

# Ch 5.1.1-2: Leave One Out Cross-validation

## Lecture 12 - CMSE 381

Prof. Mengsen Zhang

Michigan State University

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Dept of Computational Mathematics, Science & Engineering

Mon, Feb 14, 2025

# Announcements

**Last time:**

- Exam

**Announcements:**

- Exam 1 grades.... hopefully soon
- HW #4 will be posted later next week.
  - ▶ Changed Deadline! Due Wednesday Mar 2nd.

	W	2/12	<b>Midterm #1</b>		
12	F	2/14	Leave one out CV	5.1.1, 5.1.2	
13	M	2/17	k-fold CV	5.1.3	
14	W	2/19	More k-fold CV	5.1.4-5	
15	F	2/21	k-fold CV for classification	5.1.5	
16	M	2/24	Subset selection	6.1	
17	W	2/26	Shrinkage: Ridge	6.2.1	
18	F	2/28	Shrinkage: Lasso	6.2.2	HW #4 Due Sun 3/2
	M	3/3	Spring Break		
	W	3/5	Spring Break		
	F	3/7	Spring Break		
19	M	3/10	PCA	6.3	
20	W	3/12	PCR	6.3	
	F	3/14	<b>Review</b>		HW #5 Due Sun 3/16
	M	3/17	<b>Midterm #2</b>		

# Covered in this lecture

- Validation Set
- LOO CV

# Section 1

Validation set

# What's the problem?

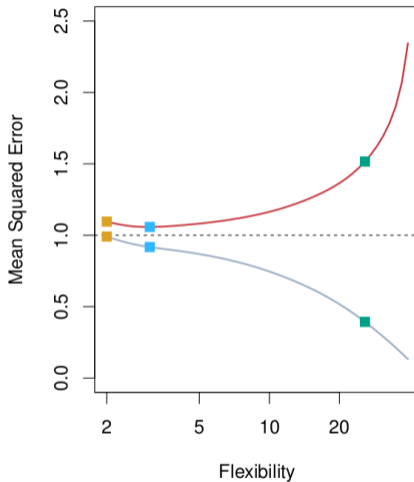
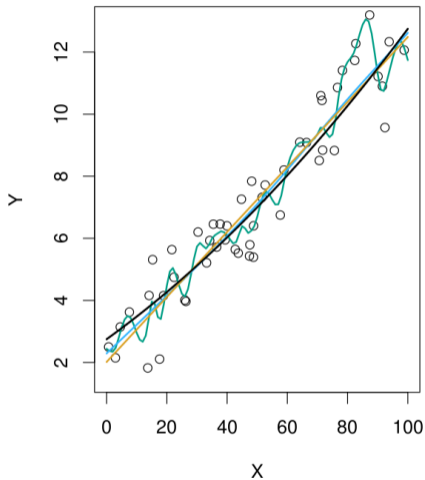
- How well is my ML method doing? *Model Assessment*
- Which method is best for our data?
- How many features should I use? Which ones? *Model selection*
- What is the uncertainty in the learned parameters?

# Training Error vs Testing Error

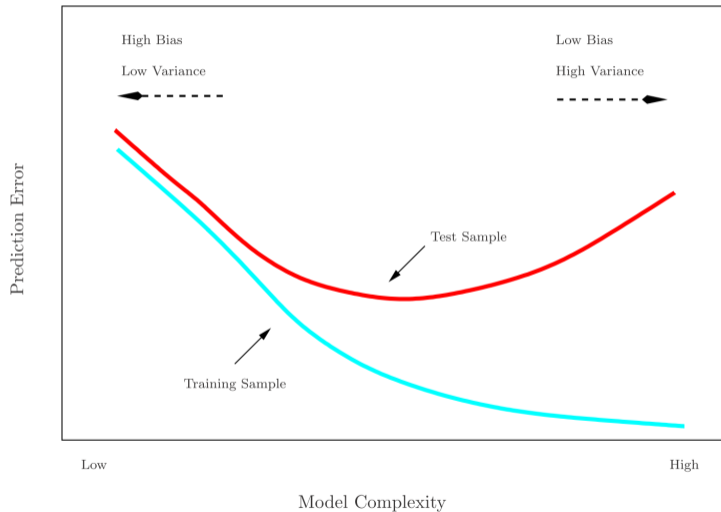
**Training Error**

**Testing Error**

# Throw-back

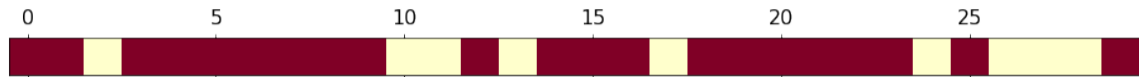


# Model tradeoffs



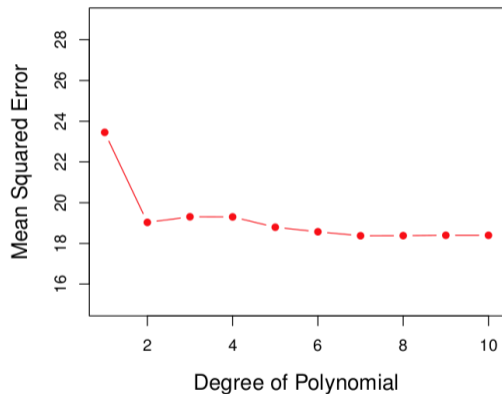


# Validation set approach



- Divide randomly into two parts:
  - ▶ Training set
  - ▶ Validation/Hold-out/Testing set
- Fit model on training set
- Use fitted model to predict response for observations in the test set
- Evaluate quality (e.g. MSE)

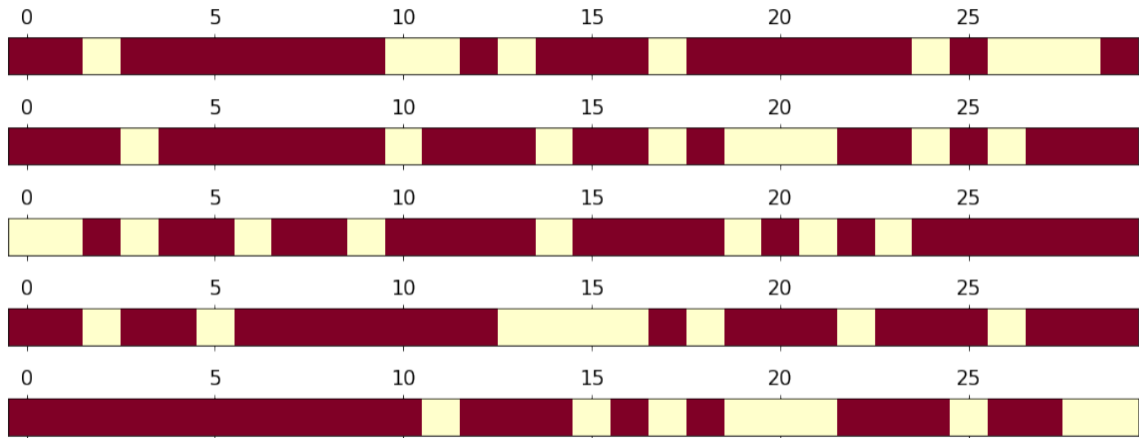
## Example with the auto data



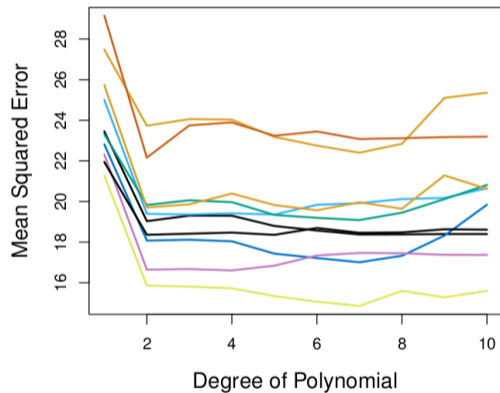
Predicting mpg using horsepower:

$$\text{mpg} = \beta_0 + \beta_1 \text{hp} + \beta_2 \text{hp}^2 + \cdots + \beta_p \text{hp}^p$$

# Rinse and repeat



## Again example with auto data

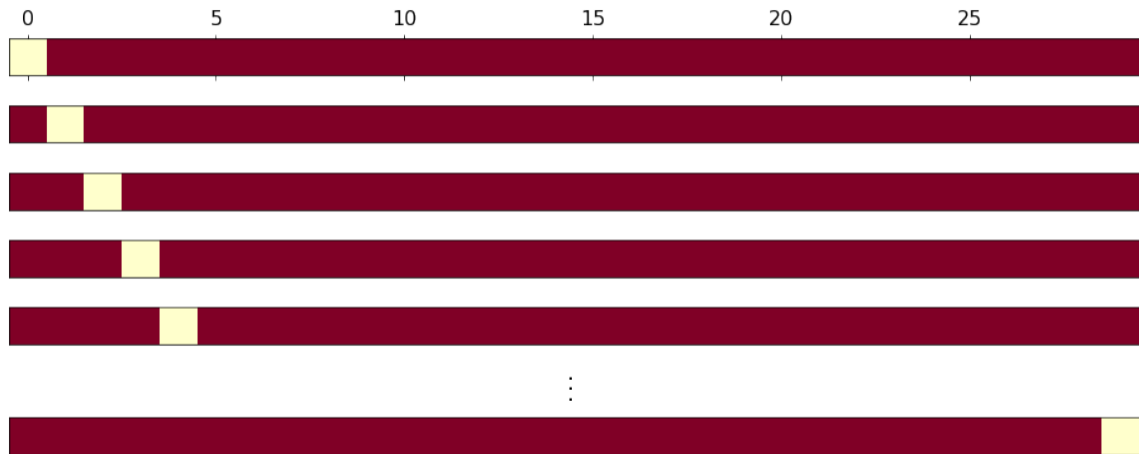


# Coding example in jupyter notebook

## Section 2

# Leave-One-Out Cross-Validation (LOOCV)

# The idea



## The idea in mathy words

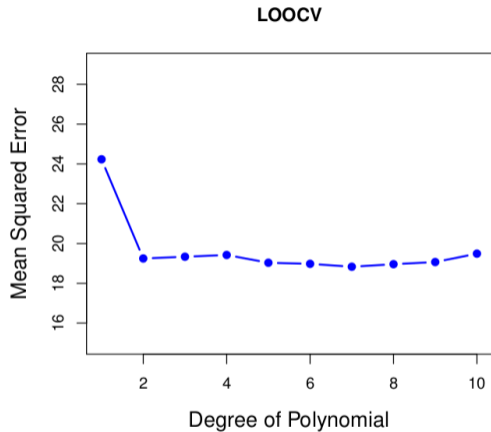
- Remove  $(x_1, y_1)$  for testing.
- Train the model on  $n - 1$  points:  
 $\{(x_2, y_2), \dots, (x_n, y_n)\}$
- Calculate  $\text{MSE}_1 = (y_1 - \hat{y}_1)^2$
  
- Remove  $(x_2, y_2)$  for testing.
- Train the model on  $n - 1$  points:  
 $\{(x_1, y_1), (x_3, y_3), \dots, (x_n, y_n)\}$
- Calculate  $\text{MSE}_2 = (y_2 - \hat{y}_2)^2$
  
- Rinse and repeat

Return the score:

$$CV_{(n)} = \frac{1}{n} \sum_{i=1}^n \text{MSE}_i$$



## Again example with auto data



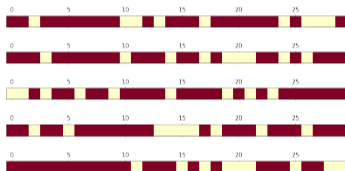
Do the LOOCV coding section

# LOOCV Pros and Cons

**Advantages:**

**Disadvantages:**

## Validation set



## LOO-CV



## LOO-CV Score

$$CV_{(n)} = \frac{1}{n} \sum_{i=1}^n \text{MSE}_i$$

# Next time

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