## Module 23 - Naive Bayes



DSC 40A, Summer 2023

Agenda
Classification.

Classification and conditional independence.

Naive Bayes.
Y probabilistic approach to supervised learning

## Recap: Bayes' theorem, independence, and conditional independence

- Bayes' theorem: $P(A \mid B)=\frac{P(A) P(B \mid A)}{P(B)}$.
- $A$ and $B$ are independent if $P(A \cap B)=P(A) \cdot P(B)$.
- $A$ and $B$ are conditionally independent given $C$ if $P((A \cap B) \mid C)=P(A \mid C) \cdot P(B \mid C)$.
- In general, there is no relationship between independence and conditional independence.


## Classification

## Taxonomy of machine learning



## 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.
- The difference is that the response variable is now categorical.
- Categories are called classes.
- Example classification problems:
$\underset{\text { class }}{\text { cmance }} \rightarrow$ Deciding whether a patient has kidney disease. $\rightarrow$ binary
multichss" $\rightarrow$ Identifying handwritten digits.
$\Rightarrow$ Determining whether an avocado is ripe. $\rightarrow$ binort
$\xrightarrow[\text { class }]{\text { imblance }} \rightarrow$ Predicting whether credit card activity is fraudulent.

Example: avocados
6 out of sample
You have a green-black avocado, and want to know if it is ripe.


Question: Based on this data, would you predict that your avocado is ripe or unripe?
$S$ queen black

$$
3 / s \text { ripe }
$$

## 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:

$$
\begin{aligned}
& P(\text { ripe lgreen-black })=\frac{3}{5} \\
& P(\text { unripelgreen-black })=\frac{2}{5}
\end{aligned}
$$

Then, predict the class with a larger probability.

## Estimating probabilities

- We would like to determine $P$ (ripelgreen-black) and $P$ (unripelgreen-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.
$P($ ripe lgreen-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 |

$$
\begin{aligned}
& P(\text { ripelgreen-black })=3 / 5 \\
& P(\text { unripelgreen-black })=2 / 5
\end{aligned}
$$

## Bayes' theorem for classification

$\Rightarrow$ 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:

- More generally:

$$
\begin{gathered}
P(\text { class|features })=\frac{\begin{array}{c}
P(\text { class }) \cdot P(\text { features } \mid c l a s s)
\end{array}}{P(\text { features })} \\
\text { qreen black }
\end{gathered}
$$

- What's the point?
- Usually, it's not possible to estimate P(class|features) directly from the data we have.
- Instead, we have to estimate $P$ (class), $P$ (features|class), and $P$ (features) separately.

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?


## 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 |
| :--- | :--- |
|  | (class $\mid$ features $)=\frac{P(\text { class }) \cdot P(\text { features } \mid \text { class })}{P(\text { features })}$ |
| bright green |  |
| 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 |

## 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 |

$P($ class $\mid$ features $)=\frac{P(\text { class }) \cdot P(\text { features } \mid \text { class })}{P(\text { features })}$
Shortcut: Both probabilities have the same denominator. The larger one is the one with the larger numerator.

$P$ (ripelgreen-black)<br>$P$ (unripelgreen-black)

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 |
| Cofar | Varkfa |  |  |

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 $P$ (ripe|features) and $P$ (unripe|features) and choose the class with the larger probability.

$$
\begin{gathered}
P(\text { ripelfirm, green-black, Zutano) } \\
P(\text { unripelfirm, green-black, Zutano) }
\end{gathered}
$$

## 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 $P$ (ripelfirm, green-black, Zutano) and $P$ (unripe|firm, green-black, Zutano) are undefined.

## A simplifying assumption

- We want to find $P$ (ripelfirm, green-black, Zutano), but there are no firm green-black Zutano avocados in our dataset.
- Bayes' theorem tells us this probability is equal to
$P\left(\right.$ ripe ${ }^{\text {firm, }}$ green-black, Zutano $)=\frac{P(\text { ripe }) \cdot P(\text { firm, green-black, Zutanolripe })}{P(\text { firm, green-black, Zutano })}$
- Key idea: Assume that features are conditionally independent given a class (e.g. ripe).
$P($ firm, green-black, Zutano|ripe $)=P$ (firm $\mid$ ripe $) \cdot P($ green-black $\mid$ ripe $) \cdot P($ Zutano $\mid$ 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?

$$
\begin{aligned}
& P(\text { ripe firm, green-black, Zutano })=\frac{P(\text { ripe }) \cdot P(\text { firm, green-black, Zutanolripe })}{}
\end{aligned}
$$

$$
\begin{aligned}
& 7 / 11 \cdot 1 / 7 \cdot 3 / 7 \cdot 2 / 7=\frac{42}{3723}=6 / 539
\end{aligned}
$$

$$
\begin{aligned}
& 4 / 113 / 4 \cdot 2 / 4 \cdot 2 / 4 \\
& 3 / 44
\end{aligned}
$$

## 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?
$P($ unripelfirm, green-black, Zutano $)=\frac{P(\text { unripe }) \cdot P(\text { firm, green-black, Zutanolunripe })}{P(\text { firm, green-black, Zutano })}$

## Conclusion

- The numerator of $P$ (ripe firm, green-black, Zutano) is $\frac{6}{539}$.
- The numerator of $P$ (unripe|firm, green-black, Zutano) is $\frac{6}{88} \cdot 3 / 44$

Both probabilities have the same denominator, $P$ (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.

Naive Bayes

## Naive Bayes classifier

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

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

- For each class, we compute the numerator using the naive 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

## (1) na.ive

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 $\checkmark$
- 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 | Has | unripe |
| purple-black | soft | Hass | ripe |
| bright green | firm | Zutano | unripe |
| green-black | soft | Zutano | ripe |
| purple-black | soft | Has | ripe |
| green-black | soft | Zutano | ripe |
| green-black | firm | Hass | unripe |
| purple-black | medium | Hass | ripe |

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


$$
7 / 11 \cdot 4 / 7 \cdot 3 / 3 \cdot 5
$$



$$
4 / 11
$$

## Uh oh...

- There are no soft unripe avocados in the data set.

The estimate $P($ soft $\mid$ unripe $) \approx \frac{\# \text { soft unripe avocados }}{\# \text { unripe avocados }}$ is 0 .

- The estimated numerator, $P$ (unripe) $\cdot P$ (soft, green-black, Hass|unripe) $=P$ (unripe) . $P$ (soft|unripe) $\cdot P$ (green-black|unripe) $\cdot P$ (Hass|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:

$$
\begin{aligned}
P(\text { soft lunripe }) & \approx \frac{\text { \# soft unripe }}{\# \text { soft unripe }+ \text { \# medium unripe }+\# \text { firm unripe }} \\
P(\text { medium lunripe }) & \approx \frac{\text { \# medium unripe }}{\# \text { soft unripe }+ \text { \# medium unripe }+\# \text { firm unripe }} \\
P(\text { firm lunripe }) & \approx \frac{\text { \# firm unripe }}{\# \text { soft unripe }+ \text { \# medium unripe }+ \text { \# firm unripe }}
\end{aligned}
$$

- With smoothing:

$$
\begin{aligned}
P(\text { soft lunripe }) & \approx \frac{\text { \# soft unripe }+1}{\# \text { soft unripe }+1+\# \text { medium unripe }+1+\# \text { firm unripe }+1} \\
P(\text { medium } \mid \text { unripe }) & \approx \frac{\# \text { medium unripe }+1}{\# \text { soft unripe }+1+\# \text { medium unripe }+1+\# \text { firm unripe }+1} \\
P(\text { firm |unripe }) & \approx \frac{\# \text { firm unripe }+1}{\# \text { soft unripe }+1+\# \text { medium unripe }+1+\# \text { firm unripe }+1}
\end{aligned}
$$

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


## 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. Using Naive Bayes, with smoothing, would you predict that your avocado is ripe or unripe?

## Summary

## Summary

- 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 $P$ (class|features) for all possible classes.
- It uses Bayes' theorem:

$$
P(\text { class } \mid \text { features })=\frac{P(\text { class }) \cdot P(\text { features } \mid c l a s s)}{P(\text { features })}
$$

- It also uses a simplifying assumption, that features are conditionally independent given a class:
$P($ features $\mid$ class $)=P\left(\right.$ feature ${ }_{1} \mid$ class $) \cdot P\left(\right.$ feature ${ }_{2} \mid$ class $) \cdot \ldots$

