Probability and Statistics for Computer Science



"A major use of probability in statistical inference is the updating of probabilities when certain events are observed" – Prof. M.H. DeGroot

Credit: wikipedia

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Counting: how many ways?

if we put 7 hats (indistinguishable) on 7 people out of 10 people randomly?

Warm up: which is larger?

P(ANB) or P(A|B) P(B) < 1A) P(AnB) B) P(AIB) c) None $P(A(B) = P(A \cap B)$ P(A)

Last time

More probability calculation using counting Conditional probability Mutiplication rule;

Objectives

%Conditional Probability



- # Bayes rule
- % Total probability
- % Independence

Conditional Probability

* The probability of A given B



$$\begin{vmatrix} P(A|B) = \frac{P(A \cap B)}{P(B)} \\ P(B) \neq 0 \end{vmatrix}$$

The line-crossed area is the new sample space for conditional P(A|B)

Joint Probability Calculation

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



Joint Probability Calculation

$\Rightarrow P(A \cap B) = P(A|B)\underline{P(B)}\text{prior}$



 $p(A^{c}|B) = ? (-P(A|B))$ $P(A^{c} \cap P) = ?$ (- PLAND) A: meat B: soup D: juice $P(soup \cap meat) =$ P(meat|soup)P(soup) $= 0.5 \times 0.8 = 0.4$

Joint Probability Calculation



Symmetry of joint event in terms of conditional prob.

$$P(A|B) = \frac{P(A \cap B)}{P(B)} \quad P(B) \neq 0$$

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

Symmetry of joint event in terms of conditional prob.

$\mathcal{B} \cap A = A \cap B$ $P(B \cap A) = P(A \cap B)$



P(A|B)P(B) = P(B|A)P(A)

The famous **Bayes** rule

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

The famous **Bayes** rule

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

Thomas Bayes (1701-1761)

Bayes rule: lemon cars

There are two car factories, **A** and **B**, that supply the same dealer. Factory A produced 1000 cars, of which 10 were lemons. Factory B produced 2 cars and both were lemons. You bought a car that turned out to be a lemon. What is the probability that it came from factory B? it cane F Ix: Fec B P(En, En, En). P(G) $P(E_{X}|E_{N}) = ? =$

Bayes rule: lemon cars

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 $P(B|L) = \frac{P(L|B)P(B)}{P(L)} = \frac{I \times I}{I}$

Bayes rule: lemon cars

Given the above information, what is the probability that it came from factory A?

P(A|L) = ?

Total probability:

to workows 80% Sunny to work out 30% Rainy H Summ/ KRainny = 3 P(Rainy Nowt) P(Sun Dut) P(work out) = ? 75%×80% + 45%×30%



 $A^{c} \rightarrow rainy$ P(B|A) = $P(B|A^{c}) =$ P(A) = P(A) = P(A) =

$P(B) = P(A \cap B) + P(A^{c} \cap B)$ = P(B|A)P(A)+P(B|A^{c}) · P(A^{c})

Total probability:

Total probability

$P(B) = P(A_1 \cap B) + P(A_2 \cap B) + P(A_3 \cap B)$ = P(B|A_1)P(A_1) + P(B|A_2)P(A_2) + P(B|A_3)P(A_3)



Total probability general form

 $P(B) = \sum_{i} P(B|A_{i})$ = $\sum_{i} P(B|A_{i}) P(A_{i})$



A; AR are d:sjoint A; $A_k = \phi$ $j \neq k$

Total probability:



P(meat) = ?0.8 x 0.5 + 0.2 x 0.3 p (soup | meat) = ? P(soup n meat) Y (meat) 0.8×0.5 0.8×0.5+0.2×0.3



Bayes rule using total prob.

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

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

$$\sum_{i} P(B|A;)P(A;)$$

$$A; nA; = \phi \rightarrow d; joint$$

$$if i \neq j$$

Bayes rule: rare disease test

There is a blood test for a rare disease. The frequency of the disease is 1/100,000. If one has it, the test confirms it with probability 0.95. If one doesn't have, the test gives false positive with probability 0.001. What is P(D|T), the probability of having disease given a positive test result?

$$\begin{split} P(D|T) &= \frac{P(T|D)P(D)}{P(T)} & \text{Using total prob.} \\ &= \frac{P(T|D)P(D)}{P(T|D)P(D)} + P(T|D^c)P(D^c) \end{split}$$

Bayes rule: rare disease test

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 $0.95 \times \frac{1}{10^5} + 0.001 (1-7(0))$ $P(T|0^{-1}=0.001)$

Independence

%One definition:

P(A|B) = P(A) orP(B|A) = P(B)

Whether A happened doesn't change the probability of B and vice versa

Independence: example

* Suppose that we have a fair coin and it is tossed twice. let A be the event "the first toss is a head" and B the event "the two outcomes are the same." A: HX B: HH TT $P(A(B) = P(A) P(A(B)) = \frac{P(AB) P(AH)}{P(B) P(AH)}$ $P(A) = \frac{1}{2}$ P(A|B) = P(A)* These two events are independent!

Independence

*****Alternative definition



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

Testing Independence:

Suppose you draw one card from a standard deck of cards. E₁ is the event that the card is a King, Queen or Jack. E₂ is the event the card is a Heart. Are E_1 and 404 E₂ independent? P(E(nEv) = $P(E_1) =$ のしだい)

Pairwise independence is not mutual independence in larger context



$$A = A_1 \cup A_2; P(A)$$

$$B = A_1 \cup A_3; P(B)$$

 $C = A_1 \cup A_4; P(C)$

*P(ABC) is the shorthand for $P(A \cap B \cap C)$

$$P(A_1) = P(A_2) = P(A_3) = P(A_4) = 1/4$$

$$P(Ang) = P(A_i) = \frac{1}{4} < \frac{P(A)P(a)}{1 + \frac{1}{4}}$$

$$P(Anc) = P(A_i) = \frac{1}{4}$$

$$P(Gnc) = P(A_i) = \frac{1}{4}$$

$$(A nGnc) = P(A_i) = \frac{1}{4}$$

$$\frac{1}{4} P(A)P(B)P(c)$$

$$= \frac{1}{4}$$

Mutual independence

* Mutual independence of a collection of events $A_1, A_2, A_3...A_n$ is :

$$P(A_j | A_j A_k \dots A_p) = P(A_i)$$

$$j, k, \dots p \neq i$$

% It's very strong independence!

Probability using the property of Independence: Airline overbooking (1)

* An airline has a flight with 6 seats. They always sell 7 tickets for this flight. If ticket holders show up independently with probability, p, what is the probability that the flight is overbooked ?

7 showed up

$$P(A, nA unas -.. nA n) = p^{7}$$

= $P(A, nA unas -.. nA n) = p^{7}$

Probability using the property of Independence: Airline overbooking (2)

* An airline has a flight with 6 seats. They always sell 8 tickets for this flight. If ticket holders show up independently with probability p, what is the probability that exactly 6 people showed up?

 $P(6 \text{ people showed up}) = \begin{pmatrix} 8 \\ 6 \end{pmatrix} P \begin{pmatrix} -p \end{pmatrix}^{2} \\ \begin{pmatrix} 6 \\ 6 \end{pmatrix} P \begin{pmatrix} -p \end{pmatrix}^{2} \\ \begin{pmatrix} 6 \\ 6 \end{pmatrix} P \begin{pmatrix} -p \end{pmatrix}^{2} \\ \begin{pmatrix} 6 \\ 6 \end{pmatrix} P \begin{pmatrix} -p \end{pmatrix}^{2} \\ \begin{pmatrix} 6 \\ 6 \end{pmatrix} P \begin{pmatrix} -p \end{pmatrix}^{2} \\ \begin{pmatrix} 6 \\ 6 \end{pmatrix} P \begin{pmatrix} -p \end{pmatrix}^{2} \\ \begin{pmatrix} 6 \\ 6 \end{pmatrix} P \begin{pmatrix} -p \end{pmatrix}^{2} \\ \begin{pmatrix} 6 \\ 6 \end{pmatrix} P \begin{pmatrix} -p \end{pmatrix}^{2} \\ \begin{pmatrix} 6 \\ 6 \end{pmatrix} P \begin{pmatrix} -p \end{pmatrix}^{2} \\ \begin{pmatrix} 6 \\ 6 \end{pmatrix} P \begin{pmatrix} -p \end{pmatrix}^{2} \\ \begin{pmatrix} 6 \\ 6 \end{pmatrix} P \begin{pmatrix} -p \end{pmatrix}^{2} \\ \begin{pmatrix} 6 \\ 6 \end{pmatrix} P \begin{pmatrix} -p \end{pmatrix}^{2} \\ \begin{pmatrix} 6 \\ 6 \end{pmatrix} P \begin{pmatrix} -p \end{pmatrix}^{2} \\ \begin{pmatrix} 6 \\ 6 \end{pmatrix} P \begin{pmatrix} -p \end{pmatrix}^{2} \\ \begin{pmatrix} 6 \\ 6 \end{pmatrix} P \begin{pmatrix} -p \\ 6 \end{pmatrix} P \begin{pmatrix} -$

Probability using the property of Independence: Airline overbooking (3)

** An airline has a flight with 6 seats. They always sell 8 tickets for this flight. If ticket holders show up independently with probability p, what is the probability that the flight is overbooked ?

P(overbooked) =

$$\frac{8}{\Sigma}$$
 $\binom{8}{u}$ $p^{u}(-p)^{8-u}$
u=7

Assignments

Module week3 on Canvas

* Next time: Random variable

Additional References

- * Charles M. Grinstead and J. Laurie Snell "Introduction to Probability"
- Morris H. Degroot and Mark J. Schervish "Probability and Statistics"

See you next time

See You!

