

CS 2810 Day 23 April 15

Admin:

10 mins for TRACE

Quiz3_02: problem 2.iv & 3.iv

Bayes Rule

- binary problems
- parametric likelihoods
- bayes rule & independence

TRACE

TRACE feedback helps me be a better teacher for you all.

TRACE feedback helps NU identify strong / weak teachers.

Please take a few minutes to give feedback about what worked and what didn't.

To zoom students just joining, we'll get started in 6-7 minutes. Please take this time to fill out the trace survey on the course. Thank you!

CONDITIONAL PROB

(INTUITION: "ZOOM INTO CONTEXT B")
SEE LAST ICA DAY 22

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

PROB A HAPPENS
GIVEN CONDITION B

PROB A, B HAPPEN
TOGETHER

PROB B HAPPENS

CONDITIONAL PROB (USEFUL ALGEBRAIC MANIPULATION)

PROB A HAPPENS
GIVEN CONDITION B

PROB A, B HAPPEN
TOGETHER

PROB B HAPPENS

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

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

Takeaway: Multiplying a conditional probability by the probability of condition yields a "full joint" probability of all variables happening together. (we'll see it works for more than just two vars A, B too!)

BAYES RULE

(GLORIFIED CONDITIONAL PROBABILITY)

SEE PREVIOUS "TAKEAWAY"

$$P(A|B)P(B) = P(AB) = P(B|A)P(A)$$

\Rightarrow

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

Notice: this formula "swaps" the order of the conditioning: $P(A|B)$ on left $P(B|A)$ on right
Its typical in a Bayes question to be given variables in one order while question asks for other.

MARGINALIZING (REMEMBER THIS?)

B=1 SHAPE IS BLUE
C=1 SHAPE IS CIRCLE



	B=0	B=1
C=0	x_1 $\frac{1}{5}$	x_3 $\frac{1}{5}$
C=1	x_4, x_5 $\frac{2}{5}$	x_2 $\frac{1}{5}$

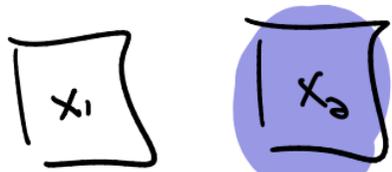
$$\begin{aligned} P(B=1) &= P(B=1, C=0) + P(B=1, C=1) \\ &= \frac{1}{5} + \frac{1}{5} \end{aligned}$$

Remember: To compute $P(B)$ we can sum $P(B, A)$ for all outcomes in sample space of A

$$P(B=b) = \sum_a P(B=b, A=a)$$

MARGINALIZING (WITH A CONDITIONAL)

$B=1$ SHAPE IS BLUE
 $C=1$ SHAPE IS CIRCLE



$$P(C=0) = 2/5$$

$$P(B=1|C=0) = 1/2$$

$$P(B=0|C=0) = 1/2$$



$$P(C=1) = 3/5$$

$$P(B=1|C=1) = 1/3$$

$$P(B=0|C=1) = 2/3$$

$$P(B=1) = \sum_c P(B=1|C=c) P(C=c)$$
$$= P(B=1|C=1) P(C=1) + P(B=1|C=0) P(C=0)$$

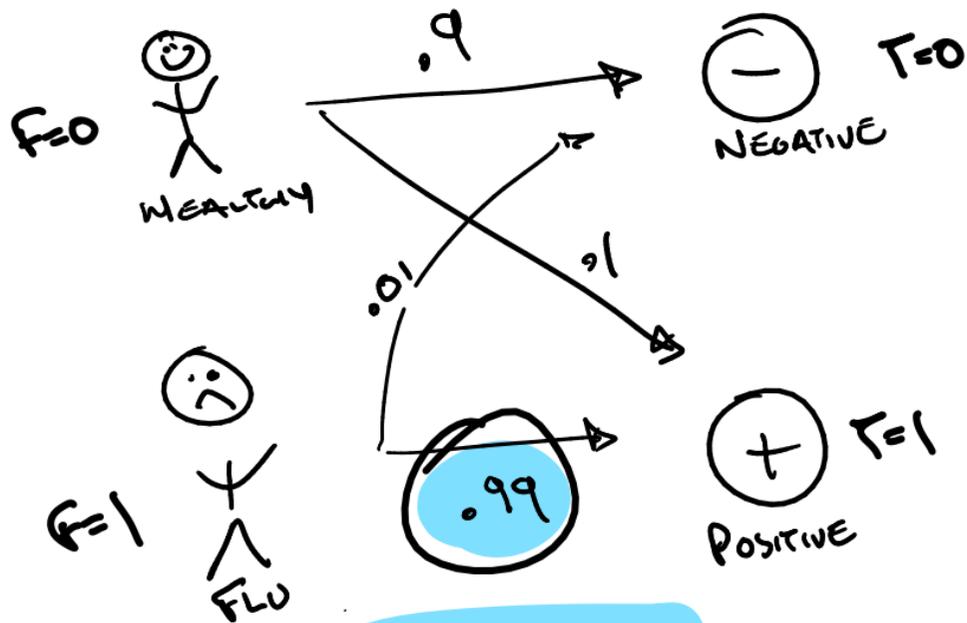
$$= P(B=1|C=1) P(C=1) +$$

$$P(B=1|C=0) P(C=0)$$

$$= 1/3 \cdot 3/5 + 1/2 \cdot 2/5$$

$$= 1/5 + 1/5 = 2/5$$

BAYES RULE Ex



$$P(T=1|F=1) = .99$$

Given flu occurs in .04 of population, what is the probability one has flu given they test positive?

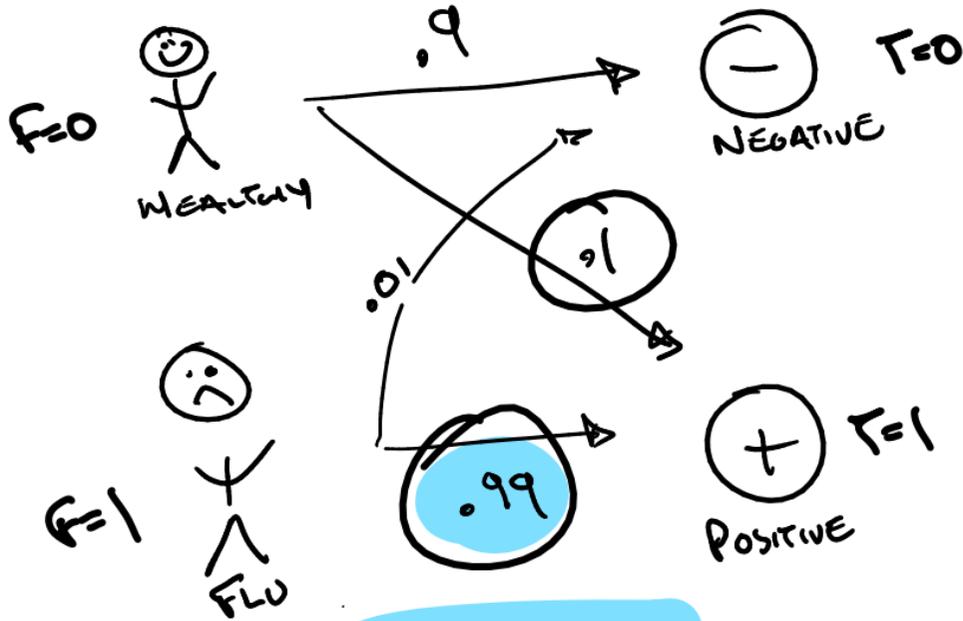
$$P(F=1) = .04$$

$$P(F=1|T=1) = \frac{P(T=1|F=1)P(F=1)}{P(T=1)}$$

$$= \frac{.99 \cdot .04}{.99 \cdot .04 + .1 \cdot .94}$$

BAYES RULE Ex

Given flu occurs in .04 of population, what is the probability one has flu given they test positive?



$$P(T=1|F=1) = .99$$

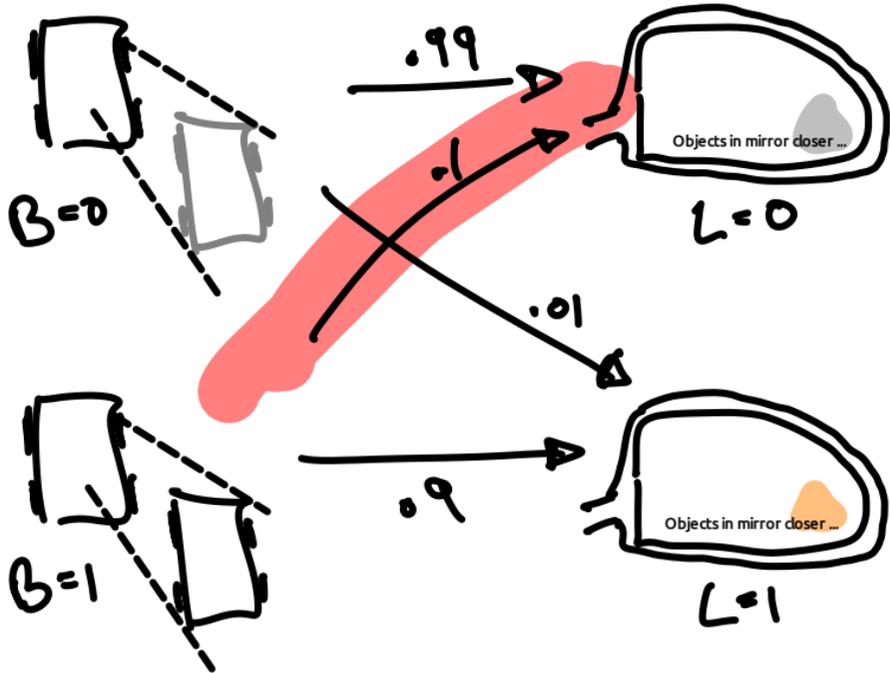
$$\begin{aligned} P(T=1) &= P(T=1|F=1)P(F=1) + \\ &P(T=1|F=0)P(F=0) \\ &= .99 \cdot .04 + .1 \cdot .94 \end{aligned}$$

In Class Assignment 1

L=1 LIGHT ON

B=1 CAR BLIND SPOT

A blind spot monitor produces a warning light (L=1) when it estimates that a car is in one's blind spot (B=1). Given that the light is off, what's the probability that a car is one's blind spot? (Assume that a car is in your blindspot .02 of the time while driving.)



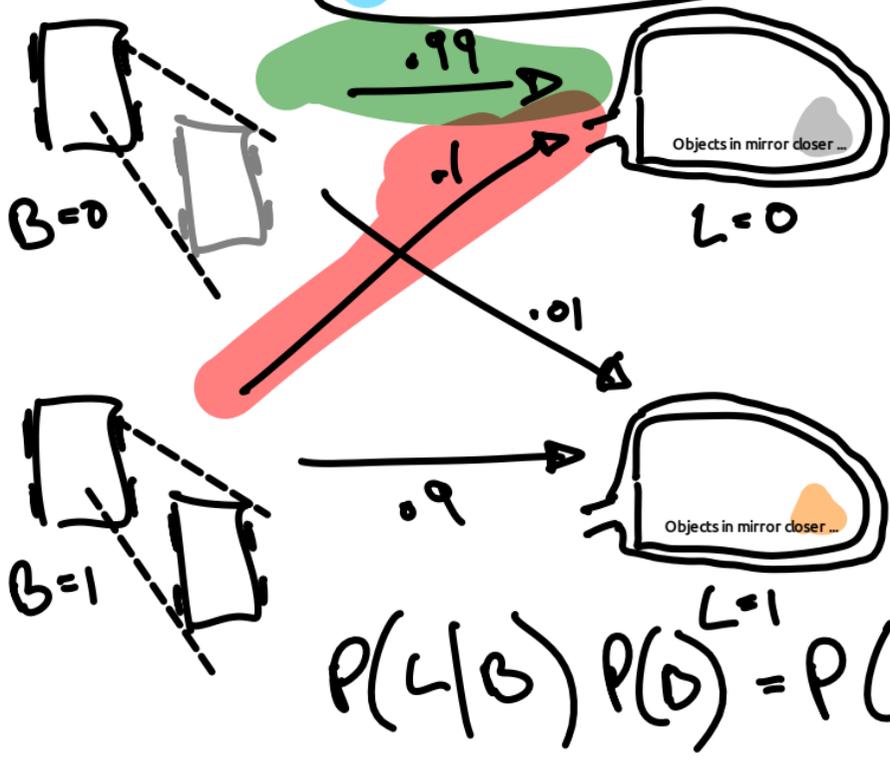
$$P(B=1 | L=0) = \frac{P(L=0 | B=1) P(B=1)}{P(L=0)}$$
$$= \frac{(.1)(.02)}{(.99)(.98) + (.1)(.02)}$$

In Class Assignment 1

L=1 LIGHT ON

B=1 CAR BLIND SPOT

A blind spot monitor produces a warning light (L=1) when it estimates that a car is in one's blind spot (B=1). Given that the light is off, what's the probability that a car is one's blind spot? (Assume that a car is in your blindspot .02 of the time while driving.)



$$\begin{aligned} P(L=0) &= P(L=0, B=0) + P(L=0, B=1) \\ &= P(L=0|B=0)P(B=0) + P(L=0|B=1)P(B=1) \\ &= (.99)(.98) + (.1)(.02) \end{aligned}$$

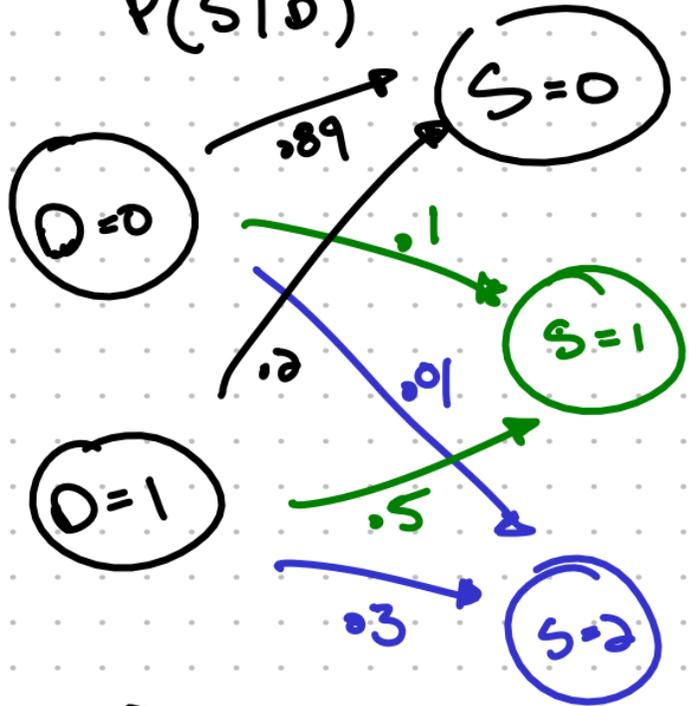
Making Bayes more useful (non-binary A, B variables):

Bayes is applicable in problems where each variable has more than 2 states too!

(see quick example on next page)

BAYES PRACTICE

$$P(S|D)$$



$$P(D=1) = 0.09$$

D=0 NO GOLD DEPOSIT

D=1 GOLD DEPOSIT

S=0 NO GOLD IN STREAM

S=1 LITTLE GOLD IN STREAM

S=2 MUCH GOLD IN STREAM

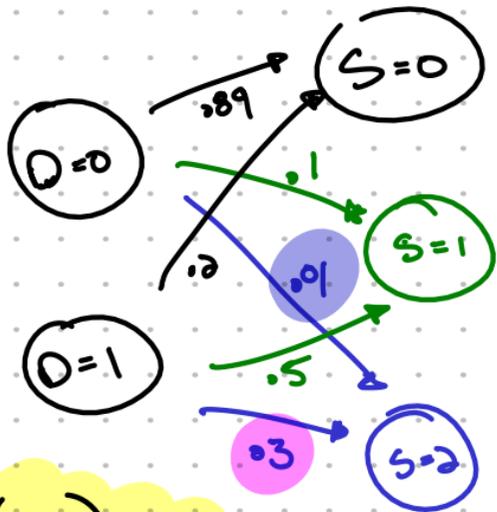
WHAT IS PROB OF
DEPOSIT GIVEN MUCH
GOLD IN STREAM?

$$P(D=1 | S=2) = \frac{P(S=2 | D=1) P(D=1)}{P(S=2)}$$

$$= \frac{P(S=2 | D=1) P(D=1)}{\sum_d P(S=2 | D=d) P(D=d)}$$

$$= \frac{P(S=2 | D=1) P(D=1)}{P(S=2 | D=0) P(D=0) + P(S=2 | D=1) P(D=1)}$$

$$= \frac{(.3 \cdot .09)}{(.01 \cdot .89 + .3 \cdot .09)} \approx .752$$



$$P(D=1) = .09$$

Making Bayes More Useful (parametric likelihoods):

Let's build models / problems where the conditional $P(B|A)$ is parametric

- binomial distribution**
 - poisson**
- 

If a stream is near a gold deposit, one typically finds a gold nugget after an hour of sifting.

$$P(X|D=1) = \text{Poisson}(\lambda=7)$$

If a stream is not near a gold deposit, one typically finds a gold nugget after a full day of sifting work (10 hours).

$$P(X|D=0) = \text{Poisson}(\lambda=0.7)$$

1% of streams are near gold deposits.

$$P(D=1) = 0.01$$

If we find 3 nuggets after 7 hours of sifting a particular stream, what's the probability that this stream is near a gold deposit?

$X = \#$ NUGGETS FOUND IN 7 HRS SIFTING

$D = 1$ IF STREAM NEAR DEPOSIT



If a stream is near a gold deposit, one typically finds a gold nugget after an hour of sifting.

$$P(X|D=1) = \text{Poisson}(\lambda=7)$$

If a stream is not near a gold deposit, one typically finds a gold nugget after a full day of sifting work (10 hours).

$$P(X|D=0) = \text{Poisson}(\lambda=0.7)$$

1% of streams are near gold deposits.

$$P(D=1) = 0.01$$

If we find 3 nuggets after 7 hours of sifting a particular stream, what's the probability that this stream is near a gold deposit?

$$P(D=1|X=3) = \frac{P(X=3|D=1)P(D=1)}{P(X=3)}$$



$$P(x=3) = P(x=3 | D=0) + P(x=3 | D=1)$$

$$= P(x=3 | D=0) P(D=0) + P(x=3 | D=1) P(D=1)$$

$$P(D=1 | x=3) = \frac{P(x=3 | D=1) P(D=1)}{P(x=3 | D=0) P(D=0) + P(x=3 | D=1) P(D=1)}$$

$$= \frac{P(x=3 | D=1) P(D=1)}{P(x=3 | D=0) P(D=0) + P(x=3 | D=1) P(D=1)}$$

$$\approx \frac{(.052)(.01)}{(.0283)(.99) + (.052)(.01)} \approx .0182$$

BAYES RULE TERMS HAVE NAMES

T TARGET VARIABLE OF INTEREST

(GOLD DEPOSIT NEAR)

E EVIDENCE

(# NUGGETS)

likelihood: probability of evidence
under each possible target outcome

"a priori" / prior
distribution of target variable
before observing any evidence

$$P(T|E) = \frac{P(E|T)P(T)}{P(E)}$$

The diagram shows the equation $P(T|E) = \frac{P(E|T)P(T)}{P(E)}$. The term $P(T|E)$ is highlighted in a red cloud. The term $P(E|T)$ is highlighted in a green cloud, with an arrow pointing to the definition of likelihood. The term $P(T)$ is highlighted in a purple cloud, with an arrow pointing to the definition of "a priori" / prior distribution. The denominator $P(E)$ is not highlighted.

"a posteriori" / posterior
the probability of target variable
after observing the evidence

If a stream is near a gold deposit, one typically finds a gold nugget after an hour of sifting.

If a stream is not near a gold deposit, one typically finds a gold nugget after a full day of sifting work (10 hours).



1% of streams are near gold deposits.

Prior

Likelihood

If we find 3 nuggets after 7 hours of sifting a particular stream, what's the probability that this stream is near a gold deposit?

POSTERIOR

$X = \# \text{ NUGGETS} \mid 7 \text{ HRS EVIDENCE}$
 $G = 1 \text{ DEPOSIT TARGET}$

In Class Assignment

$$P(N|C=0) = \text{BINOM}(n=5, p=.05)$$

In a typical box of chocolates, only 5% of chocolates are coconut flavored.

In a "coconut special", 50% of the chocolates are coconut flavored.

If one selects 5 chocolates out of a box and observes that 3 are coconut flavored, what's the probability that the box is a "coconut special" box?

$$P(N|C=1) = \text{BINOM}(n=5, p=.5)$$

Assume that coconut special boxes are as common as typical chocolate boxes.

$$P(C=1|N=3)$$

$C=1$ BOX OF CHOCOLATE IS COCONUT SPECIAL
 0 IT'S NOT

$N = \#$ OF COCONUT CHOCOLATES IN SAMPLE OF 5 CHOCOLATES FROM BOX



$$P(c=1|N=3) = \frac{P(N=3|c=1)P(c=1)}{P(N=3)}$$

$$\begin{aligned} P(N=3) &= P(N=3, c=0) + P(N=3, c=1) \\ &= P(N=3|c=0)P(c=0) + P(N=3|c=1)P(c=1) \end{aligned}$$

INDEPENDENCE + CONDITIONAL PROB

INDEPENDENCE

INTUITION:

Random variables x, y are independent if observing any outcome of one doesn't impact our beliefs about the other.

ALGEBRA:

FOR EACH OUTCOME PAIR x, y

$$P(X=x \ Y=y) = P(X=x) P(Y=y)$$

Bayes Rule shows the equivalence of the algebraic and intuitive definitions above!

INDEPENDENCE + CONDITIONAL PROBS

INDEPENDENCE

INTUITION:

Random variables x, y are independent if observing any outcome of one doesn't impact our beliefs about the other.

ALGEBRA:

FOR EACH OUTCOME PAIR x, y

$$P(X=x, Y=y) = P(X=x)P(Y=y)$$

$$P(X|Y) = \frac{P(X, Y)}{P(Y)} = \frac{P(X)P(Y)}{P(Y)} = P(X)$$

Notice that $P(X|Y) = P(X)$. Observing Y has no impact on the prob of X !