

# Lecture 23 – Classification and Conditional Independence



DSC 40A, Fall 2022 @ UC San Diego

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# Agenda

- ▶ Classification.
- ▶ Classification and conditional independence.

# Classification

# Taxonomy of machine learning

## Taxonomy of Machine Learning



Labeled Data

Reward

Unlabeled Data

Supervised Learning

Reinforcement Learning  
(not covered)

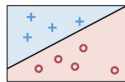
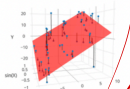
Unsupervised Learning

Quantitative Response

Categorical Response

Regression

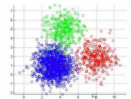
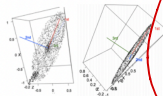
Classification



Alpha Go

Dimensionality Reduction

Clustering



# Classification problems

- ▶ Like with regression, we're interested in making predictions based on data we've already collected (called **training data**).
- ▶ The difference is that the response variable is **categorical**.
- ▶ Categories are called classes.
- ▶ Example classification problems:
  - ▶ Deciding whether a patient has kidney disease.
  - ▶ Identifying handwritten digits.
  - ▶ Determining whether an avocado is ripe.
  - ▶ Predicting whether credit card activity is fraudulent.

## Example: avocados

*feature*

You have a green-black avocado, and want to know if it is ripe.

*label*

color	ripeness
bright green	unripe
green-black	ripe
purple-black	ripe
green-black	unripe
purple-black	ripe
bright green	unripe
green-black	ripe
purple-black	ripe
green-black	ripe
green-black	unripe
purple-black	ripe

**Question:** Based on this data, would you predict that your avocado is ripe or unripe?

$\frac{3}{5}$  ripe

$\frac{2}{5}$  unripe

→ predict ripe

## Example: avocados

You have a green-black avocado, and want to know if it is ripe. Based on this data, would you predict that your avocado is ripe or unripe?

color	ripeness
bright green	unripe
green-black	ripe
purple-black	ripe
green-black	unripe
purple-black	ripe
bright green	unripe
green-black	ripe
purple-black	ripe
green-black	ripe
green-black	unripe
purple-black	ripe

**Strategy:** Calculate two probabilities:

$$P(\text{ripe}|\text{green-black}) = \frac{\# \text{ripe green-black}}{\# \text{green-black}}$$

$$P(\text{unripe}|\text{green-black}) = \frac{3}{5}$$

Then, predict the class with a **larger** probability.

#

$$= \frac{\# \text{unripe-gb}}{\# \text{gb}} = \frac{2}{2}$$

# Estimating probabilities

- ▶ We would like to determine  $P(\text{ripe}|\text{green-black})$  and  $P(\text{unripe}|\text{green-black})$  for all avocados in the universe.
- ▶ All we have is a single dataset, which is a **sample** of all avocados in the universe.
- ▶ We can estimate these probabilities by using sample proportions.

$$P(\text{ripe}|\text{green-black}) \approx \frac{\# \text{ ripe green-black avocados in sample}}{\# \text{ green-black avocados in sample}}$$

- ▶ Per the **law of large numbers** in DSC 10, larger samples lead to more reliable estimates of population parameters.



## Example: avocados

You have a green-black avocado, and want to know if it is ripe. Based on this data, would you predict that your avocado is ripe or unripe?

color	ripeness
bright green	unripe
green-black	ripe
purple-black	ripe
green-black	unripe
purple-black	ripe
bright green	unripe
green-black	ripe
purple-black	ripe
green-black	ripe
green-black	unripe
purple-black	ripe

$$P(\text{ripe}|\text{green-black}) =$$

$$\frac{3}{5}$$

$$P(\text{unripe}|\text{green-black}) =$$

$$\frac{2}{5}$$

## Bayes' theorem for classification

- ▶ Suppose that  $A$  is the event that an avocado has certain features, and  $B$  is the event that an avocado belongs to a certain class. Then, by Bayes' theorem:

$A$ : feature

$$P(B|A) = \frac{P(B) \cdot P(A|B)}{P(A)}$$

$B$ : belongs to a certain

- ▶ More generally: class

$$P(\text{class}|\text{features}) = \frac{P(\text{class}) \cdot P(\text{features}|\text{class})}{P(\text{features})}$$

- ▶ What's the point?

$P(\text{feature}|\text{class})$

- ▶ Usually, it's not possible to estimate  $P(\text{class}|\text{features})$  directly from the data we have.
- ▶ Instead, we have to estimate  $P(\text{class})$ ,  $P(\text{features}|\text{class})$ , and  $P(\text{features})$  separately.

## Example: avocados

R: ripe

G: Green-black

You have a green-black avocado, and want to know if it is ripe. Based on this data, would you predict that your avocado is ripe or unripe?

color	ripeness
bright green	unripe
✓ green-black ✓	ripe
purple-black	ripe
green-black ✓	unripe
purple-black	ripe
bright green	unripe
✓ green-black ✓	ripe
purple-black	ripe
✓ green-black ✓	ripe
green-black ✓	unripe
purple-black	ripe

$$P(\text{class}|\text{features}) = \frac{P(\text{class}) \cdot P(\text{features}|\text{class})}{P(\text{features})}$$

$$P(R|G) = \frac{P(R) \cdot P(G|R)}{P(G)}$$

$$\textcircled{1} P(R) = \frac{7}{11}$$

$$\textcircled{2} P(G|R) = \frac{3}{7}$$

$$\textcircled{3} P(G) = \frac{5}{11}$$

$$= \frac{\cancel{7/11} \cdot \cancel{3/7}}{\cancel{5/11}} = \frac{3}{5}$$

## Example: avocados

You have a green-black avocado, and want to know if it is ripe. Based on this data, would you predict that your avocado is ripe or unripe?

color	ripeness
bright green	unripe ✓
green-black	ripe
purple-black	ripe
✓ green-black	unripe ✓
purple-black	ripe
bright green	unripe ✓
green-black	ripe
purple-black	ripe
green-black	ripe
✓ green-black	unripe ✓
purple-black	ripe

$$P(R|G)$$

$$P(\text{class}|\text{features}) = \frac{P(\text{class}) \cdot P(\text{features}|\text{class})}{P(\text{features})}$$

$$P(\bar{R}|G) = \frac{P(\bar{R}) P(G|\bar{R})}{P(G)}$$

$$= \frac{\frac{4}{11} \cdot \frac{2}{4}}{5/11}$$

$$= \frac{2}{5}$$

## Example: avocados

You have a green-black avocado, and want to know if it is ripe. Based on this data, would you predict that your avocado is ripe or unripe?

color	ripeness
bright green	unripe
green-black	ripe
purple-black	ripe
green-black	unripe
purple-black	ripe
bright green	unripe
green-black	ripe
purple-black	ripe
green-black	ripe
green-black	unripe
purple-black	ripe

$$P(\text{class}|\text{features}) = \frac{P(\text{class}) \cdot P(\text{features}|\text{class})}{P(\text{features})}$$

**Shortcut:** Both probabilities have the same denominator. The larger one is the one with the larger numerator.

proportional

$$\propto \frac{7}{11} \cdot \frac{3}{7} = \frac{3}{11} \quad \text{ripe!}$$

$$P(\text{unripe}|\text{green-black})$$
$$\propto \frac{4}{11} \cdot \frac{2}{4} = \frac{2}{11}$$

## **Classification and conditional independence**

## Example: avocados, but with more features

color	softness	variety	ripeness
bright green	firm	Zutano	unripe
green-black	medium	Hass	ripe
purple-black	firm	Hass	ripe
green-black	medium	Hass	unripe
purple-black	soft	Hass	ripe
bright green	firm	Zutano	unripe
green-black	soft	Zutano	ripe
purple-black	soft	Hass	ripe
green-black	soft	Zutano	ripe
green-black	firm	Hass	unripe
purple-black	medium	Hass	ripe

You have a firm green-black Zutano avocado. Based on this data, would you predict that your avocado is ripe or unripe?

3 features

undefined

$$\frac{\# \text{ ripe firm Gb zutano}}{\# \text{ firm Gb zutano}} = \frac{0}{0}$$

## Example: avocados, but with more features

color	softness	variety	ripeness
bright green	firm	Zutano	unripe
green-black	medium	Hass	ripe
purple-black	firm	Hass	ripe
green-black	medium	Hass	unripe
purple-black	soft	Hass	ripe
bright green	firm	Zutano	unripe
green-black	soft	Zutano	ripe
purple-black	soft	Hass	ripe
green-black	soft	Zutano	ripe
green-black	firm	Hass	unripe
purple-black	medium	Hass	ripe

You have a firm green-black Zutano avocado. Based on this data, would you predict that your avocado is ripe or unripe?

**Strategy:** Calculate  $P(\text{ripe}|\text{features})$  and  $P(\text{unripe}|\text{features})$  and choose the class with the **larger** probability.

$$P(\text{ripe}|\text{firm, green-black, Zutano})$$

$$P(\text{unripe}|\text{firm, green-black, Zutano})$$



## Example: avocados, but with more features

color	softness	variety	ripeness
bright green	firm	Zutano	unripe
green-black	medium	Hass	ripe
purple-black	firm	Hass	ripe
green-black	medium	Hass	unripe
purple-black	soft	Hass	ripe
bright green	firm	Zutano	unripe
green-black	soft	Zutano	ripe
purple-black	soft	Hass	ripe
green-black	soft	Zutano	ripe
green-black	firm	Hass	unripe
purple-black	medium	Hass	ripe

You have a firm green-black Zutano avocado. Based on this data, would you predict that your avocado is ripe or unripe?

**Issue:** We have not seen a firm green-black Zutano avocado before.

This means that  $P(\text{ripe}|\text{firm, green-black, Zutano})$  and  $P(\text{unripe}|\text{firm, green-black, Zutano})$  are undefined.

## A simplifying assumption

- ▶ We want to find  $P(\text{ripe}|\text{firm, green-black, Zutano})$ , but there are no firm green-black Zutano avocados in our dataset.
- ▶ Bayes' theorem tells us this probability is equal to

$$P(\text{ripe}|\text{firm, green-black, Zutano}) = \frac{P(\text{ripe}) \cdot P(\text{firm, green-black, Zutano}|\text{ripe})}{P(\text{firm, green-black, Zutano})}$$

- ▶ **Key idea: Assume** that features are **conditionally independent** given a class (e.g. ripe).

$$P(\text{firm, green-black, Zutano}|\text{ripe}) = P(\text{firm}|\text{ripe}) \cdot P(\text{green-black}|\text{ripe}) \cdot P(\text{Zutano}|\text{ripe})$$

$$P(\text{class} | \text{features}) = \frac{P(\text{class}) P(\text{features} | \text{class})}{P(\text{features})}$$

$$P(A \cap B | C) = P(A | C) P(B | C)$$

$$P(\text{firm, green-black, zutano} | \text{ripe})$$

Assumption

$$\underline{\underline{\downarrow}} \quad P(\text{firm} | \text{ripe}) P(\text{gb} | \text{ripe}) P(\text{zutano} | \text{ripe})$$

## Example: avocados, but with more features

color	softness	variety	ripeness
bright green	firm	Zutano	unripe
green-black	medium	Hass	ripe
purple-black	firm	Hass	ripe
green-black	medium	Hass	unripe
purple-black	soft	Hass	ripe
bright green	firm	Zutano	unripe
green-black	soft	Zutano	ripe
purple-black	soft	Hass	ripe
green-black	soft	Zutano	ripe
green-black	firm	Hass	unripe
purple-black	medium	Hass	ripe

You have a firm green-black Zutano avocado. Based on this data, would you predict that your avocado is ripe or unripe?

$$P(\text{ripe} | \text{firm, green-black, Zutano}) = \frac{P(\text{ripe}) \cdot P(\text{firm, green-black, Zutano} | \text{ripe})}{P(\text{firm, green-black, Zutano})}$$

$$\begin{aligned} &\propto P(\text{ripe}) \cdot P(\text{firm} | \text{ripe}) \cdot P(\text{gb} | \text{ripe}) \\ &\quad \cdot P(\text{Zutano} | \text{ripe}) = \frac{7}{11} \cdot \frac{1}{7} \cdot \frac{3}{7} \cdot \frac{2}{7} \\ &= \frac{6}{1539} \end{aligned}$$

## Example: avocados, but with more features

color	softness	variety	ripeness
bright green	firm	Zutano	unripe
green-black	medium	Hass	ripe
purple-black	firm	Hass	ripe
green-black	medium	Hass	unripe
purple-black	soft	Hass	ripe
bright green	firm	Zutano	unripe
green-black	soft	Zutano	ripe
purple-black	soft	Hass	ripe
green-black	soft	Zutano	ripe
green-black	firm	Hass	unripe
purple-black	medium	Hass	ripe

You have a firm green-black Zutano avocado. Based on this data, would you predict that your avocado is ripe or unripe?

$$P(\text{unripe} | \text{firm, green-black, Zutano}) = \frac{P(\text{unripe}) \cdot P(\text{firm, green-black, Zutano} | \text{unripe})}{P(\text{firm, green-black, Zutano})}$$

$$\propto P(\text{ripe}) \cdot P(\text{firm} | \text{unripe}) \cdot P(\text{gb} | \text{unripe}) \\ \cdot P(\text{Zutano} | \text{unripe}) = \frac{4}{11} \cdot \frac{3}{4} \\ \cdot \frac{2}{4} \cdot \frac{2}{4} = \frac{3}{44} = \frac{6}{88}$$

## Conclusion

- ▶ The numerator of  $P(\text{ripe}|\text{firm, green-black, Zutano})$  is  $\frac{6}{539}$ .
- ▶ The numerator of  $P(\text{unripe}|\text{firm, green-black, Zutano})$  is  $\frac{6}{88}$ .
  - ▶ Both probabilities have the same denominator,  $P(\text{firm, green-black, Zutano})$ .
  - ▶ Since we're just interested in seeing which one is larger, we can ignore the denominator and compare numerators.
- ▶ Since the numerator for unripe is **larger** than the numerator for ripe, we **predict that our avocado is unripe**.

## Summary

## Summary

$$\text{Assumption: } P(\text{feature}_1, f_2 | \text{class}) \\ = P(f_1 | \text{class}) P(f_2 | \text{class})$$

- ▶ In classification, our goal is to predict a discrete category, called a **class**, given some features.
- ▶ The Naive Bayes classifier works by estimating the numerator of  $P(\text{class} | \text{features})$  for all possible classes.