

Regularization





- Regularization seeks to solve a few common model issues by:
 - Minimizing model complexity
 - Penalizing the loss function
 - Reducing model overfitting (add more bias to reduce model variance)



- In general, we can think of regularization as a way to reduce model overfitting and variance.
 - Requires some additional bias
 - Requires a search for optimal penalty hyperparameter.





- Three main types of Regularization:
 - L1 Regularization
 - LASSO Regression
 - L2 Regularization
 - Ridge Regression
 - Combining L1 and L2
 - Elastic Net





- L1 regularization adds a penalty equal to the absolute value of the magnitude of coefficients.
 - Limits the size of the coefficients.
 - Can yield sparse models where some coefficients can become zero.





 L1 regularization adds a penalty equal to the absolute value of the magnitude of coefficients.

$$\sum_{i=1}^{n} \left(y_i - \beta_0 - \sum_{j=1}^{p} \beta_j x_{ij} \right)^2 + \lambda \sum_{j=1}^{p} |\beta_j| = RSS + \left| \lambda \sum_{j=1}^{p} |\beta_j| \right|$$



- L2 regularization adds a penalty equal to the square of the magnitude of coefficients.
 - All coefficients are shrunk by the same factor.
 - Does not necessarily eliminate coefficients.





 L2 regularization adds a penalty equal to the square of the magnitude of coefficients.

$$\sum_{i=1}^{n} \left(y_i - \beta_0 - \sum_{j=1}^{p} \beta_j x_{ij} \right)^2 + \lambda \sum_{j=1}^{p} \beta_j^2 = RSS + \left[\lambda \sum_{j=1}^{p} \beta_j^2 \right]$$



 Elastic net combines L1 and L2 with the addition of an alpha parameter deciding the ratio between them:

$$\frac{\sum_{i=1}^{n} (y_i - x_i^J \hat{\beta})^2}{2n} + \lambda \left(\frac{1 - \alpha}{2} \sum_{j=1}^{m} \hat{\beta}_j^2 + \alpha \sum_{j=1}^{m} |\hat{\beta}_j| \right)$$



- These regularization methods do have a cost:
 - Introduce an additional hyperparameter that needs to be tuned.
 - A multiplier to the penalty to decide the "strength" of the penalty.

