

6.4

H - у 1 сумки
¬H - у 2 сумки

E - МАХ20

$$P(H) = 0,4 = \frac{2}{5}$$

$$P(E|H) = \frac{4}{8} = \frac{1}{2}$$

$$P(\neg H) = 0,6 = \frac{3}{5}$$

$$P(E|\neg H) = \frac{4}{6} = \frac{2}{3}$$

$$P(E) = \frac{2}{5} \cdot \frac{1}{2} + \frac{3}{5} \cdot \frac{2}{3}$$

" $\frac{1}{5}$ " $\frac{2}{5}$

$$P(H|E) = \frac{P(H) \cdot P(E|H)}{P(E)} = \frac{\frac{1}{5}}{\frac{1}{5} + \frac{2}{5}} = \frac{1}{3} < \frac{2}{5} \text{ " } P(H)$$

Верно

6.2

$$P\left(\begin{matrix} x_1 \\ x_2 \end{matrix}\right) = 0,4 \mathcal{N}\left(\begin{matrix} 10 \\ 2 \end{matrix}\right), \begin{matrix} 1 \\ 0 \end{matrix}\right) + 0,6 \mathcal{N}\left(\begin{matrix} 0 \\ 0 \end{matrix}\right), \begin{matrix} 8,4 \\ 2,0 \\ 1,7 \end{matrix}\right)$$

a) $P(x_1) = 0,4 \cdot \mathcal{N}(x_1 | 10, 1) + 0,6 \mathcal{N}(x_1 | 0, 8,4)$

$P(x_2) = 0,4 \mathcal{N}(2, 1) + 0,6 \mathcal{N}(0, 1,7) \checkmark$ Bsp 10

b) $E(x_1) = 0,4 \cdot 10 + 0,6 \cdot 0 = 4 \checkmark$

$E(x_2) = 0,4 \cdot 2 + 0,6 \cdot 0 = 0,8 \checkmark$

Modes $x_1: 10, 0$

Modes $x_2: 2, 0$

Medians:

$\int_{-\infty}^a P(x) dx = \frac{1}{2} \quad ?$

c) $E(x) = \int P(x) x dx =$

$= 0,4 \cdot \begin{bmatrix} 10 \\ 2 \end{bmatrix} + 0,6 \begin{bmatrix} 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 4 \\ 4/5 \end{bmatrix}$ Theorem 6.12

Mode:

$\frac{dP(x)}{dx} =$?

$$x_0 \sim N(\mu_0, \Sigma_0) \quad x_t = Ax_{t-1} + w, \quad w \sim N(0, Q)$$

$$y_t = Cx_t + v, \quad v \sim N(0, R)$$

[6.5]

a) Gaussian form for $p(x_1, x_2, \dots, x_T)$
 \because Gauss is linear

$$x_2 = Ax_1 + w = A(Ax_0 + w) + w = A^2x_0 + (A+1)w$$

$$x_3 = Ax_2 + w = A^3x_0 + (A^2+A+1)w$$

$$p(x_0) = N_0(\mu_0, \Sigma_0)$$

$$p(x_1) = \dots$$

$$x_1 = Ax_0 + w \sim N$$

$$x_0 \sim N_0 \quad w \sim N(0, Q)$$

$$x_2 \sim N$$

b) $\int p(x_t | y_1, \dots, y_t) = N(\mu_t, \Sigma_t)$
 1) $p(x_{t+1} | \vec{y}) = ?$
 $p(Ax_t + w | y_1, \dots, y_t) = ?$

a) Form of $p(x_0, x_1, \dots, x_T)$ is $N^S \times N^S \times \dots \times N^S$ is Gaussian \because joint is N^S .
 Beweis

b) $\int p(x_t | y_1, \dots, y_t) = N(\mu_t, \Sigma_t)$

1) $p(x_{t+1} | y_1, \dots, y_t) = ?$
 $p(x_{t+1} | \vec{y}) = \frac{p(x_{t+1}, \vec{y})}{p(\vec{y})} = \frac{p(Ax_t + w, \vec{y})}{p(\vec{y})}$
 $= \frac{p(Ax_t, \vec{y}) \cdot p(w, \vec{y})}{p(\vec{y})} = ?$

b) $\int p(\vec{x}_t | \vec{y}_1, \dots, \vec{y}_t) = N(\mu_t, \Sigma_t)$

$p(\vec{x}_{t+1} | \vec{y}_{1:t}) = N(\vec{x}_{t+1} | \mu_t, \Sigma_t)$
 \Downarrow
 $Ax_t + w$

[E.3] $p(x|M) = M^x (1-M)^{1-x}$, $x \in \{0, 1\}$
 not / compiles

Beta conj. prior

$$p(M|x) = \frac{p(x|M) \cdot p(M)}{p(x)}$$

posterior prior

Beta prior:
 $p(M|\alpha, \beta) \propto M^{\alpha-1} (1-M)^{\beta-1}$

$$p(x|M) \cdot p(M|\alpha, \beta) = M^x (1-M)^{1-x} \cdot M^{\alpha-1} (1-M)^{\beta-1} = M^{x+\alpha-1} (1-M)^{\beta-x-1}$$

$$= M^{x+\alpha-1} (1-M)^{\beta-x-1} = p(M|x+\alpha, \beta-x+1)$$

$$p(X|M) = \prod_{i=1}^N M^{x_i} (1-M)^{1-x_i}$$

$$p(M|x) \propto p(M) p(X|M) =$$

Beta

$$= p(M|\alpha + \sum x_i, \beta + N - \sum x_i)$$

Beweis