

Regularization

Afternoon

Goals

- Be able to...
 - explain ridge regression and lasso regression
 - tune the bias/variance of a regression model
 - choose the best regularization hyperparameter for regression

Issues with Ordinary Linear Regression

- High dimensions -> high variance
- High variance -> overfitting -> :(...
- And yet we may want to include dimensions/features/interactions, if they're helpful

Ordinary Linear Regression


Find betas to minimize RSS/loss function:

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



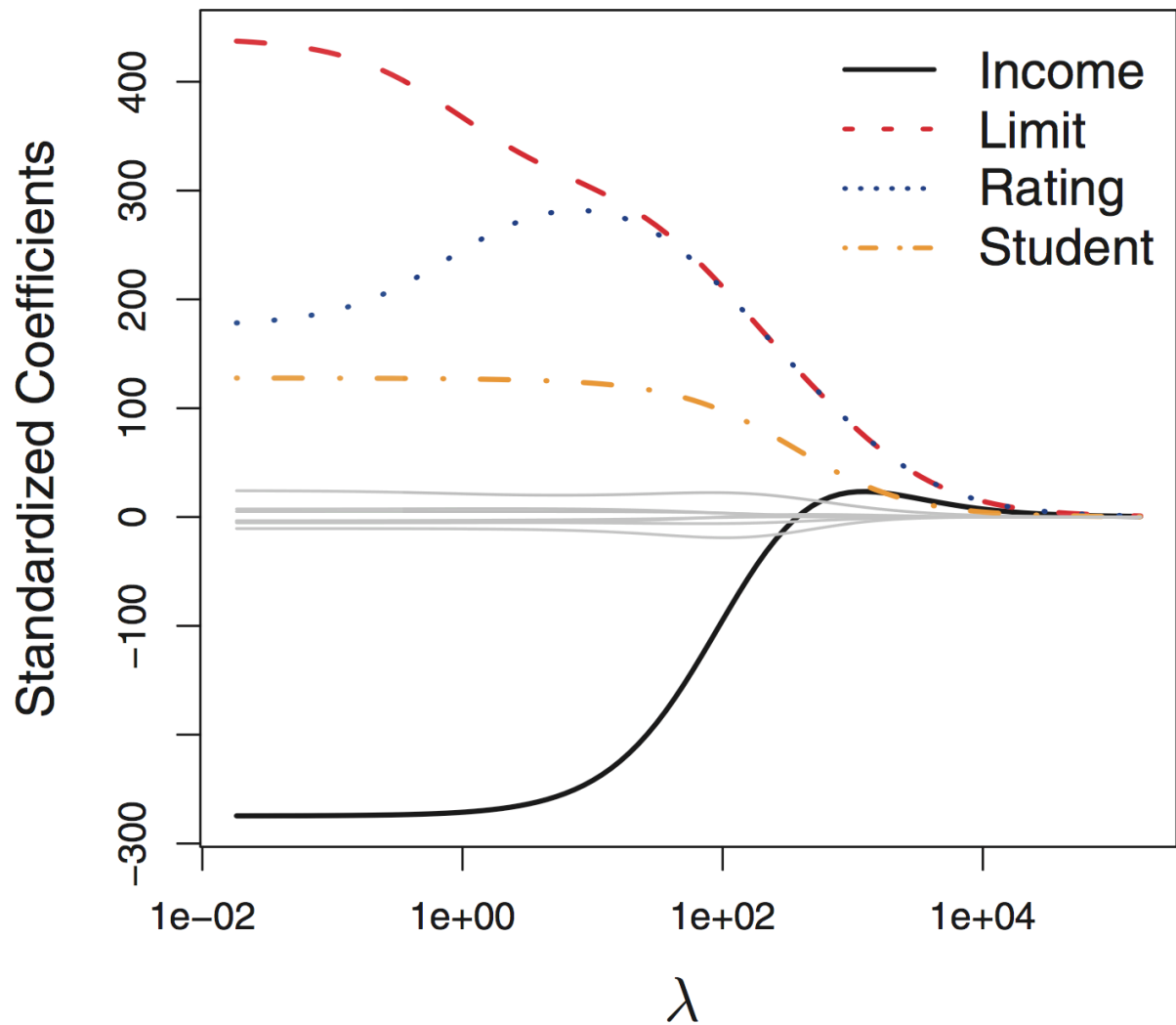
Ridge Regression

Find betas to minimize loss function:

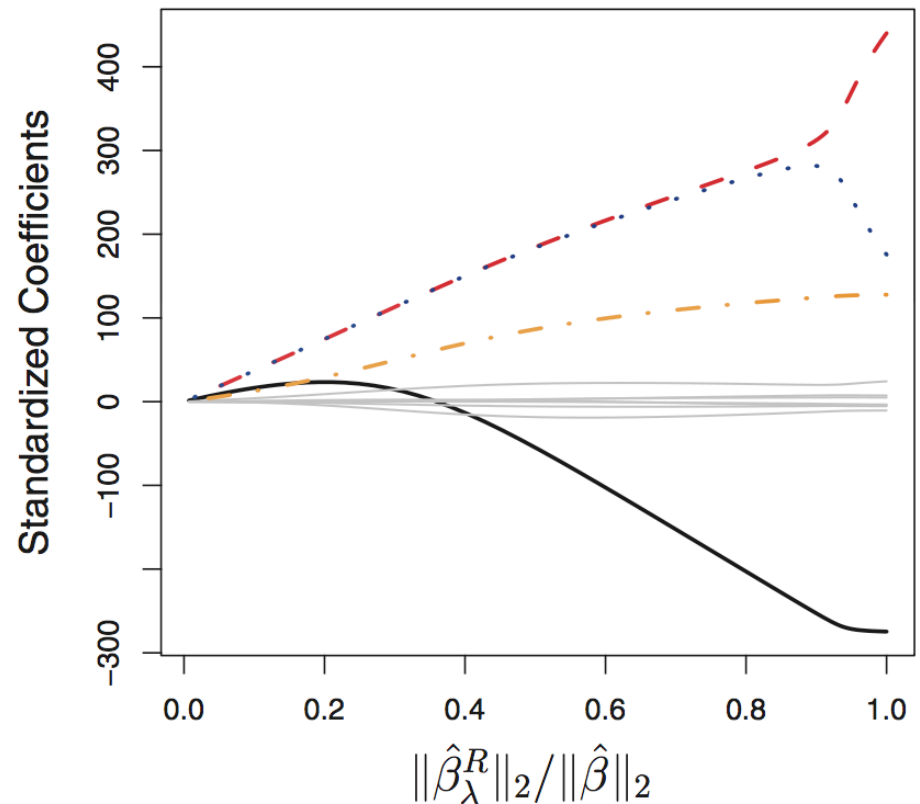
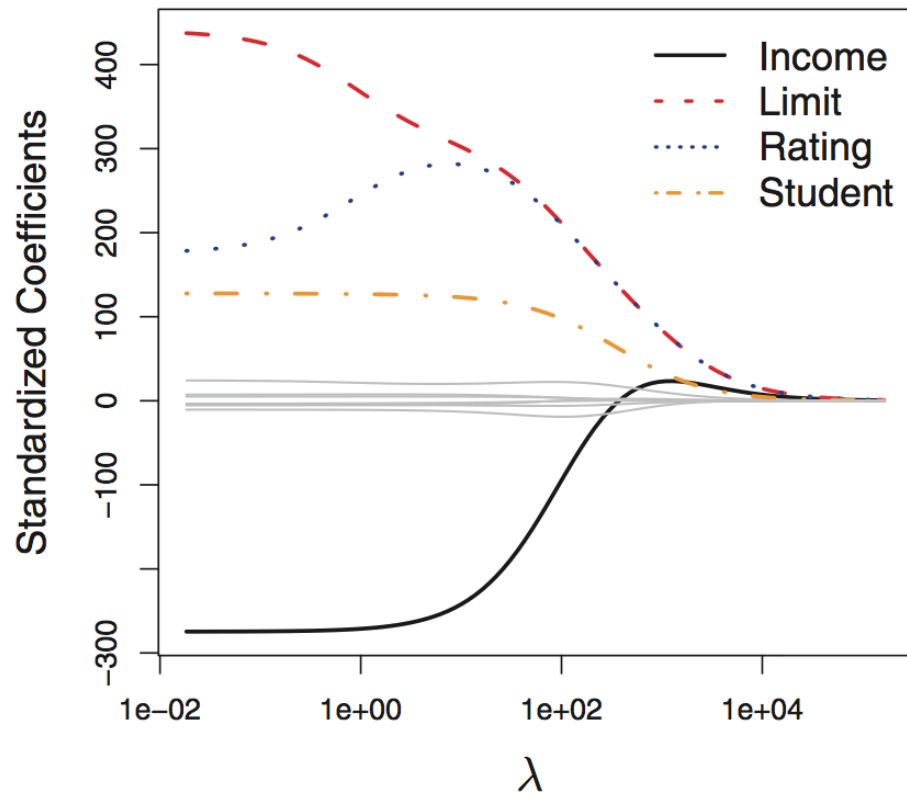
$$\sum_{i=1}^n \left(y_i - \beta_0 - \sum_{j=1}^p \beta_j x_{ij} \right)^2 + \lambda \sum_{j=1}^p \beta_j^2$$


your friend, the hyperparameter

Ridge Regression



Ridge Regression



ℓ_2 (“ell two”) norm: $\|\beta\|_2 = \sqrt{\sum_{j=1}^p \beta_j^2}$

x axis on the right: amount that the ridge regression coefficient estimates have been shrunk towards zero; a small value indicates that they have been shrunk very close to zero

Ridge Regression

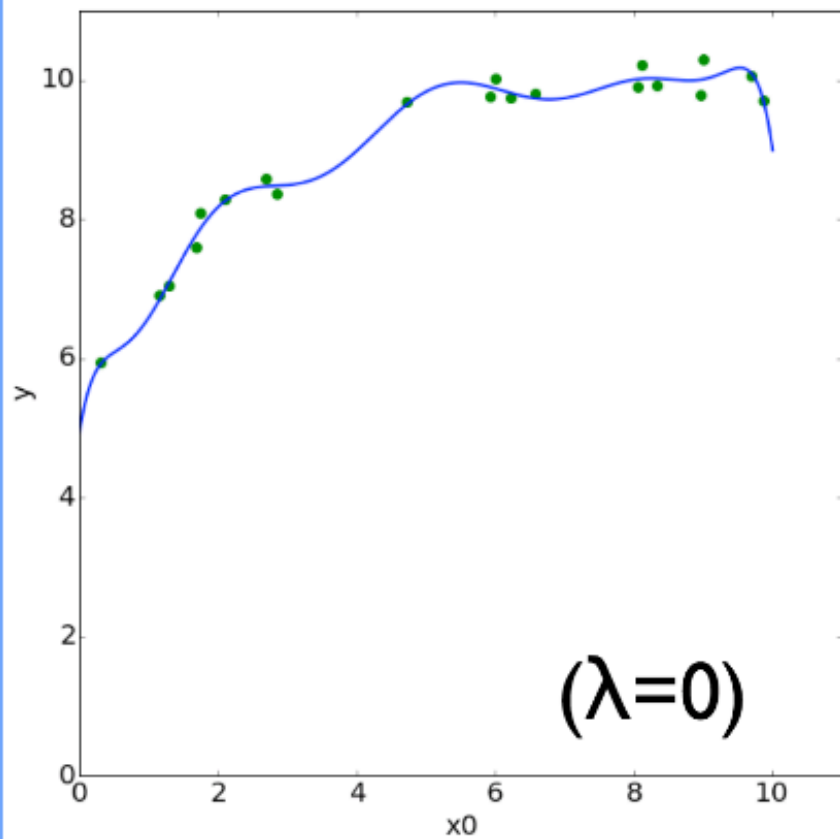
Warning: when using ridge regression, scale matters!
Why? (Consider units to measure salary)

Standardize your predictors:

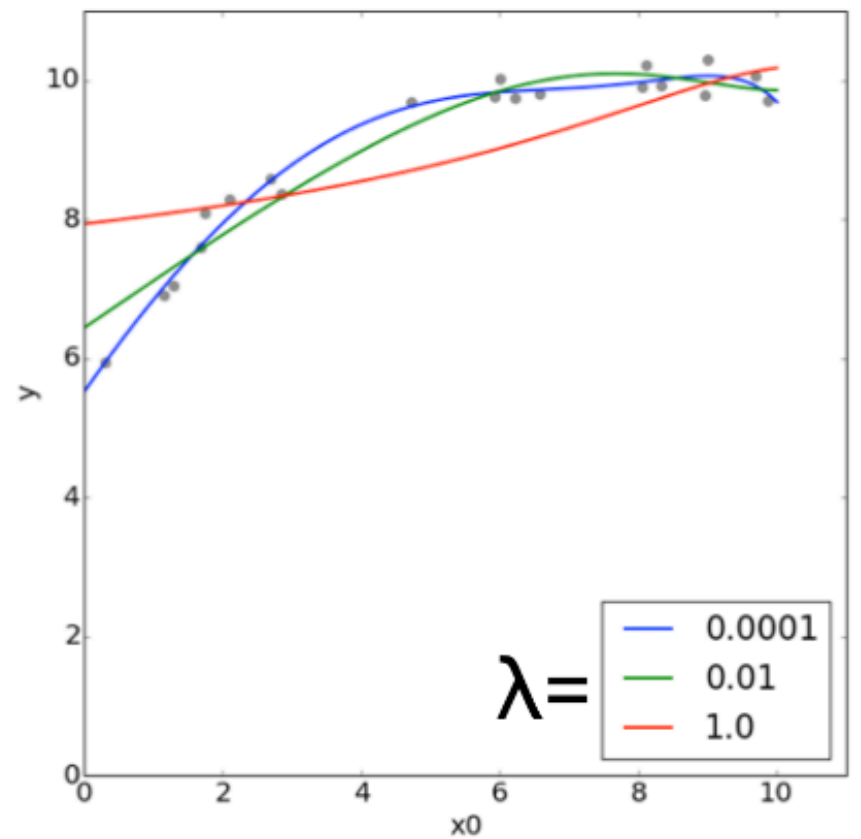
$$\tilde{x}_{ij} = \frac{x_{ij}}{\sqrt{\frac{1}{n} \sum_{i=1}^n (x_{ij} - \bar{x}_j)^2}}$$

Ridge Regression

Linear Regression




Ridge Regression



Lasso Regression

Find betas to minimize loss function:

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


remember, ridge:

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

your friend, the hyperparameter

Ridge and Lasso

Ridge:

minimize RSS:

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

l2 (“ell two”) norm:

$$\|\beta\|_2 = \sqrt{\sum_{j=1}^p \beta_j^2}$$

Lasso:

minimize RSS:

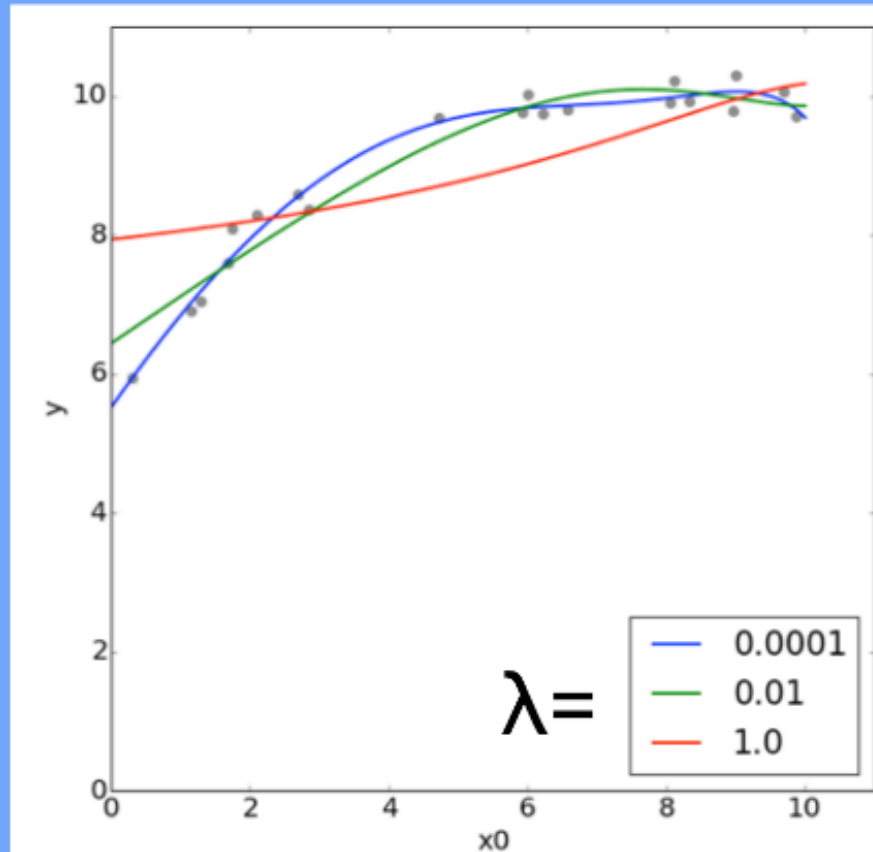
$$\sum_{i=1}^n \left(y_i - \beta_0 - \sum_{j=1}^p \beta_j x_{ij} \right)^2 + \lambda \sum_{j=1}^p |\beta_j|$$

l1 (“ell one”) norm:

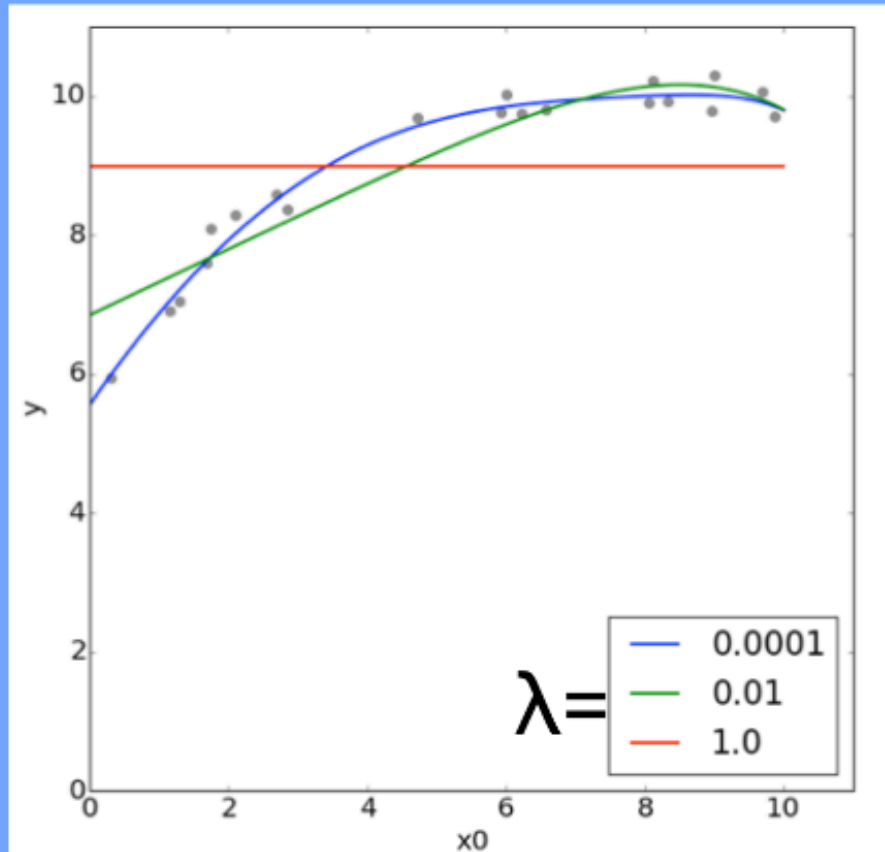
$$\|\beta\|_1 = \sum |\beta_j|$$

Ridge and Lasso

Ridge Regression



Lasso Regression



Ridge

vs.

Lasso

