

$$x_1 \dots x_n \quad (x_i, y_i)$$

$$x_i \in \mathbb{R}^n \quad y_i \in \mathbb{R}$$

$$\hat{f}(x_i) = y_i + \xi$$

$$\theta^T x_i = \hat{y}_i + \xi$$

$$1) L(y_i, \hat{y}_i)$$

$$\|y_i - \hat{y}_i\|_2$$

$$\|y_i - \theta^T x_i\|_2$$

$$2) \theta_0 \sim N(0, 1)$$

$$3) \theta_{n+1} = \theta_n - \alpha \cdot \nabla_{\theta} L(y_i, \hat{y}_i)$$

Установим связь:

$$x_1, x_2 \dots x_n, \quad P(X|\theta)$$

$$\theta = \arg \max_{\theta} P(x_1, x_2 \dots x_n | \theta), \quad \Theta$$

i.i.d.

$$\Theta \text{ для } \max_{\theta} \prod_{i=1}^n P(x_i | \theta) =$$

likelihood

$$= \arg_{\theta} \max \sum_{i=1}^n \log P(x_i | \theta)$$

H T
5 3

$$P(X | \theta) = \theta^x (1-\theta)^{1-x}$$

$$\hat{\theta} = \arg_{\theta} \max \sum_{i=1}^n \log \left(\theta^x (1-\theta)^{1-x} \right) =$$

$$= \sum_{i=1}^n x_i \log \theta + (1-x_i) \log (1-\theta) =$$

$$= \sum_{i=1}^n [x_i \log \theta - x_i \log (1-\theta)] + n \log (1-\theta) =$$

$$\frac{\partial}{\partial \theta}$$

$$\sum_{i=1}^n \left[\frac{x_i}{\theta} + \frac{x_i}{1-\theta} \right] - \frac{n}{1-\theta} = 0.$$

$$\sum_{i=1}^n \frac{x_i}{\theta - \theta^2} - \frac{\theta n}{\theta - \theta^2}$$

$$\hat{\theta} = \frac{1}{n} \sum_{i=1}^n x_i \quad \frac{1}{8} \cdot 5 = \frac{5}{8}$$

H T
8 0.

$$\hat{\theta} = 1$$

$$P(\theta | x) = \frac{P(x | \theta) P(\theta)}{P(x)}$$

↑ likelihood. ↑ Prior.
↑ evidence

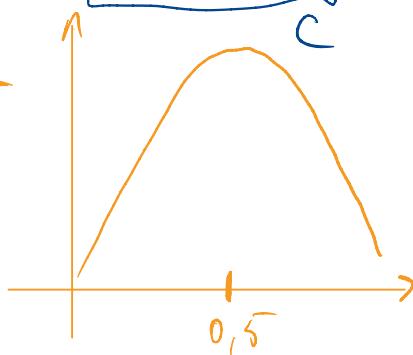
aposteriori.

$$\hat{\theta} = \arg \max_{\theta} \prod_{i=1}^n P(x_i | \theta) P(\theta) \quad \text{MAP.}$$

x_i i.i.d.

$$P(\theta | \alpha, \beta) = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha) \Gamma(\beta)} \theta^{\alpha-1} (1-\theta)^{\beta-1}$$

$$\alpha = \beta = 3$$



$$P(\theta) = C \cdot \theta^2 (1-\theta)^2$$

$$\begin{aligned}\hat{\theta} &= \arg \max_{\theta} \left[\prod_{i=1}^n P(x_i | \theta) \right] \cdot P(\theta) = \\ &= \sum_{i=1}^n \log \left(\theta^{x_i} (1-\theta)^{1-x_i} \right) + \log P(\theta) =\end{aligned}$$

$$= \left[\sum_{i=1}^n x_i \log \theta + (1-x_i) \log(1-\theta) \right] + 2 \log \theta + \\ + 2 \log(1-\theta)$$

$$\frac{\partial}{\partial \theta}$$

$$\left[\sum_{i=1}^n \frac{x_i}{\theta - \theta^2} \right] - \frac{\theta n}{\theta - \theta^2} + \frac{2}{\theta} - \frac{2}{1-\theta} =$$

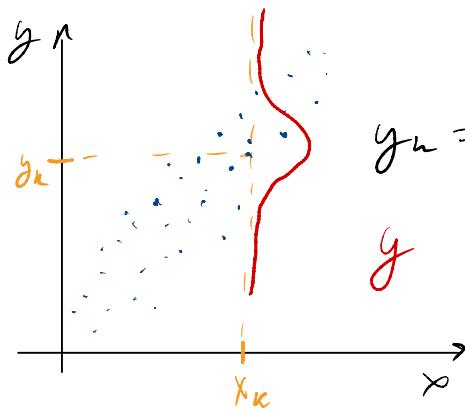
$$= \sum_{i=1}^n \frac{x_i}{\theta - \theta^2} - \frac{\theta n - 2 + 4\theta}{\theta - \theta^2}$$

$$\sum_{i=1}^n x_i - \theta(n+4) + 2$$

$$\theta = \frac{1}{n+4} \left[\sum_{i=1}^n x_i \right] + 2$$

$\frac{10}{12}$

$x_i \ y_i$



$$y_n = \hat{f}(x_n, \theta) + \epsilon \quad \epsilon \sim N(0, \sigma^2)$$

$$y \sim N(\hat{f}(x, \theta), \sigma^2).$$

$$P(y | M, \theta) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(y - \mu)^2}{2\sigma^2}}$$

$$\hat{\theta} = \arg \max \sum_{i=1}^n \left[\log \frac{1}{\sqrt{2\pi}\sigma} - \frac{1}{2\sigma^2} (y_i - \hat{f}(x_i, \theta))^2 \right]$$

$$= \arg \max_{\theta} -\frac{1}{2\sigma^2} \sum_{i=1}^n [y_i - \hat{f}(x_i, \theta)]^2 + n \log \frac{1}{\sqrt{2\pi}\sigma}$$

MSE

$$\hat{\theta} = \arg \max_{\theta} \prod_{i=1}^n P(x_i | \theta) \cdot P(\theta)$$

$$\theta \sim N(0, \sigma^2)$$

$$\frac{\sqrt{\lambda}}{\sqrt{(2\pi)^m} G^2} \cdot e^{-\frac{1}{2G^2}(y_i - \theta^T x_i)^2 - \frac{\lambda}{2}\theta^T \theta}$$

$$= \underset{\theta}{\operatorname{arg\,max}} -\frac{1}{2G^2} \sum_{i=1}^n (y_i - \theta^T x_i)^2 - \frac{\lambda}{2} \theta^T \theta$$

MSE L2 Reg.

$$P(x|\eta) = h(x) g(\eta) \cdot e^{\eta^T u(x)}$$

$$P(x|M) = M^x (1-M)^{1-x}$$

$$P(x|u) = e^{x \ln u + (1-x) \ln (1-u)} =$$

$$= (1-u) e^{\ln(\frac{u}{1-u})x} \quad \delta(x)$$

$$\eta = \ln \frac{u}{1-u} \quad u = \frac{1}{1 + \exp(-\eta)}$$

$$P(x|\eta) = g(-\eta) e^{\eta x}$$

$$u(x) = x \quad g(\eta) = \sigma(-\eta)$$

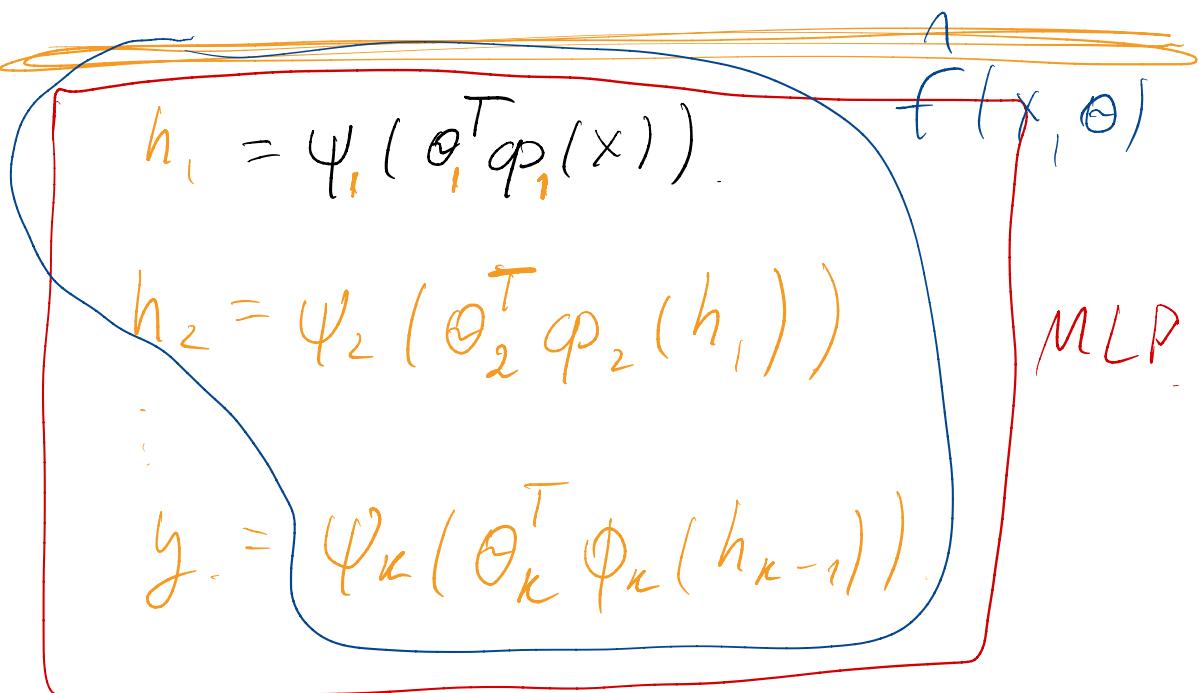
$$h(x) = 1$$

GLM.

$$y = \psi(\theta^T \varphi(x))$$

↑ ↑
activation basis function.

ψ^{-1} - link function.



- 1) Конное море.
- 2) Слоновье море
- 3) Море паранеприб.