

Local Learning



e.g. local linear regression

Idea: Only build model based on local points near x_0 .

Estimate: $\alpha(x_0)$, $\beta(x_0) = (\beta_1(x_0), \dots, \beta_p(x_0))^T$

Under the least square model we see

$$(\hat{\alpha}, \hat{\beta}) = \underset{(\alpha, \beta)}{\operatorname{argmin}} \sum_{i=1}^N \left\{ w_i \left[y_i - \alpha(x_i) - \beta(x_i)^T x_0 \right]^2 \right\}$$

Ex: $d=0$, local constant regression

$$\hat{\alpha}(x_0) = \sum_{i=1}^N w_i y_i / \sum_{i=1}^N w_i \quad \text{Kernel-Nearest Neighbors}$$

$$= \sum_{i=1}^N \left(\frac{w_i}{\sum_{j=1}^N w_j} \right) y_i \equiv \sum_{i=1}^N \ell_i(x_0) y_i \equiv \ell^T(x_0) Y$$

So all together we have

$$\hat{y} = L^T Y \quad L = (\ell(x_1) \cdots \ell(x_N))$$

Local Linear Regression

$$(\alpha(x_0), \beta(x_0)) = \underset{(\alpha, \beta)}{\text{arg min}} = \sum_{i=1}^n w_i^\sigma (y_i - \alpha(x_0) - \beta(x_0)x_i)^2$$

$$\begin{pmatrix} \hat{\alpha}(x_0) \\ \hat{\beta}(x_0) \end{pmatrix} = (X^T W^{(\sigma)} X)^{-1} X^T W^{(\sigma)} y$$

Choice of Weights

$w_i^{(\sigma)} = D(x_i - x_0)$ that has certain properties.

i.e. $\int D(u) du = 1, -D(u) = D(-u)$

Epanechnikov Kernel: $D(u) = \frac{3}{4} (1 - u^2)$

Gaussian: $D(u) = \frac{1}{\sqrt{2\pi}} e^{-u^2/2}$

The weights then are given $w_i^{(\sigma)} = K_\lambda(x_i, x_0) = \frac{1}{\lambda} D\left(\frac{x_i - x_0}{\lambda}\right)$

Definition of Sample Training Error

$$\overline{\text{err}} = \frac{1}{N} \sum_{i=1}^n L(y_i, \hat{f}(x_i))$$

$$\text{Err} = L_{(x, y)}(y, \hat{f}(x))$$