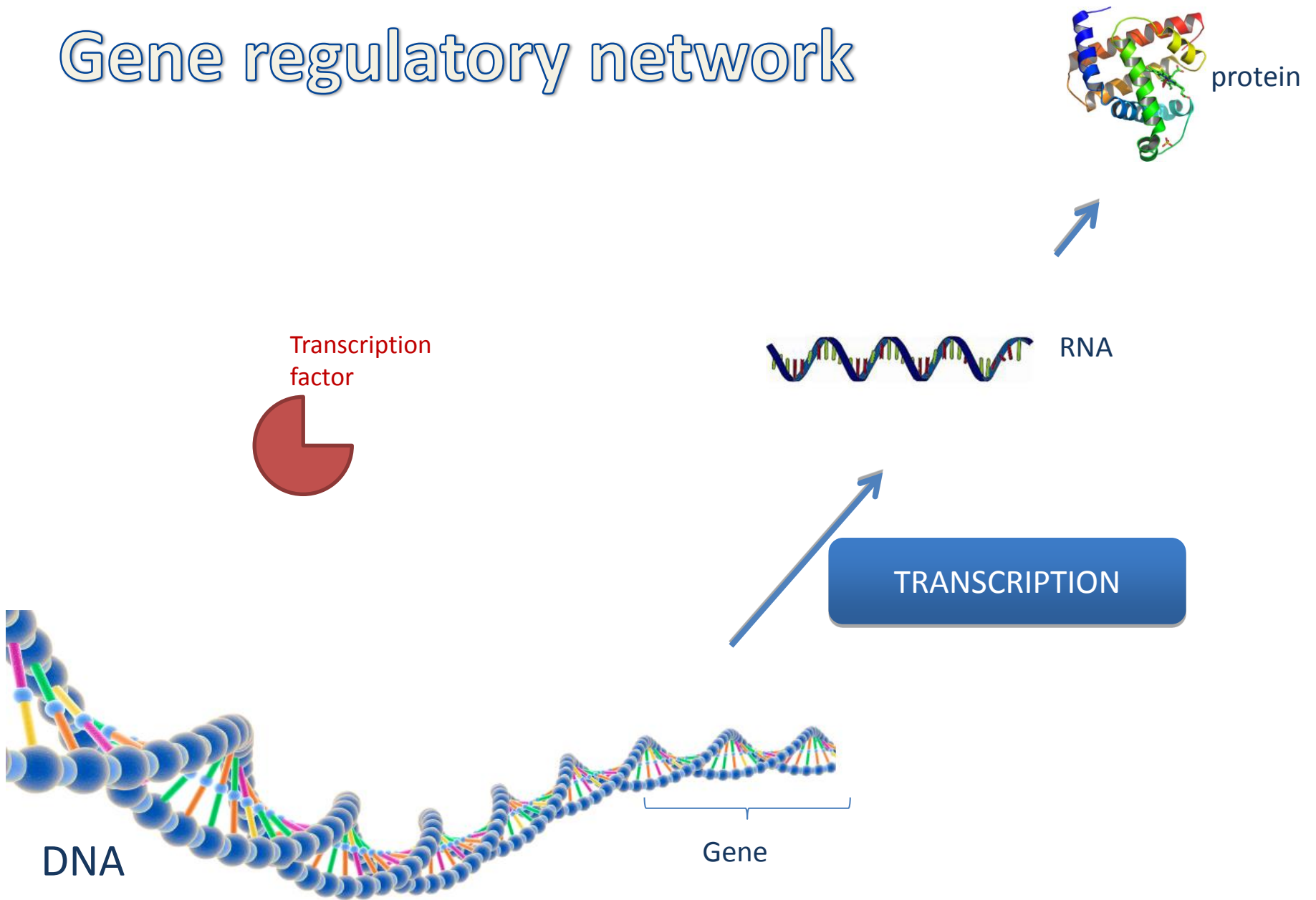
TRANSCRIPTION



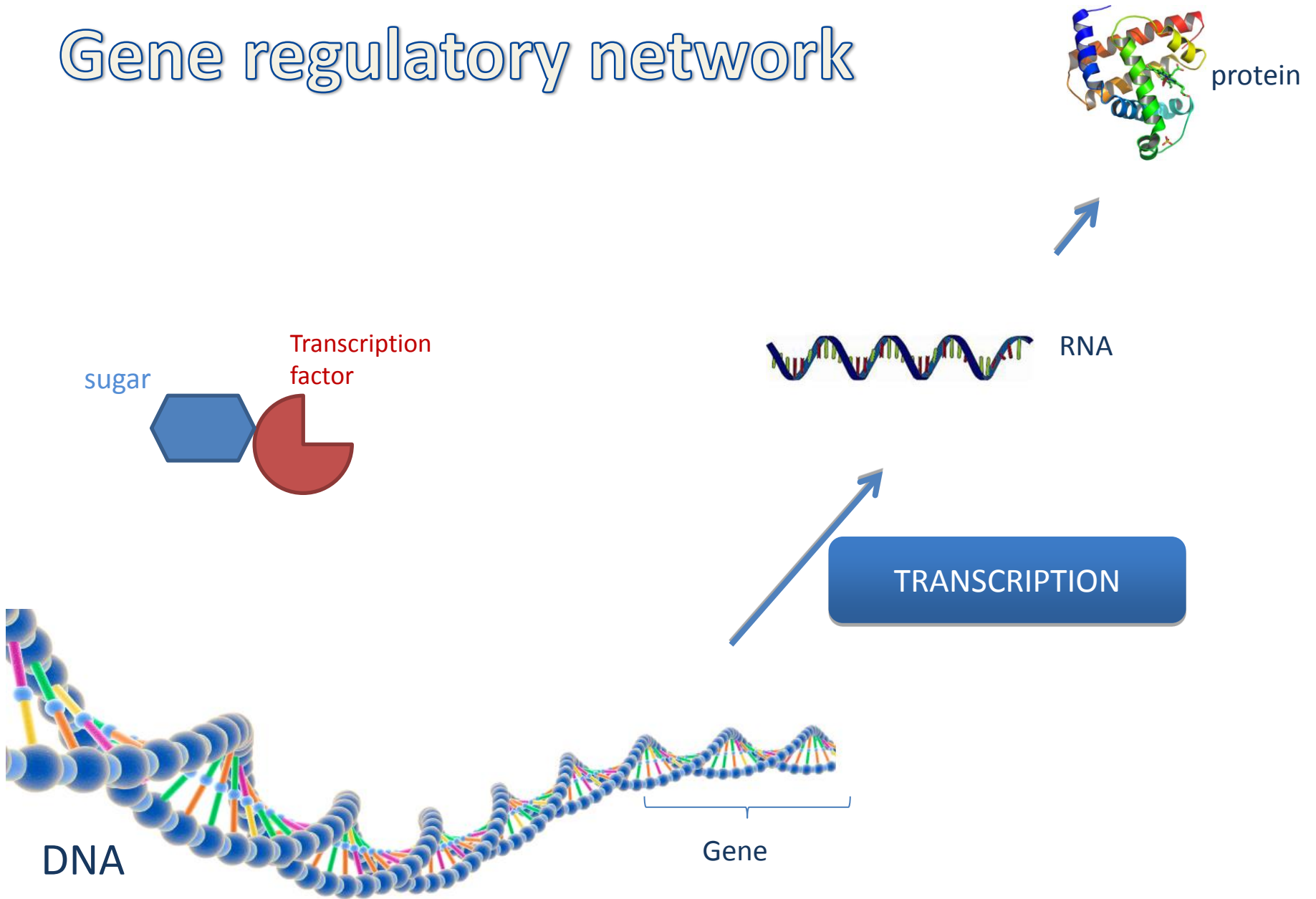
DNA

Gene: information needed to produce one kind of proteins

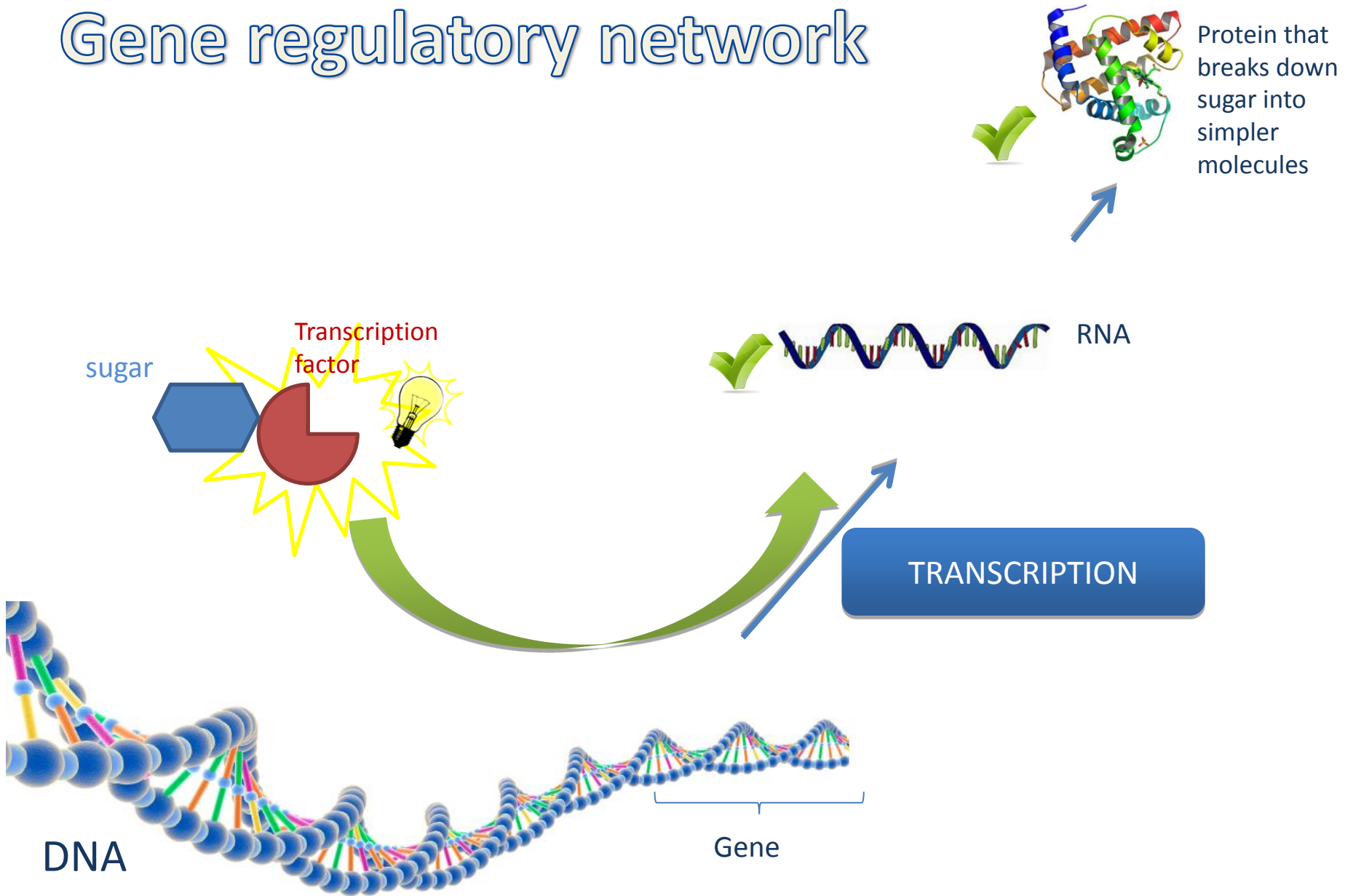
Gene regulatory network



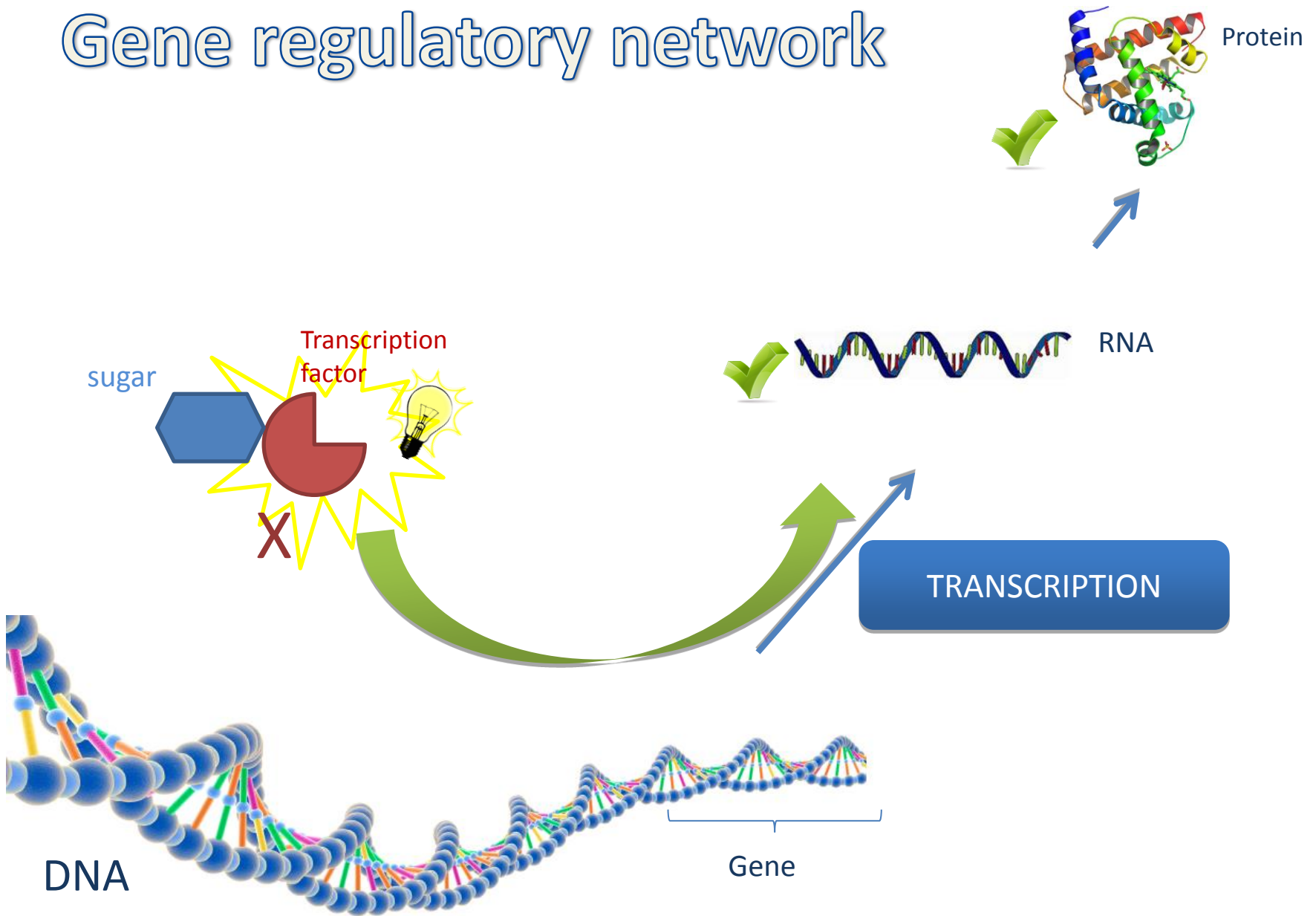
Gene regulatory network



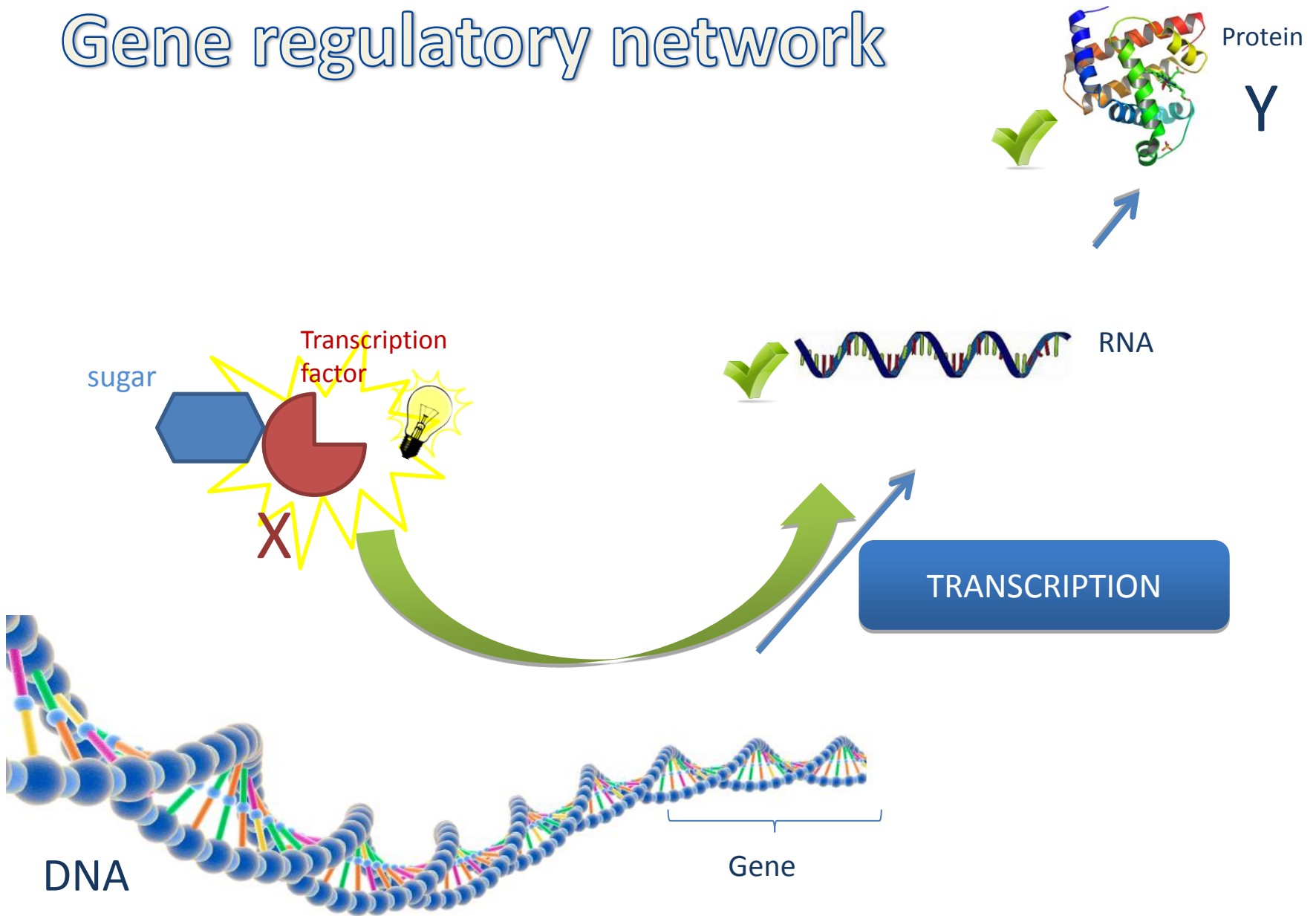
Gene regulatory network



Gene regulatory network

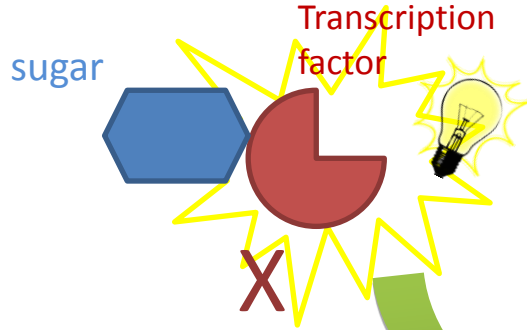
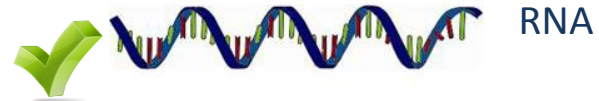
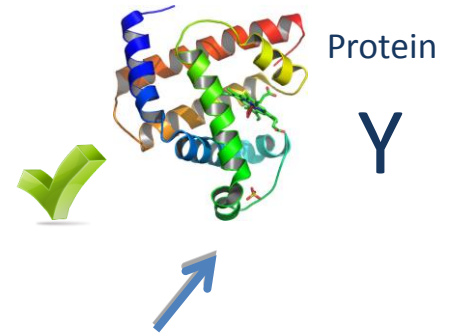


Gene regulatory network



Gene regulatory network

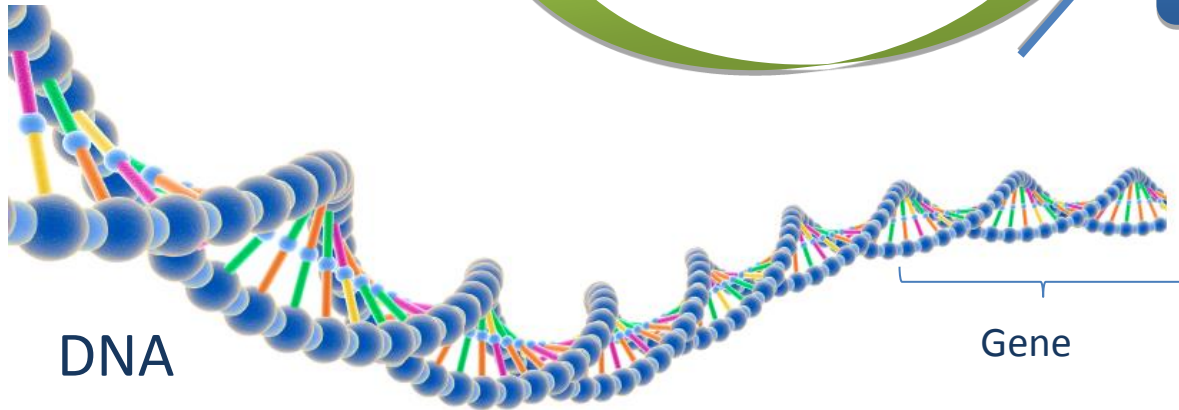
X increases the production of Y



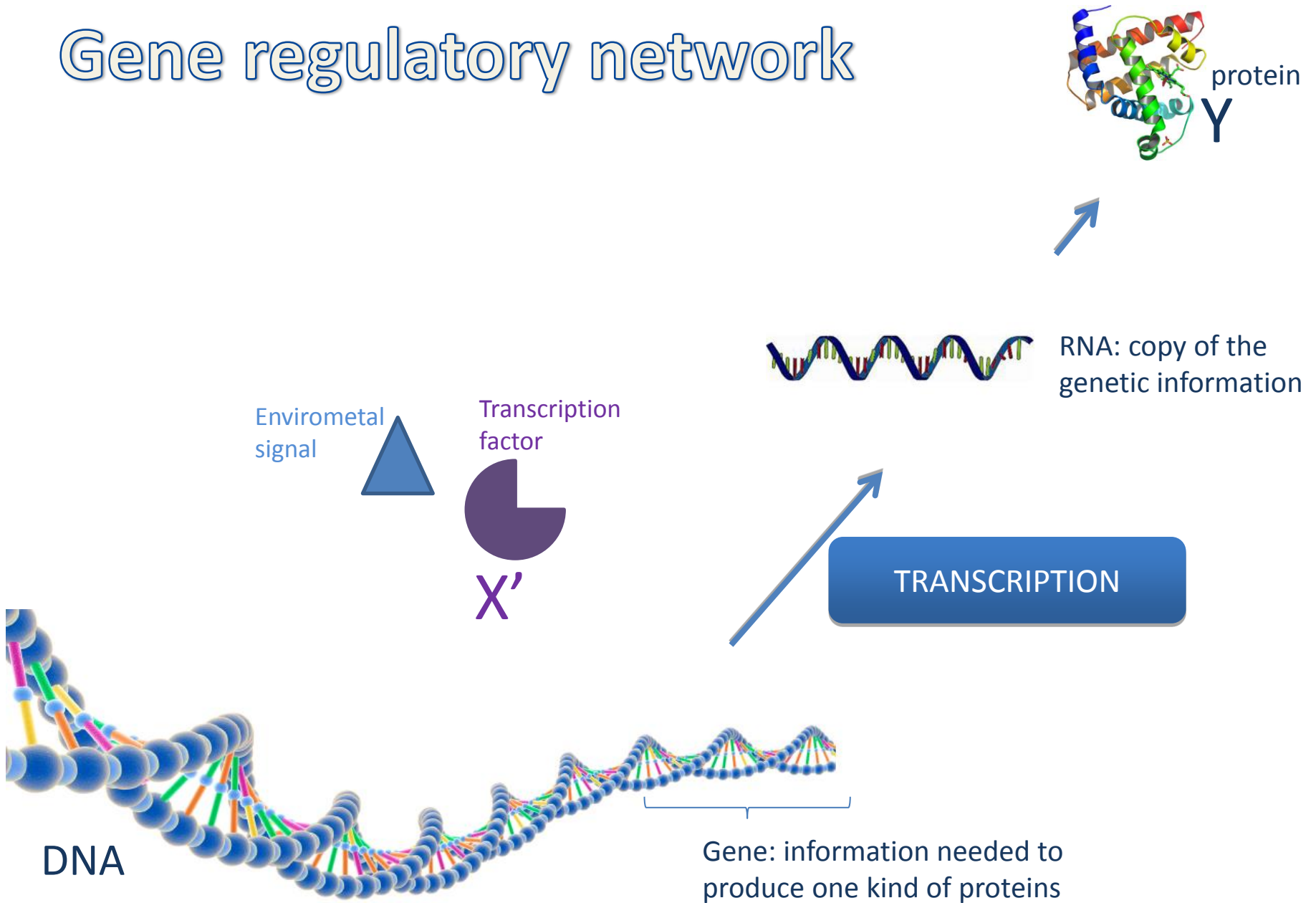
TRANSCRIPTION

DNA

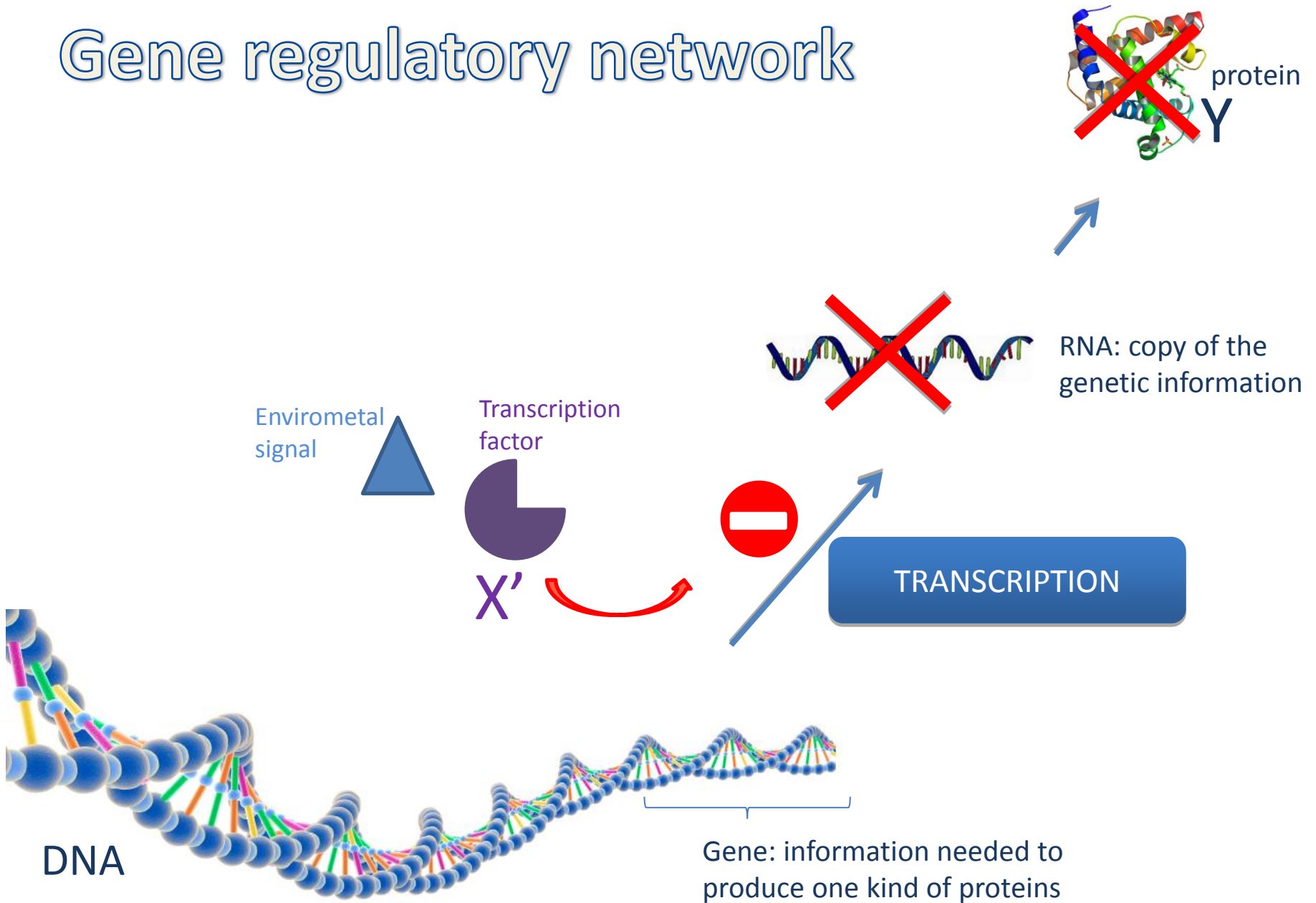
Gene



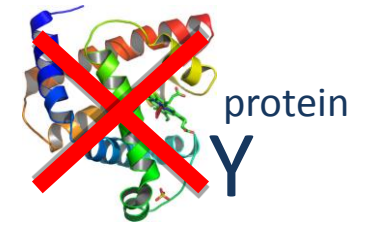
Gene regulatory network



Gene regulatory network



Gene regulatory network

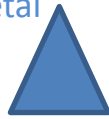


RNA: copy of the genetic information



X' decreases the production of Y

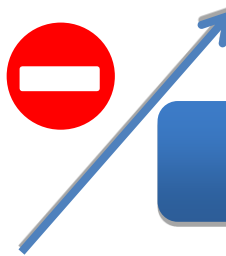
Envirometal signal



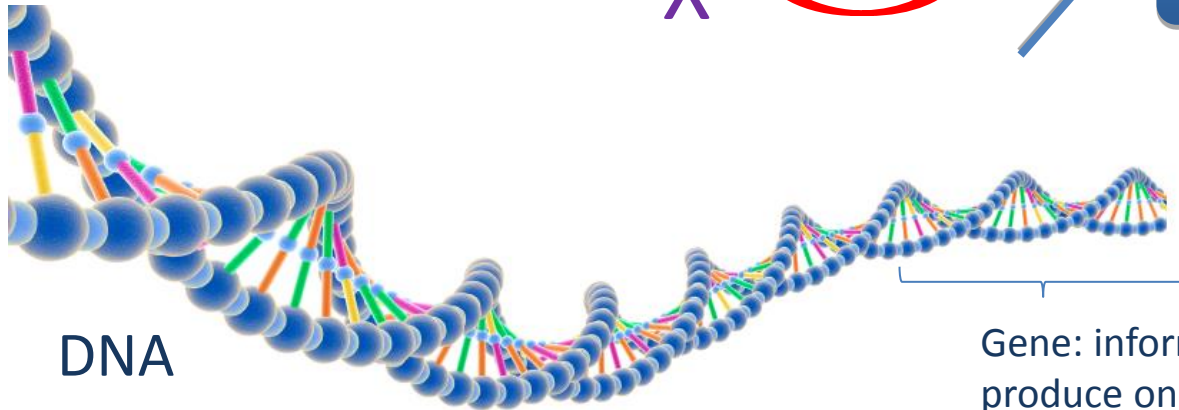
Transcription factor



X'



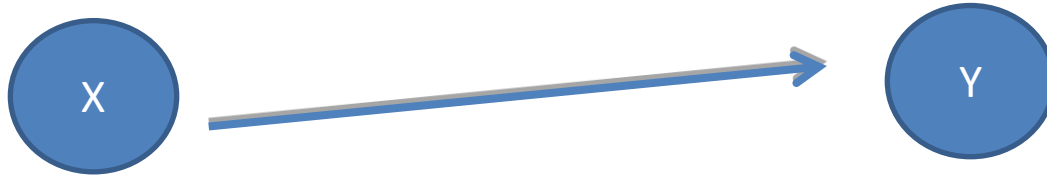
TRANSCRIPTION



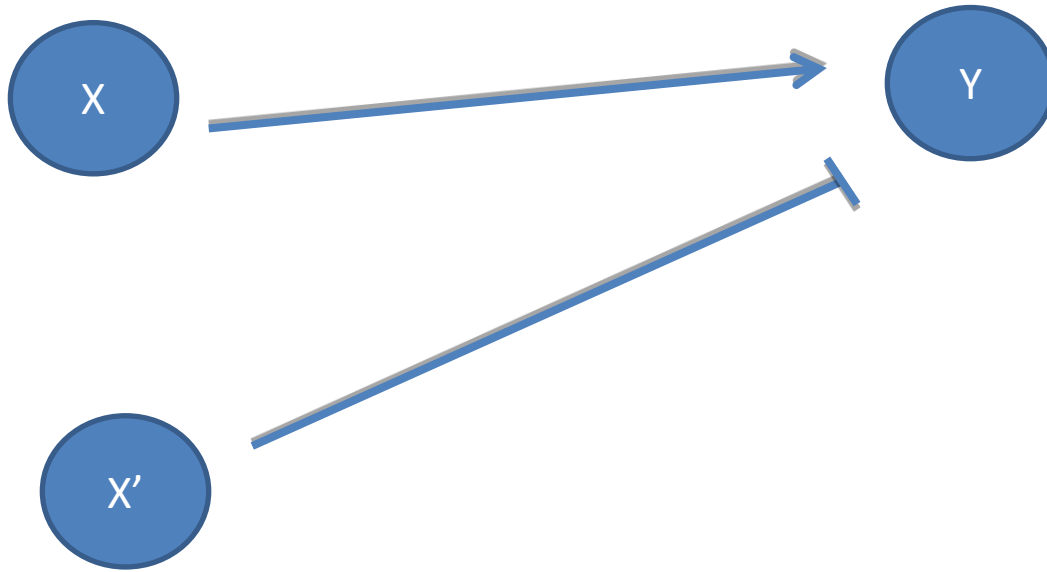
DNA

Gene: information needed to produce one kind of proteins

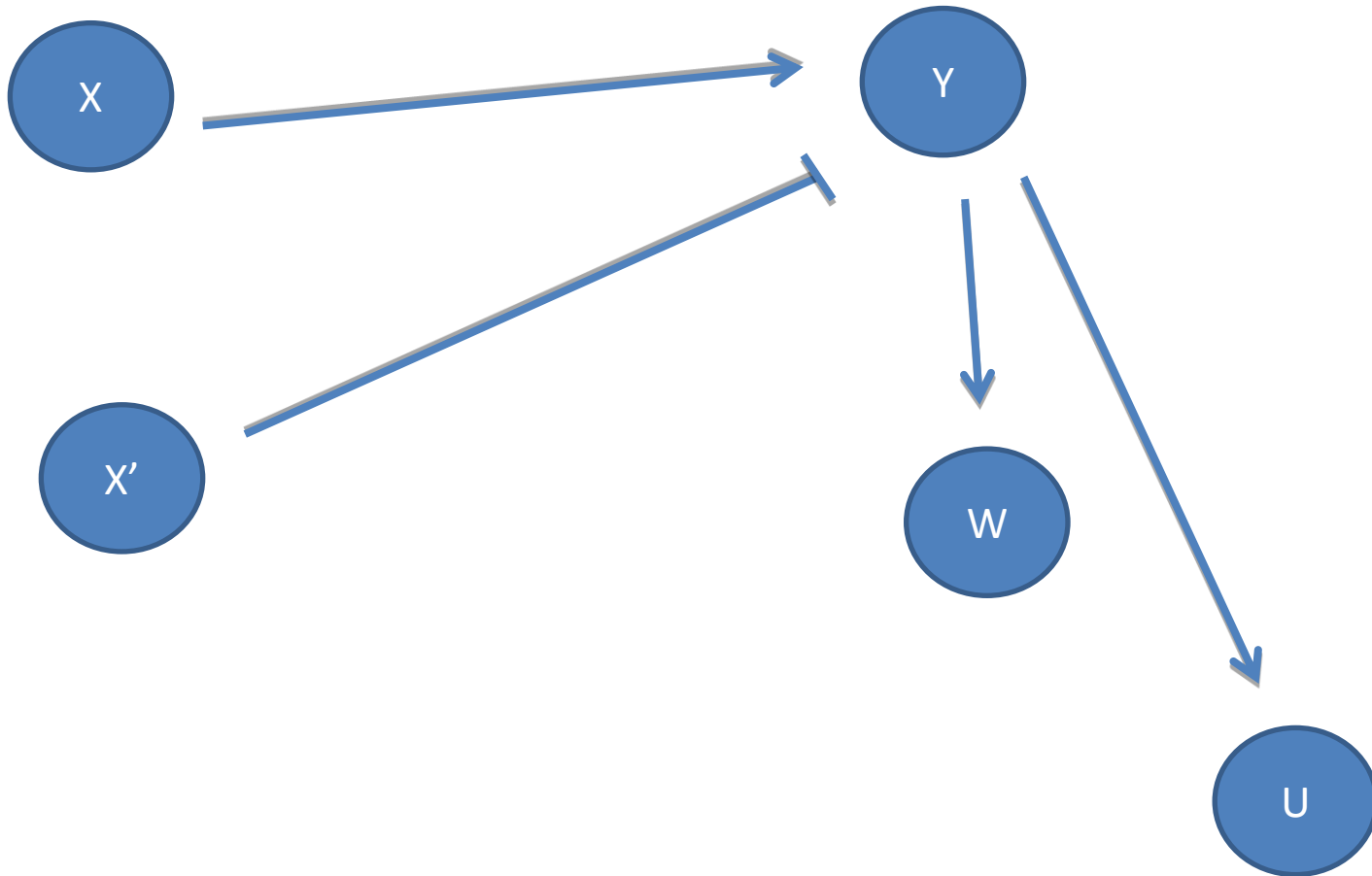
Gene regulatory network



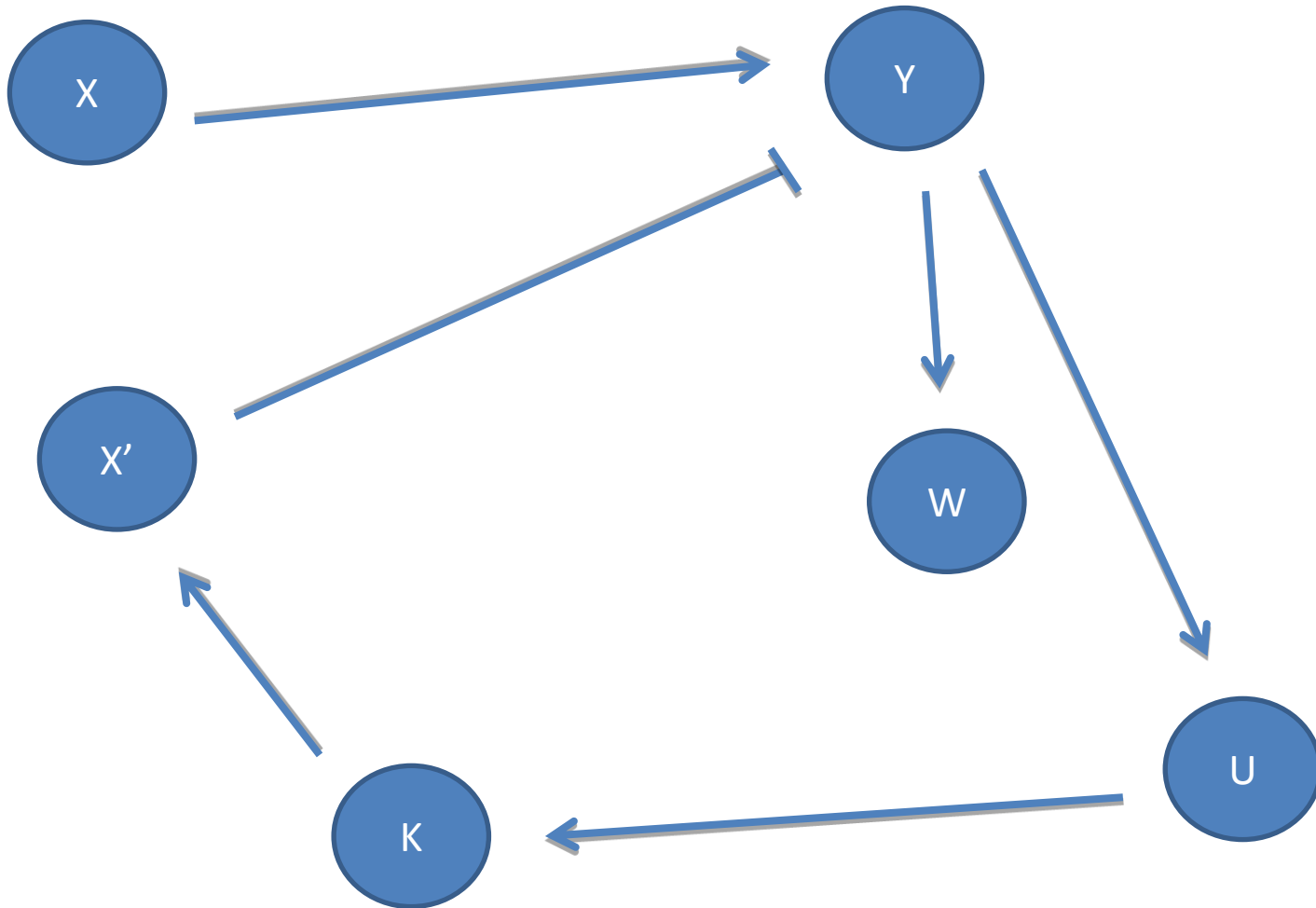
Gene regulatory network



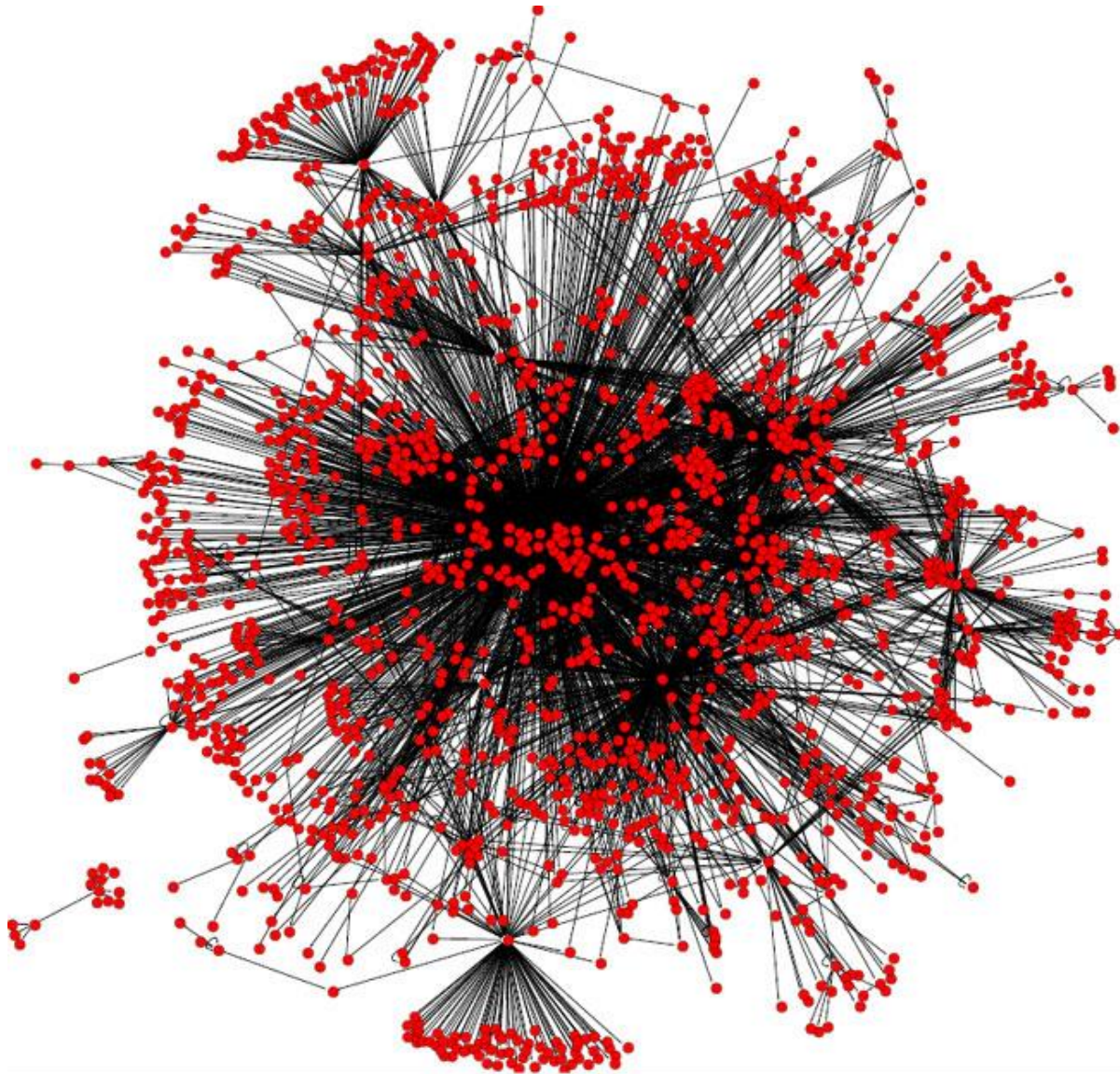
Gene regulatory network



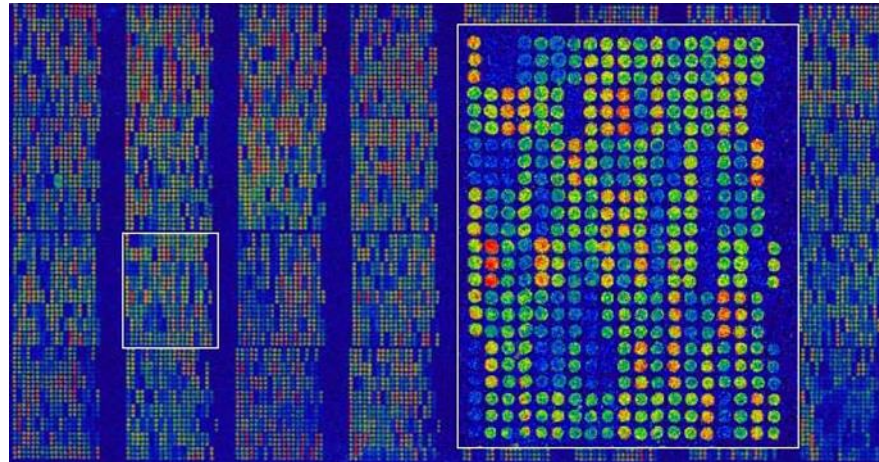
Gene regulatory network



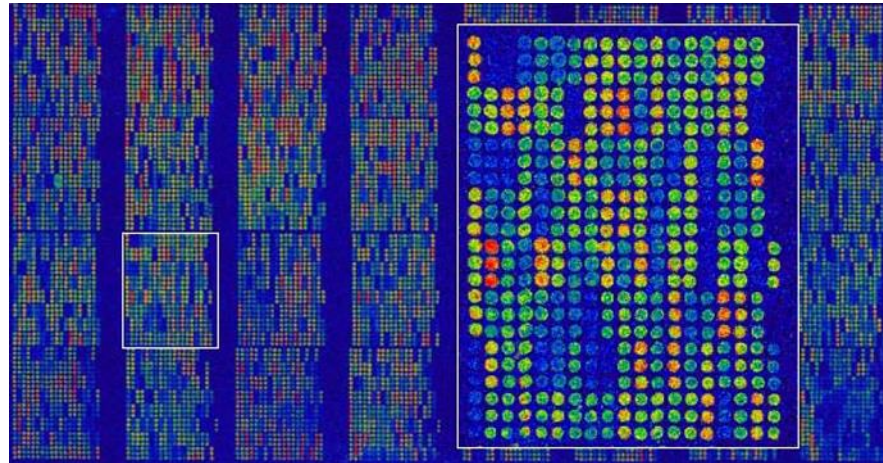
E. coli regulatory network



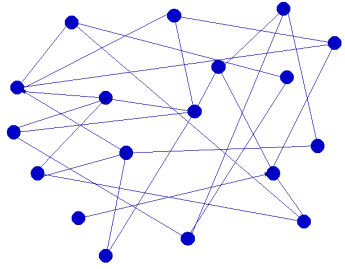
The development of high-throughput data-collection techniques allows for the simultaneous interrogation of the status of cell's components



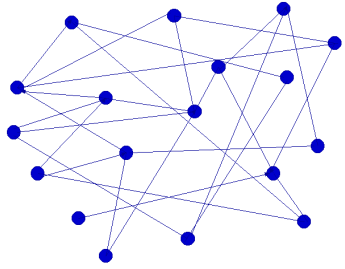
The development of high-throughput data-collection techniques allows for the simultaneous interrogation of the status of cell's components



Big data

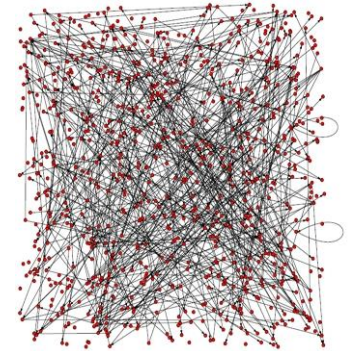


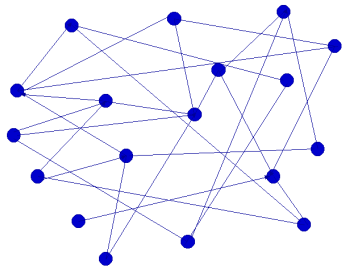
Challenge: network reconstruction



Challenge: network reconstruction

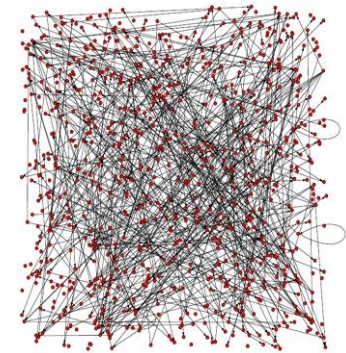
Difficulty: large size (many components)
of biological systems





Challenge: network reconstruction

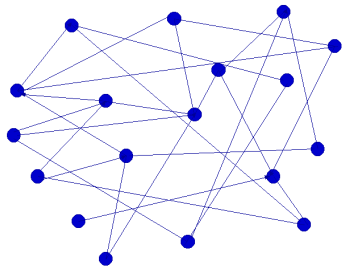
Difficulty: large size (many components)
of biological systems



Statistical
physics

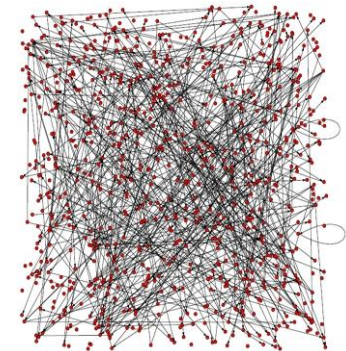


find approximate methods to
solve the problem



Challenge: network reconstruction

Difficulty: large size (many components)
of biological systems



Statistical
physics



find approximate methods to
solve the problem



Most of the current reconstruction techniques do
not exploit the temporal information of the data

GOAL:



Statistical
mechanics

GOAL:



Statistical
mechanics



develop new inference techniques that exploit the temporal
structure of the data

GOAL:



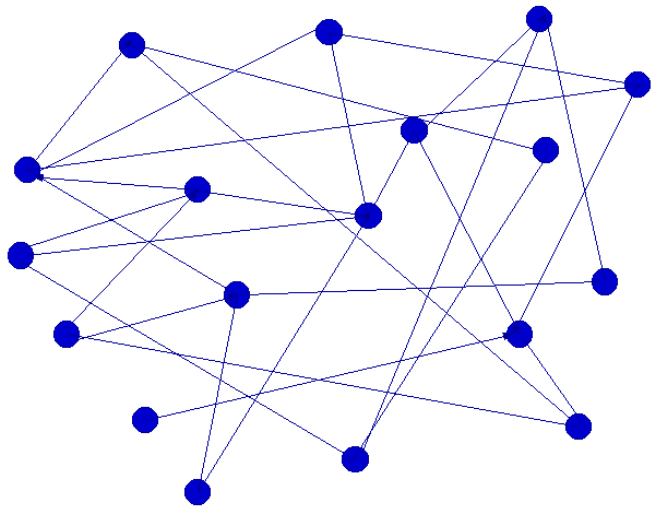
Statistical
mechanics

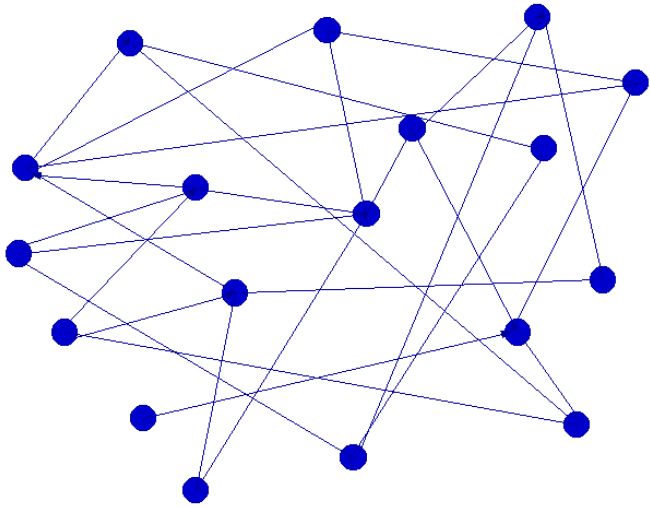


develop new inference techniques that exploit the temporal
structure of the data

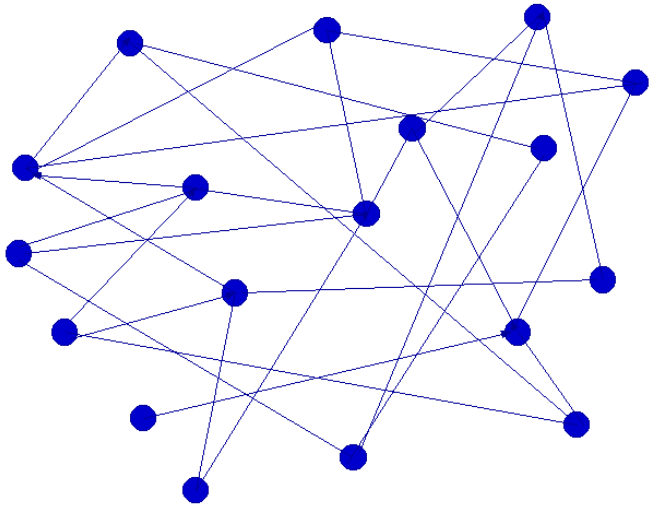


improve the quality of biological network reconstruction





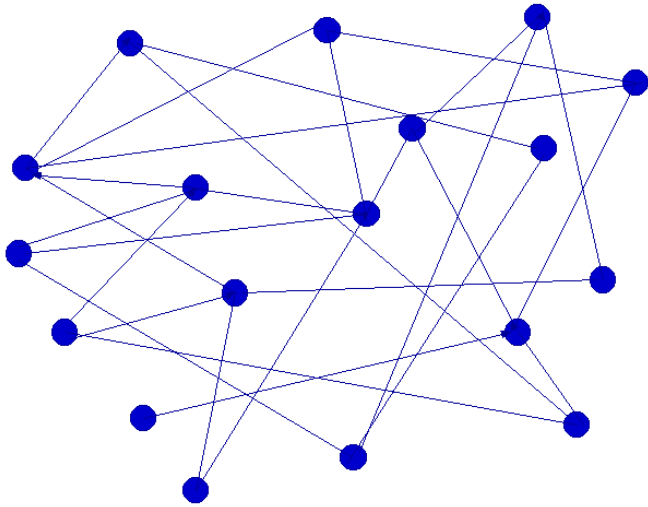
New method to predict the state of the system knowing the value of the links (results agree well with simulations of small systems)



New method to predict the state of the system knowing the value of the links (results agree well with simulations of small systems)



Faster algorithm



New method to predict the state of the system knowing the value of the links (results agree well with simulations of small systems)



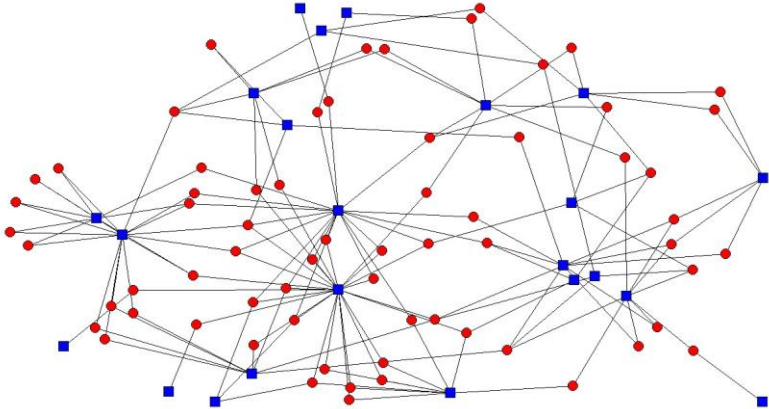
Faster algorithm



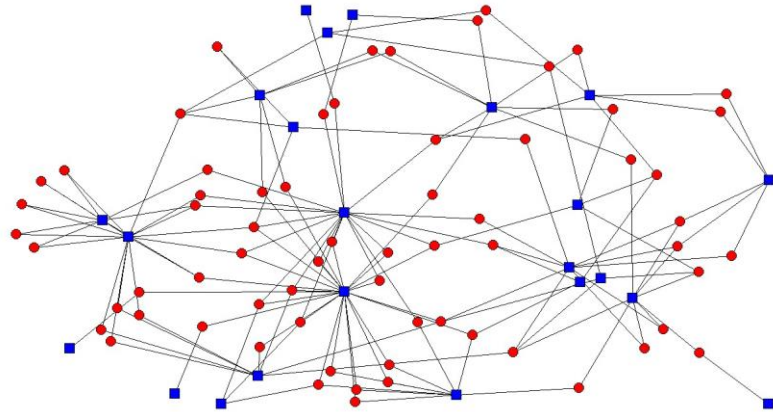
To do:

Inference problem: optimally predict the links

Hidden nodes

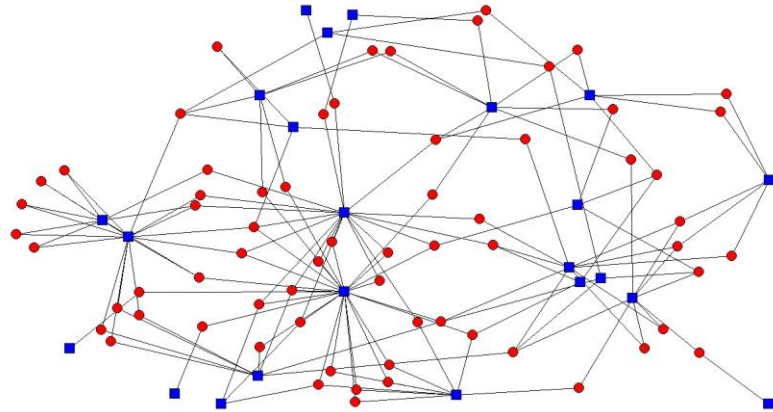


Hidden nodes



Optimally predict the state of the unobserved variables (results agree well with simulations of small systems)

Hidden nodes

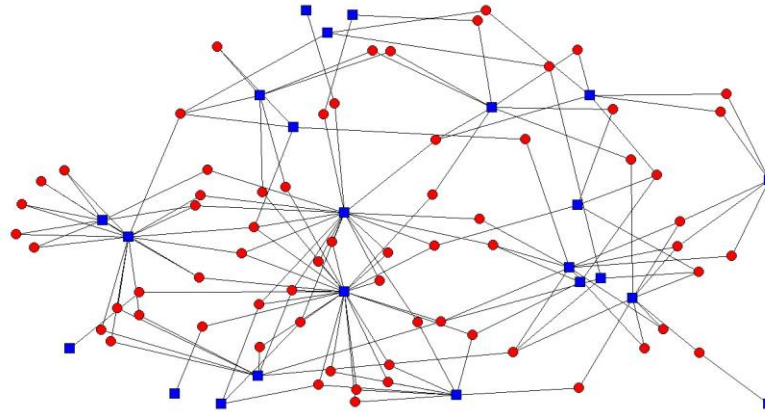


Optimally predict the state of the unobserved variables (results agree well with simulations of small systems)



Quality of the prediction on the hidden nodes varying the sistem's parameters

Hidden nodes



Optimally predict the state of the unobserved variables (results agree well with simulations of small systems)

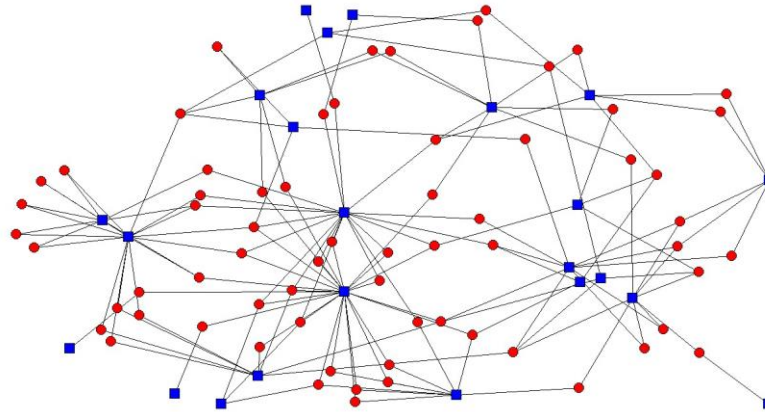


Quality of the prediction on the hidden nodes varying the sistem's parameters



Algorithm to infer the links

Hidden nodes



Optimally predict the state of the unobserved variables (results agree well with simulations of small systems)



Quality of the prediction on the hidden nodes varying the sistem's parameters



Algorithm to infer the links



To do:

Study different kinds of networks, continuous variables

● Courses and schools (besides those organized by NETADIS):

TU Berlin:

- Stochastic Systems
- Monte Carlo methods in Artificial Intelligence and Machine Learning
- Probabilistic and Bayesian Modelling in Machine Learning and Artificial Intelligence

Bernstein Center for Computational Neuroscience, Berlin:

Benjamin Lindner's seminars on "Stochastic Aspects of Neurobiological Problems"

ICTP, Trieste: School on Large Scale Problems in Machine Learning and Workshop on Common Concepts in Machine Learning and Statistical Physics, August 2012.

● Main collaborations: Prof. Sollich's group and Dr. Roudi group

Thanks for your attention!