

# SIR

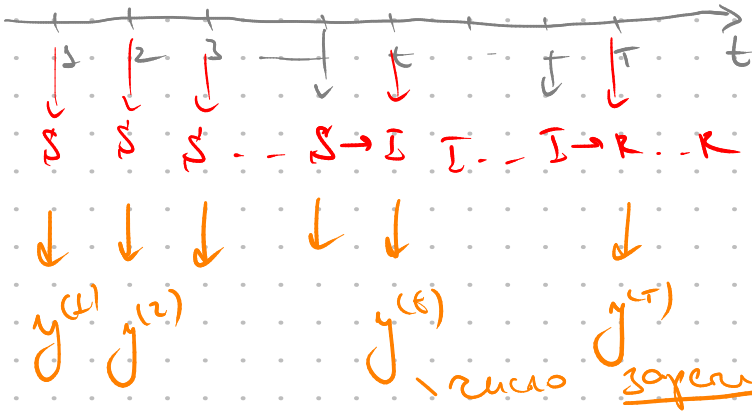
susceptible - infected - recovered

$$\bar{x}_j = (x_j^{(1)}, \dots, x_j^{(T)})$$

$$X = \{x_1, \dots, x_N\}$$

$$S = \{S, I, R\}$$

$$t: x_j^{(t)} \in \{S, I, R\}$$



$$S^{(t)} = |\{j \in \{1, \dots, N\} | x_j^{(t)} = S\}|$$

$$I^{(t)} = \dots \dots \dots I$$

$$R^{(t)} = \dots \dots \dots R$$

$$S^{(t)} + I^{(t)} + R^{(t)} = N$$

формулы в зависимости от t

1)  $p(x_j^{(t)} = I) = \pi_j$ ,  $p(x_j^{(t)} = S) = 1 - \pi_j$      $\bar{\pi} = (\pi_1, \dots, \pi_N)$

2)  $p(x_j^{(t)} \in y^{(t)} | x_j^{(t)} = I) = p$

$$p(y^{(t)} | I^{(t)}, p) = \text{Binomial}(y^{(t)} | I^{(t)}, p)$$

3)  $p(x_j^{(t+1)} = R | x_j^{(t)} = I) = \mu$      $p(x_j^{(t+1)} = I | x_j^{(t)} = I) = 1 - \mu$

4)  $\beta = p(\text{заражение от одного контакта})$

$$p(x_j^{(t+1)} = S | x_j^{(t)} = S) = (1 - \beta)^{I^{(t)}}$$

$$\bar{\theta} = \{\bar{\pi}, p, \mu, \beta\}$$

$$p(x_j^{(t+1)} = I | x_j^{(t)} = S) = 1 - (1 - \beta)^{I^{(t)}}$$

$$p(x_j^{(t+1)} | x_j^{(t)}) = \begin{matrix} S \\ I \\ R \end{matrix} \begin{pmatrix} (1-\beta)^{I^{(t)}} & 1 - (1-\beta)^{I^{(t)}} & 0 \\ 0 & 1 - \mu & \mu \\ 0 & 0 & 1 \end{pmatrix} \begin{matrix} S \\ I \\ R \end{matrix}$$

$$p(x, y | \bar{\theta}) = p(x^{(1)} | \bar{\pi}) p(y^{(1)} | x^{(1)}, p) \cdot p(x^{(2)} | x^{(1)}, \mu, \beta) \dots$$

$$X^{(t)} = \{x_1^{(t)}, \dots, x_N^{(t)}\}$$

$$\dots p(x^{(T)} | x^{(T-1)}, \mu, \beta) p(y^{(T)} | x^{(T)}, p) =$$

$$= \left( \prod_{i=1}^N \pi^{I[x_i^{(1)}=1]} (1-\pi)^{I[x_i^{(1)}=0]} \right) \times \left( \prod_{t=1}^T \left[ \left( \prod_{i=1}^N p(y^{(t)} | x_i^{(t)}, \beta, \mu) \right) p(y^{(t)}) \right] \right) \times \left( \prod_{t=1}^{T-1} \prod_{i=1}^N p(x_i^{(t+1)} | x_i^{(t)}, I^{(t)}, \beta, \mu) \right)$$

$$p(\bar{\theta} | y) = p(\bar{\pi}, \beta, \mu | y^{(1)}, \dots, y^{(T)}) \xrightarrow{\text{max}}_{\bar{\pi}, \beta, \mu}$$

$$\propto p(\theta) p(y | \theta) = p(\theta) \int p(x, y | \theta) dx$$

EM-algorithm

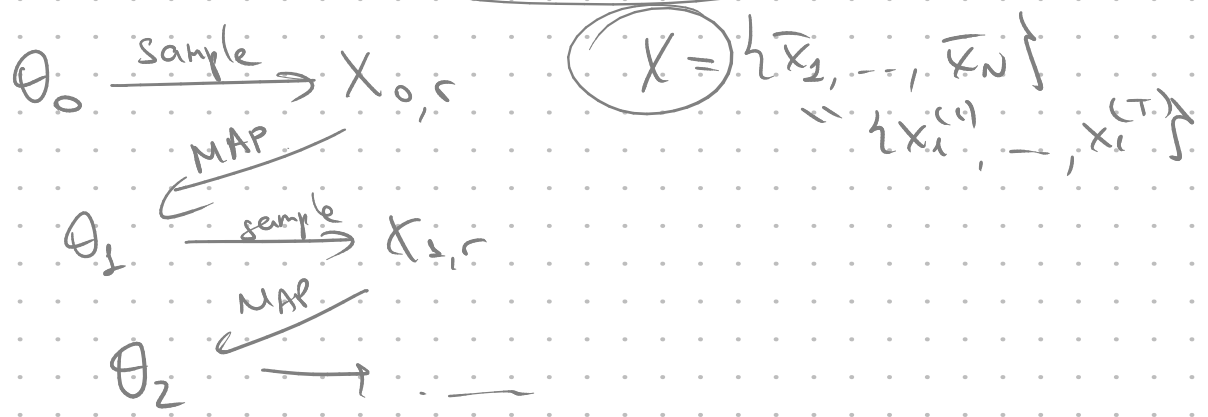
$$L(\theta, \theta_k) = E_{p(x | \theta_k, y)} [ p(x, y | \theta) p(\theta) ] \xrightarrow{\theta} \text{max}$$

Monte Carlo EM

$\theta_{k+1} = \text{argmax}_{\theta} [ \dots ]$

$$L(\theta, \theta_k) \approx \frac{1}{R} \sum_{r=1}^R p(x_r, y | \theta) p(\theta), \text{ where}$$

$$x_r \sim p(x | \theta_k, y)$$



$$X = \{ \bar{x}_1, \dots, \bar{x}_N \}$$

$$= \{ x_i^{(1)}, \dots, x_i^{(T)} \}$$

Sampling

$$X \sim p(x | \theta, y)$$

$$\{ \bar{x}_1, \bar{x}_2, \dots, \bar{x}_N \}$$

$$\bar{x}_j \sim p(\bar{x}_j | \theta, y, \underline{\bar{x}}_{-j})$$

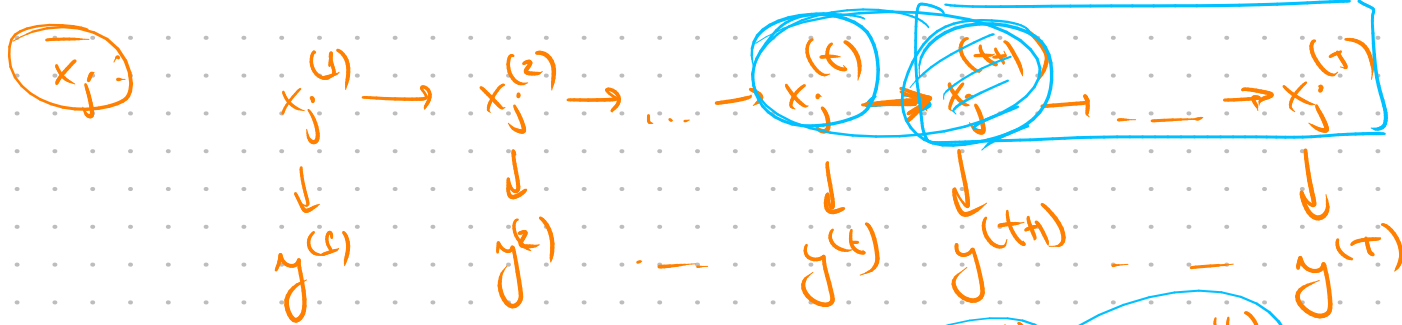
Gibbs sampling

$$j=1 \quad \bar{x}_1 \sim \dots$$

$$j=2 \quad \bar{x}_2 \sim \dots$$

$$\vdots$$

$$j=N \quad \bar{x}_N \sim \dots$$

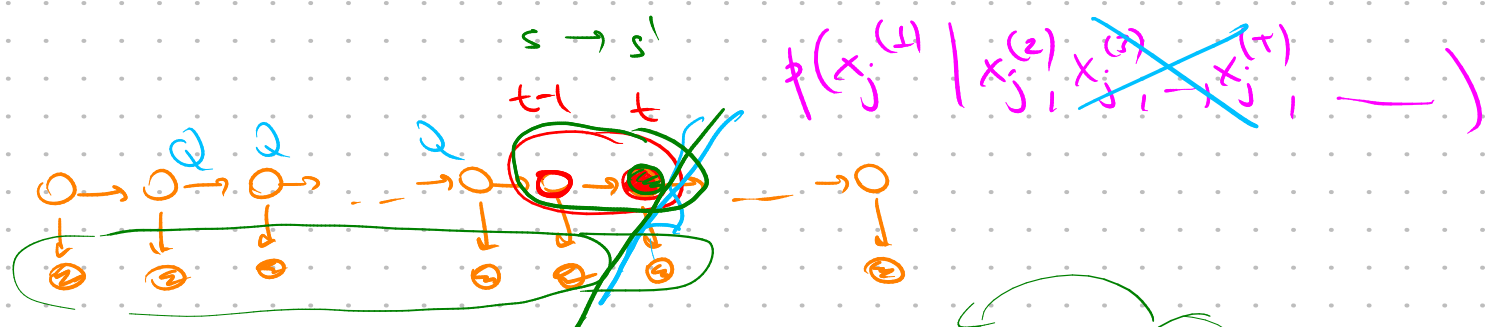


$$p(x_j^{(t+1)} | x_j^{(t)}, \theta, X_{-j}) = \begin{pmatrix} (1-\beta) I^{(t)} & 1-(1-\beta) I^{(t)} & 0 \\ 0 & 1-\mu & \mu \\ 0 & 0 & 1 \end{pmatrix}$$

$$\bar{x}_j \sim p(\bar{x}_j | \theta, X_{-j}, \mathbf{y})$$

Stochastic Viterbi algorithm

$$p(\bar{x}_j | X_{-j}, \mathbf{y}, \theta) = p(x_j^{(T)} | \dots) p(x_j^{(T-1)} | x_j^{(T)}, \dots) p(x_j^{(T-2)} | x_j^{(T-1)}, x_j^{(T)}, \dots)$$



$$q_{j,s,s'}^{(t)} = p(x_j^{(t)} = s', x_j^{(t-1)} = s | y_{-j}^{(1)}, \dots, y_j^{(t)}, X_{-j}, \theta)$$

$$p(x_j^{(t)} = s | x_j^{(t+1)} = s', y_{-j}^{(1)}, \dots, y_j^{(t)}, y_j^{(t+1)}, \dots, y_j^{(T)}, X_{-j}, \theta) \propto q_{j,s,s'}^{(t+1)}$$

$$S = \begin{pmatrix} S & I \\ R & I \end{pmatrix} \begin{matrix} (t) \\ q_{j,s,s'} \end{matrix}$$

$$q_{j,s,s'}^{(t)} \propto p(x_j^{(t)} = s', x_j^{(t+1)} = s, y_j^{(t)} | y_{-j}^{(1)}, \dots, y_j^{(t-1)}, X_{-j}, \theta) =$$

$$= p(x_j^{(t)} = s' | x_j^{(t-1)} = s, y_{-j}^{(1)}, \dots, y_j^{(t-1)}, X_{-j}, \theta) \times p(y_j^{(t)} | x_j^{(t)} = s', x_j^{(t-1)} = s, y_{-j}^{(1)}, \dots, y_j^{(t-1)}, X_{-j}, \theta)$$

(ir)  $\times p(x_j^{(t-1)} = s | y^{(t-1)}, x_{-j}, \theta) = \sum_{s''} p(x_j^{(t-1)} = s'' | y^{(t-1)}, x_{-j}, \theta)$

$= p(s' | s, x_j, \theta) \times \text{Binom}(y^{(t)} | I_{-j} + [s' = I], j) \times \sum_{s''} q_{j, s'', s}^{(t-1)}$

Stoch. Vit. algi

- creba karpelo:

$Q_j^{(t)}, Q_j^{(t)}, \dots, Q_j^{(t)}$  ye  $Q_j^{(t)} = s \binom{I_{-j} + [s' = I]}{j, s, s'}$

- creba harebo:

$p(x_j^{(t)} = s | x_j^{(t+1)} = s', x_{-j}, \theta, y) \propto q_{j, s, s'}^{(t+1)}$

Gibbs sampling:

- loop  
- for  $j = 1 \dots N$

-  $\bar{x}_j \sim \text{stoch Vit}(X_{-j})$

$\bar{\pi} = (\pi, \dots, \pi)$

$p(\theta | y, x) \propto p(\theta) p(x, y | \theta) = p(p) p(p) p(\mu) p(\pi) p(x, y | \theta)$

$\theta \rightarrow \max$

Beta( $p | a_p, b_p$ )  
 $\propto p^{a_p-1} (1-p)^{b_p-1}$

$\log p(\theta | y, x) = \text{const} + (a_p - 1) \log p + (b_p - 1) \log(1-p) +$   
 $+ (a_p - 1) \log p + (b_p - 1) \log(1-p) + (a_\mu - 1) \log \mu + (b_\mu - 1) \log(1-\mu) +$   
 $+ (a_\pi - 1) \log \pi + (b_\pi - 1) \log(1-\pi) +$   
 $+ \sum_{i=1}^N \left( [x_i^{(t)} = I] \log \pi + [x_i^{(t)} = S] \log(1-\pi) \right) +$   
 $+ \sum_{t=1}^T \left( y^{(t)} \log p + (I^{(t)} - y^{(t)}) \log(1-p) \right) +$   
 ~~$\sum_{t=1}^{T-1} \sum_{i=1}^N [x_i^{(t)} = R] \cdot \log 1 +$~~

$$+ \sum_{t=1}^{T-1} \sum_{i=1}^N [x_i^{(t)} = I] \cdot \left( [x_i^{(t+1)} = R] \log \mu + [x_i^{(t+1)} = I] \log(1-\mu) \right) +$$

$$+ \sum_{t=1}^{T-1} \sum_{i=1}^N [x_i^{(t)} = S] \cdot \left( [x_i^{(t+1)} = S] \cdot I^{(t)} \cdot \log(1-\beta) + [x_i^{(t+1)} = I] \cdot \left( P_i^{(t)} \log \beta + N_i^{(t)} \log(1-\beta) \right) \right)$$

presence-only data

$$P_i^{(t)} + N_i^{(t)} = I^{(t)}, \quad x_i^{(t)} = I \Rightarrow P_i^{(t)} \geq 1$$

# korr. → запущен # korr. → не зап.

$$E[P_i^{(t)} | x_i^{(t+1)} = I] = I^{(t)} \cdot p(\text{запущен от 1 korr.} | \geq 1 \text{ запущ.}) = I^{(t)} \cdot \frac{\beta}{1 - (1-\beta)^{I^{(t)}}}$$

$$P_i^{(t)} := I^{(t)} \cdot \frac{\beta}{1 - (1-\beta)^{I^{(t)}}}$$

$$N_i^{(t)} := I^{(t)} - P_i^{(t)}$$

$$a'_\pi = a_\pi + \sum_{i=1}^N [x_i^{(t)} = I], \quad b'_\pi = b_\pi + \sum_{i=1}^N [x_i^{(t)} = S]$$

$$= a_\pi + I^{(t)}, \quad = b_\pi + S^{(t)}$$

$$a'_y = a_y + \sum_{t=1}^T y^{(t)}, \quad b'_y = b_y + \sum_{t=1}^T (I^{(t)} - y^{(t)})$$

$$a'_\mu = a_\mu + \sum_{t=1}^{T-1} \sum_{i=1}^N [x_i^{(t)} = I, x_i^{(t+1)} = R], \quad b'_\mu = b_\mu + \sum_{t=1}^{T-1} \sum_{i=1}^N [x_i^{(t)} = I, x_i^{(t+1)} = I]$$

$$a'_\beta = a_\beta + \sum_{t=1}^{T-1} \sum_{i=1}^N P_i^{(t)} \cdot [x_i^{(t)} = S, x_i^{(t+1)} = I],$$

$$b'_\beta = b_\beta + \sum_{t=1}^{T-1} \sum_{i=1}^N \left( [x_i^{(t)} = S, x_i^{(t+1)} = S] \cdot I^{(t)} + [x_i^{(t)} = S, x_i^{(t+1)} = I] \cdot N_i^{(t)} \right)$$

# EM-algorithm:

- for  $k=1, \dots$

- E-war: Gibbs sampling given  $\theta_k$

- loop  $j=1, \dots, N, 1, \dots, N, \dots$

stoch.  
Viterbi  
alg.

-  $\bar{x}_j \sim p(\bar{x}_j | X_{-j}, y, \theta_k)$ :

- for  $t=1..T$ : compute  $Q_j^{(t)}$

- for  $t=T..1$ :  $x_j^{(t)} \sim p(x_j^{(t)} | x_j^{(t+1)}, y, X_{-j}, \theta)$

- save  $X_r$

- M-war:

-  $\theta_{k+1} = \text{argmax } p(\theta_{k+1} | X, y)$

- presence-only data

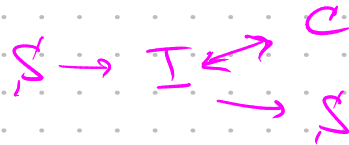
## SIR extensions

①  $SIR \rightarrow SEIR$   $S \rightarrow E \rightarrow I \rightarrow R$

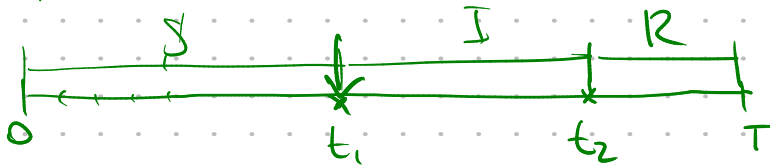
$SEIS$   $S \rightarrow E \rightarrow I \rightarrow S$

$SEIRS$   $S \rightarrow E \rightarrow I \rightarrow R \rightarrow S$

$SIC_S$  - carrier

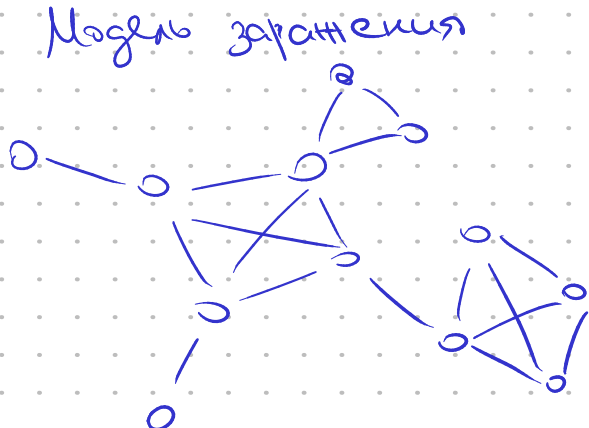


② Генерация бьеса



$R_0 = \beta \cdot E[\# \text{ contacts}]$   
 (Note:  $\beta$  is circled in pink and labeled 'берисе', and  $E[\# \text{ contacts}]$  is circled in blue and labeled 'количество')

③ Модель заражения



$(1-\beta)^{I^{(t)}}$   
 $(1-\beta)^{I^{(t)}} \cap \text{Nei}(x_j)$