# Data, Environment and Society: Lecture 15: Classification

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#### **Announcements**

### **Today**

- Introduction to classification
  - Corresponding reading: ISLR Ch 4.1 through 4.3.1, and also Ch 2.2.3.
- Classification in practice: water quality violations in California

#### **Next week**

- Lab: you'll collaboratively write study guides for each other.
- ▶ Lecture on Tuesday: I'll go over the review slides, and you can ask questions.
- Exam on Thursday!

# Remember qualitative variables? This is how you do it

$$x_1 = \begin{cases} 1, & \text{Likes split pea.} \\ 0, & \text{otherwise.} \end{cases}$$
 $x_2 = \begin{cases} 1, & \text{Likes minestrone.} \\ 0, & \text{otherwise.} \end{cases}$ 
 $x_3 = \begin{cases} 1, & \text{Doesn't like soup.} \\ 0, & \text{otherwise.} \end{cases}$ 

Question: What about the "other" category?

# Remember qualitative variables? This is how you do it

Question: What about the "other" category?

**Answer:** The answers are mutually exclusive, so if  $x_1, x_2, x_3$  are all zero, then the answer must be "other".

# Cooked-up example: Predicting age with qualitative variables

$$AGE_{i} = \beta_{0} + \beta_{1}x_{i,1} + \beta_{2}x_{i,2} + \beta_{3}x_{i,3} + \beta_{4}x_{i,4} + \epsilon_{i}$$

where  $x_1, x_2, x_3$  are defined on previous slide and  $x_4$  is how spicy the respondent likes their food.

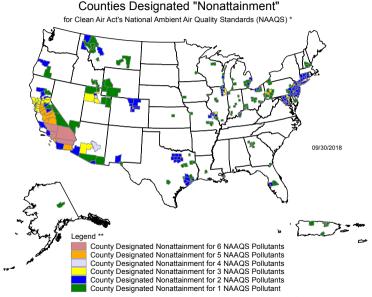
With these variables the qualitative predictors are just modifying the intercept.

- ▶ When  $x_1 = x_2 = x_3 = 0$  (i.e. the answer is "other") then the intercept is  $\beta_0$ .
- Otherwise the intercept is  $\beta_0 + \beta_i$  where *i* is the *x* variable that is nonzero.

# What it you want to *predict* a categorical variable?

### Examples:

- Clean air act attainment status for a region (attainment or non-attainment)
- Is an area going to be a candidate for a new refinery in the next 10 years
- Disease presented in emergency dept of a hospital



# Why not linear regression?

For example: EPA Criteria Pollutants. For each region *i* indicate nonattainment as follows:

- ▶  $y_i = 1 \rightarrow Ozone$
- $y_i = 2 \rightarrow PM2.5$
- ▶  $y_i = 3 \rightarrow PM10$
- $y_i = 4 \rightarrow \text{Sulfur Dioxide}$
- ▶  $y_i = 5 \rightarrow \text{Lead}$
- $y_i = 6 \rightarrow \text{Carbon Monoxide}$
- ▶  $y_i = 7 \rightarrow \text{Nitrogen Dioxide}$

Then,

$$\mathbf{y}_i = \beta \mathbf{x}_i + \epsilon$$

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- $ightharpoonup v_i = 1 \rightarrow Ozone$
- $ightharpoonup y_i = 2 \rightarrow PM2.5$
- $\mathbf{v}_i = \mathbf{3} \rightarrow \mathsf{PM10}$
- $y_i = 4 \rightarrow \text{Sulfur Dioxide}$
- ▶  $y_i = 5 \rightarrow \text{Lead}$
- $y_i = 6 \rightarrow \text{Carbon Monoxide}$
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The problem: a different ordering of variables would imply a different relationship between statuses, different models, and different predictions.

# Predicting categorical variables the right way: Similar coding to using them as predictors

For example: EPA Criteria Pollutants:

- ▶  $y_{i,1}$  → Ozone status
- ▶  $y_{i,2}$  → PM2.5 status
- $ightharpoonup y_{i,3} 
  ightharpoonup PM10 status$
- $ightharpoonup y_{i,4} 
  ightharpoonup Sulfur Dioxide status$
- $ightharpoonup y_{i,5} 
  ightharpoonup Lead status$
- $ightharpoonup y_{i,6} 
  ightharpoonup Carbon Monoxide status$
- ▶  $y_{i,7}$  → Nitrogen Dioxide status

where *i* indexes the region of interest, i.e. observations

Then, for each,

$$y_{i,j} = egin{cases} 1, & ext{Nonattainment status} \ 0, & ext{Attainment status} \end{cases}$$

where j indexes criteria pollutants

How do you get a model to output a  $\{0, 1\}$  result?

# How do you get a model to output a $\{0, 1\}$ result?

**First,** build your model so that its output estimates a *probability* that a given outcome happens.

For example, a model might give:

p(PM-2.5 nonattainment) = 0.734 and p(PM-2.5 attainment) = 0.266

(the probabilities add to one).

# How do you get a model to output a {0, 1} result? (ctd)

**Second**, we use something called the *Bayes Classifier*.

### Bayes classifier:

For a given observation  $x_i$ , choose j as the value for which

$$Pr(Y = j | X = x_i)$$

is largest.

More formally,

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### Bayes classifier:

For a given observation  $x_i$ , choose j as the value for which

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is largest.

More formally,

$$\hat{y}_i = \max_{j \in \mathcal{J}} \Pr(Y = j | X = x_i)$$

where  $\mathcal{J}$  is the set of possible (mutually exclusive) outcomes.

### Classification error rate

If the "true model" for  $Pr(Y = j | X = x_i)$  is known, then using the Bayes classifier will provide the lowest possible error rate.

The *error rate* quantifies how frequently a model mis-classifies a categorical variable.

Let  $I(\cdot)$  denote the *indicator function*:

### Classification error rate

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The *error rate* quantifies how frequently a model mis-classifies a categorical variable.

Let  $I(\cdot)$  denote the *indicator function*:

$$I(y_i \neq \hat{y}_i) = egin{cases} 1, & ext{when } y_i \neq \hat{y}_i. \ 0, & ext{otherwise.} \end{cases}$$

$$\Rightarrow$$
 error rate  $=\frac{1}{n}\sum_{i=1}^{n}I(y_i\neq\hat{y}_i)$ 

#### Location

Allegheny County, PA Cleveland, OH Delaware County, PA Imperial County, CA Lebanon County, PA South Coast Air Basin, CA Plumas County, CA Sacramento County, CA San Francisco, CA San Joaquin Valley, CA West Silver Valley, ID Luzerne County. PA Lancaster County, PA

### Actual status

Non-attainment Non-attainment Non-attainment Non-attainment Non-attainment Non-attainment Non-attainment Attainment Attainment Non-attainment Non-attainment Attainment Attainment

### Predicted (hypothetical)

Non-attainment Non-attainment Non-attainment Attainment **Attainment** Non-attainment Non-attainment Non-attainment Non-attainment Non-attainment Non-attainment

Attainment

Attainment

 $I(y_i \neq \hat{y}_i)$ 

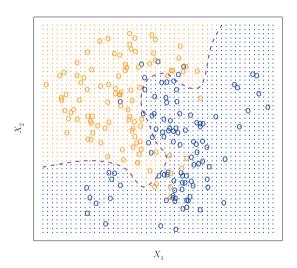
Location	Actual status	Predicted (hypothetical)	$I(y_i \neq \hat{y}_i)$
Allegheny County, PA	Non-attainment	Non-attainment	0
Cleveland, OH	Non-attainment	Non-attainment	0
Delaware County, PA	Non-attainment	Non-attainment	0
Imperial County, CA	Non-attainment	Attainment	
Lebanon County, PA	Non-attainment	Attainment	
South Coast Air Basin, CA	Non-attainment	Non-attainment	
Plumas County, CA	Non-attainment	Non-attainment	
Sacramento County, CA	Attainment	Non-attainment	
San Francisco, CA	Attainment	Non-attainment	
San Joaquin Valley, CA	Non-attainment	Non-attainment	
West Silver Valley, ID	Non-attainment	Non-attainment	
Luzerne County, PA	Attainment	Attainment	
Lancaster County, PA	Attainment	Attainment	

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Cleveland, OH	Non-attainment	Non-attainment	0
Delaware County, PA	Non-attainment	Non-attainment	0
Imperial County, CA	Non-attainment	Attainment	1
Lebanon County, PA	Non-attainment	Attainment	1
South Coast Air Basin, CA	Non-attainment	Non-attainment	0
Plumas County, CA	Non-attainment	Non-attainment	0
Sacramento County, CA	Attainment	Non-attainment	
San Francisco, CA	Attainment	Non-attainment	
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San Joaquin Valley, CA	Non-attainment	Non-attainment	0
West Silver Valley, ID	Non-attainment	Non-attainment	0
Luzerne County, PA	Attainment	Attainment	0
Lancaster County, PA	Attainment	Attainment	0

 $\rightarrow$  Overall error rate is  $\frac{1}{13} \cdot 4 = .31$ 

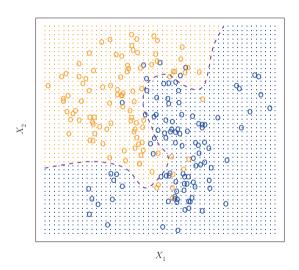
# Finally, the decision boundary



Along the boundary the probability of one outcome equals the probability of the other.

If the true model  $Pr(Y = j | X = x_i)$  is known, we call this the...

# Finally, the decision boundary



ISLR Fig 2.13

Along the boundary the probability of one outcome equals the probability of the other.

If the true model  $Pr(Y = j | X = x_i)$  is known, we call this the... Bayes decision boundary.

While not all methods directly estimate  $Pr(Y = j | X = x_i)$ , they *all do* try to *estimate* the Bayes decision boundary.

# KNN for categorical variables

It's pretty simple! Estimate the probabilities as

K = number of neighboring *training* points to consider  $\mathcal{N}_0 =$  set of K *training* points closest to observation  $x_0$ .

...Then apply the Bayes decision rule to choose the outcome variable.

# KNN for categorical variables

It's pretty simple! Estimate the probabilities as

$$\Pr(Y = j | X = x_0) = \frac{1}{K} \sum_{i \in \mathcal{N}_0} I(y_i = j)$$

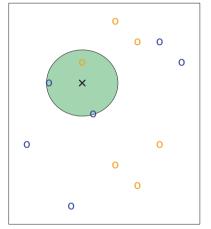
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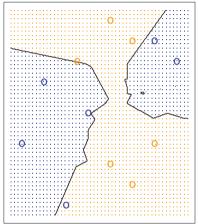
...Then apply the Bayes decision rule to choose the outcome variable.

$$\hat{y}_i = \arg\max_{j \in \mathcal{J}} \Pr(Y = j | X = x_i)$$

where  $x_i$  is any feasible point in the space of independent variables.

# Simple KNN Example, K = 3

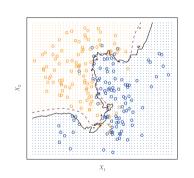


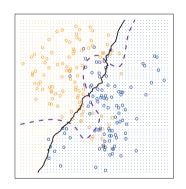


ISLR Fig 2.14

# Which has the highest *K*? Which has the lowest?

**Dashed** = Bayes decision boundary **Solid** = KNN estimate of Bayes decision boundary

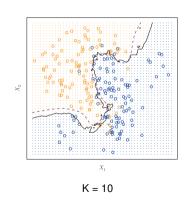


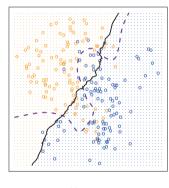


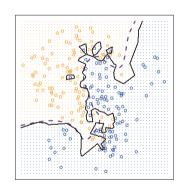


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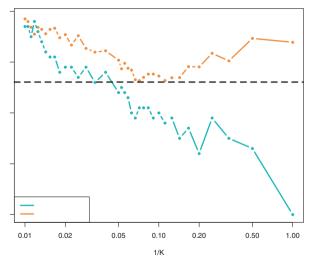






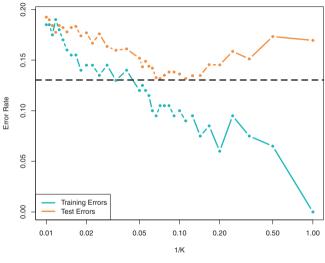
K = 1

# KNN test and training error



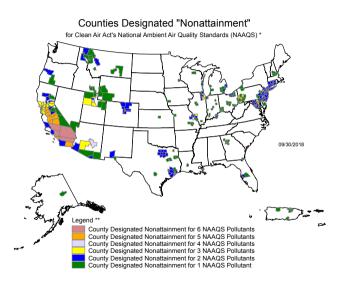
**ISLR 2.17** 

# KNN test and training error

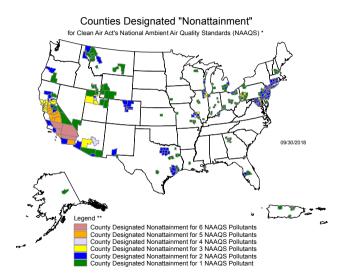


**ISLR 2.17** 

# How would KNN perform with nonattainment areas?



# How would KNN perform with nonattainment areas?



- Challenge: If we only use location as independent variables, the Bayes decision boundary is very complex!
- But if we use other independent variables (a simple one would be local emissions) we might do ok.