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## **1. Simple Linear Regression**

### **1.1**

Mean  $x = 2$

Mean  $y = 170/3$

$\text{Cov}(x,y) = (1-2)*(50-170/3) + (2-2)*(55-170/3) + (3-2)*(65-170/3) = 15$

$\text{Var}(x) = 2$

$\beta_1 = \text{Cov} / \text{Var} = 7.5$

$\beta_0 + \epsilon = 41.67$

$\hat{y} = 41.6 + 7.5x$

### **1.2**

$X=5 \quad \hat{y} = 79.1$

$X=10 \quad \hat{y} = 116.6$

### **1.3**

$\text{RSS} = 0.83^{**2} + 1.67^{**2} + 0.83^{**2} = 2.7889$

### **1.4**

L1:  $\text{RSS} = 0.83^{**2} + \lambda * 7.5^{**2} + 1.67^{**2} + \lambda * 7.5^{**2} + 0.83^{**2} + \lambda * 7.5^{**2}$

L2:  $\text{RSS} = 0.83^{**2} + \lambda * 7.5 + 1.67^{**2} + \lambda * 7.5 + 0.83^{**2} + \lambda * 7.5$

### **1.5**

Regularization is to avoid overfitting by punishing extreme value of  $\beta$  which occurs where there are extreme noise and avoid the slope coefficient to be too large.

## **2. Multiple linear regression**

### **2.1**

$X = [[1,1,1], [1,4,1], [1,6,0], [1,8,2], [1,10,1]]$

$X^T = [[1,1,1,1,1], [1,4,6,8,10], [1,1,0,2,1]]$

$Y = [52, 63, 62, 91, 75]$

$X^T * X = \begin{bmatrix} 5 & 29 & 5 \\ 29 & 217 & 31 \\ 5 & 31 & 7 \end{bmatrix}$

$X^T * y = [343 \ 2154 \ 372]$

$(X^T * X)^{-1} = \begin{bmatrix} 1.19230769 & -0.1025641 & -0.3974359 \\ -0.1025641 & 0.02136752 & -0.02136752 \\ -0.3974359 & -0.02136752 & 0.52136752 \end{bmatrix}$

$\beta = [40.19230769 \ 2.8974359 \ 11.6025641]$

$y = 40.19 + 2.897x_1 + 11.6x_2$

### **2.2**

$x_1 = 5, x_2 = 1 \quad y = 66.275$

### **2.3**

$\hat{y} = [54.687, 63.378, 57.572, 86.566, 80.76]$

$$RSS = 2.687^2 + 0.378^2 + 4.428^2 + 4.434^2 + 5.76^2 = 79.8$$

## 2.4

$$RSS = (2.687^2 + \lambda * (2.897^2 + 11.6^2)) + \\ (0.378^2 + \lambda * (2.897^2 + 11.6^2)) + \\ (4.428^2 + \lambda * (2.897^2 + 11.6^2)) + \\ (4.434^2 + \lambda * (2.897^2 + 11.6^2)) + \\ (5.76^2 + \lambda * (2.897^2 + 11.6^2))$$

Large  $\lambda$ :

Benefit - strong regularization and reduce the magnitude of the coefficients and the complexity of the model.

Drawbacks - too much punishment may lead to under fitting.

Small  $\lambda$ :

Benefit - allow the model to realize the complex relation between variables.

Drawbacks - may not work so efficiently and can not avoid overfitting or simplify the model efficiently.