

Recal :

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

also:

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

Ex.

A random individual has Prob. .05 of having a certain disease

— If a Person has the disease, then Prob. of a Positive Test

.99. — If the Person does —

not have disease, then

Prob. of Positive test is .10

- what is the false positive Prob.?
- what is the false negative Prob.?

Let

$T = \{\text{test is positive}\}$

$D = \{\text{person has disease}\}$

we are given

$$P(T|D) = .99$$

$$P(T^c|D) = .01$$

$$P(T|D^c) = .10$$

$$P(T^c|D^c) = .90$$

$$P(D) = .05$$

$$P(D^c) = .95$$

we must find:

$$P(T \cap D^c) \quad (\text{false positive})$$

$$P(T^c \cap D) \quad (\text{false negative})$$

we have

$$P(T \cap D^c) = P(T|D^c) \cdot P(D^c)$$

$$= (.10)(.95) = \boxed{.095}$$

and

$$P(T^c \cap D) = P(T^c|D) \cdot P(D)$$

$$= (.01)(.05) = \boxed{.0005}$$

Recall

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

Generalize this

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

Proof

$$\text{RHS} = \cancel{P(A)} \cdot \frac{P(A \cap B)}{\cancel{P(A)}} \cdot \frac{P(A \cap B \cap C)}{\cancel{P(A \cap B)}}$$

$$= P(A \cap B \cap C) = \text{LHS} \quad \blacksquare$$

Exercise Prove for 4 sets

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

Multiplication rule

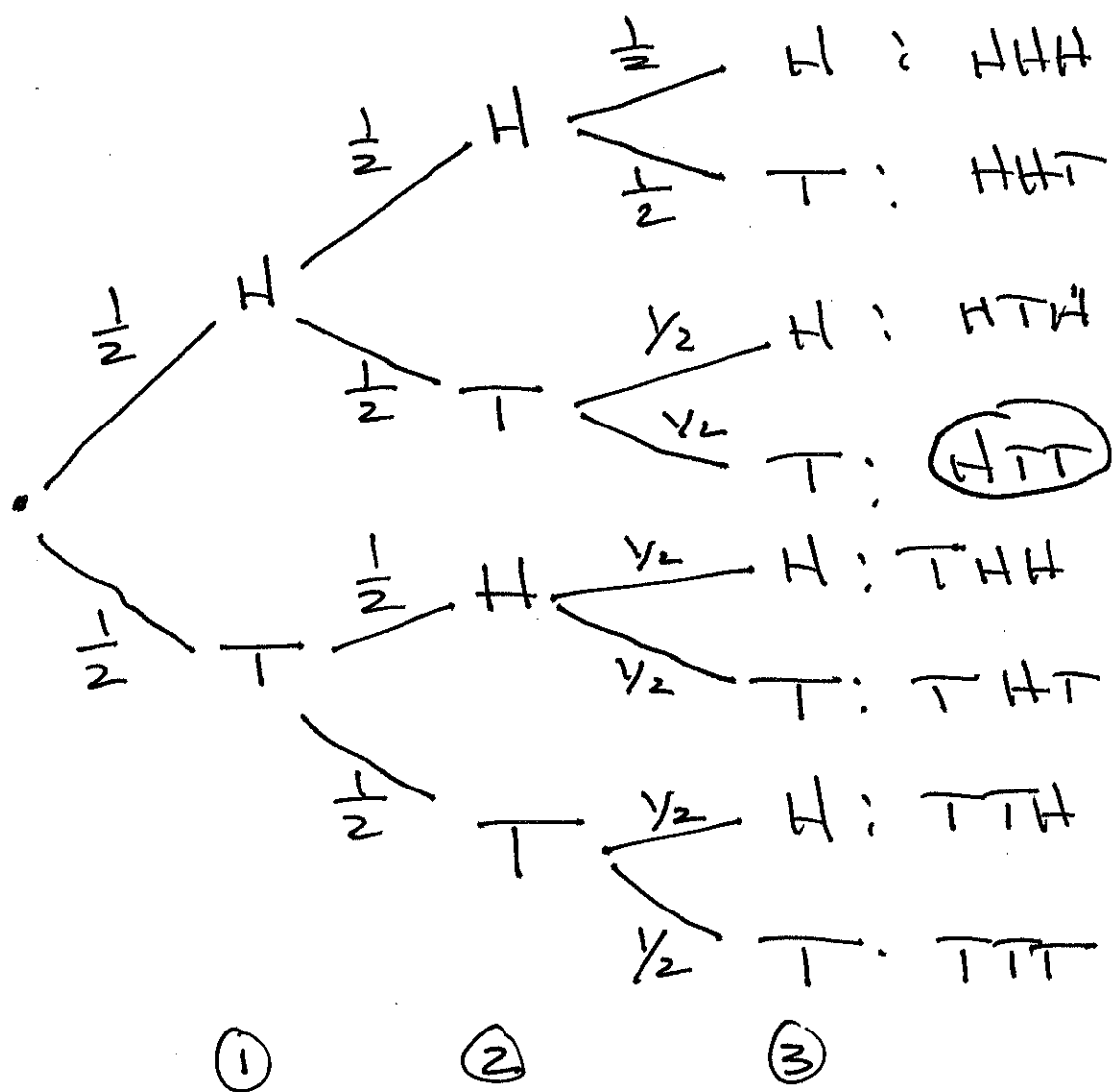
$$P\left(\bigcap_{i=1}^n A_i\right) = P(A_1) \cdot P(A_2|A_1) \cdot P(A_3|A_1 \cap A_2) \\ \dots \cdot P(A_n | \bigcap_{j=1}^{n-1} A_j)$$

i.e.

$$P\left(\bigcap_{i=1}^n A_i\right) = \prod_{i=1}^n P\left(A_i \mid \bigcap_{j=1}^{i-1} A_j\right)$$

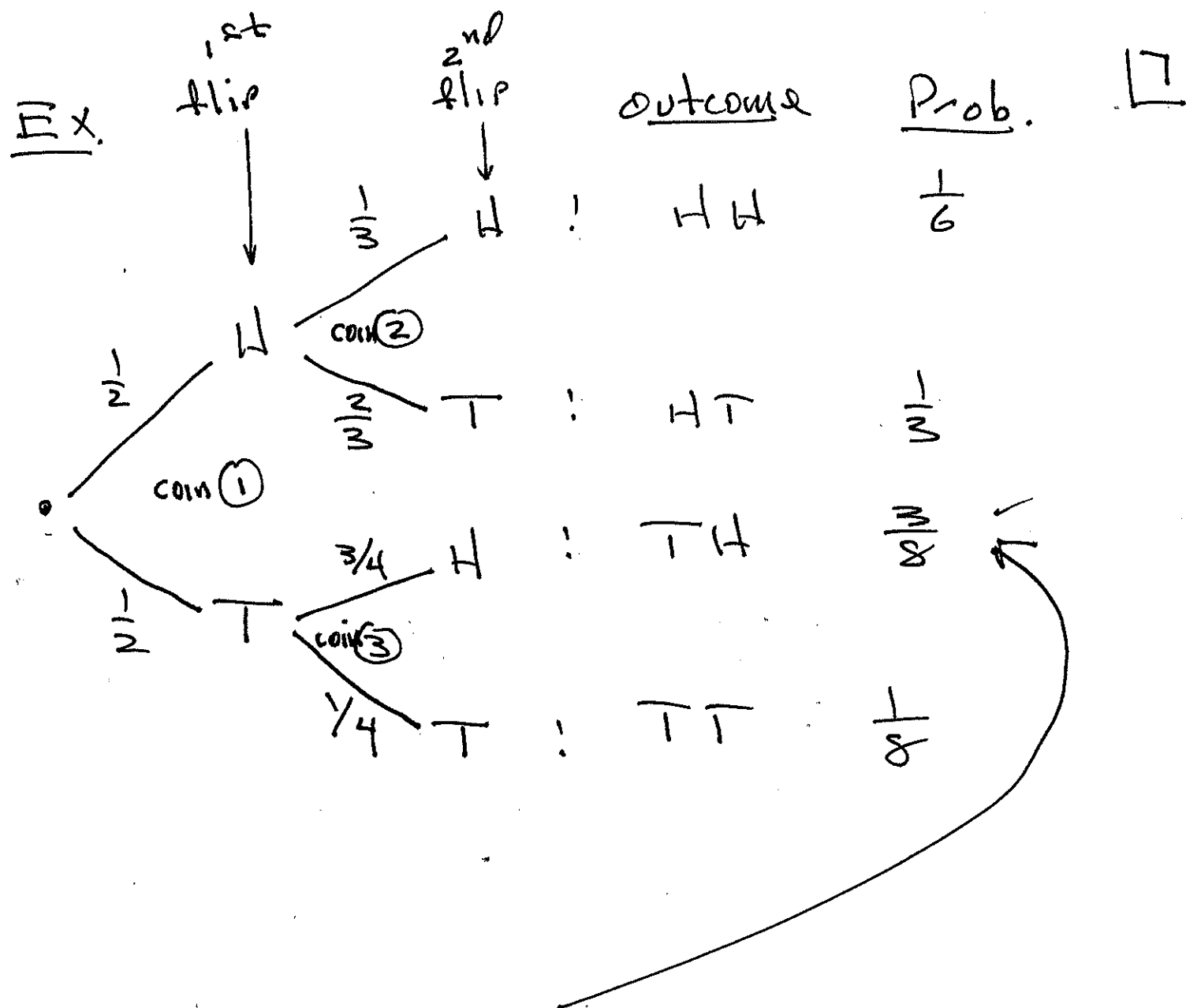
This rule justifies the practice of multiplying probabilities down branches of a tree diagram.

Ex toss 3 fair coins



$$\frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{8}$$

This also works for loaded coins.



$$P(1^{\text{st}} T \text{ and } 2^{\text{nd}} H)$$

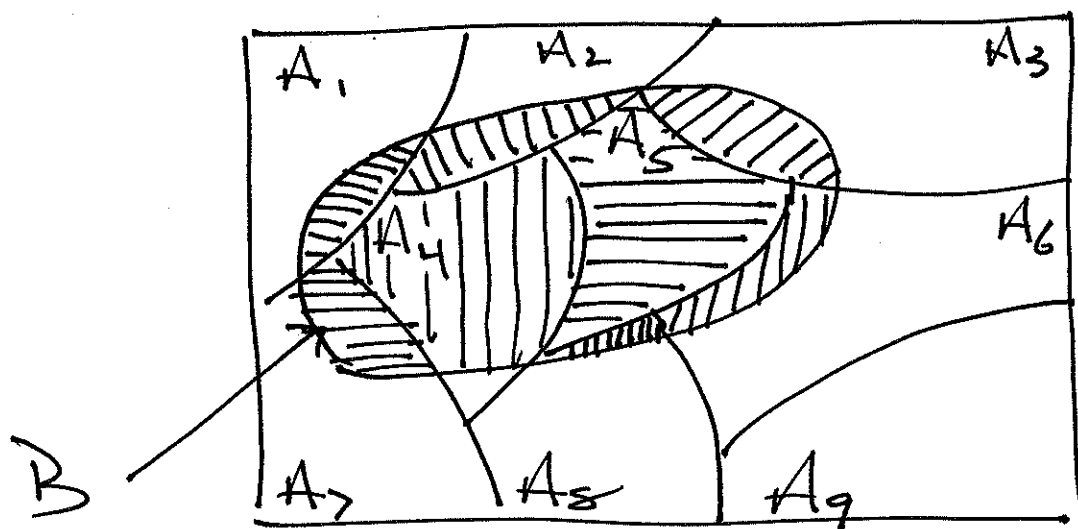
$$= P(1^{\text{st}} T) \cdot P(2^{\text{nd}} H | 1^{\text{st}} T)$$

$$= \frac{1}{2} \cdot \frac{3}{4} = \boxed{\frac{3}{8}}$$

1.4 Total Probability, Bayes rule

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Let $A_1, A_2, \dots, A_n$ is a
Partition of Ω :



Ω

Assume
 $P(A_i) > 0$
for $i = 1, \dots, n$

Total Probability theorem

$$P(B) = P(A_1 \cap B) + P(A_2 \cap B) + \dots + P(A_n \cap B)$$

$$= P(A_1) \cdot P(B|A_1) + \dots + P(A_n) \cdot P(B|A_n)$$

Ex. Roll a "fair" 6-sided die. If result is in $\{1, 2, 3\}$, roll again, otherwise stop. What is the probability the the sum (of all rolls) is ≥ 5 ?

- Let $A_1 = \{1^{st} \text{ roll is } 1\}$ $P(A_1) = \frac{1}{6}$
- $A_2 = \{1^{st} \text{ roll is } 2\}$ $\frac{1}{6}$
- $A_3 = \{ \text{" " " } 3 \}$ $\frac{1}{6}$
- $A_4 = \{ \text{" " " } 4 \}$ $\frac{1}{6}$
- $A_5 = \{ \text{" " " } 5 \}$ $\frac{1}{6}$
- $A_6 = \{ \text{" " " } 6 \}$ $\frac{1}{6}$

let $B = \{ \text{sum} \geq 5 \}$

- if A_1 occurs, B occurs iff $2^{\text{nd}} \in \{4, 5, 6\}$
 .. A_2 .., B iff $2^{\text{nd}} \in \{3, 4, 5, 6\}$
 .. A_3 .., B iff $2^{\text{nd}} \in \{2, 3, 4, 5, 6\}$
 .. A_4 .., B does not occur
 .. A_5 .., B occurs
 .. A_6 .., B occurs

Thus

$$P(B|A_1) = \frac{3}{6} = \frac{1}{2}$$

$$P(B|A_2) = \frac{4}{6} = \frac{2}{3}$$

$$P(B|A_3) = \frac{5}{6}$$

$$P(B|A_4) = 0$$

$$P(B|A_5) = P(B|A_6) = 1$$

Thus

$$P(B) = P(A_1) \cdot P(B|A_1) + \dots + P(A_6) \cdot P(B|A_6)$$

$$= \frac{1}{2} \cdot \frac{1}{6} + \frac{2}{3} \cdot \frac{1}{6} + \frac{5}{6} \cdot \frac{1}{6} + 0 \cdot \frac{1}{6} + 1 \cdot \frac{1}{6} + 1 \cdot \frac{1}{6}$$

$$= \frac{1}{6} \left(\frac{1}{2} + \frac{2}{3} + \frac{5}{6} + 0 + 1 + 1 \right)$$

$$= \frac{1}{6} \cdot \frac{24}{6} = \boxed{\frac{2}{3}}$$

Exercise enumerate full sample space, and do it without total Probability.

Bayes' Rule

Let $A_1, A_2, \dots, A_n$ be a Partition of Ω (i.e. Pairwise disjoint, and union = Ω .) Assume $P(A_i) > 0$ for $i = 1, \dots, n$. Then for each i

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

$$= \frac{P(A_i) \cdot P(B|A_i)}{P(A_1) \cdot P(B|A_1) + \dots + P(A_n) \cdot P(B|A_n)}$$

Proof

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

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

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

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

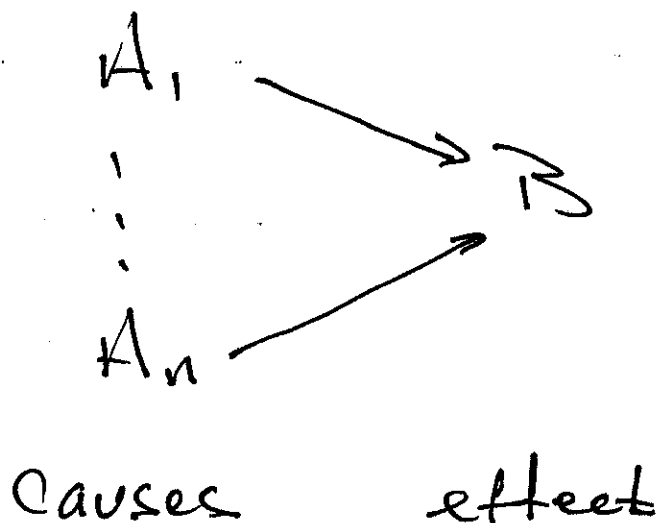
Apply this to A_i , then use
total Prob. thm on $P(B)$. 

Bayes' rule is used for inference questions:

causes: $A_1, A_2, \dots, A_n$

effect: B

we observe effect, and infer causes.



$$A_i \rightarrow B : P(B|A_i)$$

Causation ↑
(strength of causation)

$$A_i \leftarrow B : P(A_i|B)$$

Inference ↑
(strength of inference)

$$P(A_i|B) = \frac{\overbrace{P(A_i)}^{\text{Prior}} \cdot \overbrace{P(B|A_i)}^{\text{Likelihood}}}{\underbrace{P(B)}_{\text{Marginal}}}$$

Posterior ↑
Marginal

Ex. 'Disease' example again

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$T = \{ \text{test positive} \}$

$D = \{ \text{has disease} \}$

$$P(T|D) = .99$$

$$P(T|D^c) = .10$$

$$P(D) = .05, P(D^c) = .95$$

find $P(\text{disease} | \text{test})$, i.e.

$$P(D|T) = \frac{P(D) \cdot P(T|D)}{P(T)}$$

$$= \frac{P(D) \cdot P(T|D)}{P(D) \cdot P(T|D) + P(D^c) \cdot P(T|D^c)}$$

$$= \frac{(.05)(.99)}{(.05)(.99) + (.95)(.10)}$$

$$= \boxed{.3426}$$