

Basic Model

$$\begin{aligned} \mathbf{x}_{t+1} &= \mathbf{A}\mathbf{x}_t + \mathbf{w}_t = \mathbf{A}\mathbf{x}_t + \mathbf{w}_\bullet & \mathbf{w}_\bullet &\sim \mathcal{N}(0, \mathbf{Q}) \\ \mathbf{y}_t &= \mathbf{C}\mathbf{x}_t + \mathbf{v}_t = \mathbf{C}\mathbf{x}_t + \mathbf{v}_\bullet & \mathbf{v}_\bullet &\sim \mathcal{N}(0, \mathbf{R}) \end{aligned}$$

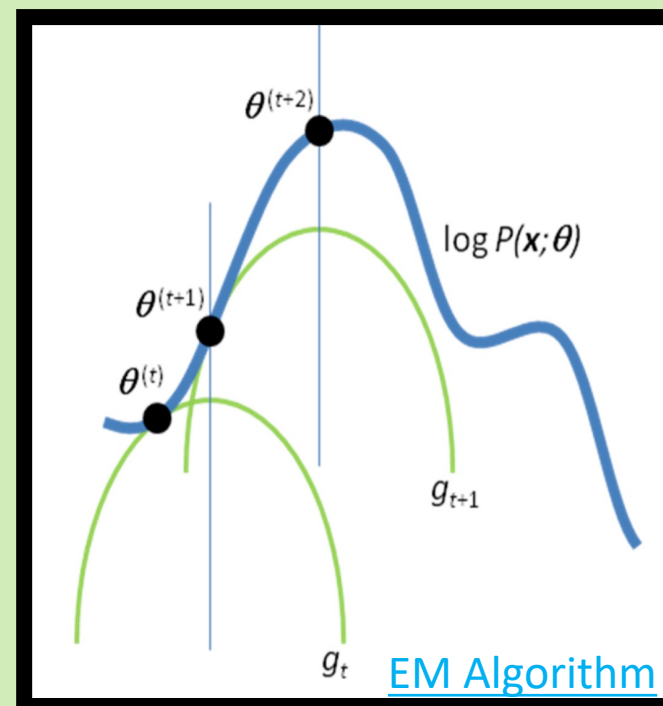
EM algorithm: Learning

$$\begin{aligned} \log \int_{\mathbf{X}} P(\mathbf{Y}, \mathbf{X} | \theta) d\mathbf{X} &= \log \int_{\mathbf{X}} Q(\mathbf{X}) \frac{P(\mathbf{X}, \mathbf{Y} | \theta)}{Q(\mathbf{X})} d\mathbf{X} \\ &\geq \int_{\mathbf{X}} Q(\mathbf{X}) \log \frac{P(\mathbf{X}, \mathbf{Y} | \theta)}{Q(\mathbf{X})} d\mathbf{X} \\ &= \int_{\mathbf{X}} Q(\mathbf{X}) \log P(\mathbf{X}, \mathbf{Y} | \theta) d\mathbf{X} - \int_{\mathbf{X}} Q(\mathbf{X}) \log Q(\mathbf{X}) d\mathbf{X} \\ &= \mathcal{F}(Q, \theta) \end{aligned}$$

↗ increase

E step $Q_{k+1} \leftarrow \arg \max_Q \mathcal{F}(Q, \theta_k)$

M step $\theta_{k+1} \leftarrow \arg \max_{\theta} \mathcal{F}(Q_{k+1}, \theta)$



Filtering & smoothing

$$\begin{aligned} P(\mathbf{x}_t | \{\mathbf{y}_1, \dots, \mathbf{y}_t\}) \\ P(\mathbf{x}_t | \{\mathbf{y}_1, \dots, \mathbf{y}_\tau\}) \end{aligned}$$

Basic Model

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Static data modelling (variants)

$$\begin{aligned}\mathbf{x}_\bullet &= \mathbf{w}_\bullet & \mathbf{w}_\bullet &\sim \mathcal{N}(\mathbf{0}, \mathbf{I}) \\ \mathbf{y}_\bullet &= \mathbf{C}\mathbf{x}_\bullet + \mathbf{v}_\bullet & \mathbf{v}_\bullet &\sim \mathcal{N}(\mathbf{0}, \mathbf{R})\end{aligned}$$

Factor Analysis

$$\mathbf{R} = \begin{bmatrix} r_1 & 0 & 0 & \dots & 0 \\ 0 & r_2 & 0 & \dots & 0 \\ 0 & 0 & r_m & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & r_s \end{bmatrix}$$

SPCA

PCA

$$\mathbf{R} = \epsilon \mathbf{I} \quad \mathbf{R} = \lim_{\epsilon \rightarrow 0} \epsilon \mathbf{I}$$

Time-series modelling

Kalman Filter

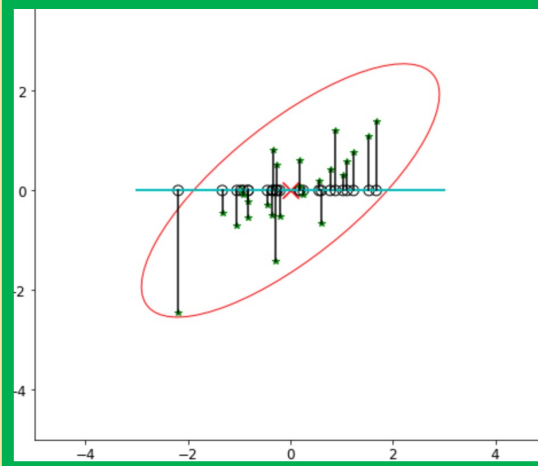
$$\begin{aligned}\mathbf{x}_{t+1} &= \mathbf{A}\mathbf{x}_t + \mathbf{w}_t = \mathbf{A}\mathbf{x}_t + \mathbf{w}_\bullet & \mathbf{w}_\bullet &\sim \mathcal{N}(\mathbf{0}, \mathbf{Q}) \\ \mathbf{y}_t &= \mathbf{C}\mathbf{x}_t + \mathbf{v}_t = \mathbf{C}\mathbf{x}_t + \mathbf{v}_\bullet & \mathbf{v}_\bullet &\sim \mathcal{N}(\mathbf{0}, \mathbf{R})\end{aligned}$$

PCA

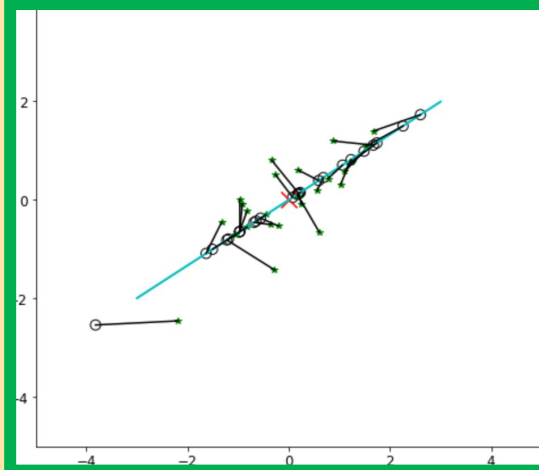
$$\mathbf{x}_\bullet = \mathbf{w}_\bullet \quad \mathbf{w}_\bullet \sim \mathcal{N}(\mathbf{0}, \mathbf{I})$$

$$\mathbf{y}_\bullet = \mathbf{C}\mathbf{x}_\bullet + \mathbf{v}_\bullet \quad \mathbf{v}_\bullet \sim \mathcal{N}(\mathbf{0}, \mathbf{R}) \quad \mathbf{R} = \lim_{\epsilon \rightarrow 0} \epsilon \mathbf{I}$$

E step



M step



Init: $\mathbf{C} = [1, 0]$
Generate \mathbf{X}

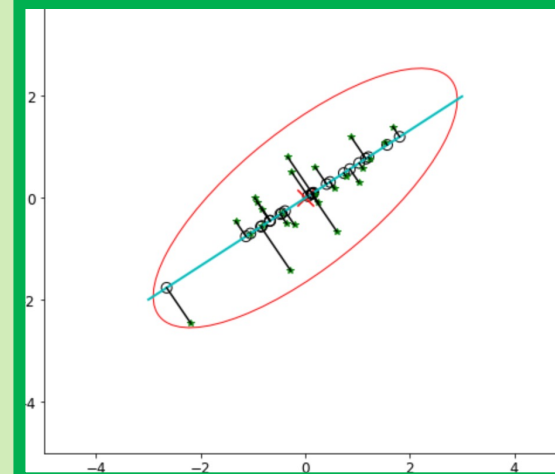
M: update \mathbf{C}

PCA Simulation

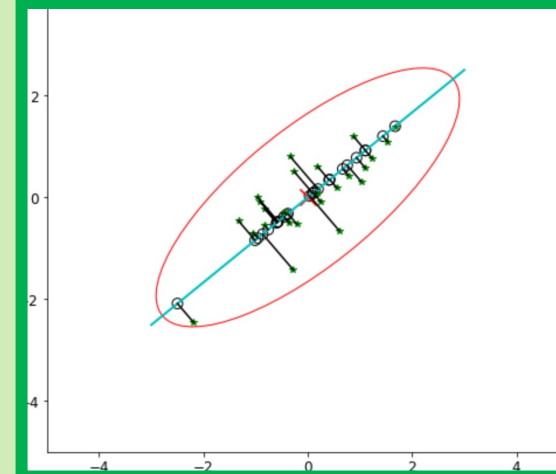
$$\mathbf{E}: \quad \mathbf{X} = (\mathbf{C}^T \mathbf{C})^{-1} \mathbf{C}^T \mathbf{Y}$$

$$\mathbf{M}: \quad \mathbf{C}^{\text{new}} = \mathbf{Y} \mathbf{X}^T (\mathbf{X} \mathbf{X}^T)^{-1}$$

E step



M step



E: update \mathbf{X}

M: update \mathbf{C}

[code](#)

ChIP-Seq (button to turn on or not)

(expressed or not) RNA-Seq

COBRA

Find TF target

TSS

Target gene

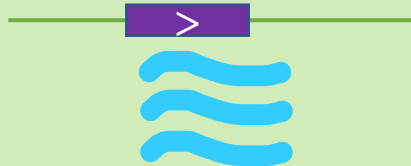
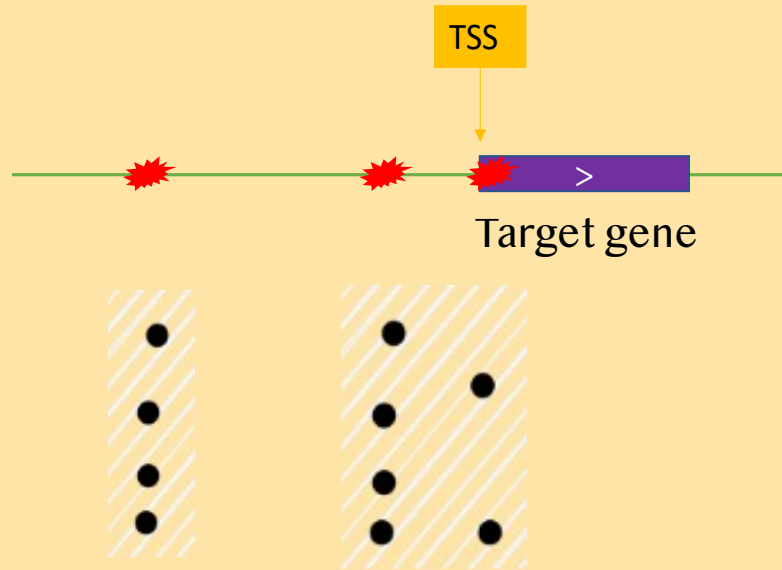
Before

After

ChIP-seq

RNA-seq

merge

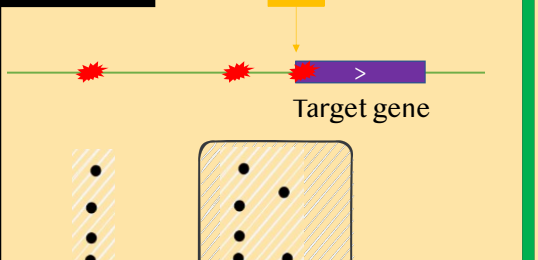


Preprocess


Data

TSS


Target gene



DBSCAN

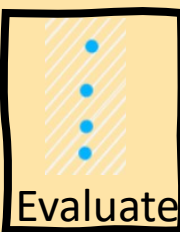


cluster1
cluster2



Analysis

Evaluate



$X \sim \mathcal{N}(\mu, \sigma^2)$

Hypotheses

$H_0 : \sigma^2 \leq \sigma_0^2$
 $H_1 : \sigma^2 > \sigma_0^2$
 $\hat{\sigma}^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2$

Test statistic

$\sim \chi^2_{n-1}$
 $\chi^2 = \frac{1}{\sigma_0^2} \sum_{i=1}^n (X_i - \bar{X})^2 = \frac{(n-1)\hat{\sigma}^2}{\sigma_0^2}$

result

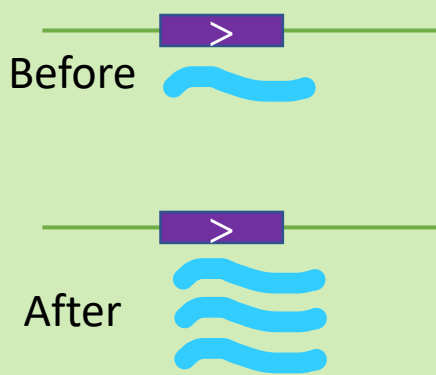
P(turn on)

Preprocess

Before

After

Fold Change = 3 : 1

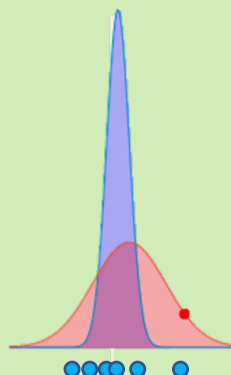


result

P(expressed)

Analysis: GMM

Log Fold Change = log 3

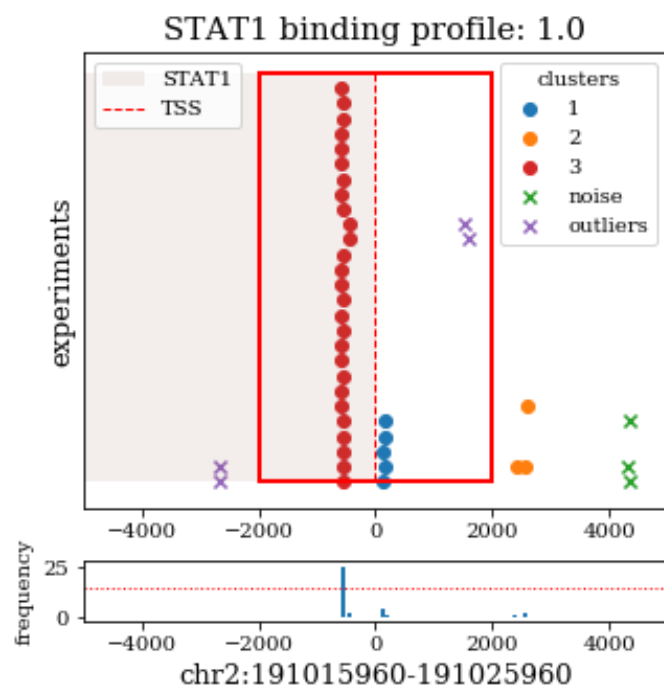


P(turn on) P(expressed)

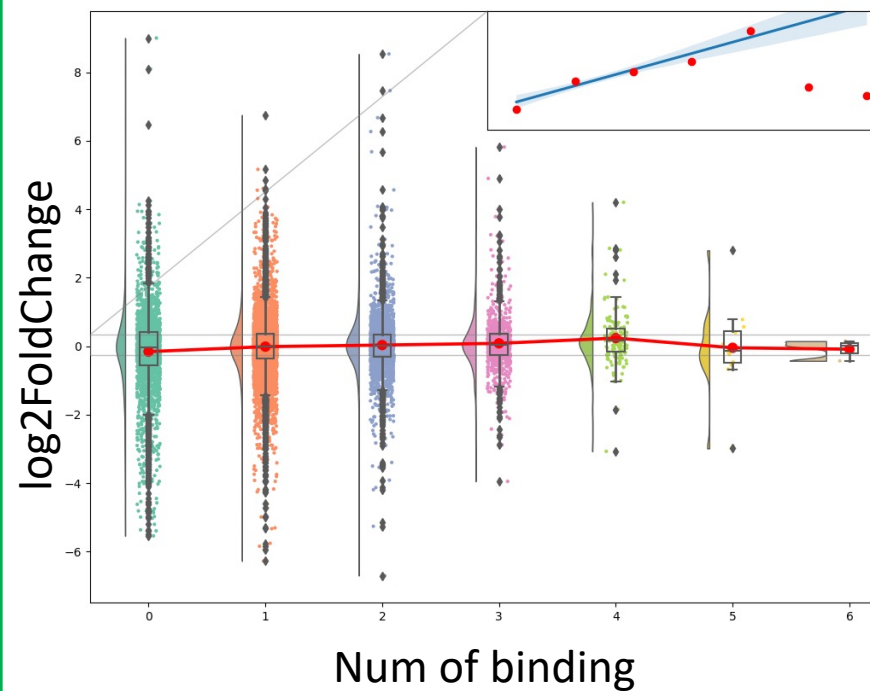
Independent

COBRA

ChIP-Seq



ChIP-Seq + RNA-Seq



Comparison

