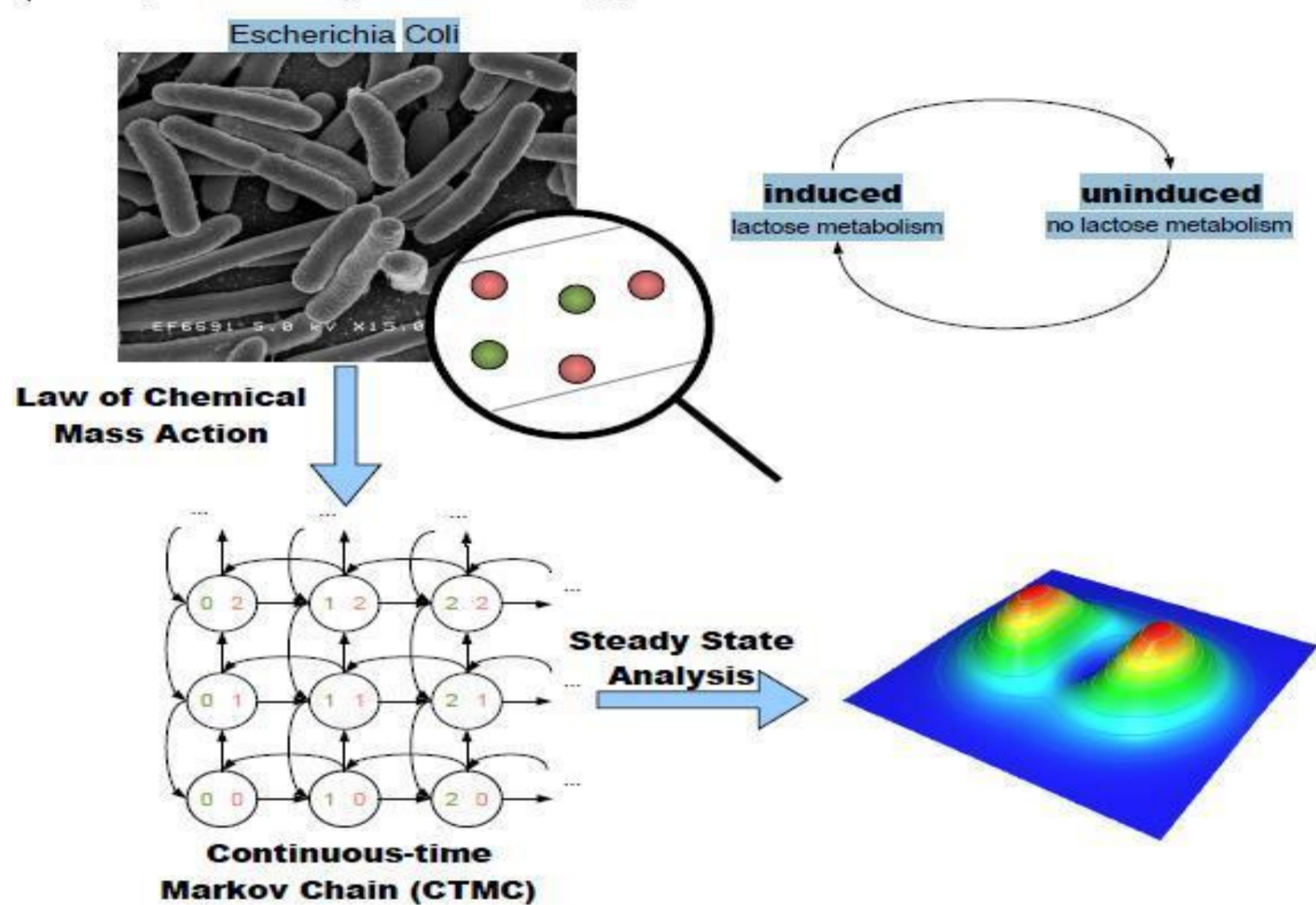


## Introduction

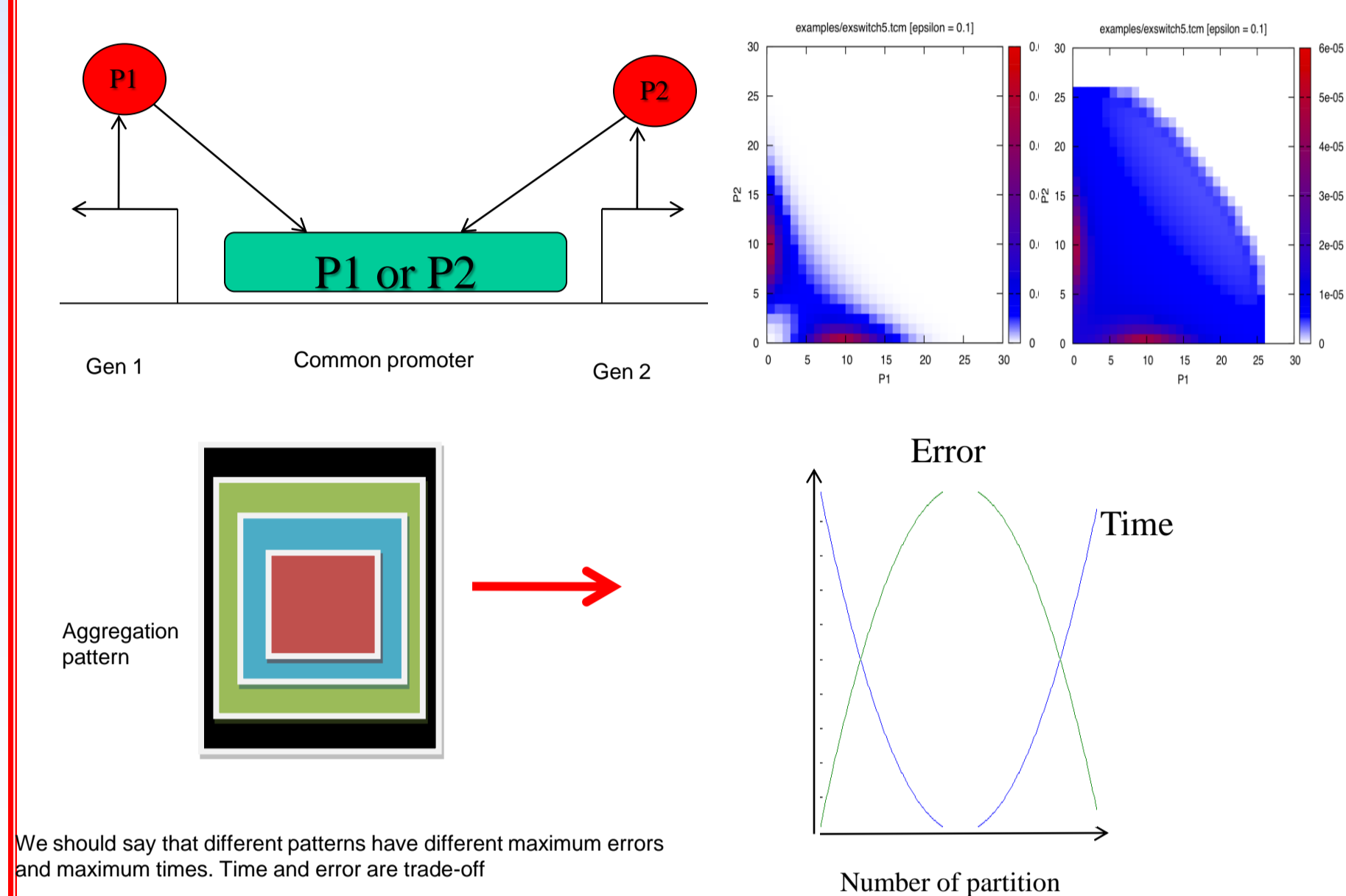
### Stochastic Process:

The complexity of living systems has led to a rapidly increasing interest in modeling and analysis of biochemically reacting systems. Since biochemical reactions occur randomly, we need kind of stochastic process. Markov chain is stochastic model which at any time  $t$  one can see current state of system by defining appropriate parameters of our model.

#### Example from Systems Biology:



## Results



We should say that different patterns have different maximum errors and maximum times. Time and error are trade-off

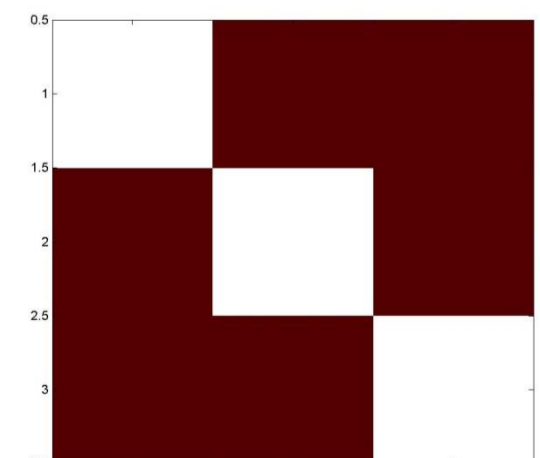
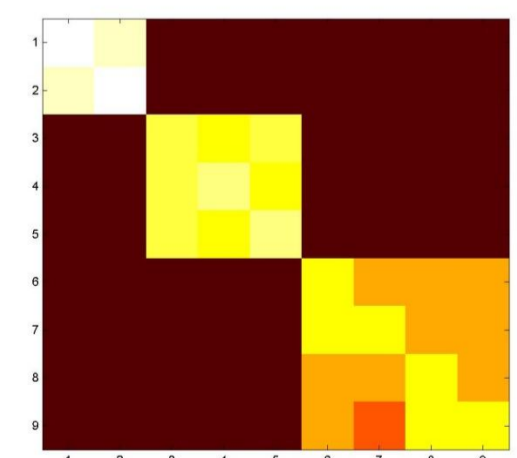
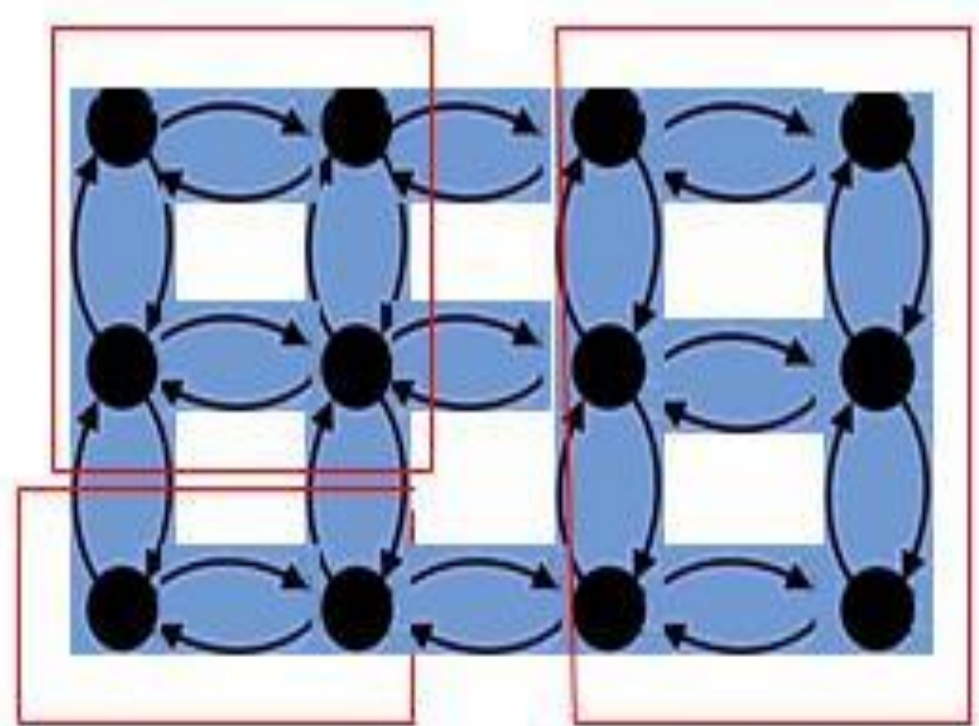
## Methods

In chemical reactions modeling because of discreteness of system, state space is too large and computing of steady states is too computing expensive or almost impossible. Aggregation method is one way for fast computing of steady state solution of stochastic systems.

### Aggregation :

Markov chain defined by  $P$  with state space  $i \in I$  can be aggregated to a Markov chain with a smaller state space  $A_i \in \Delta$  and a transition matrix  $R$

Normal Steady state computing :  $\pi_j = \lim_{t \rightarrow \infty} \pi_j(t) = \lim_{t \rightarrow \infty} p_{ij}(t)$   $\pi \cdot Q = 0$  or  $P \cdot \pi = \pi$



### Aggregated steady state solution

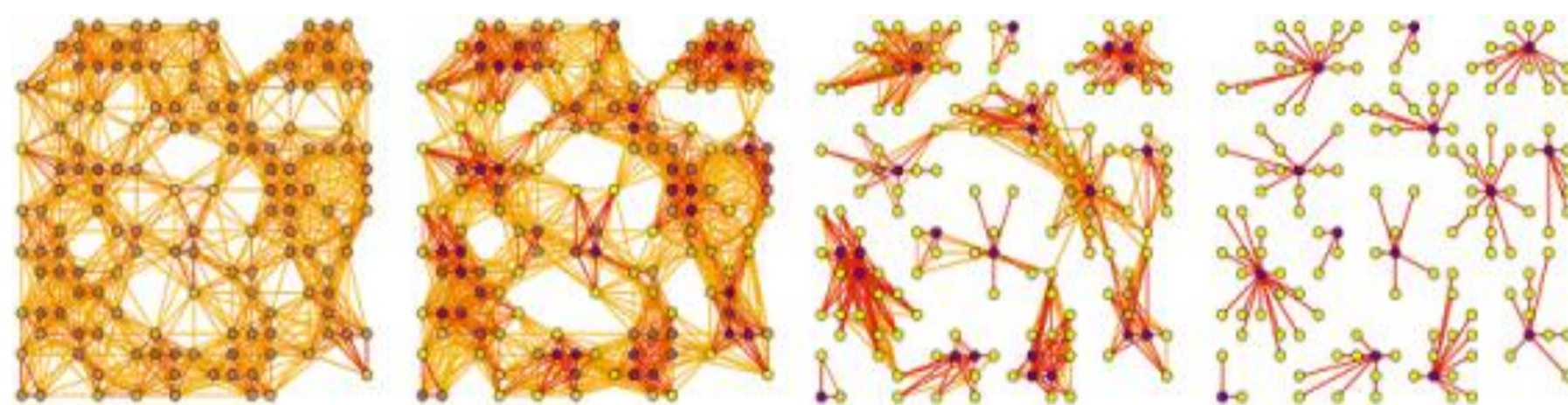
1. compute of each macro state independently using normal steady state formulas
2. Compute  $Q$  for macro-state Markov chain, steady state solution for this markov chain
3. Consider final steady state of each state as multiply of corresponding macro-state and its own steady state from (1)

With aggregation one for sure would have error but by choosing right states in one partition the solution is not far from exact one.

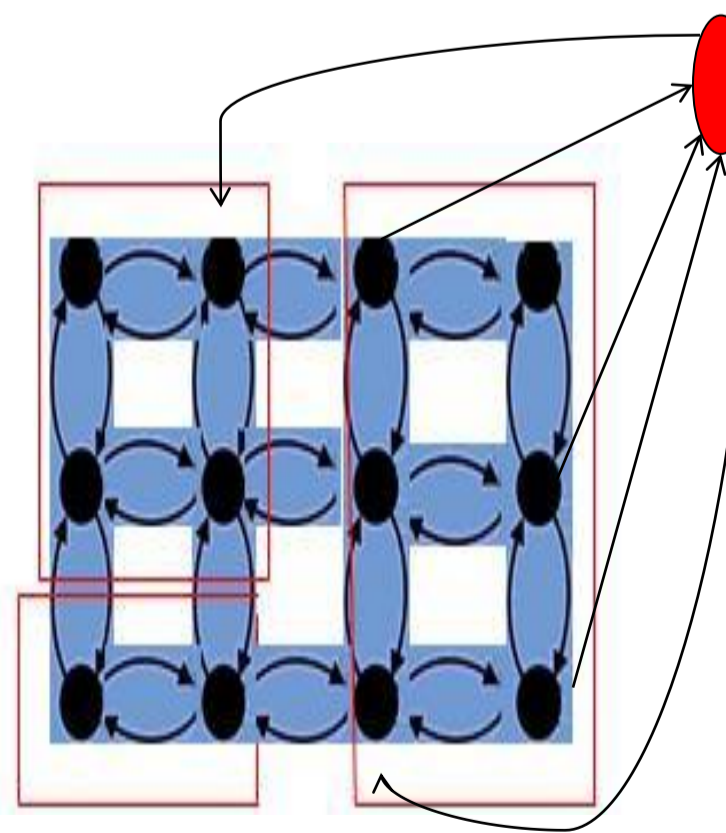
## Future Works

Idea: A good aggregation should group fast subsystems and slow transition between them, which is exactly clustering problem, so

one can finding optimal aggregation pattern using spectral or markov clustering.



Adding dummy state to markov chain as a reprensive of other state s which already ignored. To direct border state's transition to the dummy state and redirect to one of the states in our boundary.



## REFERENCES

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2. Richard L. Tweedie. Sufficient conditions for regularity, recurrence and ergodicity of markov process, *Mathematical Proceedings of the Cambridge Philosophical Society*, 78(01):125-136, 1975
3. P.-J. Courtois and P Semal. Bounds for the positive eigenvectors of nonnegative matrices and for their approximations by decomposition. *J. ACM*, 31(4):804-825, 1984.