

Lasso Regression. (LI)

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Only difference b/w Ridge and Lasso is Lasso consider $|w|$ mod. of Coefficient.

$$\therefore \text{MSE} = \sum_{i=1}^n (y_i - \hat{y}_i)^2 + \lambda |w|$$

by increase alpha or λ coefficient can be zero.

Indirectly we can use Lasso to feature selection for high dimensional data.

key points.

- 1) Coefficient can be zero. This property can be used to remove unnecessary column. (feature selection)
 - 2) Coef having higher value decrease fastly and intermediate value of λ give good feature selection.
- * Lasso shows sparsity, which means making coef of zero.

$$L = \sum_{i=1}^N (\tilde{x}_i - \hat{y}_i)^2 + \lambda |m|.$$

$$b = \bar{y} - m\bar{x}$$

$$L = \sum_{i=1}^N (x_i - \hat{x}_i)^2 + \lambda |m|.$$

mathematical
convenience

$$= \sum_{i=1}^N (x_i - mx_i - \bar{y} + m\bar{x})^2 + 2\lambda |m|$$

$$\frac{\partial L}{\partial m} = 2 \sum (x_i - mx_i - \bar{y} + m\bar{x})(-x_i + \bar{x}) + 2\lambda = 0$$

$$-2 \sum [(x_i - \bar{x}) - m(x_i - \bar{x})](x_i - \bar{x}) + 2\lambda = 0$$

$$- \sum [(x_i - \bar{x})(x_i - \bar{x}) - m(x_i - \bar{x})^2] + \lambda = 0$$

$$- \sum (x_i - \bar{x})(x_i - \bar{x}) + m \sum (x_i - \bar{x})^2 + \lambda = 0$$

$$m \sum (x_i - \bar{x})^2 = \sum (x_i - \bar{x})(x_i - \bar{x}) - \lambda$$

$$m = \frac{\sum (x_i - \bar{x})(x_i - \bar{x}) - \lambda}{\sum (x_i - \bar{x})^2}$$

Loss

for $m > 0$

$$m = \frac{\sum (x_i - \bar{x})(x_i - \bar{x}) - \lambda}{\sum (x_i - \bar{x})^2}$$

for $m = 0$

$$m = \frac{\sum (x_i - \bar{x})(x_i - \bar{x})}{\sum (x_i - \bar{x})^2}$$

for $m < 0$

$$m = \frac{\sum (x_i - \bar{x})(x_i - \bar{x}) + \lambda}{\sum (x_i - \bar{x})^2}$$

Important question is that why Lasso create sparsity and Ridge don't answer is in formula only

- Ridge :
$$m = \frac{\sum (x_i - \bar{y})(x_i - \bar{x})}{\sum (x_i - \bar{x})^2 + \lambda}$$

as alpha is at denominator. \therefore for m to be zero value in ~~numerator~~ numerator has to be zero. (In most of the cases numerator is positive)

- Lasso.

alpha value is in numerator. therefore by changing alpha value we can easily manipulate m . which create sparsity there are three cases where $m < 0$, $m = 0$, $m > 0$.