Lecture 17

Naïve Bayes

DSC 40A, Summer 2024

Announcements

- Homework 8, the final homework, is due tomorrow.
  - It's short: only two questions.

The Final Exam is on Friday, September 6th!

- The Final Exam is on **Friday, September 6th from 11:30AM-2:30PM** in WLH 2113.
- 180 minutes, on paper, no calculators or electronics.
    - **You are allowed to bring two double-sided index cards (4 inches by 6 inches) of notes that you write by hand (no iPad).**
- Content: All lectures (including this week), homeworks (including HW 8), and groupworks.
- Prepare by practicing with old exam problems at practice.dsc40a.com.
    - There are tons of past probability exams, searchable by topic.
    - Check out the advice page for study strategies.
- No formal review session but lots of office hours this week - come through!

Agenda

- Classification.
- Classification and conditional independence.
- Naïve Bayes.

Recap: Bayes' Theorem, independence, and conditional independence

- Bayes' Theorem describes how to update the probability of one event, given that another event has occurred.

new old input A

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

$\mathbb{P}(A|B) = \mathbb{P}(A)$

equiv

$\mathbb{P}(B|A) = \mathbb{P}(B)$

- A and B are **independent** if:

$$\mathbb{P}(A \cap B) = \mathbb{P}(A) \cdot \mathbb{P}(B)$$

- A and B are **conditionally independent** given C if:

$$\mathbb{P}((A \cap B)|C) = \mathbb{P}(A|C) \cdot \mathbb{P}(B|C)$$

  - In general, there is no relationship between independence and conditional independence.

Question 🤔

Answer at q.dsc40a.com

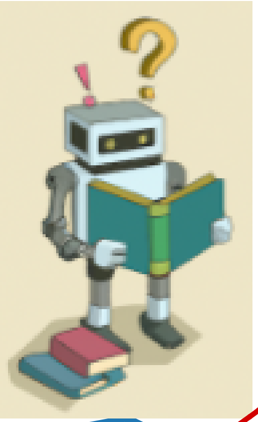
Remember, you can always ask questions at q.dsc40a.com!

If the direct link doesn't work, click the "🤔 Lecture Questions" link in the top right corner of dsc40a.com.

Classification

When you already know the "right answer" in your dataset
⇒ y variable

Taxonomy of machine learning



Labeled data

Unlabeled data

Reward

Supervised learning

Quantitative prediction

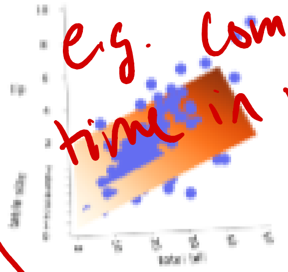
Categorical prediction

today!

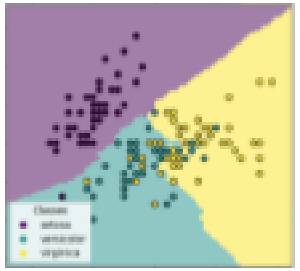
Regression

Predict a real number y_i

e.g. Commute time (in minutes)

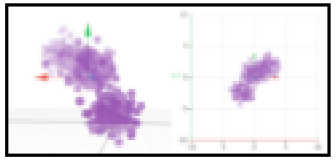


Classification

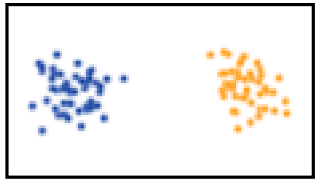


Unsupervised learning

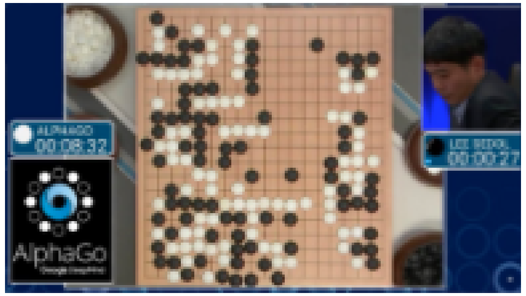
Dimensionality reduction



Clustering



Reinforcement learning



AlphaGo

Classification problems

- Like with regression, we're interested in making predictions based on data (called **training data**) for which we know the value of the response variable. y
- The difference is that the response variable is now **categorical**.
  y
- Categories are called **classes**.
- Example classification problems:
  - Deciding whether a patient has kidney disease.
  - Identifying handwritten digits.
  - Determining whether an avocado is ripe. today
  - Predicting whether credit card activity is fraudulent.
  - Predicting whether you'll be late to school or not.

Example: Avocados

color (x)	ripeness (y)
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 ✅

"total" training data

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

→ more of the G-B avocados are ripe than unripe

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

Of the 5 G-B avocados I've seen:

3 are ripe

2 are unripe

3 > 2 so I'll predict that my new avocado is ripe!

Example: Avocados

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 ✅

new

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

Strategy: Calculate two probabilities:

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

$$\mathbb{P}(\text{unripe}|\text{green-black}) = \frac{2}{5}$$

total # of G-B avocados

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

$\frac{3}{5} > \frac{2}{5}$ so ripe seems more likely.

predict ripe → H(green-black) = ripe

Estimating probabilities

population parameters

- We would like to determine $\mathbb{P}(\text{ripe}|\text{green-black})$ and $\mathbb{P}(\text{unripe}|\text{green-black})$ 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. sample statistic

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

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

Example: Avocados

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 ✅

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

Predict!

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

$$\mathbb{P}(\text{unripe}|\text{green-black}) = \frac{2}{5}$$

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:

$$\mathbb{P}(\overset{\text{ripe}}{\downarrow}\!B|\underset{\text{green-black}}{\uparrow}\!A) = \frac{\mathbb{P}(B) \cdot \mathbb{P}(A|B)}{\mathbb{P}(A)}$$

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

- More generally:

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

- What's the point?
  - Usually, it's not possible to estimate $\mathbb{P}(\text{class}|\text{features})$ directly.
  - Instead, we often have to estimate $\mathbb{P}(\text{class})$, $\mathbb{P}(\text{features}|\text{class})$, and $\mathbb{P}(\text{features})$ separately.

Example: Avocados

$\frac{3}{7} = \frac{\frac{3}{11}}{\frac{7}{11}} \Rightarrow \frac{\mathbb{P}(\text{green-black} \cap \text{ripe})}{\mathbb{P}(\text{ripe})}$

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 ✅

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

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

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

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

same as before!

Example: Avocados

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 ✅

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

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

$$\mathbb{P}(\text{unripe} \mid \text{green-black}) = \frac{\mathbb{P}(\text{unripe}) \cdot \mathbb{P}(\text{g-b} \mid \text{unripe})}{\mathbb{P}(\text{g-b})} = \frac{\frac{4}{11} \cdot \frac{2}{4}}{\frac{5}{11}} = \boxed{\frac{2}{5}}$$

Same!

Example: Avocados

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 ✅

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

fixed for a given input

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

5/11

Shortcut: Both probabilities have the same denominator, so the larger probability is the one with the **larger numerator**.

proportional to

$$\mathbb{P}(\text{ripe}|\text{green-black}) \propto \mathbb{P}(\text{ripe}) \cdot \mathbb{P}(\text{g-b} \mid \text{ripe})$$

$= \frac{3}{11}$ bigger!

$$\mathbb{P}(\text{unripe}|\text{green-black}) \propto \mathbb{P}(\text{unripe}) \cdot \mathbb{P}(\text{g-b} \mid \text{unripe})$$

$= 2/11$

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

x y

new

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

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

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

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

$$= \frac{\#\ \text{ripe, firm, g-b, Zutano}}{\#\ \text{firm, g-b, Zutano}} = \frac{0}{0} = ???$$

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

Issue: We have not seem a firm green-black Zutano avocado before, which means that the following probabilities are undefined:

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

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

$\mathbb{P}(A \cap B | C) = \mathbb{P}(A|C) \cdot \mathbb{P}(B|C)$ Cond. indp!

A simplifying assumption

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

$\mathbb{P}(\text{features} \mid \text{class})$

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

$\mathbb{P}(\text{class} \mid \text{features})$

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

"ripe" is the given

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

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?

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

problem

$$\propto \mathbb{P}(\text{ripe}) \cdot \mathbb{P}(\text{firm} \wedge \text{g-b} \wedge \text{Zutano} \mid \text{ripe})$$

$$= \mathbb{P}(\text{ripe}) \cdot \mathbb{P}(\text{firm} \mid \text{ripe})\, \mathbb{P}(\text{g-b} \mid \text{ripe})\, \mathbb{P}(\text{Zutano} \mid \text{ripe})$$

$$= \frac{7}{11} \cdot \frac{1}{7} \cdot \frac{3}{7} \cdot \frac{2}{7} = \boxed{\frac{6}{539}}$$

assuming! features conditionally independent given class

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?

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

$$\propto \mathbb{P}(\text{unripe})\, \mathbb{P}(\text{firm} \mid \text{unripe})\, \mathbb{P}(\text{g-b} \mid \text{unripe})\, \mathbb{P}(\text{Zutano} \mid \text{unripe})$$

$$= \frac{4}{11} \cdot \frac{3}{4} \cdot \frac{2}{4} \cdot \frac{2}{4} = \boxed{\frac{3}{44} = \frac{6}{88}}$$

Conclusion

- The numerator of $\mathbb{P}(\text{ripe}|\text{firm, green-black, Zutano})$ is $\frac{6}{539}$.
- The numerator of $\mathbb{P}(\text{unripe}|\text{firm, green-black, Zutano})$ is $\frac{6}{88}$.
- Both probabilities have the same denominator, $\mathbb{P}(\text{firm, green-black, Zutano})$.
- 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** ❌.

$$\frac{6}{8} > \frac{6}{10}$$

$$.75 > 0.6$$

Naïve Bayes

The Naïve Bayes classifier

- We want to predict a class, given certain features.
- Using Bayes' Theorem, we write:

apply cond. indp to this

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

- For each class, we compute the numerator using the **naïve assumption of conditional independence of features given the class**.
- We estimate each term in the numerator based on the training data.
- We predict the class with the largest numerator.
    - Works if we have multiple classes, too!

Dictionary

Definitions from Oxford Languages · Learn more

na·ive

/näˈēv/

adjective

(of a person or action) showing a lack of experience, wisdom, or judgment.
"the rather naive young man had been totally misled"

- (of a person) natural and unaffected; innocent.
  "Andy had a sweet, naive look when he smiled"

  Similar: innocent | unsophisticated | artless | ingenuous | inexperienced

- of or denoting art produced in a straightforward style that deliberately rejects sophisticated artistic techniques and has a bold directness resembling a child's work, typically in bright colors with little or no perspective.

Example: Avocados, again

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 soft, green-black, Hass avocado. Based on this data, would you predict that your avocado is ripe or unripe?

wins!

$$\mathbb{P}(\text{ripe} \mid \text{soft}, \text{g-b}, \text{Hass}) \propto$$

$$\mathbb{P}(\text{ripe})\, \mathbb{P}(\text{soft} \mid \text{ripe})\, \mathbb{P}(\text{g-b} \mid \text{ripe})\, \mathbb{P}(\text{Hass} \mid \text{ripe})$$

$$= \frac{7}{11} \cdot \frac{4}{7} \cdot \frac{3}{7} \cdot \frac{5}{7}$$

$$\mathbb{P}(\text{unripe} \mid \text{soft}, \text{g-b}, \text{Hass}) \propto$$

$$\mathbb{P}(\text{unripe})\ \mathbb{P}(\text{soft} \mid \text{unripe})\ \mathbb{P}(\text{g-b} \mid \text{unripe}) \cdot \mathbb{P}(\text{Hass} \mid \text{unripe})$$

$$= \frac{4}{11} \cdot \frac{0}{4} \cdot \frac{2}{4} \cdot \frac{2}{4} = \boxed{0}$$

Uh oh!

- There are no soft unripe avocados in the data set.
- The estimate $\mathbb{P}(\text{soft}|\text{unripe}) \approx \dfrac{\#\text{ soft unripe avocados}}{\#\text{ unripe avocados}}$ is 0.
- The estimated numerator:

  0

  $$\mathbb{P}(\text{unripe}) \cdot \mathbb{P}(\text{soft, green-black, Hass}|\text{unripe}) = \mathbb{P}(\text{unripe}) \cdot \mathbb{P}(\text{soft}|\text{unripe}) \cdot \mathbb{P}(\text{green-black}|\text{unripe}) \cdot \mathbb{P}(\text{Hass}|\text{unripe})$$

  is also 0.
- But just because there isn't a soft unripe avocado in the data set, doesn't mean that it's impossible for one to exist!
- **Idea**: Adjust the numerators and denominators of our estimate so that they're never 0.

Smoothing

- **Without** smoothing:

Sum up to 1

$$
\begin{cases}
\mathbb{P}(\text{soft}|\text{unripe}) \approx \dfrac{\overbrace{\#\text{ soft unripe}}^{=0}}{\underbrace{\#\text{ soft unripe} + \#\text{ medium unripe} + \#\text{ firm unripe}}_{\#\text{ unripe avocados}}} \\
\mathbb{P}(\text{medium}|\text{unripe}) \approx \dfrac{\#\text{ medium unripe}}{\#\text{ soft unripe} + \#\text{ medium unripe} + \#\text{ firm unripe}} \\
\mathbb{P}(\text{firm}|\text{unripe}) \approx \dfrac{\#\text{ firm unripe}}{\#\text{ soft unripe} + \#\text{ medium unripe} + \#\text{ firm unripe}}
\end{cases}
$$

- **With** smoothing:

Adds to 1

$$
\begin{cases}
\mathbb{P}(\text{soft}|\text{unripe}) \approx \dfrac{\#\text{ soft unripe} + 1}{\#\text{ soft unripe} + 1 + \#\text{ medium unripe} + 1 + \#\text{ firm unripe} + 1} \\
\mathbb{P}(\text{medium}|\text{unripe}) \approx \dfrac{\#\text{ medium unripe} + 1}{\#\text{ soft unripe} + 1 + \#\text{ medium unripe} + 1 + \#\text{ firm unripe} + 1} \\
\mathbb{P}(\text{firm}|\text{unripe}) \approx \dfrac{\#\text{ firm unripe} + 1}{\#\text{ soft unripe} + 1 + \#\text{ medium unripe} + 1 + \#\text{ firm unripe} + 1}
\end{cases}
$$

→ can't be 0 any more!

- When smoothing, we add 1 to the count of every group whenever we're estimating a conditional probability.

Example: Avocados, with smoothing only cond. prob.

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 soft green-black Hass avocado. Based on this data, would you predict that your avocado is ripe or unripe?

$$P(\text{ripe} \mid \text{soft}, \text{g-b}, H)$$

$$\propto P(\text{ripe})\, P(\text{soft} \mid \text{ripe})\, P(\text{g-b} \mid \text{ripe}) \cdot P(\text{Hass} \mid \text{ripe})$$

$$= \underbrace{\frac{7}{11}}_{\text{not smoothed}} \cdot \frac{4+1}{7+3} \cdot \frac{3+1}{7+3} \cdot \frac{5+1}{7+2} = \frac{14}{165}$$

soft, med, firm

g-b, b-g, p-b

Hass, Zutano

Example: Avocados, with smoothing

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 soft green-black Hass avocado. Based on this data, would you predict that your avocado is ripe or unripe?

$P(\text{unripe} \mid \text{soft}, \text{g-b}, \text{Hass})$

$\propto P(\text{unripe}) \cdot P(\text{soft} \mid \text{unripe}) \cdot P(\text{g-b} \mid \text{unripe}) \cdot P(\text{Hass} \mid \text{unripe})$

$= \frac{4}{11} \cdot \frac{0+1}{4+3} \cdot \frac{2+1}{4+3} \cdot \frac{2+1}{4+2} = \boxed{\frac{6}{539}}$

g-b, p-b, bg

H, Z

using proportionality trick

ripe numerator : $\frac{14}{165}$

$>$

unripe numerator : $\frac{6}{539}$

Predict a soft, green-black, Hass avocado is

$\boxed{\text{Ripe}}$

Summary

Summary

- In classification, our goal is to predict a discrete category, called a **class**, given some features.
- The Naïve Bayes classifier uses Bayes' Theorem:

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

- And works by estimating the numerator of $\mathbb{P}(\text{class}|\text{features})$ for all possible classes.
- It also uses a simplifying assumption, that features are conditionally independent given a class:

$$\mathbb{P}(\text{features}|\text{class}) = \mathbb{P}(\text{feature}_1|\text{class}) \cdot \mathbb{P}(\text{feature}_2|\text{class}) \cdot \ldots$$