Conditional Probability

- Let A and B be events P(B) > 0
- $P(A|B) = \frac{P(A \cap B)}{P(B)}$

Multiplication Laws

- 1. $P(A \cap B) = P(A|B)P(B)$
- 2. Let A_1 through A_n be a sequence of events
- Then $P(\cap A_i) = P(A_1)P(A_2|A_1)...P(A_n|A_1 \cap ...A_{n-1})$

Law of Total Probability

- Let B_1 through B_n be a partition of Ω
- Let $\bigcup B_i = \Omega$, $B_i \cap B_j = \emptyset$, $i \neq j$
- Then for any event A, $P(A) = \sum P(A_i|B_i)P(B_i)$

Example

- $A = \{Randomly drawn person will vote in the US\}$
- $B_i = \{\text{Randomly drawn person resides in state i}\}, i = 1 \text{ through } 50$
- $P(A) = \sum P(A|B_i)P(B_i)$
- $P(B_i)$ = number of people in state i / number of people in the US

Example: Occupational Mobility

- Occupations are categorized into 3 groups, upper, middle, and lower
- $u_1 = \{\text{Randomly drawn father's occupation is in u}\}$
- $u_2 = \{\text{Randomly drawn son's occupation is in u}\}$
- $M_1 = \{\text{Randomly drawn father's occupation is in M}\}$
- $M_2 = \{\text{Randomly drawn son's occupation is in M}\}$
- $L_1 = ...$
- $L_2 = ...$
- We are given

	U_2	M_2	L_2
$\overline{U_1}$.45	.48	.07
M_1	.05	.7	.25
L_1	.01	.5	.49

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- $P(U_1) = .1, P(M_1) = .4, P(L_1) = .5, ...$
- Law of total probability = $P(U_2|U_1)P(U_1) + P(U_2|M_1)P(M_1) + P(U_2|L_1)P(L_1)$
- Partition is U_1, M_1, L_1
- Note: $P(U_2) + P(M_2) + P(L_2) = 1$

How can we flip conditional probabilities?

- eg $P(U_1|U_2)$
- eg $P(U_1|U_2)$ $P(U_1|U_2) = \frac{P(U_1 \cap U_2)}{P(U_2)}$
- Notice $P(U_1 \cap U_2) = P(U_2|U_1)P(U_1)$
- Notice $P(U_1|U_2) = \frac{P(U_2|U_1)P(U_1)}{P(U_2)} = \frac{.45*.1}{.07}$
- This is an illustration of Bayes' rule
- Two formulations
 - 1. For any events A and B, P(A), P(B) > 0

 - We have $P(B|A) = \frac{P(A|B)P(B)}{P(A)}$ Note P(B|A)P(A) = P(A|B)P(B)
 - 2. Let B_1 to B_n be a partition of Ω
 - Then for my fixed $i \in 1 \to n$

 - $-P(B_{i}|A) = \frac{P(A|B_{i})P(B_{i})}{P(A)}$ $-P(B_{i}|A) = \frac{P(A|B_{i})P(B_{i})}{\sum P(A|B_{l})P(B_{l})}$

Example: Lie Detector Test

- $L = \{ \text{Person taking the test is lying} \}$
- $T = \{ \text{Person taking the test is telling the truth} \}$
- $D_+ = \{\text{Positive test}\}$
- $D_{-} = \{\text{Negative alarm}\}$
- According to manufacturers data, we have $P(D_{+}|L) = .88$, $P(D_{-}|L) = .12$, $P(D_{+}|T) = .14$, $P(D_{-}|T) = .14$
- Let's try to calculate the probability that the person is telling the truth given that there was a false alarm
- $P(T|D_+) = ?$
- Let's assume $P(T)=.99,\ P(L)=.01$ $P(T|D_+)=\frac{P(D_+|T)P(T)}{P(D_+)}=\frac{P(D_+|T)P(T)}{P(D_+|T)P(T)+P(D_+|L)P(L)}$
- Probability of a false positive: $P(T|D_+) = .94$
- This is an issue when screening cuz false positives are too high, therefore these lie detector tests are controversial

Example: Gambler's ruin

- A gambler has k dollars
- They want to win N dollars
- 0 < k < N
- The gambler plays the following game as many times as necessary until he goes bankrupt or reaching N dollars
- The game is flipping a fair coin where heads implies win \$1 and tails implies lose \$1
- A = {gambler goes bankrupt}
- Want to calculate P(A) = ?
- Use $P_k(A)$ for P(A)
- Want to derive a formula for $P_k(A)$ in terms of k and N
- $B = \{1st \text{ toss is heads}\}$

Techniques "first step analysis"

- $P_k(A) = P_k(A|B)P_k(B) + P_k(A|B^C)P(B^C)$
- Law of total probability with partition B, B^C

- $\bullet = .5P_k(A|B) + .5P_k(A|B^C)$ $\bullet = .5P_{k+1}(A) + .5P_{k-1}(A)$ $\bullet \text{ Let } P_k = P_k(A), \text{ ie } P_k = .5P_{k+1} + .5P_{k-1} \text{ for any k}$
 - This is a recurrence relation

 - This is a recurrence reason $\to 2P_k = P_{k+1} + P_{k-1}$ $\to P_k P_{k-1} = P_{k+1} P_k \text{ for all k}$ $\text{ Let } b_k = P_k P_{k-1}$ $\text{ Key observation: } b_1 = b_2 = \dots = c \text{ where } c \text{ is a constant}$