Data Science of Text Generation 1. Taking your chances

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Bachelor in Data Science

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Random words ●○○	Probability rules	Conditional probabilities	Bayes' Theorem	Conclusion O
A funny po	em			

Normally, probability starts with an urn of coloured balls.

We start with a poem:

Do you carrot all for me? My heart beets for you, With your turnip nose And your radish face, You are a peach. If we cantaloupe, Lettuce marry: Weed make a swell pear.

consisting of 28 different words.

Task 1. Pick one random word from poem and write down.



Probability rules

Conditional probabilities

Bayes' Theorem

Conclusion

Have you magically chosen the same word?

Question:

How likely is it that at least two of you selected same word?

Choose from:

10%, 40%, 80%, 95%?



Random words ○O●	Probability rules	Conditional probabilities	Bayes' Theorem	Conclusion O
What are th	ne chances			

The 28 words in poem consist of:

- **9 food items**: carrot, beets, turnip, radish, peach, cantaloupe, lettuce, weed, pear.
- 3 body parts: heart, nose, face
- 4 verbs: do, are, marry, make
- 5 pronouns: you, me, my, your, we
- 7 others: all, for, with, and, a, if, swell

Task 2. What is the probability that you chose a food item?





The 28 words in poem consist of:

- 9 food items: carrot, beets, turnip, radish, peach, cantaloupe, lettuce, weed, pear.
- 3 body parts: heart, nose, face
- 4 verbs: do, are, marry, make
- 5 pronouns: you, me, my, your, we
- 7 others: all, for, with, and, a, if, swell

Task 2. What is the probability that you chose a food item?

$$P(\text{chose food item}) = \frac{9}{28} = 0.32$$



Probability rules

Conditional probabilities

Bayes' Theorem

Conclusion

Probability rules!



Probability rules

Conditional probabilities

Bayes' Theorem

 $\underset{\bigcirc}{\text{Conclusion}}$

Definition of probability according to Marquis de Laplace (1779)



Pierre Simon Laplace

Probability of event is ratio of
number of cases favorable, to
number of all cases possible;
when nothing leads us to expect that any one of these cases should occur more than any other, which renders them, for us, equally possible.



Probability rules

Conditional probabilities

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Conclusion

Definition of probability according to Marquis de Laplace (1779)



Pierre Simon Laplace

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number of all cases possible;
when nothing leads us to expect that any one of these cases should occur more than any other, which renders them, for us, equally possible.

In mathematical terms:

 $P(E) = \frac{\text{Number of elements in } E}{\text{Total number of elements}}$

where *E* is an event.



Random words	Probability rules ○○●○○○○○○○○○○	Conditional probabilities	Bayes' Theorem	Conclusion O
Sample sp	ace			

Consider a process with an uncertain outcome:

- amount of rain in Lugano tomorrow,
- roll of a die,
- Word chosen from funny poem.

Collection of all possible outcomes is the Sample Space.

$$S_{rain} =$$



Random words	Probability rules	Conditional probabilities	Bayes' Theorem	Conclusion O
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Collection of all possible outcomes is the Sample Space.

$$S_{\mathsf{rain}} = \{ x \mid x \ge 0 \},\$$

$$S_{die} =$$



Random words	Probability rules	Conditional probabilities	Bayes' Theorem	Conclusion O
Sample sp	ace			

Consider a process with an uncertain outcome:

- amount of rain in Lugano tomorrow,
- roll of a die,
- Word chosen from funny poem.

Collection of all possible outcomes is the Sample Space.

$$S_{\mathsf{rain}} = \{ x \mid x \ge 0 \},\$$

$$\textit{S}_{\textit{die}} = \{1, 2, 3, 4, 5, 6\}$$

 $S_{poem} = \{do, you, carrot, all, \dots, swell, pear\}$



Random words	Probability rules	Conditional probabilities	Bayes' Theorem	Conclusion O
Events				

An event is a subset of the sample space.

E = word is a verb

is an event w.r.t. Spoem, since

 $E = \{ do, are, marry, make \} \subset S_{poem}.$

The set

F = word is funny

is *not* an event w.r.t. Spoem, because $F \not\subset S_{poem}$.



Random words	Probability rules	Conditional probabilities	Bayes' Theorem	Conclusion ○
Oemskining				

Consider selecting word from poem and following two events:

- A = food item
- B = contains letter "w",

then we can combine the events as follows:

 $A \cap B =$



Random words	Probability rules ○○○○●○○○○○○○○	Conditional probabilities	Bayes' Theorem	Conclusion ○
A 1 1 1				

Consider selecting word from poem and following two events:

- A = food item
- B = contains letter "w",

$$A \cap B = \{weed\}$$

 $A \cup B =$



Random words	Probability rules	Conditional probabilities	Bayes' Theorem	Conclusion O
Osushinina				

Consider selecting word from poem and following two events:

- A = food item
- B = contains letter "w",

$$A \cap B = \{weed\}$$

 $A \cup B = \{carrot, ..., pear, we, with, swell\}$
 $A^c =$



Random words	Probability rules	Conditional probabilities	Bayes' Theorem	Conclusion O
Osushinina	a venta			

Consider selecting word from poem and following two events:

- A = food item
- B = contains letter "w",

$$A \cap B = \{weed\}$$

 $A \cup B = \{carrot, ..., pear, we, with, swell\}$
 $A^c = \{heart, nose, ..., if, swell\}$
 $B^c =$



Random words	Probability rules	Conditional probabilities	Bayes' Theorem	Conclusion O
Combining	overte			

Consider selecting word from poem and following two events:

- A = food item
- B = contains letter "w",

$$A \cap B = \{\text{weed}\}$$

$$A \cup B = \{\text{carrot}, \dots, \text{pear, we, with, swell}\}$$

$$A^c = \{\text{heart, nose, } \dots, \text{if, swell}\}$$

$$B^c = S_{\text{poem}} - \{\text{weed, we, with, swell}\}$$

$$B^c \cup B^c =$$



Random words	Probability rules	Conditional probabilities	Bayes' Theorem	Conclusion O
Combining	overte			

 (A^{c})

Consider selecting word from poem and following two events:

- A = food item
- B = contains letter "w",

$$A \cap B = \{\text{weed}\}$$

$$A \cup B = \{\text{carrot}, \dots, \text{pear, we, with, swell}\}$$

$$A^c = \{\text{heart, nose, } \dots, \text{ if, swell}\}$$

$$B^c = S_{\text{poem}} - \{\text{weed, we, with, swell}\}$$

$$A^c \cup B^c = S_{\text{poem}} - \{\text{weed}\}$$

$$\cup B^c)^c =$$



Random words	Probability rules	Conditional probabilities	Bayes' Theorem	Conclusion O
O				

()

Consider selecting word from poem and following two events:

- A = food item
- B = contains letter "w",

then we can combine the events as follows:

$$A \cap B = \{ weed \}$$

$$A \cup B = \{ carrot, \dots, pear, we, with, swell \}$$

$$A^{c} = \{ heart, nose, \dots, if, swell \}$$

$$B^{c} = S_{poem} - \{ weed, we, with, swell \}$$

$$A^{c} \cup B^{c} = S_{poem} - \{ weed \}$$

$$A^{c} \cup B^{c} = \{ weed \}$$

A general rule helpful in calculating probabilities:

$$(A \cap B) = (A^c \cup B^c)^c$$



Random words	Probability rules ○○○○○●○○○○○○○	Conditional probabilities	Bayes' Theorem	Conclusion O
Compleme	nt: "not"			

Probability that event does not happen:

$$P(E^c)=1-P(E).$$

For example, let E = word is **not** food item, then

$$P(E) = 1 - P(E^c)$$



Random words	Probability rules ○○○○○●○○○○○○○	Conditional probabilities	Bayes' Theorem	Conclusion O
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= 1 - P({word is food item})



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Probability that event does not happen:

```
P(E^c)=1-P(E).
```

For example, let E = word is **not** food item, then

$$P(E) = 1 - P(E^{c})$$

= 1 - P({word is food item})
= 1 - 9/28
= 19/28

Here, gains of "switching to complement" are not very high.

Complements are often good strategy when confronted with

- "at most x" questions, where x is high,
- "at least y" questions, where y is low.





The interaction operator is typically described as "and": $A \cap B$ means "both A, and B".

Example: Poem. Probability of selecting food item with a "w"?

$$A = \{ food item \}$$

 $B = \{ contains a "w" \}$

 $P(A \cap B) =$





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 $P(A \cap B) = P(weed)$





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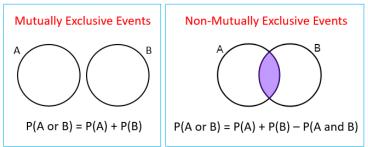
$$P(A \cap B) = P(weed)$$
$$= \frac{1}{28}$$



Random words	Probability rules	Conditional probabilities	Bayes' Theorem	Conclusion O
Unions: "o	r"			

Union operator is non-exclusive "or":

- $A \cup B$ means "or A, or B, or both".
- This corresponds to area contained in both circles:





Random words	Probability rules ○○○○○○○●○○○○	Conditional probabilities	Bayes' Theorem	Conclusion O
Example o	f union: "or"			

For example, consider again:

$$A = \{\text{food item}\}$$

 $B = \{\text{contains a "w"}\}$

so

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$
$$=$$



Random words	Probability rules ○○○○○○○●○○○○	Conditional probabilities	Bayes' Theorem	Conclusion O
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For example, consider again:

$$A = \{food item\}$$
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SO

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= 9/28 + 4/28 - 1/28

=



Random words	Probability rules ○○○○○○○●○○○○	Conditional probabilities	Bayes' Theorem	Conclusion O
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For example, consider again:

$$A = \{food item\}$$
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$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

= 9/28 + 4/28 - 1/28
= 12/28



Random words	Probability rules	Conditional probabilities	Bayes' Theorem	Conclusion O

Consider words you and your neighbour selected from poem.

What is sample space?

Spoem2 =



Consider words **you and your neighbour** selected from poem. What is **sample space**?

 $S_{\text{poem2}} = \{(\text{carrot}, \text{carrot}), \dots, (\text{carrot}, \text{pear}), \dots, (\text{pear}, \text{pear})\}$



Consider words **you and your neighbour** selected from poem. What is **sample space**?

 $S_{poem2} = \{(carrot, carrot), \dots (carrot, pear), \dots, (pear, pear)\}$

= {28 × 28 word combinations}

Let's consider the event

$$E = \{\text{both of you choose food items}\}$$

Note,

$$|E| =$$



Consider words **you and your neighbour** selected from poem. What is **sample space**?

 $S_{poem2} = \{(carrot, carrot), \dots (carrot, pear), \dots, (pear, pear)\}\$ = $\{28 \times 28 \text{ word combinations}\}$

Let's consider the event

$$E = \{\text{both of you choose food items}\}$$

Note,

$$|\boldsymbol{E}|=\boldsymbol{9}\times\boldsymbol{9}.$$

If you didn't cheat, then

$$P(E) =$$



Consider words **you and your neighbour** selected from poem. What is **sample space**?

 $S_{poem2} = \{(carrot, carrot), \dots (carrot, pear), \dots, (pear, pear)\}\$ = $\{28 \times 28 \text{ word combinations}\}$

Let's consider the event

$$E = \{\text{both of you choose food items}\}$$

Note,

$$|\boldsymbol{E}|=\boldsymbol{9}\times\boldsymbol{9}.$$

If you didn't cheat, then

$$P(E) = \frac{9 \times 9}{28 \times 28} = .10$$



Random words	Probability rules	Conditional probabilities	Bayes' Theorem	Conclusion O
Independe	anco			

There is something special about previous example:

Your word does not affect your neighbour's word So, events

- $A = \{$ your word is food item $\}$
- $B = \{$ your neighbour's word is food item $\}$

are so-called independent events.

In case of independent events, we can use

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

Example. 2 words from poem

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



Random words	Probability rules ○○○○○○○○○●○○	Conditional probabilities	Bayes' Theorem	Conclusion O
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In case of independent events, we can use

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Example. 2 words from poem

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

= $\frac{9}{28} \times \frac{9}{28} = 0.10$



Random words

Probability rules

Conditional probabilities

Bayes' Theorem

Conclusion

Have you magically chosen the same word?

Question:

How likely is it that at least two of you selected same word?

Choose from:

10%, 40%, 80%, 95%?



Random words	Probability rules ○○○○○○○○○○○●	Conditional probabilities	Bayes' Theorem	Conclusion O
Magic?				

• E = at least two words match



Random words	Probability rules ○○○○○○○○○○○●	Conditional probabilities	Bayes' Theorem	Conclusion O
Magic?				

- E = at least two words match
- E^c = no words match



Random words	Probability rules	Conditional probabilities	Bayes' Theorem	Conclusion O
Magic?				

- E = at least two words match
- *E^c* = no words match
- A_{ij} = words of person i and j do not match
- Note that we can write E^c in terms of A_{ij}:



Random words	Probability rules ○○○○○○○○○○●	Conditional probabilities	Bayes' Theorem	Conclusion ○
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- E = at least two words match
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$$E^{c} = \cap_{i,j} A_{ij}.$$

$$P(E) =$$



Random words	Probability rules	Conditional probabilities	Bayes' Theorem	Conclusion O
Magic?				

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$$P(E) = 1 - P(E^2)$$



Random words	Probability rules ○○○○○○○○○○●	Conditional probabilities	Bayes' Theorem	Conclusion O
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- E = at least two words match
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$$P(E) = 1 - P(E^2)$$

= 1 - P(\cap_{i,j}A_{ij})



Random words	Probability rules	Conditional probabilities	Bayes' Theorem	Conclusion O
Magic?				

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- Note that we can write E^c in terms of A_{ij}:

$$E^c = \cap_{i,j} A_{ij}.$$

$$P(E) = 1 - P(E^{2}) \\ = 1 - P(\cap_{i,j}A_{ij}) \\ = 1 - \prod_{i,j} P(A_{ij})$$



Random words	Probability rules ○○○○○○○○○○●	Conditional probabilities	Bayes' Theorem	Conclusion O
Magic?				

- E = at least two words match
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Ρ

- A_{ij} = words of person i and j do not match
- Note that we can write E^c in terms of A_{ij}:

$$E^{c} = \cap_{i,j} A_{ij}.$$

$$P(E) = 1 - P(E^{2})$$

= 1 - P(\cap_{i,j}A_{ij})
= 1 - \prod_{i,j}P(A_{ij})
= 1 - \left(1 - \frac{1}{28}\right)^{\begin{subarray}{c} 13 \\ 2 \end{subarray}} = 0.94



Random words

Probability rules

Conditional probabilities

Bayes' Theorem

Conclusion

Conditional Probabilities



Random words	Probability rules	Conditional probabilities	Bayes' Theorem	Conclusion ○

Dependence

Independence is great, because

- we can focus on smaller sample space
- which makes calculations easier

However, often events are not independent.

Example: Poem. Probability of selecting food item with a "w"?

$$A = \{\text{food item}\}$$

 $B = \{\text{contains a "w"}\}$

$$0.04 = \frac{1}{28} = P(A \cap B) \neq P(A)P(B) = \frac{9}{28}\frac{4}{28} = 0.05$$



How can we do simple calculations with dependent events?

Random words	Probability rules	Conditional probabilities	Bayes' Theorem	Conclusion O

Definition of conditional probability

Example. Draw 2 cards from deck without replacement.

- E = 1st card is ace
- F = 2nd card is ace

$P(E \cap F) =$



Random words	Probability rules	Conditional probabilities	Bayes' Theorem	Conclusion O

Definition of conditional probability

Example. Draw 2 cards from deck without replacement.

- E = 1st card is ace
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$$P(E \cap F) = \frac{4}{52} \frac{3}{51}$$



Random words	Probability rules	Conditional probabilities ○○●○○	Bayes' Theorem	Conclusion O

Definition of conditional probability

Example. Draw 2 cards from deck without replacement.

- E = 1st card is ace
- F = 2nd card is ace

$$P(E \cap F) = \frac{4}{52} \frac{3}{51}$$
$$= P(E)P(F|E)$$

where P(F|E) is the probability of F given E.

Definition. Conditional probability of *A* if *B* happened: $P(A \mid B) = \frac{P(A \cap B)}{P(B)}.$

Random words	Probability rules	Conditional probabilities	Bayes' Theorem	Conclusion O
Using cond	ditional probabi	ilities		

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Random words	Probability rules	Conditional probabilities ○○○●○	Bayes' Theorem	Conclusion O
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$$P(A \cap B) = P(A)P(B|A)$$

=



Random words	Probability rules	Conditional probabilities	Bayes' Theorem	Conclusion O
Using cond	ditional probabi	ilities		

$$A = \{\text{food item}\}$$

$$B = \{\text{contains a "w"}\}$$

$$P(A \cap B) = P(A)P(B|A)$$
$$= \frac{9}{28}\frac{1}{9}$$



Random words	Probability rules	Conditional probabilities	Bayes' Theorem	Conclusion O
Using cond	ditional probabi	ilities		

$$A = \{\text{food item}\}$$

$$B = \{\text{contains a "w"}\}$$

$$P(A \cap B) = P(A)P(B|A)$$
$$= \frac{9}{28}\frac{1}{9}$$
$$= \frac{1}{28}$$



Random words	Probability rules	Conditional probabilities ○○○○●	Bayes' Theorem	Conclusion O

Multiplication rule

For three events A, B, C (not necessarily independent),

$$P(A \cap B \cap C) = P(A) \times P(B \mid A) \times P(C \mid A \cap B)$$

Example. Consider taking a 3 cards from a pack of cards. What is the probability that they are all aces?

$$A_i = Ace$$
 in *i*th draw

$$P(A_1 \cap A_2 \cap A_3) =$$



Random words	Probability rules	Conditional probabilities ○○○○●	Bayes' Theorem	Conclusion O

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Random words	Probability rules	Conditional probabilities ○○○○●	Bayes' Theorem	Conclusion O

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$$P(A_{1} \cap A_{2} \cap A_{3}) = P(A_{1}) \times P(A_{2} | A_{1}) \times P(A_{3} | A_{1} \cap A_{2}) \\ = \frac{4}{52} \times$$

Random words	Probability rules	Conditional probabilities ○○○○●	Bayes' Theorem	Conclusion O

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 = Ace in *i*th draw

$$P(A_1 \cap A_2 \cap A_3) = P(A_1) \times P(A_2 \mid A_1) \times P(A_3 \mid A_1 \cap A_2) \\ = \frac{4}{52} \times \frac{3}{51} \times$$

Random words	Probability rules	Conditional probabilities ○○○○●	Bayes' Theorem	Conclusion O

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For three events A, B, C (not necessarily independent),

$$P(A \cap B \cap C) = P(A) \times P(B \mid A) \times P(C \mid A \cap B)$$

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$$A_i$$
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$$P(A_{1} \cap A_{2} \cap A_{3}) = P(A_{1}) \times P(A_{2} | A_{1}) \times P(A_{3} | A_{1} \cap A_{2})$$

= $\frac{4}{52} \times \frac{3}{51} \times \frac{2}{50}$

Random words	Probability rules	Conditional probabilities ○○○○●	Bayes' Theorem	Conclusion O

Multiplication rule

For three events A, B, C (not necessarily independent),

$$P(A \cap B \cap C) = P(A) \times P(B \mid A) \times P(C \mid A \cap B)$$

Example. Consider taking a 3 cards from a pack of cards. What is the probability that they are all aces?

$$A_i = Ace$$
 in *i*th draw

$$P(A_1 \cap A_2 \cap A_3) = P(A_1) \times P(A_2 \mid A_1) \times P(A_3 \mid A_1 \cap A_2) \\ = \frac{4}{52} \times \frac{3}{51} \times \frac{2}{50} \\ = 0.00018$$

Random words

Probability rules

Conditional probabilities

Bayes' Theorem

Conclusion

Bayes' Theorem



Probability rules

Conditional probabilities

Bayes' Theorem

Conclusion

Who was Reverend Bayes?



Reverend Thomas Bayes (c.1702 - 17 April 1761)

- English mathematician and Presbyterian minister
- 1743: Elected Fellow of Royal Society.
- 1761: Thomas Bayes dies
- 1763: Essay Towards Solving a Problem in the Doctrine of Chances read before Royal Society of London
- 20th century: famous for solving problem of "inverse probability

Random words	Probability rules	Conditional probabilities	Bayes' Theorem ○○●○○○○	Conclusion O
Bayes' The	orem			

Assume we get data *D* from the true state S_0 of reality:

$$D = \{ data \}$$

 $S = \{ state of the world. \}$

Question. Given data *D* what is our belief in $S \in \{S_0, S_1, \dots, S_n\}$?

Note that typically $P(D | S_i)$ is easy.

Bayes' Theorem

$$P(S \mid D) = \frac{P(D \mid S)P(S)}{\sum_{i=0}^{n} P(D \mid S_i)P(S_i)}$$

Probabilities $P(S_i)$ need to be assumed known *a priori*.

God's not playing dice, but flipping coins...

Imagine that at beginning of time, God flips a fair coin:

- If heads, then God creates two universes: one with black-haired people, other with blond haired people.
- If tails, then God creates one black-haired universe.

Now suppose that you are living in black-haired universe.

Then what is probability of God's coin having landed heads?



Random words	Probability rules	Conditional probabilities	Bayes' Theorem ○○○○●○○	Conclusion O
God's flipp	ing coins			

- $E = \{$ Living in a black-haired universe. $\}$
- $F = \{\text{Heads}\}$

P(F|E) =





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$$= \frac{0.5 \times 0.5}{0.5 \times 0.5 + 1}$$





God's flipping coins...

- $E = \{$ Living in a black-haired universe. $\}$
- $F = \{\text{Heads}\}$

$$P(F|E) = \frac{P(FE)}{P(E)} \\ = \frac{P(E|F)P(F)}{P(E|F)P(F) + P(E|F^c)P(F^c)} \\ = \frac{0.5 \times 0.5}{0.5 \times 0.5 + 1 \times 0.5} \\ = 1/3$$



Random words	Probability rules	Conditional probabilities	Bayes' Theorem ○○○○○●○	Conclusion ○
Continuent	enelucia			

Sentiment analysis

We get a piece of text (e.g. tweet) and we want to know:

Does it express a positive or negative sentiment?

• Let's consider following dictionary:

 $C = \{$ of, great, kind, weird, stuff, mean $\}$

- Two two sentiments: $S \in \{\text{positive, negative}\}$
- Conditional probabilities *P*(*word* | *sentiment*) are:

word	positive	negative
of	0.1	0.1
great	0.3	0.1
kind	0.3	0.1
weird	0.1	0.3
stuff	0.1	0.2
mean	0.1	0.2



Random words	Probability rules	Conditional probabilities	Bayes' Theorem ○○○○○○●	Conclusion O
Tweet: "we	eird kind of stuf	f "		

Define our events:

- *W_i* = *i*th word (*i* = 1, 2, 3, 4)
- N = negative sentiment



Random words	Probability rules	Conditional probabilities	Bayes' Theorem ○○○○○○●	Conclusion O
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Define our events:

- *W_i* = *i*th word (*i* = 1, 2, 3, 4)
- N = negative sentiment

$$P(N \mid W_1 \cap \ldots \cap W_4) = \frac{P(W_1 \cap \ldots \cap W_4 \mid N)P(N)}{P(W_1 \cap \ldots \cap W_4 \mid N)P(N) + P(W_1 \cap \ldots \cap W_4 \mid N^c)P(N^c)}$$



Random words	Probability rules	Conditional probabilities	Bayes' Theorem ○○○○○○●	Conclusion O
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= 0.3×



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= 0.3×0.1×0.1×0.2×



Random words	Probability rules	Conditional probabilities	Bayes' Theorem ○○○○○○●	Conclusion O
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$$P(N \mid W_1 \cap \ldots \cap W_4) = \frac{P(W_1 \cap \ldots \cap W_4 \mid N)P(N)}{P(W_1 \cap \ldots \cap W_4 \mid N)P(N) + P(W_1 \cap \ldots \cap W_4 \mid N^c)P(N^c)} \\ = \frac{0.3 \times 0.1 \times 0.1 \times 0.2 \times 0.5}{0.3 \times 0.1 \times 0.1 \times 0.2 \times 0.5 + 0.1 \times 0.3 \times 0.1 \times 0.1 \times 0.5}$$



Random words	Probability rules	Conditional probabilities	Bayes' Theorem ○○○○○○●	Conclusion O
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- *W_i* = *i*th word (*i* = 1, 2, 3, 4)
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$$P(N \mid W_1 \cap \ldots \cap W_4) = \frac{P(W_1 \cap \ldots \cap W_4 \mid N)P(N)}{P(W_1 \cap \ldots \cap W_4 \mid N)P(N) + P(W_1 \cap \ldots \cap W_4 \mid N^c)P(N^c)} \\ = \frac{0.3 \times 0.1 \times 0.1 \times 0.2 \times 0.5}{0.3 \times 0.1 \times 0.1 \times 0.2 \times 0.5 + 0.1 \times 0.3 \times 0.1 \times 0.1 \times 0.5} \\ = 0.67$$



Random words	Probability rules	Conditional probabilities	Bayes' Theorem ○○○○○○●	Conclusion O
Tweet: "we	aird kind of stuf	f"		

Define our events:

- *W_i* = *i*th word (*i* = 1, 2, 3, 4)
- N = negative sentiment

Bayes' Theorem! Let prior probability P(N) = 0.5:

$$P(N \mid W_1 \cap \ldots \cap W_4) = \frac{P(W_1 \cap \ldots \cap W_4 \mid N)P(N)}{P(W_1 \cap \ldots \cap W_4 \mid N)P(N) + P(W_1 \cap \ldots \cap W_4 \mid N^c)P(N^c)} \\ = \frac{0.3 \times 0.1 \times 0.1 \times 0.2 \times 0.5}{0.3 \times 0.1 \times 0.1 \times 0.2 \times 0.5 + 0.1 \times 0.3 \times 0.1 \times 0.1 \times 0.5} \\ = 0.67$$

We have *secretly* made use of conditional independence! We relax this assumption in the afternoon.



Random words	Probability rules	Conditional probabilities	Bayes' Theorem	Conclusion •

Conclusion

In this class we have learned:

- Laplace's definition of probability
- Rules for combining event and probabilities
- Independence simplifies calculations.
- Conditional probabilities are also easy.
- Bayes' Theorem to learn about reality from data.

