

# Теорема о разложении.

①

Докажем, что

$$E_{x,y} E_x (y - a_x(x))^2 = \text{noise} + \text{bias} + \text{variance},$$

где

$$E_{x,y} (y - E(y|x))^2 - \text{noise}$$

$$E_{x,y} (E(y|x) - E_x a_x(x))^2 - \text{bias}$$

$$E_{x,y} E_x (a_x(x) - E_x a_x(x))^2 - \text{variance}$$

$$\begin{aligned} & E_{x,y} E_x (y - a_x(x))^2 = \\ & = E_{x,y} E_x (y - E(y|x) + E(y|x) - a_x(x))^2 = \\ & = \underbrace{E_{x,y} E_x [y - E(y|x)]^2}_{\substack{\text{noise,} \\ \text{т.к. не зависит от } x}} + E_{x,y} E_x [a_x(x) - \end{aligned}$$

$$- E_x a_x(x) + E_x a_x(x) - E(y|x)]^2 +$$

$$+ 2 E_{x,y} E_x [(y - E(y|x)) (E(y|x) - a_x(x))] =$$

$$= \text{noise} + E_{x,y} E_x [E_x a_x(x) - E(y|x)]^2 +$$

bias



$$\begin{aligned}
 & + E_{x,y} E_x \left[ a_{x,e} - E_x a_{x,e} \right]^2 + \\
 & + 2 E_{x,y} E_x \left[ (a_{x,e} - E_x a_{x,e}) (E_x a_{x,e} - E(y|x)) \right] + \\
 & + \underbrace{2 E_{x,y} (y - E(y|x)) E_x (E(y|x) - a_{x,e}(x))}_{0} =
 \end{aligned}$$

$\approx$  noise + bias + variance +

$$+ 2 E_{x,y} \left[ E_x a_{x,e} - E(y|x) \right] \underbrace{E_x [E_x a_{x,e} + a_{x,e}]}_{0}$$

$\approx$  noise + bias + variance  $\square$

$$(2) \quad a(x) = \frac{1}{M} \sum_{m=1}^M a_m(x)$$

$$\bullet \text{ bias}(a(x)) = E_{x,y} (E(y|x) - E_x \left( \frac{1}{M} \sum_{m=1}^M a_m(x) \right))^2$$

не линейности функции:  
 $\approx$  и равномерно распределены  $a_i(x)$

$$E_{x,y} \left( E(y|x) - \frac{1}{M} \cdot M E_x(a_i(x)) \right)^2 =$$

$$= \text{bias}(a_i(x)) \quad \forall i \in \{1, \dots, M\}$$

$$\begin{aligned}
 \bullet \text{ variance}(a(x)) &= E_{x,y} E_x \left( \frac{1}{M} \sum_{i=1}^M a_i(x) - \right. \\
 &\left. - E_x \frac{1}{M} \sum_{i=1}^M a_i(x) \right)^2 =
 \end{aligned}$$



$$= \frac{1}{M^2} E_{x,y} E_{x,e} \left( \sum_{i=1}^M (a_i(x) - E_{x,e}(a_i(x))) \right)^2$$

②  $a_i(x)$  - независимы  $\Rightarrow$  некоррелированы  $\Rightarrow$

$$E_{x,e} (a_i(x) - E_{x,e}(a_i(x))) (a_j(x) - E_{x,e}(a_j(x))) = 0$$

$\forall i, j : j \neq i$

③  $\frac{1}{M^2} E_{x,y} E_{x,e} M (a_i(x) - E_{x,e}(a_i(x))) =$

$$= \frac{1}{M} \text{variance}(a_i(x)) \quad \forall i \in 1, \dots, M$$

③  $X_1, \dots, X_M \sim \text{i.i.d. p.s.v. с дисперсией } \sigma^2$

$\forall i, j : i \neq j \quad \text{corr}(X_i, X_j) = \rho. \quad EX_i = a$

$$\begin{aligned} \text{D} \bar{X} &= D \left( \frac{1}{M} \sum_{i=1}^M X_i \right) = E \left( \frac{1}{M} \sum_{i=1}^M X_i - a \right)^2 = \\ &= \frac{1}{M^2} E \left( \sum_{i=1}^M (X_i - a) \right)^2 = \frac{1}{M^2} \left( \sum_{i=1}^M E(X_i - a)^2 + \right. \\ &+ \sum_{\substack{i,j=1 \\ i \neq j}}^M (X_i - a)(X_j - a) \Big) = \frac{1}{M} \sigma^2 + \frac{1}{M^2} 2C_M^2 \rho \sigma^2 = \\ &= \frac{\sigma^2}{M} + \frac{M(M-1)}{M^2} \rho \sigma^2 = \frac{\sigma^2}{M} (1 + (M-1)\rho) \\ &= \rho \sigma^2 + (1-\rho) \frac{\sigma^2}{M} \end{aligned}$$