

3/12/18

Lecture 11

Extrapolation vs Interpolation

Extrapolation can lead to "dicey" territory. (Dangerous)

You should always check that you are in the range of \mathbb{D} .

Silver calls Extrapolation "Out of Sample"

Extrapolation

$$x^* \notin \text{Range}[X] := [X_{\cdot 1, \min}, X_{\cdot 1, \max}] \times [X_{\cdot 2, \min}, X_{\cdot 2, \max}] \times \dots \times [X_{\cdot p, \min}, X_{\cdot p, \max}]$$

$$X_1 \sim f_1(x), X_2 \sim f_2(x), \dots$$

if x^* is in the "tails" of the distr's \Rightarrow extr.

($p+1$ dimensional entity)

$$\vec{y} \in \mathbb{R}^n \quad X \in \mathbb{R}^{n \times (p+1)}$$

linear, multivariate, least squares fitting

$$\vec{b} = (X^T X)^{-1} X^T \vec{y} \quad \hat{y}^* = g(\vec{x}^*) = \vec{x}^* \vec{b} = b_0 + b_1 x_1^* + \dots + b_p x_p^* \quad \vec{\hat{y}} = X \vec{b}$$

$$\vec{\hat{y}} = X \vec{b} \Rightarrow \vec{\hat{y}} = \underbrace{X(X^T X)^{-1} X^T}_{H \text{ "hat matrix"}} \vec{y} \quad (\text{hat usually means estimate})$$

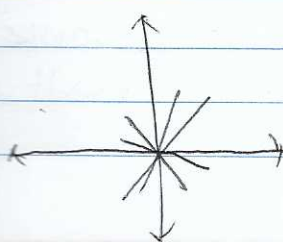
$$\vec{\hat{y}} = H \vec{y} \quad \vec{y} = g(\vec{x}) + \vec{e} \quad \vec{y} = \vec{\hat{y}} + \vec{e} \Rightarrow \vec{e} = \vec{y} - \vec{\hat{y}} \quad \hookrightarrow \text{dimension } (n \times 1)$$

$$\text{SSE} := \sum_{i=1}^n e_i^2 = \|\vec{e}\|^2$$

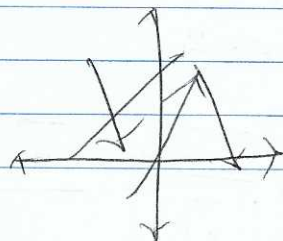
$$\text{MSE} = \frac{1}{n-(p+1)} \text{SSE} \quad \left(\text{If } p=1 \text{ then we get } \frac{1}{n-2} \text{SSE} \right)$$

\hookrightarrow (degrees of freedom)

degrees of freedom will be rank of X



If intercept is fixed
then we can choose the
slope: degrees of
freedom = 1



If we can choose
intercept and slope,
degrees of freedom = 2

$$RMSE = \sqrt{MSE} = \sqrt{\frac{1}{n-(p+1)} SSE} = \frac{\|\vec{e}\|}{\sqrt{n-p-1}} \quad R^2 := \frac{S_y^2 - S_e^2}{S_y^2}$$

(Data frames and matrices are not the same)

(Not being linearly independent is rarely a problem in real data)

R^2 = percent sample variance reduction

ANOVA = analysis of variance

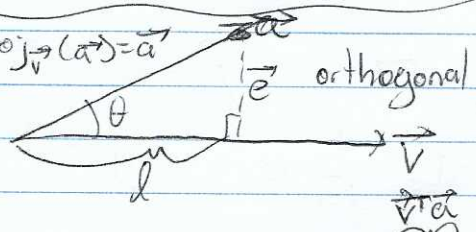
ANCOVA = analysis of covariance

You dummyify categorical variables

Calculation of R^2 works only if you have an intercept

If you add all the columns of airbags you get a column of 1's
 $\begin{matrix} \text{||} & \text{||} & & \text{||} & \text{||} \\ \text{Cars} & & & & \end{matrix}$

So if this is the case then we will have linear dependence, so one airbag type or one car type is unnecessary.

Linear Algebra $\vec{a} + \text{proj}_{\vec{v}}(\vec{a}) = \vec{a}$ 

Law of Cosines: $\cos(\theta) = \frac{\vec{a} \cdot \vec{v}}{\|\vec{a}\| \|\vec{v}\|}$

$$l = \|\vec{a}\| \cos(\theta) = \frac{\vec{a} \cdot \vec{v}}{\|\vec{v}\|}$$

$$\text{proj}_{\vec{v}}(\vec{a}) = l \vec{v}_0 = \frac{\vec{a} \cdot \vec{v}}{\|\vec{v}\|^2} \vec{v} = \frac{\vec{a}^T \vec{v} \vec{v}}{\vec{v}^T \vec{v}}$$

$$\text{proj}_{\vec{v}}(\vec{a}) = \underbrace{\frac{\vec{v} \vec{v}^T}{\vec{v}^T \vec{v}}}_{H \in \mathbb{R}^{n \times n}} \vec{a}$$

$$\vec{a} \in \mathbb{R}^n \quad \vec{v} \in \mathbb{R}^n$$

$\begin{pmatrix} H & H & H & H & (H \vec{a}) \\ \text{already projected it, so same result} \end{pmatrix}$

$$H H = \left(\frac{\vec{v} \vec{v}^T}{\|\vec{v}\|^2} \right) \left(\frac{\vec{v} \vec{v}^T}{\|\vec{v}\|^2} \right) = \frac{\vec{v} \|\vec{v}\|^2 \vec{v}^T}{(\|\vec{v}\|^2)^2} = \frac{\vec{v} \vec{v}^T}{\|\vec{v}\|^2} = H$$