

Homework 2

12/7/2020

Q1.] Ordinary Least Squares (OLS) can be expressed as a minimization problem

$$\min_{\beta} \|X\beta - y\|_2^2$$

$\hat{y} = X\beta$ is the general linear model

X is the design matrix having predictors

β is vector of regression coefficients or beta parameters.

To find the value of β above eqn can be written in the form

$y = X\beta$, where y is the dependent or response variable.

Now, $y = X\beta$ (1)

Taking left inverse on both sides

$$(X^T X)^{-1} X^T y = (X^T X)^{-1} X^T X \beta \quad (2)$$

$$\text{Now, } (X^T X)^{-1} = (X^T X)^{-1}$$

$$\therefore (X^T X)^{-1} X^T X = (X^T X)^{-1} \cdot (X^T X) = I$$

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On Simplifying, eqⁿ (2) becomes

$$\beta = (X^T X)^{-1} X^T \cdot y \quad (3)$$

for $(X^T X)^{-1}$ to exist, mandatory condition is $\text{rank}(X^T X)$ should be same as number of columns of X . This means X should not have any column which is linearly dependent on other column of X .

Thus, if there are any dependent column, the design matrix becomes close to singular and as a result, the least square estimate becomes highly sensitive to Random errors in the observed target, producing large variance.

Q2.]

Corrective measures:

- i) Design matrix should be a tall matrix, i.e., no. of rows should be more than no. of columns.
- ii) Using VIF to identify independent columns.

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- iii) Using L_2 (Ridge) and L_1 (Lasso) Regularization techniques.
 - iv) Removing highly correlated variables
 - v) Linearly combine the correlated variables, such as addition
 - vi) Use Principal Component Analysis (PCA) to find variables (Principal components) which are linearly independent.
 - vii) Using other estimators like partial least square regression.