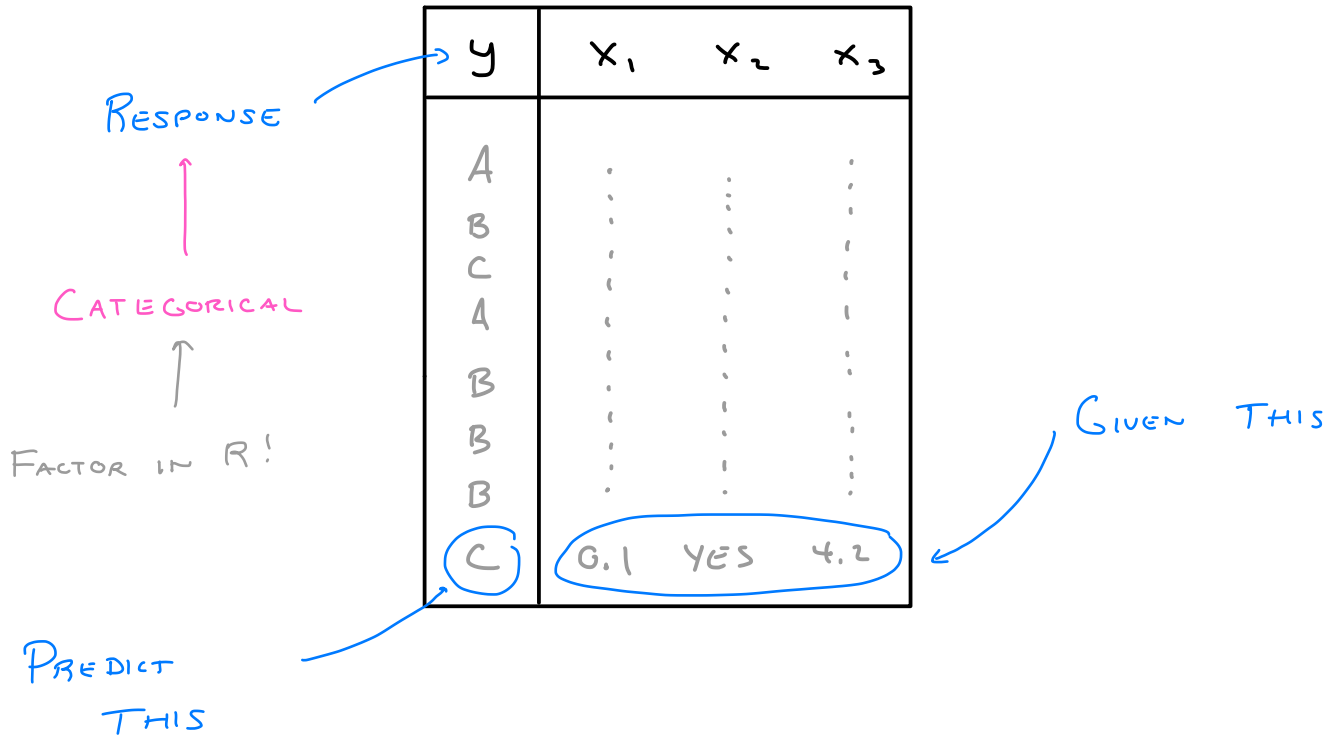


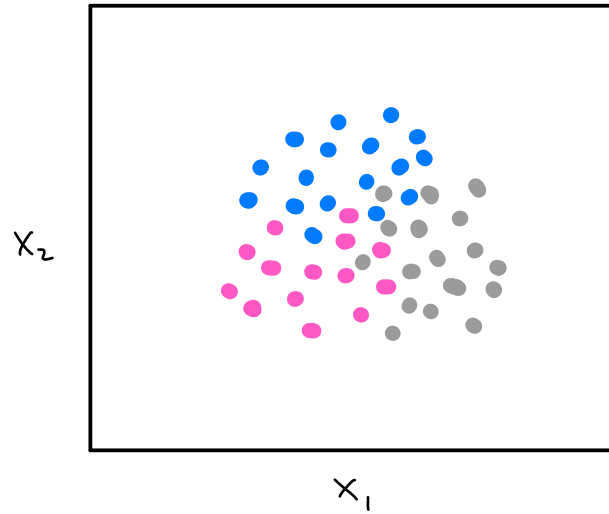
CLASSIFICATION

↳ AN INTRODUCTION

# DATA VIEW



# VISUAL DATA VIEW



# PROBABILITY VIEW

$$(X, Y) \in \mathbb{R}^P \times \{1, 2, \dots, K\}$$

Diagram illustrating the probability view of a classification problem. The input  $(X, Y)$  is drawn from the joint space  $\mathbb{R}^P \times \{1, 2, \dots, K\}$ . The label  $X$  is associated with  $P$  FEATURES, and the label  $Y$  is associated with  $K$  CATEGORIES.

FIND A CLASSIFIER  $C(x)$  THAT MINIMIZES

$$P[C(x) \neq Y] \leftarrow \text{PROBABILITY OF MISCLASSIFICATION}$$

WHERE

$$C: \mathbb{R}^P \longrightarrow \{1, 2, 3, \dots, K\}$$

Diagram illustrating the classifier function  $C$ . The input  $\mathbb{R}^P$  is labeled as INPUT FEATURES, and the output  $\{1, 2, 3, \dots, K\}$  is labeled as OUTPUT CATEGORY.

# BAYES CLASSIFIER

← MINIMIZES PROBABILITY  
OF MISCLASSIFICATION

$$C^B(x) \triangleq \underset{k \in \{1, \dots, K\}}{\text{ARGMAX}} P[Y = k | X = x]$$

GIVEN FEATURE VECTOR  $x$ , CLASSIFY OBSERVATION  
AS THE CATEGORY WITH THE HIGHEST PROBABILITY

DUH?

# EXAMPLE

$$C^B(x=0) = ?$$

$$\frac{P[X=0 \cap Y=A]}{P[X=0]}$$

		X		
		0	1	
Y	A	0.1	0.1	0.2
	B	0.2	0.1	0.3
	C	0.1	0.4	0.5

JOINT DISTRIBUTION OF (X,Y)

$$P[Y | X=0] = \begin{cases} 0.25 & y=A \\ 0.50 & y=B \\ 0.25 & y=C \end{cases}$$

CONDITIONAL PROBABILITY OF Y|X=0

0.4    0.6

MARGINAL DISTRIBUTION OF X

$$C^B(x=0) = B$$

$$C^B(x=1) = C$$

# BAYES ERROR

← AVERAGE MISCLASSIFICATION  
USING BAYES CLASSIFIER

$$1 - E_x \left[ \max_k P[Y=k | X=x] \right]$$

"IRREDUCIBLE ERROR"

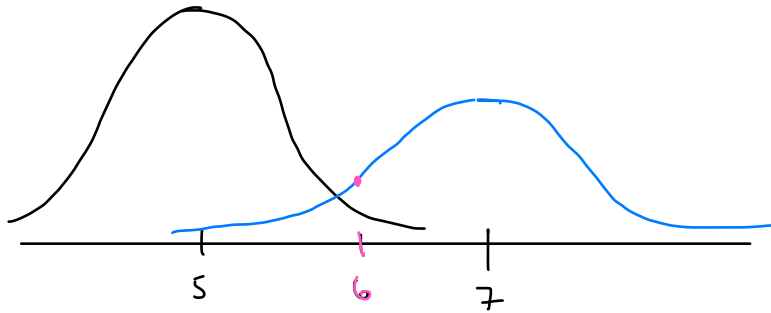
	X		
	0	1	
A	0.1 ✓	0.1 ✓	0.2
B	0.2 ✗	0.1 ✓	0.3
C	0.1 ✓	0.4 ✗	0.5
	0.4	0.6	

$$= 1 - \left[ P[Y=B | X=0] P[X=0] + P[Y=C | X=1] P[X=1] \right]$$

$$= 1 - \left[ \left( \frac{0.2}{0.4} \right) (0.4) + \left( \frac{0.4}{0.6} \right) (0.6) \right]$$

$$= 1 - [0.2 + 0.4] = \underline{0.4}$$

# EXAMPLE



$$X|Y=0 \sim N(\mu=5, \sigma=1) \quad f_0(x)$$

$$X|Y=1 \sim N(\mu=7, \sigma=2) \quad f_1(x)$$

$$\pi_0 = P[Y=0] = 0.6$$

$$\pi_1 = P[Y=1] = 0.4$$

$$C^B(X=6) = ?$$

CALCULATE

$$P[Y=0 | X=6] = \frac{\pi_0 f_0(6)}{\pi_0 f_0(6) + \pi_1 f_1(6)} = \dots$$



# IN PRACTICE

DON'T KNOW  $P[Y = k | X = x]$  !!!  
...

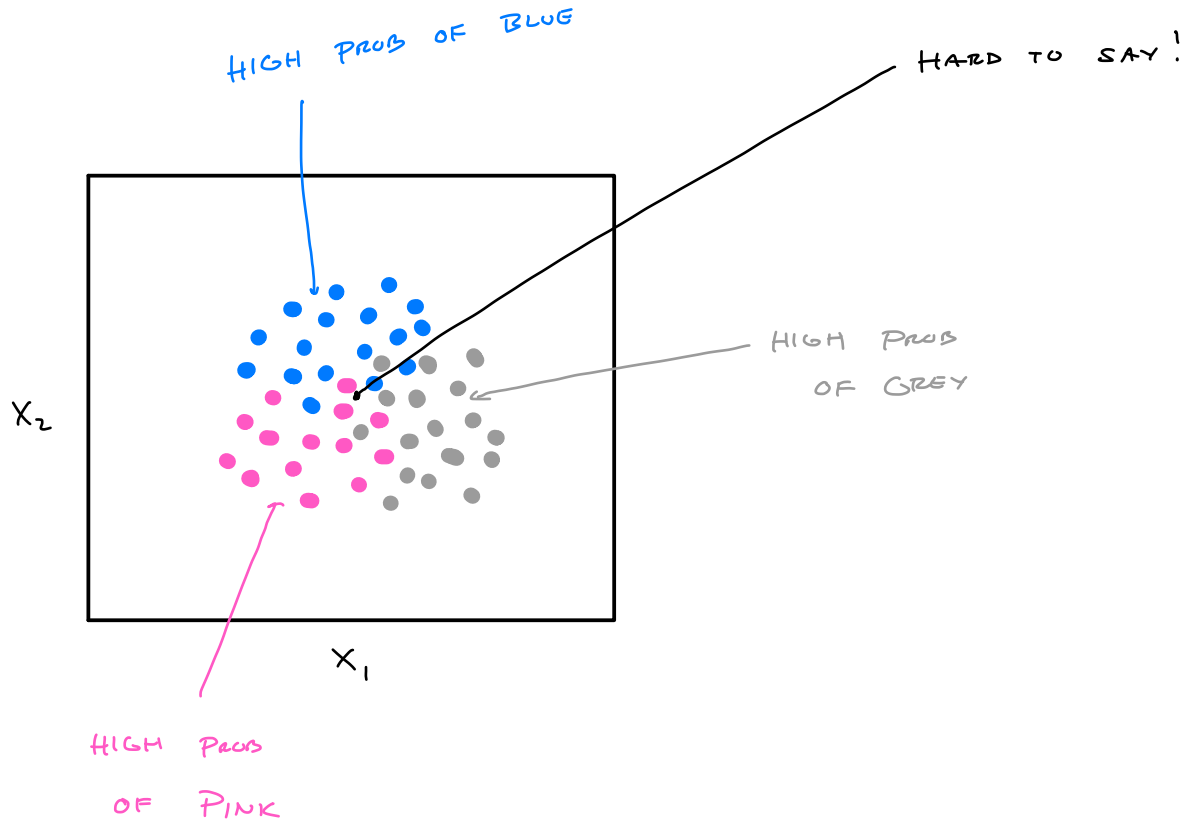
ESTIMATE IT !!!

LEARNED  
CLASSIFIER

$$\hat{C}(x) = \underset{k}{\operatorname{ARGMAX}} \underbrace{\hat{P}[Y = k | X = x]}_{\text{ESTIMATE OF CONDITIONAL PROBABILITY}}$$

A "GUESS"  
FOR  $\hat{C}^B(x)$

How?



# ESTIMATING CONDITIONAL PROBABILITIES

KNN

w/

caret :: knn3

TREES

w/

rpart :: rpart

LINEAR  
MODELS

w/

glm  
nnet :: nnet

} predict(mod, data, type)

ENSURE RESPONSE IS A FACTOR

# METRICS

would like  $P[C(x) \neq Y]$

SETTLE FOR

$$\frac{1}{n} \sum_{i=1}^n I(C(x_i) \neq y_i)$$

MISCLASSIFICATION

$$I(C(x_i) \neq y_i) = \begin{cases} 1 & C(x_i) \neq y_i \\ 0 & C(x_i) = y_i \end{cases}$$

$$\frac{1}{n} \sum_{i=1}^n I(C(x_i) = y_i)$$

ACCURACY

$C(x)$  PLACEHOLDER

$\hat{C}(x)$  LEARNED

$C^b(x)$  BAYES

# BINARY CLASSIFICATION

$Y = 0$  or  $Y = 1$   
↑                      ↑  
"NEGATIVE"        "POSITIVE"

METRICS

FP/TP

FN/TN

ETC

$$p(x) \triangleq P[Y=1 | X=x]$$

$$1-p(x) = P[Y=0 | X=x]$$

$$C^b(x) = \begin{cases} 1 & p(x) \geq 0.5 \\ 0 & \text{ELSE} \end{cases}$$