

## Supplemental Lecture

Observe

- $P(A|B) \cdot P(B) = \underbrace{P(A \cap B) = P(A)P(B|A)}_{\text{generalize this}}$
- $P(A \cap B \cap C) = P(A) \cdot P(B|A) \cdot P(C|A \cap B)$  ✓

Prove this:

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

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

$$\bullet P(A \cap B \cap C \cap D)$$

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

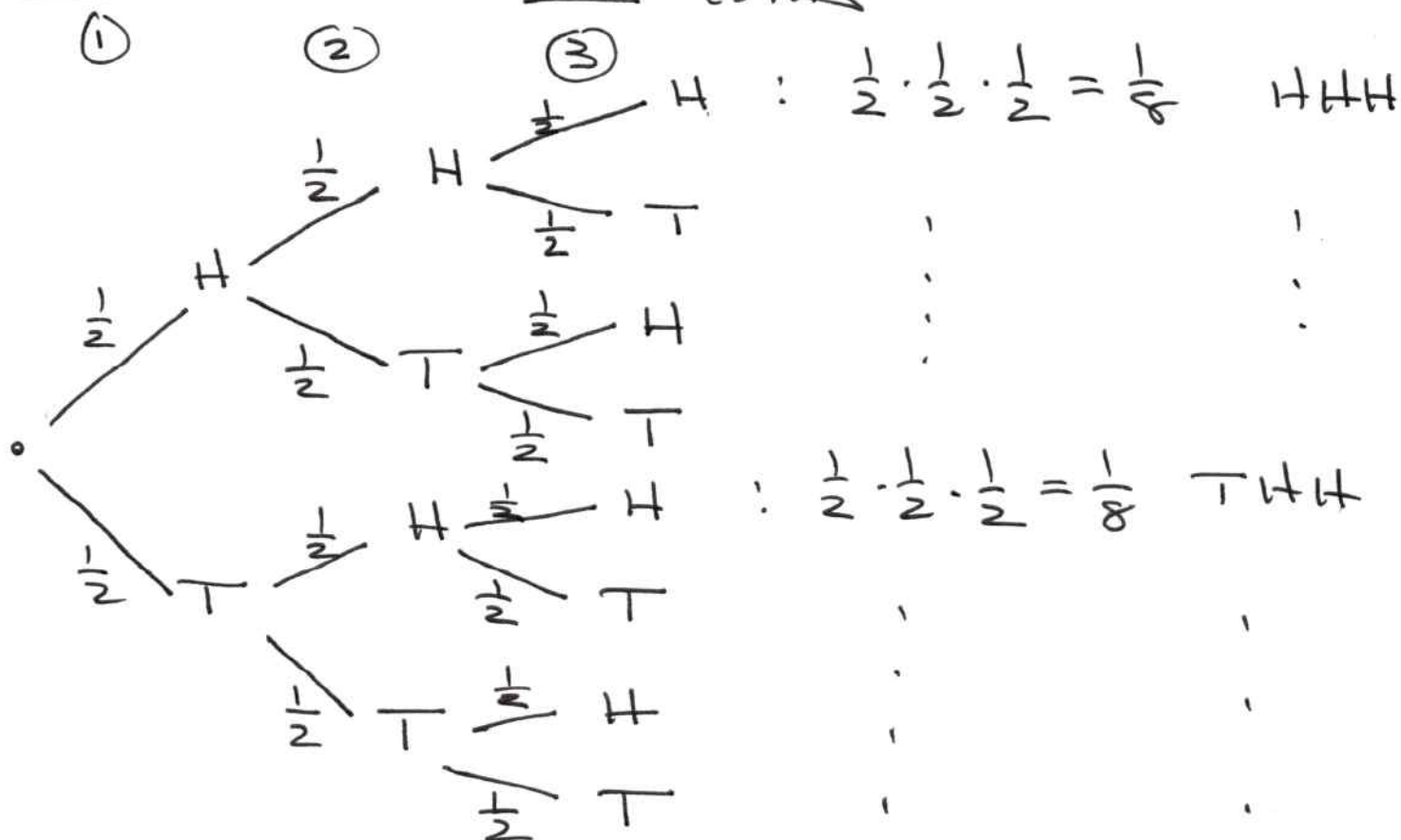
Exercise: Prove this

Multiplication Rule

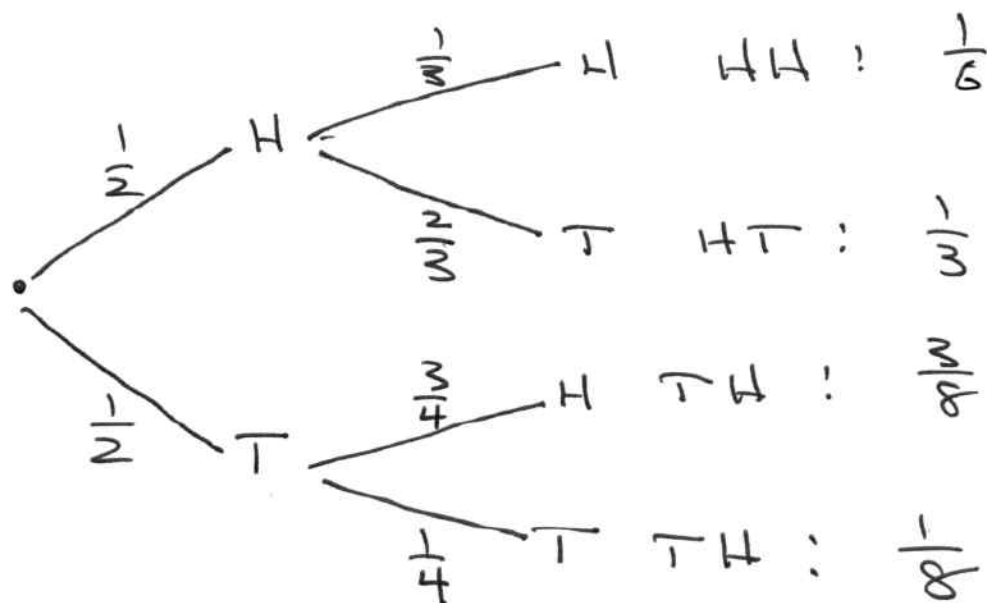
$$P\left(\bigcap_{i=1}^n A_i\right) = P(A_1) \cdot P(A_2|A_1) \cdot P(A_3|A_1 \cap A_2) \cdots \\ \cdots P\left(A_n \mid \bigcap_{i=1}^{n-1} A_i\right)$$

Common Practice: multiply Prob.  
down a tree diagram.

EX Toss 3 fair coins



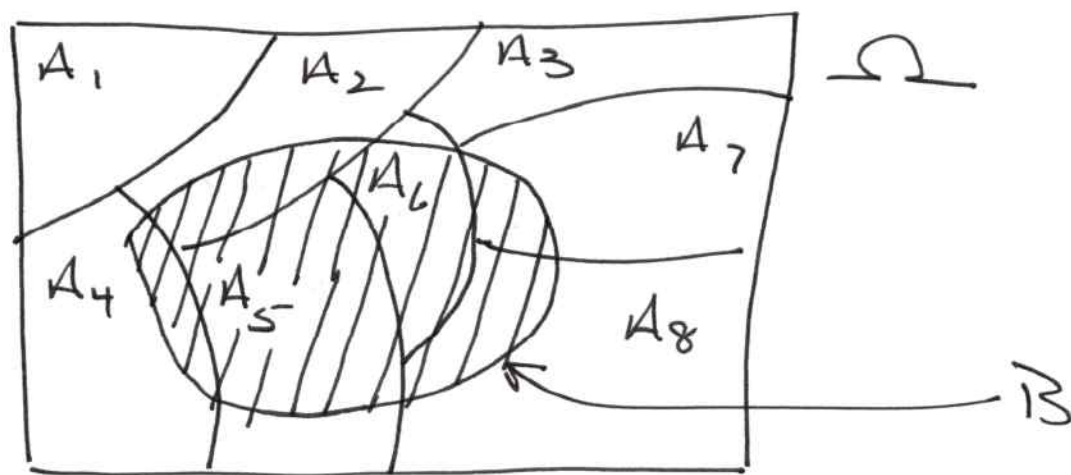
EX. 3 coins {1, 2, 3} : flip ①, if head, flip ②, if tails, flip ③



## 1.4 Total Probability, Bayes rule

Let  $A_1, A_2, \dots, A_n \subseteq \Omega$  be pairwise disjoint (i.e.  $A_i \cap A_j = \emptyset$  if  $i \neq j$ ).

Assume  $A_1 \cup A_2 \cup \dots \cup A_n = \Omega$ . We call this a Partition of  $\Omega$ .



### Total Probability Theorem

Assume  $P(A_i) > 0$  for each  $1 \leq i \leq n$ . Then for any  $B \subseteq \Omega$  we have

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

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

Ex: Roll a fair die. If result is in  $\{1, 2, 3\}$ , roll again, otherwise stop. What is the Prob. of

$$B = \{\text{sum of all rolls} \geq 5\}$$

Define

$$A_i = \{\text{1st roll is } i\} \quad (1 \leq i \leq 6)$$

$$\rightarrow P(A_i) = \frac{1}{6}$$

Note

if  $A_1$  occurs, then  $B$  occurs iff 2<sup>nd</sup>  $\in \{4, 5, 6\}$

"  $A_2$  " " " " "  $\in \{3, 4, 5, 6\}$

"  $A_3$  " " " " "  $\in \{2, 3, 4, 5, 6\}$

Thus

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

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

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

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

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

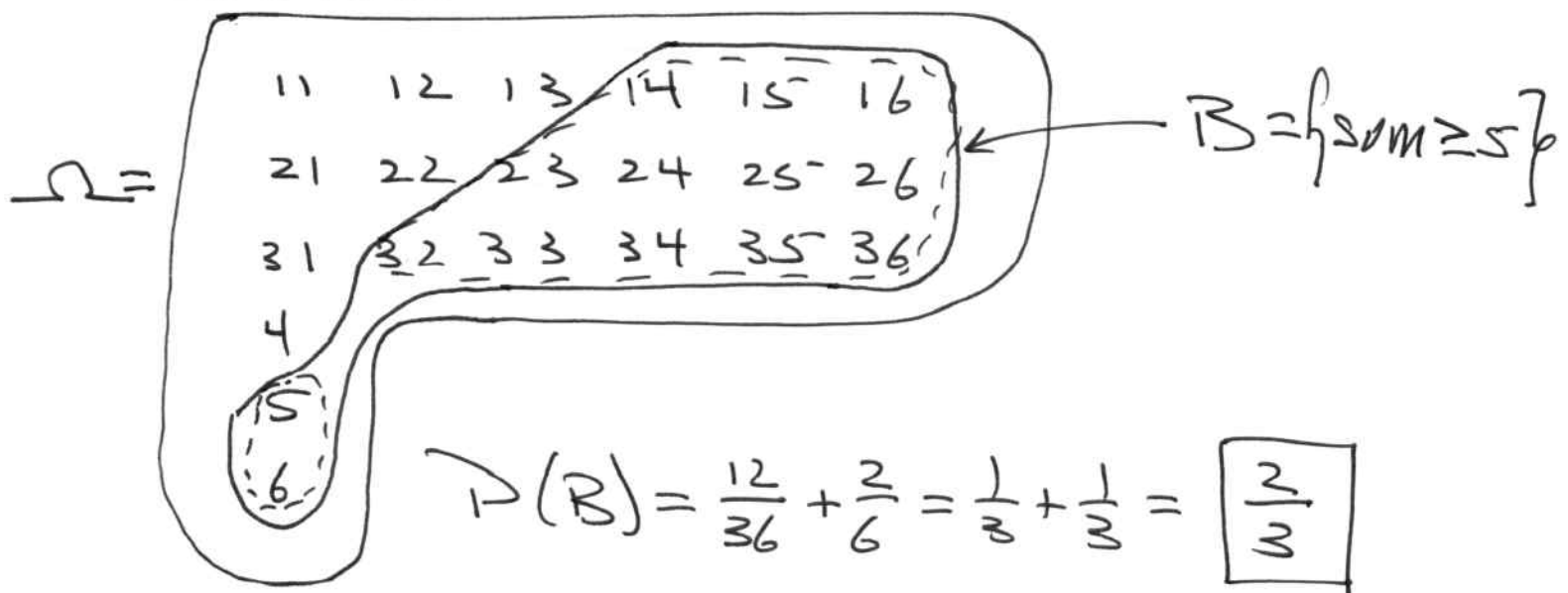
$$P(B|A_6) = 1$$

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

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

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

Another way



# Bayes' Rule

Let  $A_1, \dots, A_n$  be pairwise disjoint events forming a partition of  $\Omega$ .

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

Also assume  $B \subseteq \Omega$  with  $P(B) > 0$ .

Then for each  $i = 1, 2, \dots, n$

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

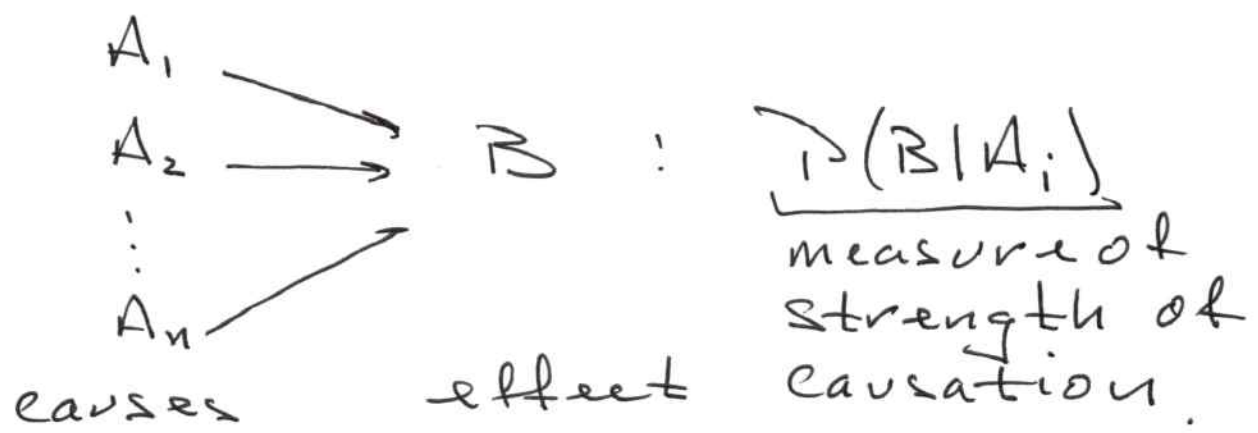
$$= \frac{P(A_i) \cdot P(B | A_i)}{P(A_1) \cdot P(B | A_1) + \dots + P(A_n) \cdot P(B | A_n)}$$

Proof: follows from defn of cond.

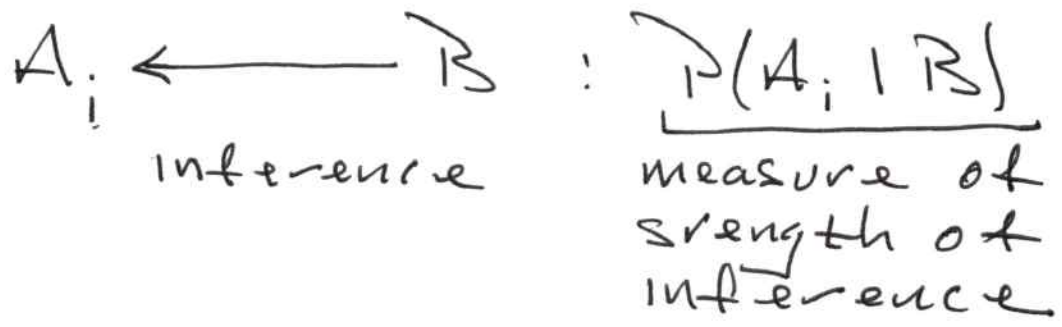
prob. and total prob. thm. ▀



Bayes' rule is used for 'inference' Problems.



We go from effect to cause





$$\underbrace{P(A_i | B)}_{\text{Posterior}} = \frac{\overbrace{P(A_i)}^{\text{Prior}} \overbrace{P(B | A_i)}^{\text{likelihood}}}{\underbrace{P(B)}_{\text{marginal}}}$$

often we use Partition:  $A \cup A^c = \Omega$

Then Bayes' rule becomes:

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

$$= \frac{P(A) \cdot P(B | A)}{P(A) \cdot P(B | A) + P(A^c) \cdot P(B | A^c)}$$

Ex. disease example again

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

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

we were given

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

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

$$P(D) = .05 \quad \therefore P(D^c) = .95$$

to find

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

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

$$= \frac{(.05)(.99)}{(.05)(.99) + (.95)(.10)} = \boxed{.3426}$$

round to 4