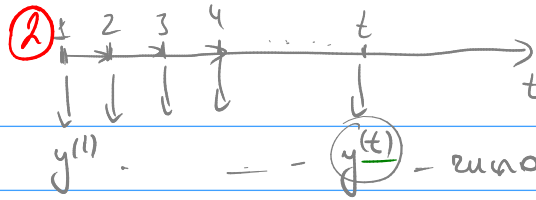


# SIR - model - susceptible - infected - recovered

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

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

①  $S \rightarrow I \rightarrow R$



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

$$i \in \{1, \dots, N\} \mid x_i^{(t)} = S$$

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

③  $\beta = p(\text{заболевание от одного контакта})$

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

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

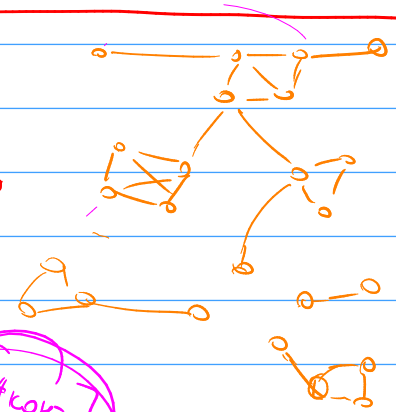
1)  $p(x_i^{(1)} = I) = \pi_i, p(x_i^{(1)} = S) = 1 - \pi_i$

2)  $p(x_i^{(t)} \in y^{(t)} \mid x_i^{(t-1)} = I) = p$

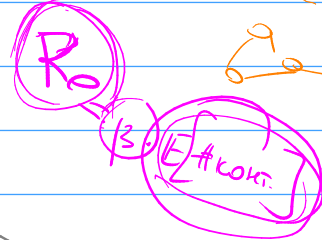
$$p(y^{(t)} \mid I^{(t)}, p) = \text{Binomial}(y^{(t)} \mid I^{(t)}, p)$$

3)  $p(x_i^{(t+1)} = R \mid x_i^{(t)} = I) = \mu$

- ① SEIR, SEIRS
- ② Непрерывное время
- ③ Модель заражения



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



$$X = \{x_i^{(t)} \mid i, t\}$$

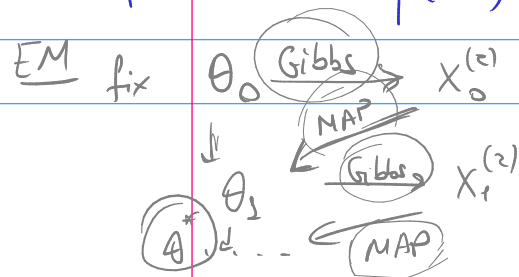
$$y = \{y^{(t)} \mid t\}$$

$$p(X, y \mid \theta) = p(x^{(1)} \mid \pi) p(x^{(2)} \mid x^{(1)}, \beta, \mu) \dots p(x^{(T)} \mid x^{(T-1)}, \beta, \mu) \cdot p(y^{(1)} \mid x^{(1)}, p) \dots p(y^{(T)} \mid x^{(T)}, p)$$

$$= \left( \prod_i \pi_i^{[x_i^{(1)}=I]} (1-\pi_i)^{[x_i^{(1)}=S]} \right) \cdot \prod_{t=1}^{T-1} \prod_i p(x_i^{(t+1)} \mid I^{(t)}, \beta, \mu) \cdot \prod_{t=1}^T p(y^{(t)} \mid I^{(t)}, p)$$

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

$$p(\theta \mid y) = \mathbb{E}_{p(x \mid \theta)} [p(\theta) p(y \mid x, \theta)] \approx \frac{1}{R} \sum_{r=1}^R p(\theta) p(y \mid x_r, \theta)$$



$$x_2 \sim p(x \mid \theta, y)$$

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

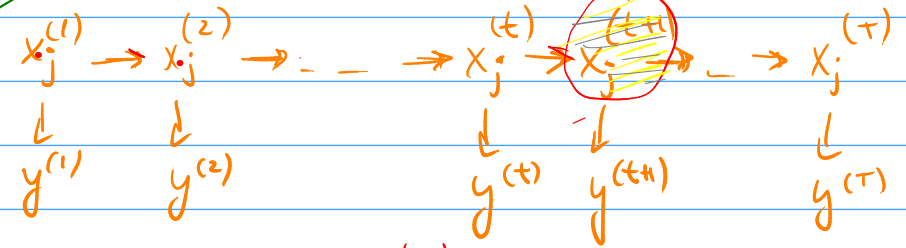
$$\approx \{\bar{x}_1, \bar{x}_2, \dots, \bar{x}_N\}$$

$$\approx (x_1^{(1)}, x_1^{(2)}, \dots, x_1^{(T)})$$

Gibbs sampling 2

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

Hidden Markov models



Stochastic Viterbi algorithm

$$p(\bar{x}_j | X_j, \bar{y}, \theta)$$

$$p(x_j^{(t)} | x_j^{(t+1)}, x_j^{(t-1)}, \bar{x}_j, \bar{y}, \theta)$$

$$= p(x_j^{(t)} | x_j^{(t+1)}, x_j^{(t-1)}, \bar{y}^{(t)}, \theta)$$

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

$$x_j^{(t)}$$

$$p(x_j^{(t)} = s | \dots) \propto \pi_{\cdot}(s) \cdot p(y_j^{(t)} | x_j^{(t)} = s)$$

$$q_{j, s', s}^{(t)} = p(x_j^{(t)} = s, x_j^{(t-1)} = s' | \bar{y}, X_j, \theta) = p(x_j^{(t)} = s | x_j^{(t-1)} = s', \bar{y} | X_j, \theta) \cdot$$

$$\cdot p(y_j^{(t)} | x_j^{(t)} = s, x_j^{(t-1)} = s', y^{(1:t-1)} | X_j, \theta)$$

$$\cdot p(x_j^{(t-1)} = s' | \bar{y}^{(t-1)}, X_j, \theta) =$$

$$\sum_{s''} p(x_j^{(t-1)} = s', x_j^{(t-2)} = s'' | \bar{y}^{(t-1)}, X_j, \theta)$$

$$= p(x_j^{(t)} = s | x_j^{(t-1)} = s', X_j, \theta) \cdot \text{Known}(y_j^{(t)} | I_{-j}^{(t)} + [s], \theta) \cdot \left( \sum_{s''} q_{j, s'', s'}^{(t-1)} \right)$$

1) Forward pass  $Q_j^{(t)} = \begin{pmatrix} s \\ I \\ R \\ \uparrow \\ q_{j, s', s}^{(t)} \end{pmatrix}$

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

2) Случай.  $X_j^{(T)}, X_j^{(T-1)}, \dots, X_j^{(1)}$



$$p(x_j^{(T)} = s | y, X_{-j}^{(T)}(\theta)) = \sum_{s'} p(x_j^{(T)} = s, x_j^{(T-1)} = s' | \dots) = \sum_{s'} q_{j, s', s}^{(T)}$$

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

Алгоритм: 1) Init  $X$  так, чтобы  $y$  могли вычисляться

- Gibbs {
- 2) Итерации:
    - шаг  $j$
    - сгрупп.  $Q_j^{(t)} = (q_{j, s', s}^{(t)})_{s, s' \in \{S, I, R\}}$
    - случай.  $\bar{x}_j \sim p(\bar{x}_j | X_{-j}, y, \theta)$  при ном.  $Q$
    - замени  $\bar{x}_j$  на выборки

3) Оценки параметров модели при фикс.  $X$   $\theta = \{p, \mu, \beta, \pi\}$

$$p(\theta | y, X) \propto \underbrace{p(\theta)}_{p(p)p(\mu)p(\beta)p(\pi)} p(y, X | \theta)$$

$$p(p) = \text{Beta}(p | a_p, b_p) \\ \dots \mu, \beta, \pi \dots$$

$$\log p(\theta | y, X) = \text{const} + (a_p - 1) \log p + (b_p - 1) \log(1 - p) + (k_\mu - 1) \log \mu + (b_\mu - 1) \log(1 - \mu) \\ + (a_\beta - 1) \log \beta + (b_\beta - 1) \log(1 - \beta) + (a_\pi - 1) \log \pi + (b_\pi - 1) \log(1 - \pi)$$

$$a_p' = a_p + \sum_{t=1}^T y^{(t)} \\ b_p' = b_p + \sum_{t=1}^T (I - y^{(t)}) \\ + \sum_{i=1}^N \left[ [x_i^{(t)} = I] \log \pi + [x_i^{(t)} = S] \log(1 - \pi) \right] \\ + \sum_{t=1}^T \left( y^{(t)} \log p + (I - y^{(t)}) \log(1 - p) \right) \\ + \sum_{i=1}^N \sum_{t=1}^{T-1} \left( [x_i^{(t)} = I, x_i^{(t+1)} = R] \log \mu + [x_i^{(t)} = I, x_i^{(t+1)} = I] \log(1 - \mu) \right)$$

$$+ \sum_{t=1}^{T-1} \sum_{i: x_i^{(t)} = S} \left( P_i^{(t)} \log \beta + N_i^{(t)} \log(1 - \beta) \right)$$

# контактов, от k-пар j-p
# контактов, от k-пар j-p
 $P_i^{(t)} + N_i^{(t)} = I^{(t)}$

Presence-only data

- если  $x_i^{(t+1)} = S$ , то  $P_i^{(t+1)} = 0$ ,  $N_i^{(t+1)} = I^{(t+1)}$
- если  $x_i^{(t+1)} = I$ , то  $P_i^{(t+1)} \geq 1$ ,  $N_i^{(t+1)} = I^{(t+1)} - P_i^{(t+1)}$

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

$$P_i^{(t)} = \begin{cases} 0, & X_i^{(t+1)} = S \\ \frac{\beta}{I^{(t)} - (1-p)I^{(t-1)}}, & X_i^{(t+1)} = I \end{cases}, \quad N_i^{(t)} = I^{(t)} - P_i^{(t)}$$

$$a_p' = a_p + \sum_{t=1}^{T-1} \sum_{i: X_i^{(t)} = S} P_i^{(t)}$$

$$b_p' = b_p + \dots N_i^{(t)}$$