

# Design and calibration of stochastic models for DNA methylation patterns

Master seminar by Andrea Kupitz

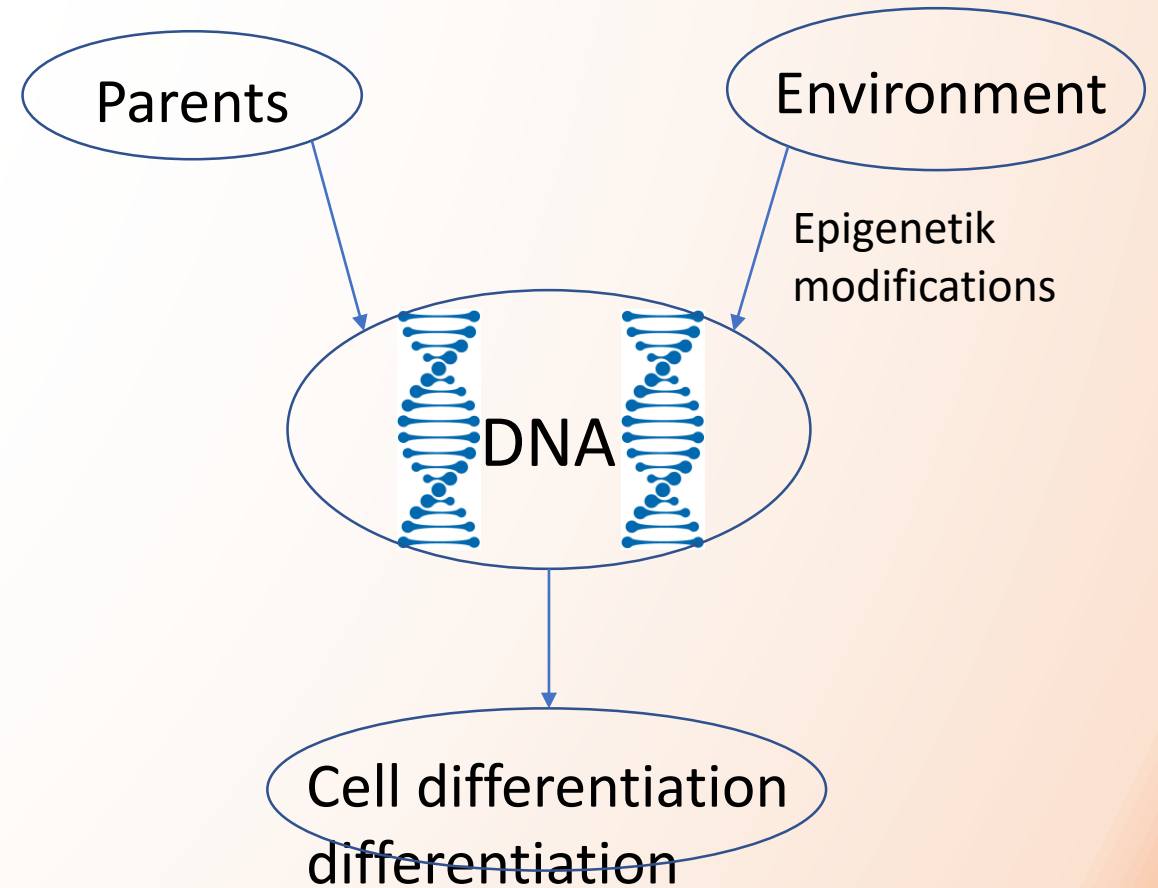
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# Overview

- Problem:
  - Method of operation of DNMTs unclear
  - Measure distances between methylation pattern distributions
- Approach:
  - MCMC algorithm to simulate cell cycle
  - Pairwise distances for ABC
- Results:
  - Parameters of MCMC difficult to identify
- Outlook:
  - Validation of ABC method

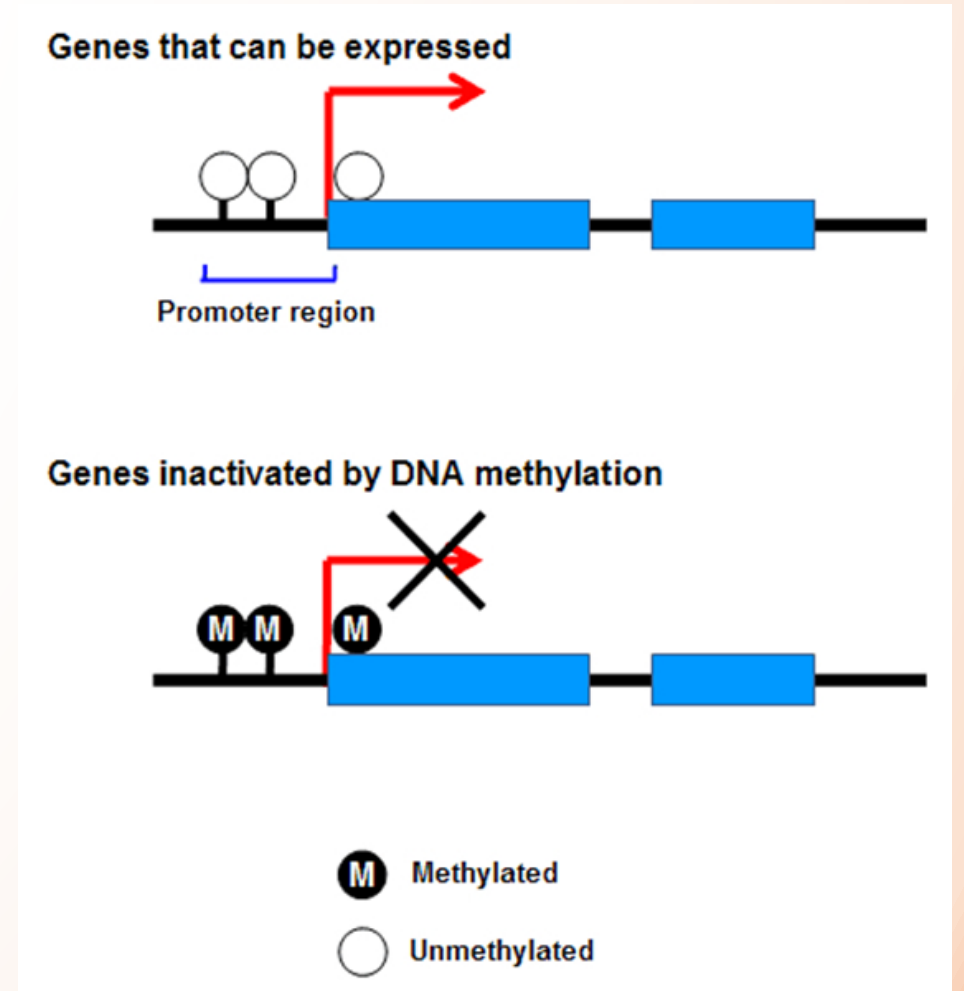
# Introduction – Epigenetics

- Histon modifications
  - Histon methylation
  - Histon acetylation
  - ...
- DNA modifications
  - **DNA methylation**



# Introduction- Methylation

- Influences gene expression
- Indicator for diseases
- Heritable



# Introduction - Methylation

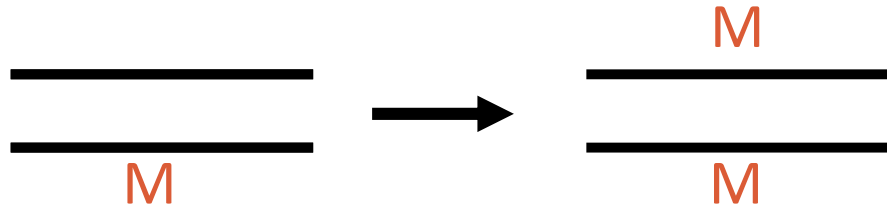


- DNA methylation at CpG
- Methylation transmitted by DNMTs
- Method of operation of DNMTs unclear

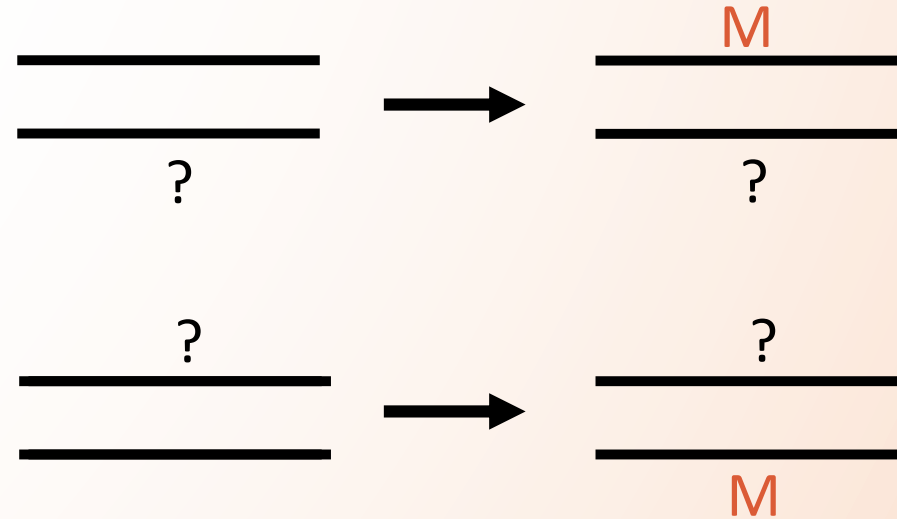
# Introduction- DNMTs

## Methylation

- Maintenance:



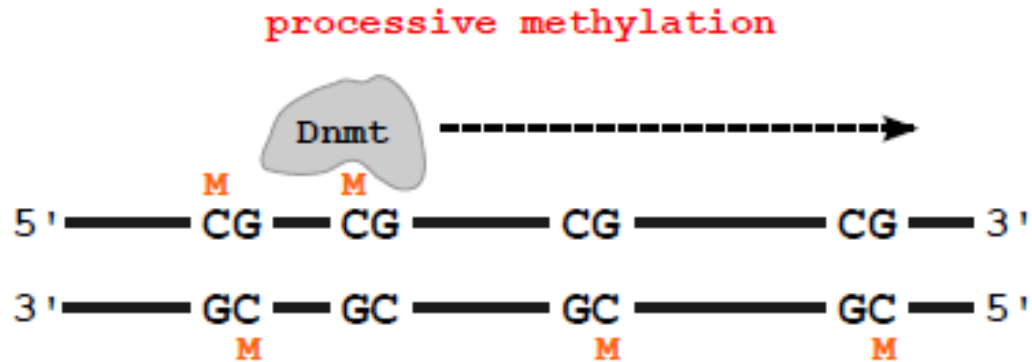
- De novo:



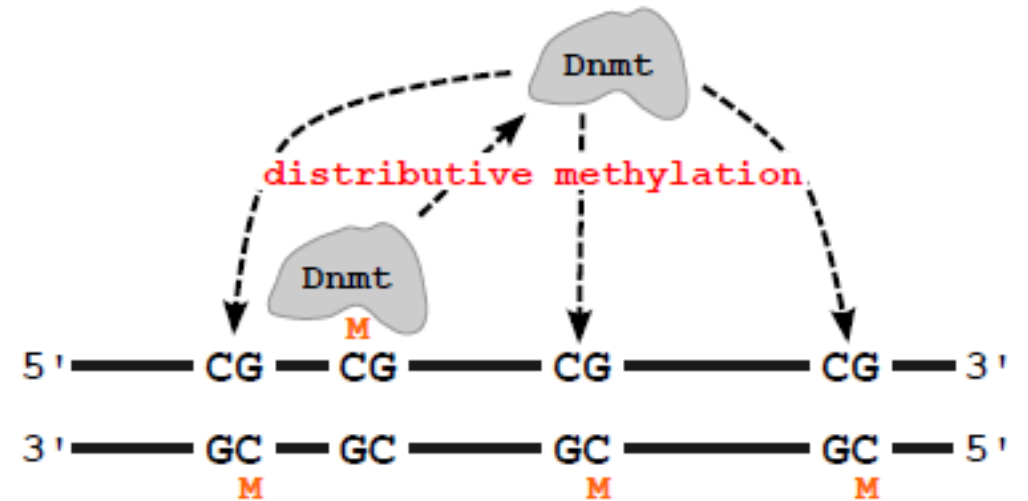
# Introduction- DNMTs

## Processivity

- High:



- Low:



# Methods – Markov Chain

$\rho$  – Dissociation probability

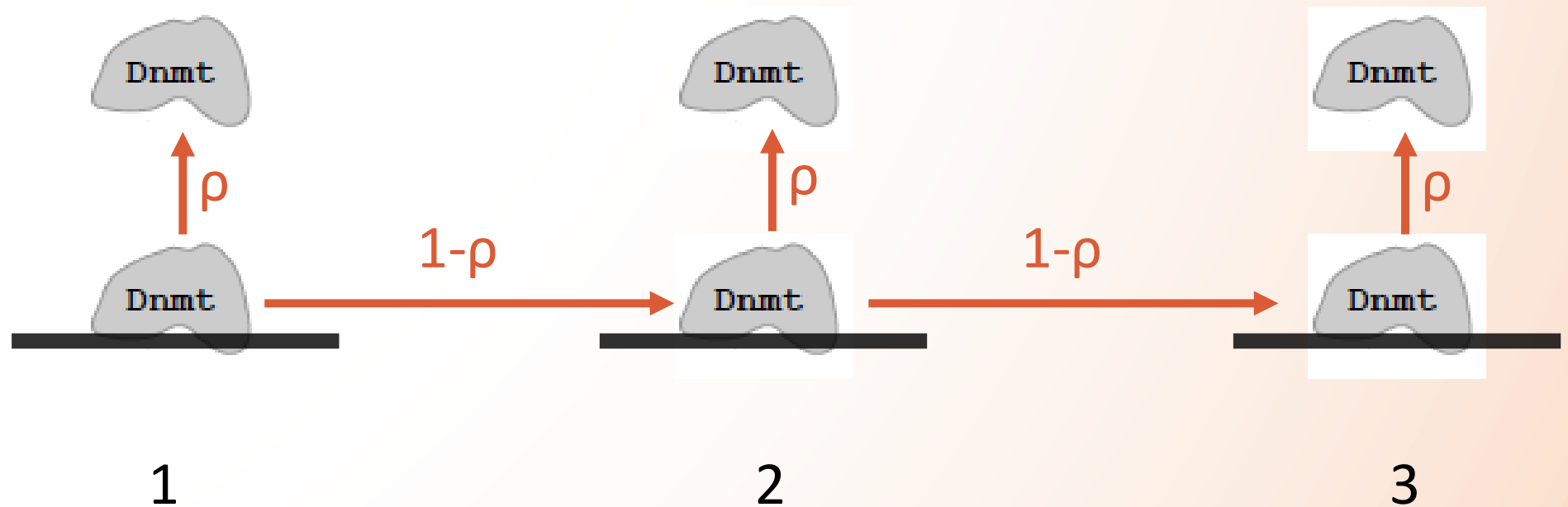
DNMT

Not bound

DNMT

Bound

Position





# Methods – Markov Chain

$\tau$  – Association probability

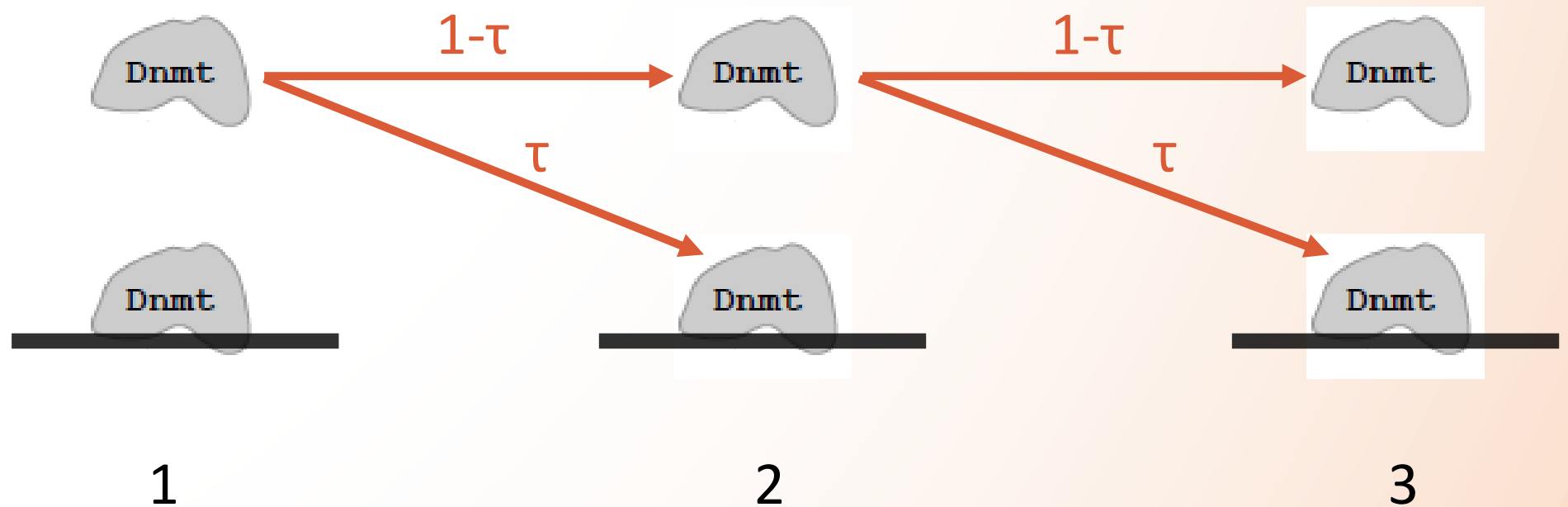
DNMT

Not bound

DNMT

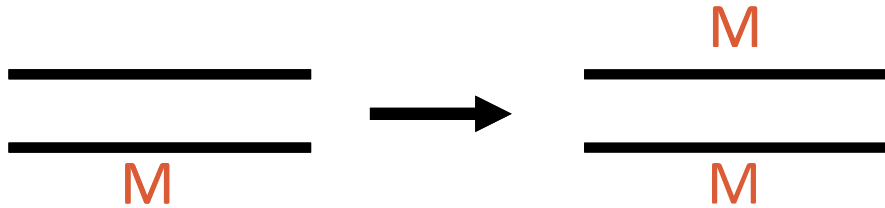
Bound

Position

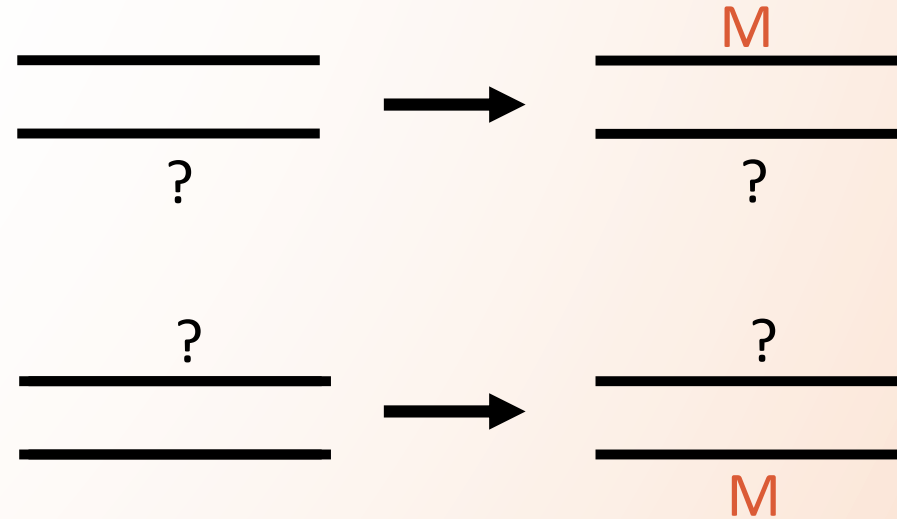


# Methods – Markov Chain

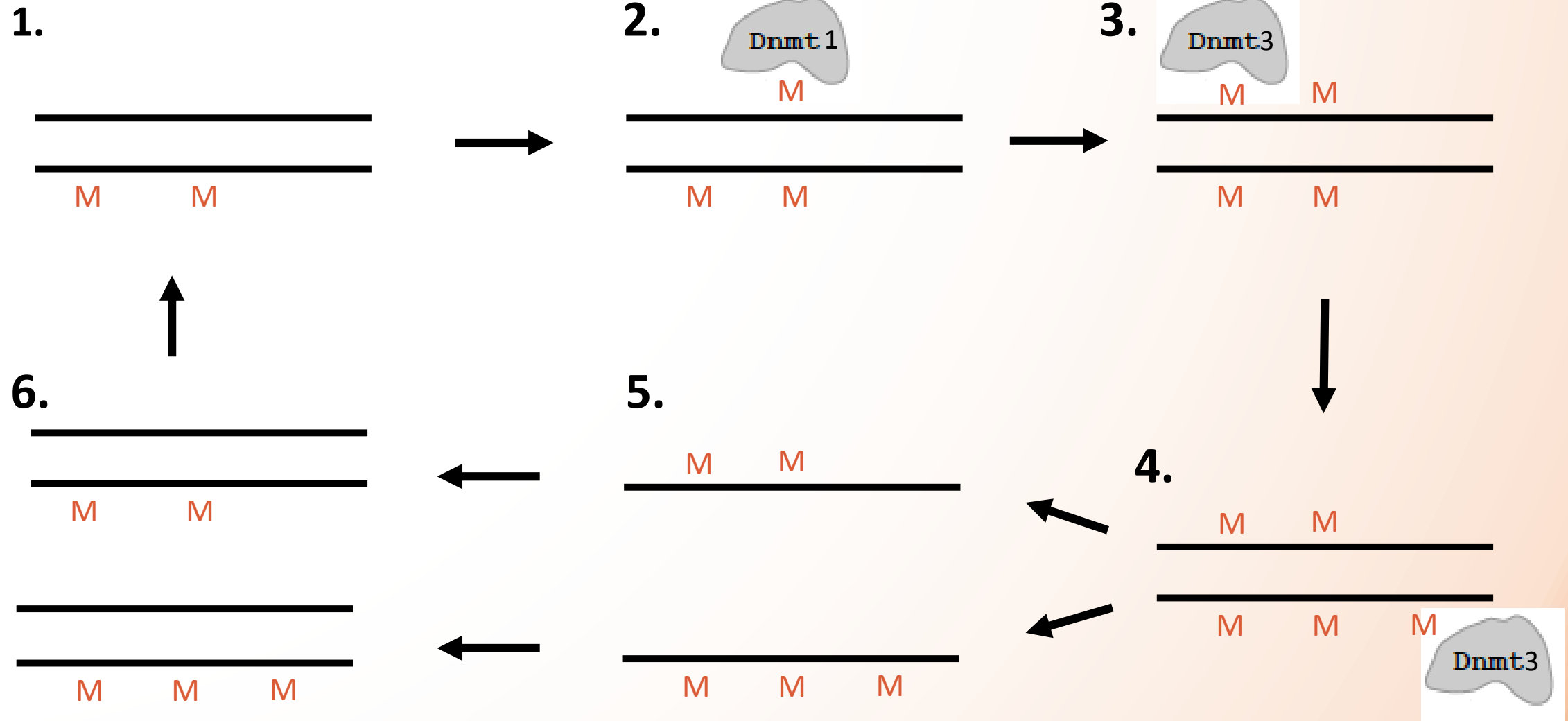
$\delta$  – De novo methylation probability



$\mu$  - Maintenance methylation probability



# Methods – Cell Division



# Methods - MCMC

## Markov chain Monte Carlo

- Given:
  - Measured methylation pattern distribution before cell divisions
  - Simulation parameters  $\rho$ ,  $\tau$ ,  $\delta$  and  $\mu$
- Aim:
  - Simulate a pattern distribution after cell division

# Methods- MCMC

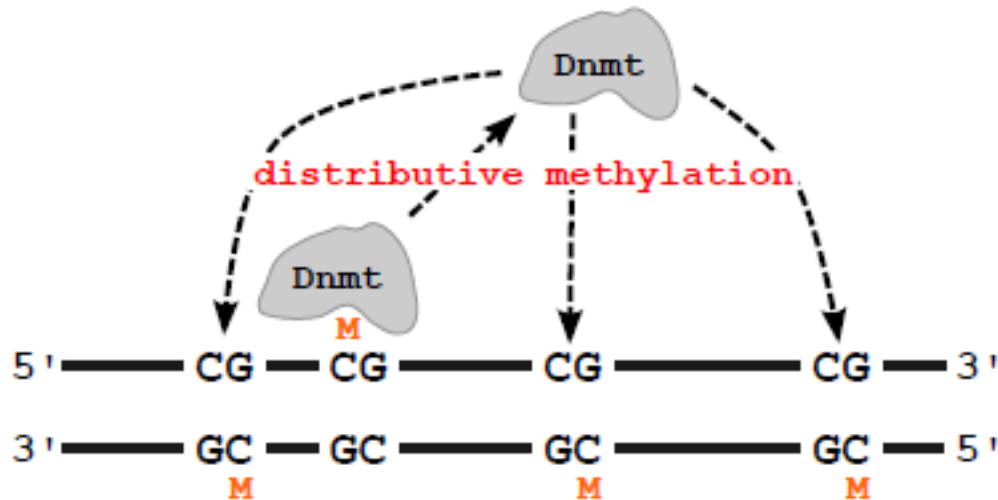
Workflow:

- 1. Sample from pattern distribution
- 2. Simulate t cell divisions
  - Draw RN and decide using  $\rho$ ,  $\tau$ ,  $\delta$  and  $\mu$  if DNMT binds/methylates
- 3. Repeat 1 and 2 10000 times

# Results- MCMC

## DMNT3 - Expectations

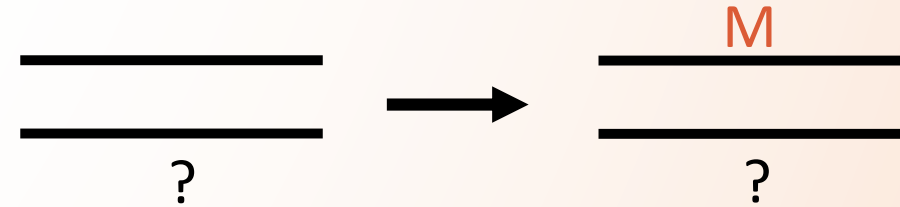
- Low processivity:



Result:

- $\tau$  high (0.89)
- $P$  low (0.28)

- De novo methylation/few maintenance methylation activity:



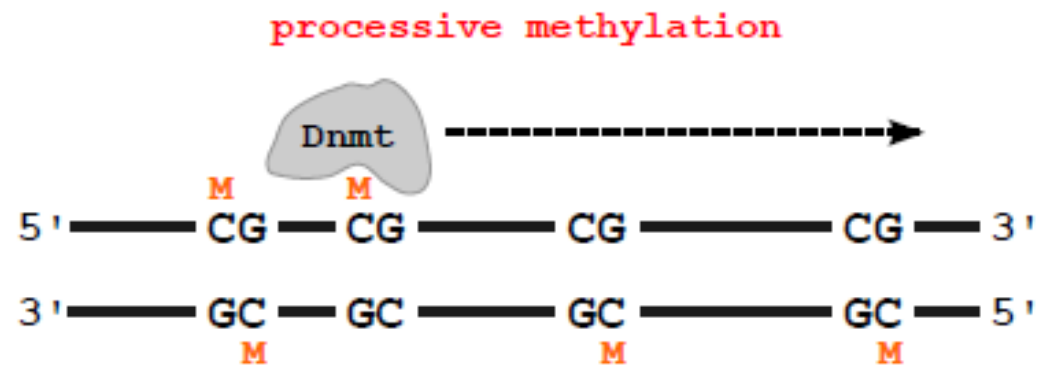
Result:

- $\mu$  high (0.78)
- $\delta$  high (1.00)

# Results- MCMC

## DMNT1 - Expectations

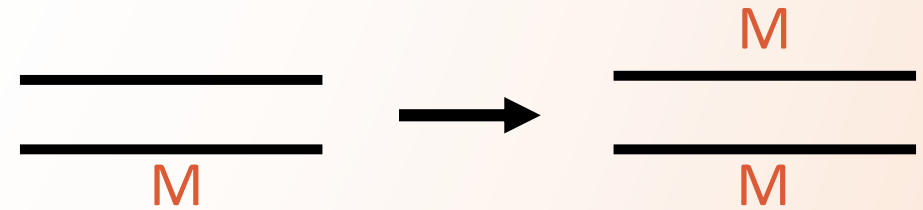
- High processivity:



Result:

- $\tau$  low (0.23)
- $P$  high (1.00)

- Maintenance methylation:



Result:

- $\mu$  high (0.75)
- $\delta$  low (0.43)

# Methods - MLE

Log-likelihood:

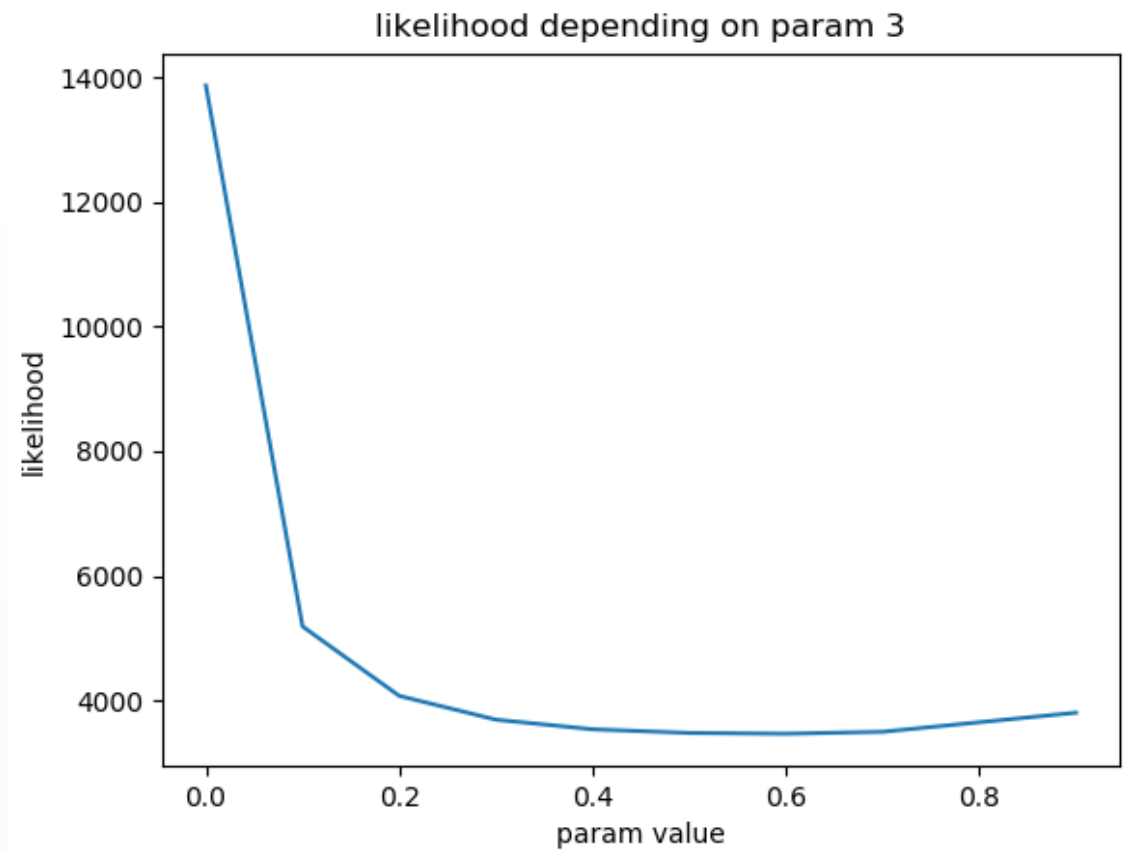
- $L(\theta) = \sum_{i=1}^{4^l} \log(\hat{\pi}_i(\theta)) * N_i$
- $\theta = (\rho, \tau, \delta, \mu)$
- $l$ : number of CpGs
- $\hat{\pi}$  : pattern distribution of simulation
- $N_i$ : occurrences of pattern  $i$  in measured data

MLE:  $\theta = \arg \max_{\theta} L(\theta)$



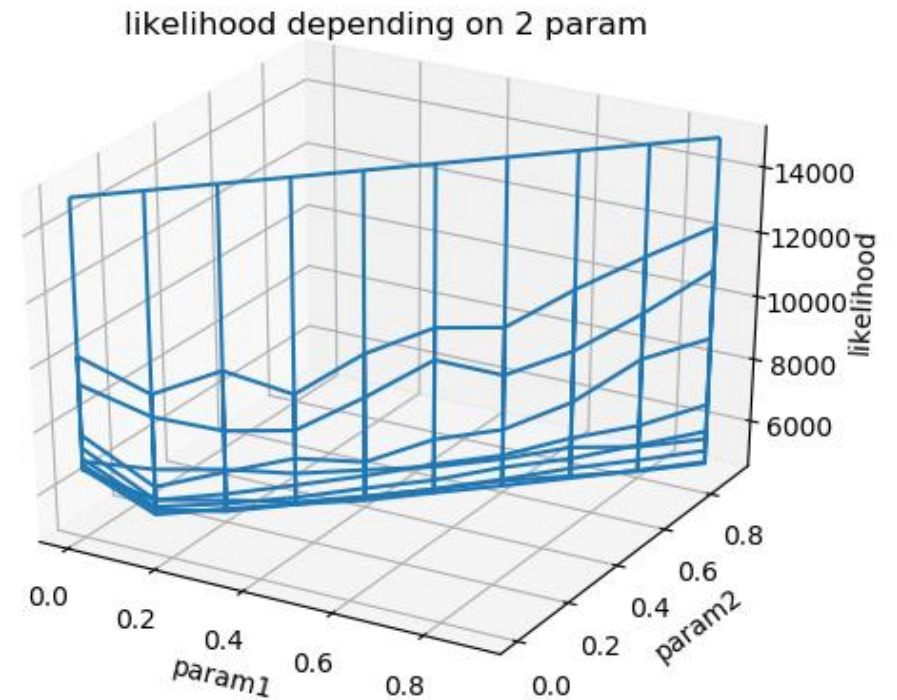
# Results- MLE

- Param 3:  $\delta$
- $\rho$ ,  $\tau$  and  $\mu$  fixed
- Neg. log-likelihood
- Large interval for  $\delta$



# Results- MLE

- Param 1:  $\rho$
- Param 2:  $\mu$
- $\tau$  and  $\delta$  fixed
- Neg. log-likelihood
- $\rho$  and  $\mu$  linear dependent



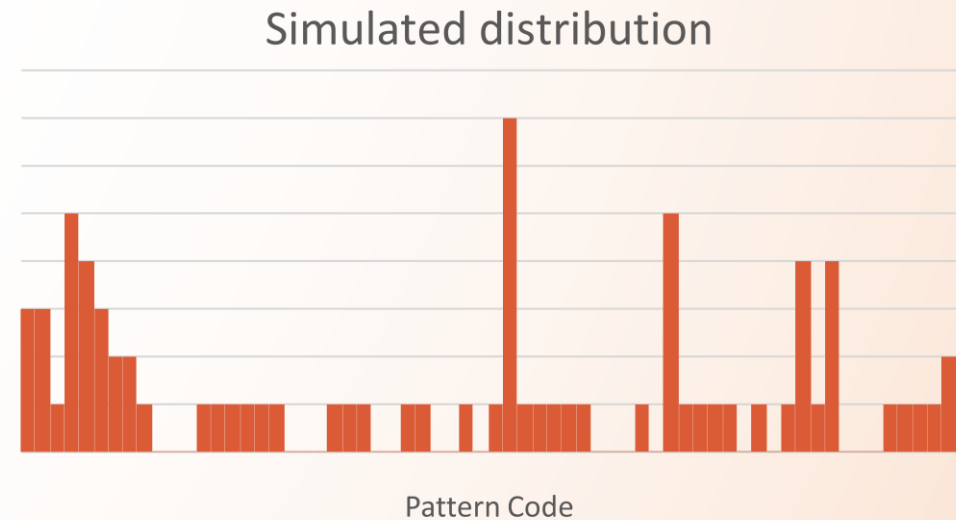
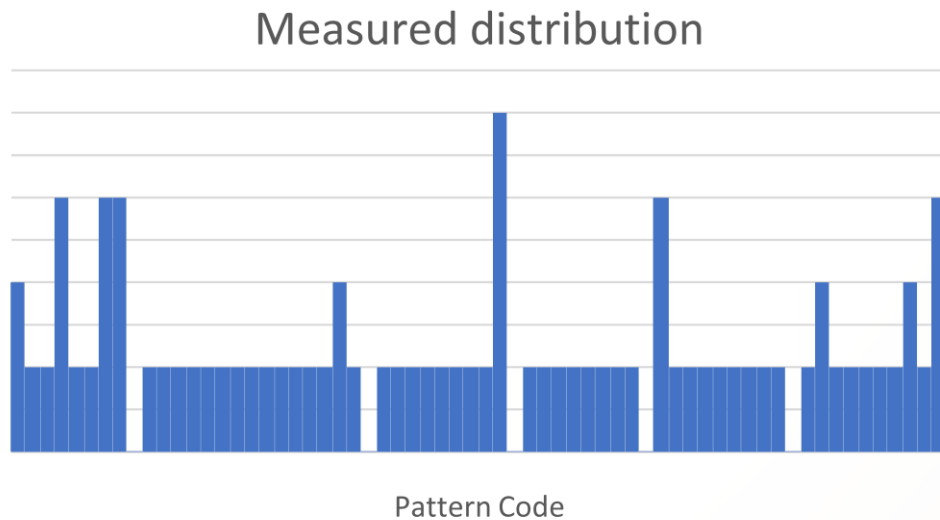
# Methods- ABC

## **Approximate Bayesian Computation**

- Given:
  - Measured methylation pattern distribution after cell division
  - Function simulating cell division
- Needed:
  - Distance between pattern distributions
- Output:
  - Best parameters

# Methods - ABC

## Pattern distributions – distance function

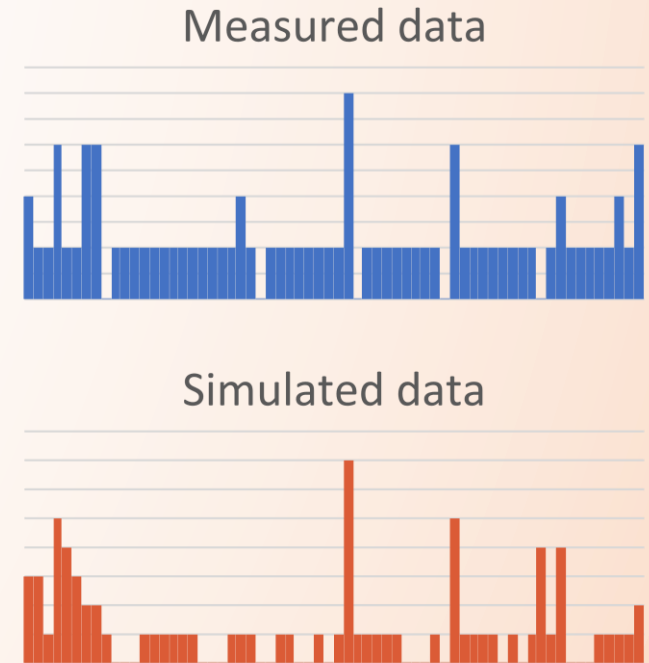


dist = ?

# Methods - ABC

## Pattern distributions – distance function

- $\text{dist}((X_1, \dots, X_{l_1}), (Y_1, \dots, Y_{l_2})) = \sum_{i=1}^{l_1} \sum_{j=1}^{l_2} w_{ij} (X_i - Y_j)^2$
- $w_{ij}$ : distance between pattern i and j



# Methods - ABC

## Methylation patterns – distance function

M	
	M
M	

M	
	M

$$\text{dist}(i, j) = w_{ij}$$

# Methods - ABC

Workflow:

- 1. Choose  $\rho$ ,  $\tau$ ,  $\delta$  and  $\mu$  randomly
- 2. Simulate cell division
- 3. Compute  $\text{dist}((X_1, \dots, X_{l_1}), (Y_1, \dots, Y_{l_2}))$
- 4. Repeat 1-3
- 5. Yield best  $\rho$ ,  $\tau$ ,  $\delta$  and  $\mu$

# Outlook

- Implement distance function  $w_{ij}$  for ABC
- Use distance function to validate results from MCMC
- Test ABC for artificial data





"There are no stupid questions, so let's also agree there are no stupid answers."

Thank you!

Any Questions?

# References

- A. Lück et al., A Stochastic Model for the Formation of Spatial Methylation Patterns
- A. Q. Fu et al., DNA Methyltransferases from Double-Stranded Methylation Patterns
- <https://de.wikipedia.org/wiki/MCMC-Verfahren>
- <https://de.wikipedia.org/wiki/DNA-Methylierung>
- [https://en.wikipedia.org/wiki/DNA\\_methyltransferase](https://en.wikipedia.org/wiki/DNA_methyltransferase)

# Pictorial sources

- Slide 2: <https://www.biomol.de/die-top-10-der-dna-farbstoffe-und-sonden.html?id=821>
- Slide 3:  
<https://www.ncc.go.jp/en/ri/division/epigenomics/project/230/20170913152903.html>
- Slide 7–9, 11, 14, 15: A. Lück et al., A Stochastic Model for the Formation of Spatial Methylation Patterns, p.3
- Slide 22:  
<https://www.cartoonstock.com/cartoonview.asp?catref=cwln5040>