

Bias - Variance decomposition

$$TSS = ESS + RSS$$

$$\begin{aligned}
 L(M) &= E_{x,y} [E_x [(y - M(X)(x))^2]] = \\
 &= \underbrace{E_{x,y} [(y - E[y|x])^2]}_{\text{шум}} + \underbrace{E_x \left[\left(\underbrace{E_x [M(X)(x)]}_{\text{Смещение}} - \underbrace{E[y|x]}_{\text{Смещение}} \right)^2 \right]}_{\text{Смещение}} + \\
 &\quad \underbrace{E_x \left[E_x \left[(M(X)(x) - E_x [M(X)(x)])^2 \right] \right]}_{\text{Разброс}}
 \end{aligned}$$

Что-то с регуляризацией

$$\boxed{y = wx}, \quad X = \{(x_1, y_1), \dots, (x_l, y_l)\}$$

$$\boxed{L(w) = \frac{1}{l} \sum_{i=1}^l (y_i - wx_i)^2 + \lambda w^2} = w(X)$$

$$M(X)(x) = w(X) x$$

$$\hat{\beta} = (X^T X)^{-1} X^T y$$

$$w(X) = \frac{\sum x_i y_i}{\sum x_i^2 + \lambda}$$

1) X - детерминированы

2) X - стохастические

Детерминированный X

$$y = f(x) + \varepsilon \quad , \quad \varepsilon - \text{iid} \quad , \quad E(\varepsilon) = 0 \quad , \quad \text{Var}(\varepsilon) = \sigma^2$$

$$y \sim (f(x), \sigma^2)$$

Пусть истинная зависимость:

$$\boxed{f(x) = \alpha x}$$

1) МЗМ. $E_{x,y}[(y - E[y|x])^2]$

$$E(y|x) = E(f(x) + \varepsilon | x) = f(x) \quad \text{Var}(x) = E(x^2) - E(x)^2$$

$$\begin{aligned} E_{x,y}[(\cancel{f(x)} + \varepsilon - \cancel{f(x)})^2] &= E_{x,y}[\varepsilon^2] = \\ &= \text{Var}(\varepsilon) + \underbrace{E(\varepsilon)^2}_0 = \sigma^2 \end{aligned}$$

2) Смещение $E_x[\underbrace{E_x[\mu(x)(x)]}_{\text{true}} - \underbrace{E[y|x]}_{\text{fit}}]^2]$

$$X = ((x_1, y_1), \dots, (x_n, y_n))$$

$$E_x[\mu(x)(x)] = E_x[\omega(x) \cdot x] = E_x[\omega(x)](x)$$

$$E_x[\omega(x)] = E_x\left[\frac{\sum x_i y_i}{\sum x_i^2 + \lambda}\right] = E_x\left[\frac{\sum x_i (f(x_i) + \varepsilon_i)}{\sum x_i^2 + \lambda}\right] =$$

$$\begin{aligned}
 &= E_{\varepsilon} \left[\frac{\sum x_i (a x_i + \varepsilon_i)}{\sum x_i^2 + \lambda} \right] = E_{\varepsilon} \left[a \frac{\sum x_i^2}{\sum x_i^2 + \lambda} + \frac{\sum x_i \varepsilon_i}{\sum x_i^2 + \lambda} \right] \\
 &= a \frac{\sum x_i^2}{\sum x_i^2 + \lambda} + \underbrace{\frac{\sum x_i E \varepsilon_i}{\sum x_i^2 + \lambda}}_{=0}
 \end{aligned}$$

$$\begin{aligned}
 &E_x[\mu(x|x)] - E[y|x] \quad \sum x_i^2 = S \\
 &= \left(a \frac{\sum x_i^2}{\sum x_i^2 + \lambda} x - a x \right) = \left(a \frac{S}{S + \lambda} x - a x \right)
 \end{aligned}$$

$$\frac{S}{S + \lambda} = 1 \Rightarrow \lambda = 0$$

$$E_x \left[(E_x[\mu(x|x)] - E[y|x])^2 \right] = \left(\frac{S}{S + \lambda} - 1 \right)^2 a^2 x^2$$

$$\text{Разброс: } \underbrace{E_x \left[E_x \left[(\mu(X|x) - E_x[\mu(X|x)])^2 \right] \right]}_{\text{Разброс}}$$

$$\text{Var}_x(\mu(X|x)) = \text{Var}_x \left(\frac{\sum x_i (a x_i + \varepsilon_i)}{\sum x_i^2 + \lambda} x \right) =$$

$$\begin{aligned}
 &= \text{Var}_X \left(\underbrace{\alpha \frac{\sum x_i^2}{\sum x_i^2 + 1}}_{\sigma^2} x + \frac{\sum x_i \varepsilon_i}{\sum x_i^2 + 1} x \right) = \\
 &= \text{Var} \left(\frac{\sum x_i \varepsilon_i}{\sum x_i^2 + 1} x \right) = \frac{x^2}{(\sum x_i^2 + 1)^2} \sum x_i^2 \text{Var}(\varepsilon_i) \quad \text{③}
 \end{aligned}$$

$$\text{Var}(\sum x_i \varepsilon_i) = \sum x_i^2 \text{Var}(\varepsilon_i)$$

$$\text{④} \quad \frac{s}{(s+1)^2} x^2 \sigma^2$$

$$E_{x,y} [E_x[(y - \mu(X)(x))^2]] = \sigma^2 +$$

$$\underbrace{\left(\frac{s}{s+1} - 1 \right)^2 \alpha^2 x^2}_0 + \underbrace{\frac{s}{(s+1)^2} x^2 \sigma^2}_{\rightarrow \min_{\lambda}}$$

$$\lambda = 0$$

$$\lambda \gg 0$$

$$\gamma = \frac{1}{s+1}$$

$$\sigma^2 + (s\gamma - 1)^2 \alpha^2 x^2 + s\gamma^2 x^2 \sigma^2 \rightarrow \min_{\lambda}$$

$$(s\alpha^2 + \sigma^2)x^2 s\gamma^2 - 2s\alpha^2 x^2 \gamma + \alpha^2 x^2 + \sigma^2$$

$$\gamma^* = \frac{1}{s + \sigma^2/\alpha^2} \quad \lambda^* = \frac{\sigma^2}{\alpha^2} > 0$$

$$\hat{d} = \hat{\omega} =$$

Деревья

1) Бэггинг

2) Бустинг

$$y = \hat{L}x$$

Бэггинг.

$$X = (x_1, y_1) \dots (x_n, y_n)$$

$$\begin{matrix} \hat{x}_1 & = & - & - & - & - & - & \mu_1 \\ \hat{x}_2 & & - & - & - & - & - & \vdots \\ \vdots & & & & & & & \vdots \\ \hat{x}_k & & - & - & - & - & - & \mu_k \end{matrix}$$

Какие деревья строим?
и признаков
и т.

При бинарной классификации

$$P(\text{ошибки}) = p$$

$$P(\text{ошибки голосования}) = f(p) \quad \forall p$$

При каких p вер-ть будет меньше p

$(1-p)^3$ - все три верно
 $3p(1-p)^2$ - один ошибся

$$1 - (1-p)^3 - 3p(1-p)^2 = p^2(3-2p)$$

$$-2p^3 + 3p^2 < p$$

$$p(-2p^2 + 3p - 1) < 0$$

$$\underbrace{-2p}_{\leq 0} \underbrace{(p-1)(p-\frac{1}{2})}_{\leq 0} < 0$$

$$p \in [0; 1]$$

$$\underbrace{\leq 0}_{\leq 0} \underbrace{\leq 0}_{\leq 0} < 0$$

$$p - \frac{1}{2} < 0$$

$$p < \frac{1}{2}$$

$$p \in (0; \frac{1}{2})$$

$$p < \frac{1}{2}$$