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ElasticNet Regression

- it is the combination of Ridge Regression and Lasso Regression.

$$L = \sum_{i=1}^n (y_i - \hat{y}_i)^2 + \lambda \|W\|^2$$

Ridge Regression

$$L = \sum_{i=1}^n (y_i - \hat{y}_i)^2 + \lambda \|W\|$$

Lasso Regression.

$$\lambda (W_1^2 + W_2^2 + \dots + W_n^2)$$

$$\lambda (|W_1| + |W_2| + |W_3| + \dots + |W_n|)$$

- here $\lambda \uparrow$ then W approaches 0
- here all columns are important.

- here $\lambda \uparrow$ then some W become zero.
- here may be all columns are not important

- ElasticNet Regression use when we can't predict which is we use Lasso or Ridge.
- here loss function for ElasticNet is:

$$L = \sum (y_i - \hat{y}_i)^2 + a \|W\|^2 + b \|W\|$$

in Scikitlearn $\lambda = a + b$

$$L1\text{-ratio} = \frac{a}{a+b} \quad \text{or} \quad L1 = \frac{a}{\lambda}$$

Default value $\lambda = 1$ and $L1\text{-ratio} = 0.5$

hence $a = 0.5$ and $b = 0.5$

$L1\text{-ratio} = 0.9$ means 90% Ridge and 10% Lasso

- if n input column multi-collinearity, then here use ElasticNet.