

YZV 231E

18.10.2021

Probability Theory & Stats

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Recap: Conditional Probability: $P(A|B) = \frac{P(A \cap B)}{P(B)}$

ex: Consider a random experiment:
measuring heights & weights of students
in a class

$$A \cap C \rightarrow P(A \cap C) = 0.22$$

	w_1 (45-50)	w_2 (50-55)	w_3	w_4	w_5	$P(H_i)$	marginal probability
H_1	0.08	0.04	.	.	0.14	$P(H_1)$	$P(A B) = ?$
H_2	0.12		.	.	0.26	$P(H_2)$	$P(A B) = \frac{P(A \cap B)}{P(B)}$
H_3	0.06				0.26	$P(H_3)$	$= \frac{0.08}{0.6} \approx 0.15$
H_4	0.02				0.22	$P(H_4)$	
H_5	0				0.12	$P(H_5)$	
assume sums							Compare to $P(A)$
$P(w_i)$							0.24
Event							
$P(A)$; $A = \{\text{student has weight } w=w_2\}$							
$B = \{\text{student has } H \geq H_3\}$							

$$H=H_3, H=H_4, H=H_5$$

$$\begin{aligned} P(A|B) &= P(A \cap B) \\ &= \frac{P(A \cap B)}{P(B)} \\ &= \frac{0.08}{0.6} \approx 0.15 \\ &\text{In this case} \\ &P(A|B) < P(A) \end{aligned}$$

Now consider another event



$$A = \checkmark \quad \{ \omega = \omega_2 \}$$

$$C = \{ H \leq H_3 \}$$

$$\text{Now: } P(A|C) > P(A)$$

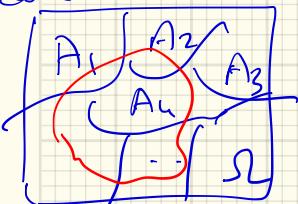
$$\text{Recall: } P\left(\bigcap_i A_i\right) =$$

$$P(A \cap B \cap C) = P(C|A \cap B) P(A \cap B) = P(C|A \cap B), P(B|A) P(A)$$

Multiplication (Prob. Chain) Rule

$$P\left(\bigcap_{i=1}^n A_i\right) = P(A_1) \cdot P(A_2|A_1) \cdot P(A_3|A_1 \cap A_2) \cdots \cdot P(A_n | \bigcap_{i=1}^{n-1} A_i)$$

Recall:



$\bigcup_i A_i = S \rightarrow$ then $\{A_i\}_{i=1}^n$ are a partition of S

$$P(B) = \sum_{i=1}^n P(B|A_i) \cdot P(A_i)$$

$$P(B \cap A_i)$$

Total Prob.
Law.

$$A_i \cap A_j = \emptyset, \forall i, j, i \neq j.$$

$$P(A|C) = \frac{P(A \cap C)}{P(C)} = \frac{0.22}{0.66} = 0.33$$

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

Independence: $A \& B$ are independent events when

occurrence of A (or B) does not influence occurrence of the other.

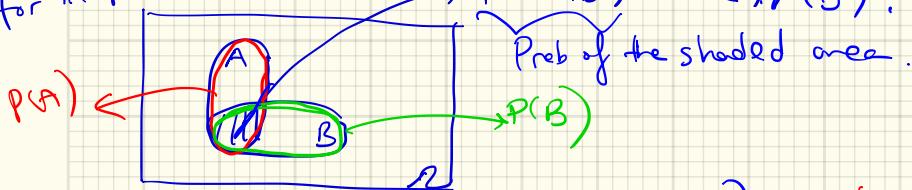
$A \& B$ independent

$$\rightarrow P(A|B) = P(A) \Rightarrow \frac{P(A \cap B)}{P(B)} \stackrel{+ \epsilon}{\underset{\Sigma}{\approx}} P(B) \neq 0 \quad \text{denominator}$$

Def: $A \& B$ are independent iff (if and only if)

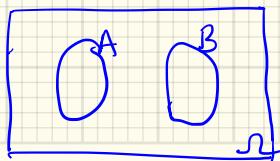
$$\rightarrow P(A \cap B) = P(A) \cdot P(B)$$

for independence, check whether $P(A \cap B) = P(A)P(B)$.



Q: Are disjoint events independent? No!

Say $A \& B$ are disjoint $\Rightarrow A \cap B = \emptyset$



$$P(A \cap B) = 0 \neq P(A) \cdot P(B)$$

$A \& B$ are NOT INDEPENDENT.

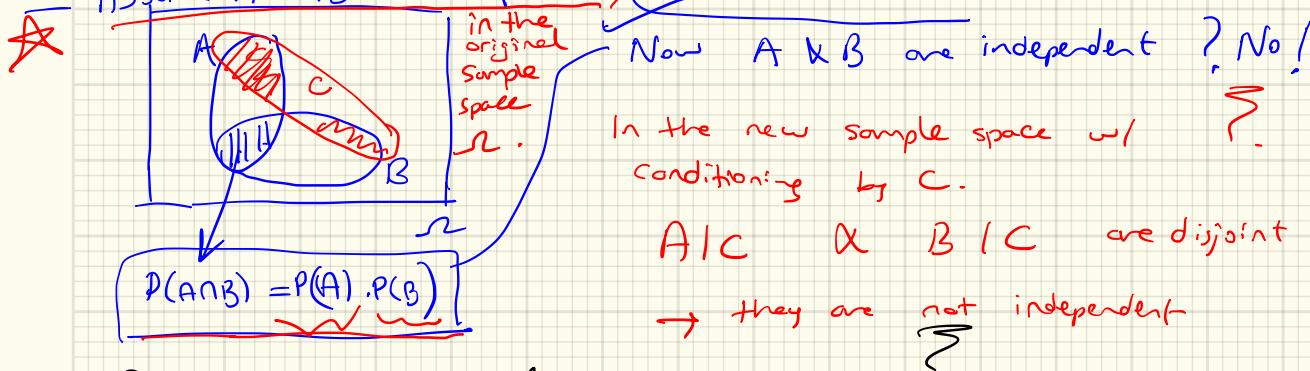
Don't confuse disjointness w/ independence.

\rightarrow Conditional Independence: (same rules apply) $A \wedge B \xrightarrow{\text{conditionally}} \text{indep} \wedge C \text{ occurred}$.

$\star A \wedge B$ are conditionally indep. $P(A \wedge B | C) = P(A | C) \cdot P(B | C)$

\therefore independence applies in a conditional universe.

Assume $A \wedge B$ are independent; (Now we are given that C occurred.)



\star Conditioning may affect independence.

\star Having independence in the original space does not imply independence in the conditional sample space.

- These statements are equivalent.

1) $A \wedge B$ are independent

2) $A^c \wedge B^c$ " "

3) $A \vee B$ " "

4) $A^c \vee B$ " "

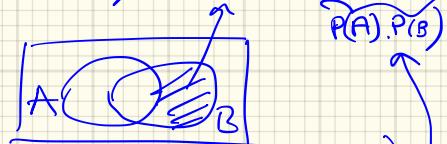
$$\Rightarrow P(A^c \cap B) = P(B)(1 - P(A))$$

$$P(A^c \cap B) = P(B) \cdot \overbrace{P(A^c)}^{P(A) \cdot P(B)}$$

$\therefore A^c \wedge B$ are independent.

Show 4):

$$P(A^c \cap B) = P(B) \cdot P(A \cap B)$$



($A \wedge B$ are independent.)
 $P(A \cap B) = P(A) \cdot P(B)$)

$$\begin{aligned} P(\Omega) &= 1 \\ P(A) &= P(A \cap B) + P(A \cap B^c) \\ P(A^c) &= 1 - P(A) \end{aligned}$$

Exercise : Show the equiv. of these 4 statements

Ex: Tossing a fair coin two times, w/ $P = 0.5$ (fair coin. $\begin{matrix} H \\ \rightarrow T. \end{matrix}$)

Events: A_1 : H on first toss

A_2 : H on 2nd toss

A_3 : same outcome on both tosses

Sample Space $\Omega = \{(S) \underbrace{\text{HH}}, \underbrace{\text{HT}}, \underbrace{\text{TH}}, \underbrace{\text{TT}}\}; |\Omega|=4$

each outcome is equally likely, $P(\cdot) = \frac{1}{4}$

$$P(A_1) = \frac{1}{2} = P(A_2) = P(A_3)$$

$$P(A_1 \cap A_2) = \frac{|\text{HH}|}{4} = P(A_1) \cdot P(A_2)$$

Q: Are A_i pairwise independent?

Yes all 3 are pairwise indep.

$$P(A_1 \cap A_3) = \frac{1}{4}$$

$$P(A_2 \cap A_3) = \frac{1}{4}$$

Q. What about mutually independent?

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

$\uparrow \quad \uparrow \quad \uparrow \quad \frac{1}{4} \neq \frac{1}{8}$

\therefore Not mutually indep. event.

\Rightarrow Pairwise indep. does not imply mutual independence. of multiple events.

multiple events independent $P(\bigcap_i A_i) = P(A_1), P(A_2), \dots, P(A_n)$

Def: A collection of events A_i are called mutually independent iff every sub-collection consists of independent events.

Multiple Random Experiments :

\rightarrow A coin toss \rightarrow one random experiment : Sample Space $S_1 = \{HT\}$

\rightarrow Two coin tosses \rightarrow two independent experiments $S_2 = \{H, T\}$

$$\rightarrow S = S_1 \times S_2 = \{HH, HT, TH, TT\}$$

\swarrow 1st toss \searrow 2nd toss

An event in S ; $A = (A_1, A_2)$ $\rightarrow P(A) = P(A_1) \cdot P(A_2) = \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4}$

A Bernoulli trial $\stackrel{\text{experiment}}{\longrightarrow}$ = One coin toss
 $\stackrel{\text{experiment}}{\longrightarrow}$ = A binary outcome experiment

\Rightarrow Bernoulli Sequence \equiv Experiments composed of sub experiments, which are INDEPENDENT.

"Success" , "Failure" $\curvearrowright \star$

$$P(\overline{A}) = p \quad ; \quad p \in [0,1]$$

$$A^c \Rightarrow 1-p = P(A^c)$$

\Rightarrow Multiple coin tosses \equiv Bernoulli sequence.

M times: (M independent coin tosses)
 say $\overbrace{\text{M coin tosses}}$ \Leftrightarrow Bernoulli seq. w/ prob of success $= p$.
 $P(H T H H T T T) = ?$
 $P.(1-p).P.p(1-p)^3 = p^3(1-p)^4$

$$P(\text{Sequence}) = P(\# \text{Heads}) (1-p)^{\# \text{Tails}}.$$

P(k Heads in n tosses)

$$= \binom{n}{k} p^k (1-p)^{n-k}$$

$\boxed{\begin{array}{ccccccc} H & T & H & H & T & H & T \\ | & | & | & | & | & | & | \end{array}}$

$P(3 \text{ Heads in } 7 \text{ tosses})$

Binomial Law.

Recall: Counting w/ replacement
w/o replacement.) (49 in Turkish lottery)

Ex: Lottery: (N4 state 44 balls) → pick 6 numbers
for your ticket.

Winning # is randomly selecting 6 #'s from 44

w/ replacement $N = 44 \text{ balls}$

ordered. $\boxed{1} \boxed{1} \boxed{1} \boxed{1} \boxed{1} \boxed{1} = N^{r=6} = \underline{44}^6 \rightarrow \frac{1}{44^6} \approx .1 \text{ in a } 7 \text{ Billion.}$

$44 \cdot 44 \cdot 44 \cdot 44 \cdot 44 \cdot 44.$

w/o replacement.

$$\boxed{44 \cdot 43 \cdot 42 \dots} = N_r = (44)_6 = \frac{44!}{(44-6)!} = \frac{44!}{38!} \geq 5 \text{ Billion.}$$

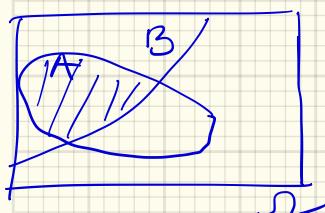
Unordered w/o replacement. $r!$ ways of ordering of r elements from total N elements.

$$\Rightarrow \frac{N!}{(N-r)! \cdot r!} = \binom{N}{r} \text{ combination.}$$

Ex: Coin tossing experiment: 10 tosses of independent coin tosses \rightarrow Bernoulli experiment.

Event $B = 3$ out of 10 tosses were "Heads" \rightarrow .

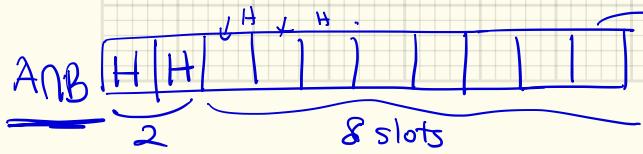
Event $A =$ The 1st two tosses were "Heads"



Given B occurred $\rightarrow P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{\binom{8}{1}}{\binom{10}{3}}$

$$|B| = \binom{10}{3} \rightarrow p^3 \cdot (1-p)^7$$

$\underbrace{\hspace{1cm}}$ prob. of each outcome in B .



\rightarrow 3rd Head is placed in those 8 slots
in $\binom{8}{1}$ ways = 8.

Binomial Prob Law: $p(\text{success}) = p$

$P(k \text{ successes in } M \text{ Bernoulli trials})$

$\equiv P(k \text{ successes in } M \text{ binary}^{\text{attempts}} \text{ in any order})$

$$= \binom{M}{k} p^k (1-p)^{M-k}$$

Geometric Prob Law: Another Bernoulli trial

$F = \text{Fail (Tails)}$

Events = $F \cdot F \cdot F \cdot \dots \cdot F \cdot S$) = ?

P (success at the k^{th} attempt in a Bernoulli experiment)

$$= (1-p)^{k-1} \cdot p$$

(Ex. 4.8) Fax machine dials a phone # that is busy
80% of the time. $\rightarrow p = 0.2, (1-p) = 0.8$

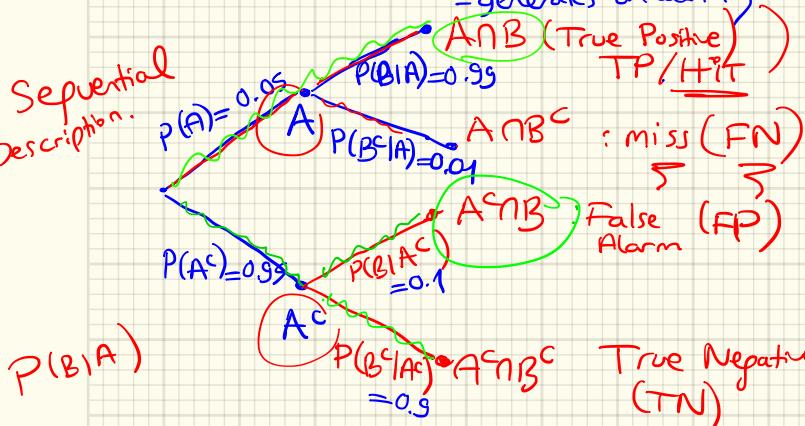
$$P(\text{success at the } k^{\text{th}} \text{ trial}) = \frac{(0.8)^8 \cdot 0.2}{(1-p)^{k-1} \cdot p}$$

Ex: [1.9 Bertsekas] Radar Detection : Detecting an Aircraft.

Event A = { aircraft present }. A^c = { aircraft not present }

$$B = \{ \text{radar fires} \\ \quad \quad \quad \text{= generates an alarm} \} \rightarrow B^c = \{ \text{radar does not gen.} \\ \quad \quad \quad \text{an alarm} \}$$

Sequential Description



$$P(B) = P(TP) + P(FP) = 0.0495 + 0.99 \times 0.1 = 0.1445$$

Q: Given that your radar fired, how likely is it that there is an airplane out there?

$$P(A|B) = ? \quad \frac{P(A \cap B)}{P(B)} = \frac{0.0495}{0.1445} = 0.34$$

This is inference!

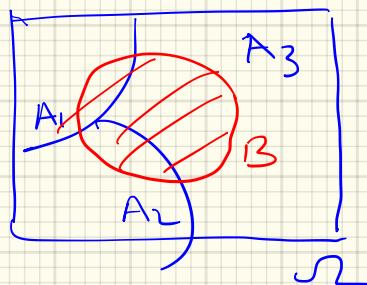
Bayes Rule (Theorem)

"Prior" probabilities : $P(A_i)$

: our initial belief about how likely each event A_i are to occur

We know $P(B|A_i) \forall i$,
"conditional" probabilities

Now, we're told that event B occurred.



$$P(A_i | B) = \frac{P(A_i \cap B)}{P(B)}$$
$$P(A_i | B) = \frac{P(B|A_i) \cdot P(A_i)}{\sum_j P(B|A_j) P(A_j)}$$

prior.
Bayes Thm (Rule).

We reversed the order of conditioning!

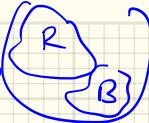
Cause A_i $\xrightarrow{\text{aircraft present}}$ B (effect) → Radar fires.

we Reversed it
In Bayes rule: Scenario A_i given B .

$A_i \xleftarrow{\text{how likely was inference}} B$: B occurred.

we are given the effect.

Ex: We have an urn w/ Red & Black balls.



(Kang) $A = \{ \text{observe 10 Red balls in a row w/replacement} \}$

Hypothesis $B = \{ \text{urn is fair} \} = (\# \text{Reds} = \# \text{Blacks})$

$$P(B|A) = ? \quad P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{0.9}{\text{prior belief}}$$

$$P(A|B) = \binom{10}{10} \left(\frac{1}{2}\right)^{10} \cdot \left(\frac{1}{2}\right)^0 = \left(\frac{1}{2}\right)^{10}$$

$$P(B|A) = \frac{P(A|B)P(B)}{P(A)} \quad P(A) = [P(A|B)P(B) + P(A|B^c)P(B^c)]$$

$$P(A|B^c) = 1 \quad (\{B^c = \{\text{unfair urn}\} = \{\text{all Red balls}\}\} \text{ for simplicity.})$$

$$P(A) = \left(\frac{1}{2}\right)^{10} (0.9) + 1 \cdot (1 - 0.9) \quad \begin{aligned} &\text{posterior prob. (after 10 Red} \\ &\text{balls are drawn) that the} \\ &\text{urn is fair is only} \end{aligned}$$

$$P(B|A) = \frac{P(A|B)P(B)}{P(A)} = \frac{\left(\frac{1}{2}\right)^{10}(0.9)}{\left(\frac{1}{2}\right)^{10}(0.9) + 0.1} \approx 0.0087 \quad 0.0087 \rightarrow$$

\rightarrow reject the assumption of a fair urn!
 \sum (hypothesis)

A Quantity "Odds Ratio": Odds against the hypothesis
(odds) $\xrightarrow{\text{OR}}$ (a/f vs $b/(n-a)$)

$$\text{odds} = \frac{P(B^c|A)}{P(B|A)} = \frac{1 - P(B|A)}{P(B|A)} = \frac{1 - 0.0087}{0.0087} = 113.$$

∴ having an unfair urn is 113 times more likely (probable) than fair urn.