

# Ch 5.1.5: $k$ -fold Cross-Validation for Classification

Lecture 12 - CMSE 381

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Mon, Oct 10, 2022

## **Last time:**

- k-fold CV

## **This lecture:**

- CV for classification

## **Announcements:**

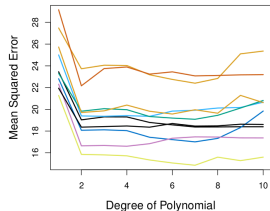
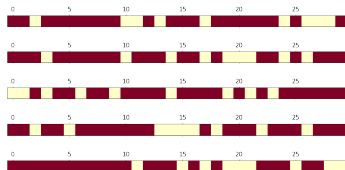
- Homework #4 is posted, Due tonight
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# Section 1

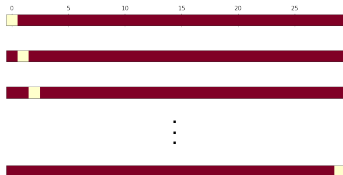
Last time

# Approximations of Test Error

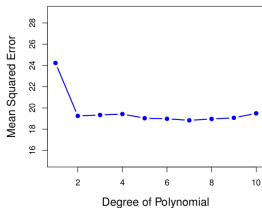
## Validation Set



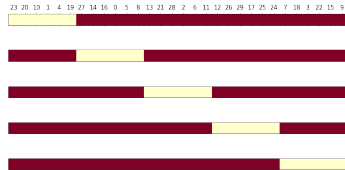
## LOOCV



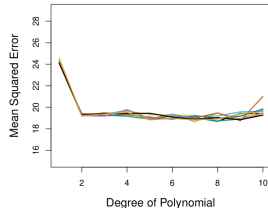
### LOOCV



## K-fold CV

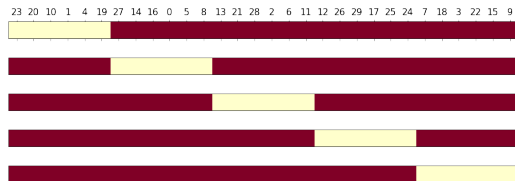


### 10-fold CV



# Definition of $k$ -fold CV

- Randomly split data into  $k$ -groups (folds)
- Approximately equal sized. For the sake of notation, say each set has  $\ell$  points
- Remove  $i$ th fold  $U_i$  and reserve for testing.
- Train the model on remaining points
- Calculate
$$\text{MSE}_i = \frac{1}{\ell} \sum_{(x_j, y_j) \in U_i} (y_j - \hat{y}_j)^2$$
- Rinse and repeat



Return

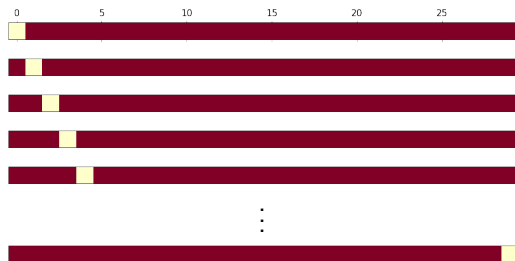
$$CV_{(k)} = \frac{1}{k} \sum_{i=1}^k \text{MSE}_i$$

## Section 2

### CV for Classification

# Setup: LOOCV

- Remove  $i$ th point  $(x_i, y_i)$  and reserve for testing.
- Train the model on remaining points
- Calculate  $\text{Err}_i = I(y_i \neq \hat{y}_i)$
- Rinse and repeat

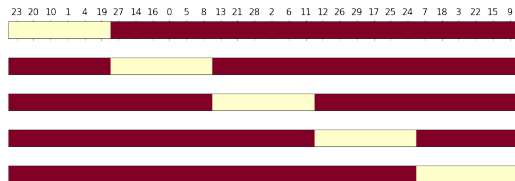


Return

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

## Setup: $k$ -fold

- Randomly split data into  $k$ -groups (folds)
- Approximately equal sized. For the sake of notation, say each set has  $\ell$  points
- Remove  $i$ th fold  $U_i$  and reserve for testing.
- Train the model on remaining points
- Calculate
$$\text{Err}_i = \frac{1}{\ell} \sum_{(x_j, y_j) \in U_i} \mathbb{I}(y_j \neq \hat{y}_j)$$
- Rinse and repeat

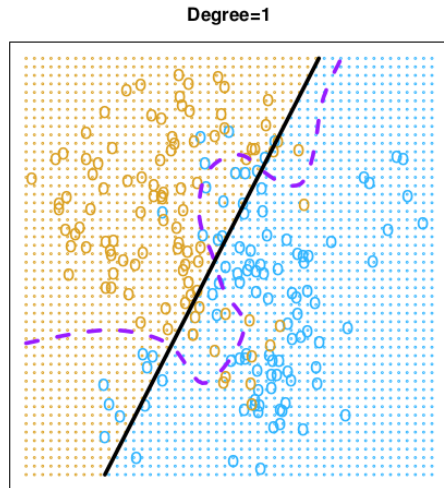


Return

$$CV_{(k)} = \frac{1}{k} \sum_{i=1}^k \text{Err}_i$$

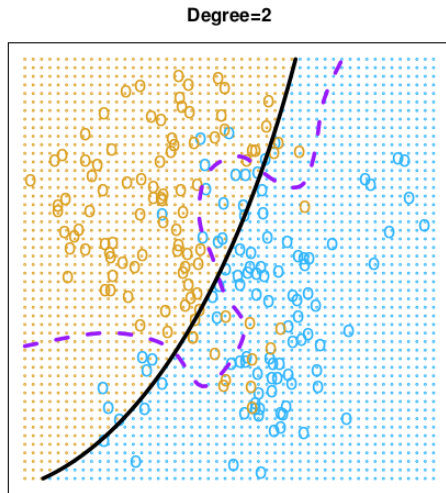


## Example on simulated data: Linear



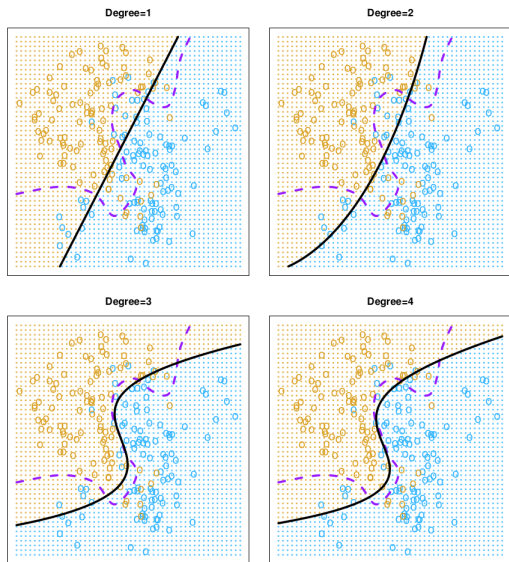
- Purple: Bayes decision boundary.
  - ▶ Error rate: 0.133
- Black: Logistic regression
  - ▶  $\log(p/(1-p)) = \beta_0 + \beta_1 X_1 + \beta_2 X_2$
  - ▶ Error rate: 0.201

## Example on simulated data: Quadratic logistic regression



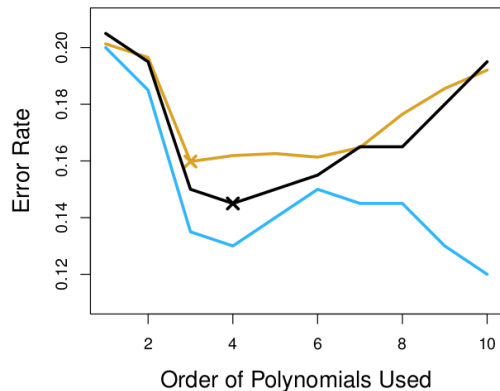
- Purple: Bayes decision boundary.
  - ▶ Error rate: 0.133
- Black: Logistic regression
  - ▶  $\log(p/(1-p)) = \beta_0 + \beta_1 X_1 + \beta_2 X_1^2 + \beta_3 X_2 + \beta_4 X_2^2$
  - ▶ Error rate: 0.197

# Example on simulated data: all the polynomials!



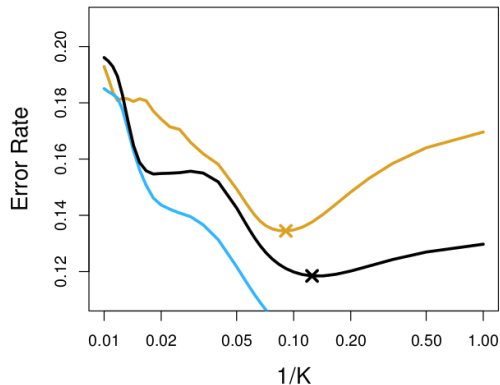
- Purple: Bayes decision boundary.
  - ▶ Error rate: 0.133
- Black: Logistic regression
  - ▶ Deg 1 Error rate: 0.201
  - ▶ Deg 2 Error rate: 0.197
  - ▶ Deg 3 Error rate: 0.160
  - ▶ Deg 4 Error rate: 0.162

# Decide degree based on CV



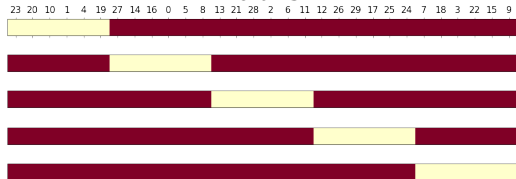
- Test error (brown)
- Training error (blue)
- 10-fold CV error (black)

# Similar game for KNN



- Test error (brown)
- Training error (blue)
- 10-fold CV error (black)

# Coding - k-fold for penguin classification section

$k$ -fold CV

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

Use  $k = 5$  or  $10$  usually

$k$ -fold CV for classification

$$\text{Err}_i = \mathbb{I}(y_j \neq \hat{y}_j)$$

$$CV_{(k)} = \frac{1}{k} \sum_{i=1}^k \text{Err}_i$$

# Next time

10	M	Oct 3	Leave one out CV	5.1.1, 5.1.2	
11	W	Oct 5	k-fold CV	5.1.3	
12	F	Oct 7	More k-fold CV,	5.1.4-5	
13	M	Oct 10	k-fold CV for classification	5.1.5	HW #4 Due
14	W	Oct 12	Resampling methods: Bootstrap	5.2	
15	F	Oct 14	Subset selection	6.1	
16	M	Oct 17	Shrinkage: Ridge	6.2.1	HW #5 Due
17	W	Oct 19	Shrinkage: Lasso	6.2.2	
18	F	Oct 21	[No class, Dr Munch out of town]		
	M	Oct 24	No class - Fall break		
19	W	Oct 26	Dimension Reduction	6.3	
20	F	Oct 28	More dimension reduction; High dimensions	6.4	HW #6 Due
	M	Oct 31	Review		
	W	Nov 2	<b>Midterm #2</b>		