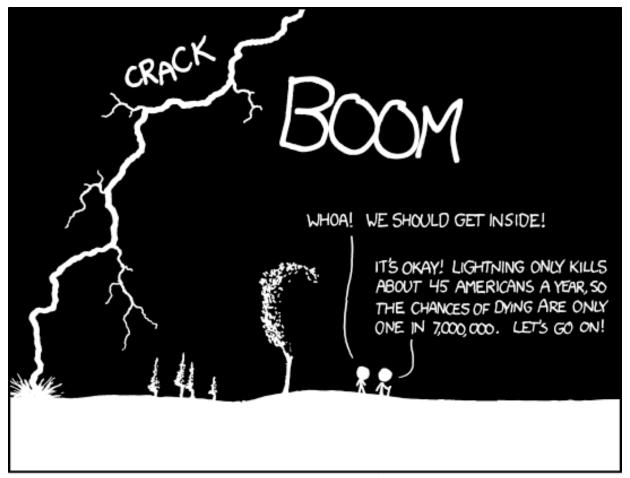
## The Tragedy of Conditional Probability



THE ANNUAL DEATH RATE AMONG PEOPLE WHO KNOW THAT STATISTIC IS ONE IN SIX.

Thanks xkcd! http://xkcd.com/795/

#### A Few Useful Formulas

For any events A and B:

$$P(A B) = P(B A)$$
 (Commutativity)  
 $P(A B) = P(A | B) P(B)$  (Chain rule)  
 $= P(B | A) P(A)$ 

$$P(A B^{c}) = P(A) - P(AB)$$
 (Intersection)

$$P(A B) \ge P(A) + P(B) - 1$$
 (Bonferroni)

# Generality of Conditional Probability

 For any events A, B, and E, you can condition consistently on E, and these formulas still hold:

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

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

$$P(A | B E) = \frac{P(B | A E) P(A | E)}{P(B | E)}$$
 (Bayes' Thm.)

- Can think of E as "everything you already know"
- Formally, P( | E) satisfies 3 axioms of probability

### Dissecting Bayes' Theorem

Recall Bayes' Theorem (common form):

"Posterior" "Likelihood" "Prior"
$$P(H \mid E) = \frac{P(E \mid H) P(H)}{P(E)}$$

- Prior: Probability of H before you observe E
- Likelihood: Probability of E given that H holds
- Posterior: Probability of H after you observe E

#### Odds

Odds of an event defined as:

$$\frac{P(A)}{P(A^c)} = \frac{P(A)}{1 - P(A)}$$

Odds of H given observed evidence E:

$$\frac{P(H \mid E)}{P(H^{c} \mid E)} = \frac{P(H) P(E \mid H) / P(E)}{P(H^{c}) P(E \mid H^{c}) / P(E)}$$
$$= \frac{P(H) P(E \mid H)}{P(H^{c}) P(E \mid H^{c})}$$

• After observing E, just update odds by:  $\frac{P(E \mid H)}{P(E \mid H^c)}$ 

#### Coins and Urns?!

- An urn contains 2 coins: A and B
  - A comes up heads with probability ¼
  - B comes up heads with probability ¾
  - Pick coin (equally likely), flip it, and it comes up heads
  - What are odds that A was picked (note: A<sup>c</sup> = B)?

$$\frac{P(A \mid heads)}{P(A^c \mid heads)} = \frac{P(A) P(heads \mid A)}{P(A^c) P(heads \mid A^c)}$$
$$= \frac{\frac{1}{2} \frac{1}{4}}{\frac{1}{2} \frac{3}{4}} = \frac{1}{3}$$

- Odds are 1/3:1 (or probability ½) that A was picked
- Note: before observing heads P(A) / P(A<sup>c</sup>) = 1:1
  - Equally likely to pick A vs. not picking A (1 out of 2 chance)

### It Always Comes Back to Dice

- Roll two 6-sided dice, yielding values D<sub>1</sub> and D<sub>2</sub>
  - Let E be event:  $D_1 = 1$
  - Let F be event:  $D_2 = 1$
- What is P(E), P(F), and P(EF)?
  - P(E) = 1/6, P(F) = 1/6, P(EF) = 1/36
  - P(EF) = P(E) P(F)  $\rightarrow$  E and F <u>independent</u>
- Let G be event:  $D_1 + D_2 = 5$  {(1, 4), (2, 3), (3, 2), (4, 1)}
- What is P(E), P(G), and P(EG)?
  - P(E) = 1/6, P(G) = 4/36 = 1/9, P(EG) = 1/36
  - P(EG) ≠ P(E) P(G) → E and G <u>dependent</u>

### Independence

Two events E and F are called <u>independent</u> if:

$$P(EF) = P(E) P(F)$$

Or, equivalently:  $P(E \mid F) = P(E)$ 

- Otherwise, they are called <u>dependent</u> events
- Three events E, F, and G independent if:

$$P(EFG) = P(E) P(F) P(G)$$
, and

$$P(EF) = P(E) P(F)$$
, and

$$P(EG) = P(E) P(G)$$
, and

$$P(FG) = P(F) P(G)$$

#### Let's Do a Proof

Given independent events E and F, prove:

$$P(E | F) = P(E | F^{c})$$

Proof:

$$P(E F^c)$$
 =  $P(E) - P(EF)$  Intersection  
=  $P(E) - P(E) P(F)$  Independence  
=  $P(E) [1 - P(F)]$  Factoring  
=  $P(E) P(F^c)$  Complement

So, E and F<sup>c</sup> independent, implying that:

$$P(E \mid F^{c}) = P(E) = P(E \mid F)$$

 Intuitively, if E and F are independent, knowing whether F holds gives us no information about E

### Generalized Independence

General definition of Independence:

Events  $E_1$ ,  $E_2$ , ...,  $E_n$  are independent if for every subset  $E_{1'}$ ,  $E_{2'}$ , ...,  $E_{r'}$  (where  $r \le n$ ) it holds that:

$$P(E_{1'}E_{2'}E_{3'}...E_{r'}) = P(E_{1'})P(E_{2'})P(E_{3'})...P(E_{r'})$$

- Example: outcomes of n separate flips of a coin are all independent of one another
  - Each flip in this case is called a "trial" of the experiment

#### Two Dice

- Roll two 6-sided dice, yielding values D<sub>1</sub> and D<sub>2</sub>
  - Let E be event:  $D_1 = 1$
  - Let F be event:  $D_2 = 6$
  - Are E and F independent? Yes!
- Let G be event:  $D_1 + D_2 = 7$ 
  - Are E and G independent? Yes!
  - P(E) = 1/6, P(G) = 1/6, P(E|G) = 1/36 [roll (1, 6)]
  - Are F and G independent? Yes!
  - P(F) = 1/6, P(G) = 1/6, P(F G) = 1/36 [roll (1, 6)]
  - Are E, F and G independent? No!
  - $P(EFG) = 1/36 \neq 1/216 = (1/6)(1/6)(1/6)$

### Generating Random Bits

- A computer produces a series of random bits,
   with probability p of producing a 1.
  - Each bit generated is an independent trial
  - E = first n bits are 1's, followed by a single 0
  - What is P(E)?
- Solution
  - P(first *n* 1's) = P(1st bit=1) P(2nd bit=1) ... P(nth bit=1) =  $p^n$
  - P(n+1 bit=0) = (1-p)
  - $P(E) = P(first \ n \ 1's) \ P(n+1 \ bit=0) = p^n \ (1-p)$

## Coin Flips

- Say a coin comes up heads with probability p
  - Each coin flip is an independent trial
- $P(n \text{ heads on } n \text{ coin flips}) = p^n$
- P(n tails on n coin flips) =  $(1 p)^n$
- P(first k heads, then n k tails) =  $p^k (1-p)^{n-k}$
- P(exactly *k* heads on *n* coin flips) =  $\binom{n}{k} p^k (1-p)^{n-k}$

#### Hash Tables

- m strings are hashed (equally randomly) into a hash table with n buckets
  - Each string hashed is an independent trial
  - E = at least one string hashed to first bucket
  - What is P(E)?
- Solution
  - $F_i$  = string i not hashed into first bucket (where  $1 \le i \le m$ )
  - $P(F_i) = 1 1/n = (n 1)/n$  (for all  $1 \le i \le m$ )
  - Event  $(F_1F_2...F_m)$  = no strings hashed to first bucket
  - $P(E) = 1 P(F_1 F_2 ... F_m) = 1 P(F_1)P(F_2)...P(F_m)$ =  $1 - ((n-1)/n)^m$
  - Similar to ≥ 1 of m people having same birthday as you

#### Yet More Hash Table Fun

- m strings are hashed (unequally) into a hash table with n buckets
  - Each string hashed is an independent trial, with probability p<sub>i</sub> of getting hashed to bucket i
  - E = At least 1 of buckets 1 to k has ≥ 1 string hashed to it
- Solution
  - F<sub>i</sub> = at least one string hashed into i-th bucket

•  $P(F_1^c F_2^c ... F_k^c) = P(\text{no strings hashed to buckets 1 to } k)$ =  $(1 - p_1 - p_2 - ... - p_k)^m$ 

• P(E) = 
$$1 - (1 - p_1 - p_2 - ... - p_k)^m$$

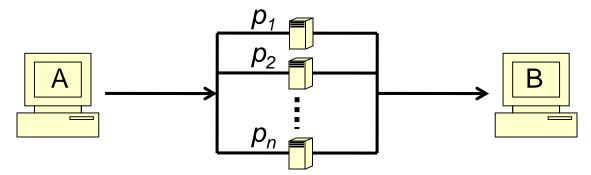
### No, Really, it's More Hash Table Fun

- m strings are hashed (unequally) into a hash table with n buckets
  - Each string hashed is an independent trial, with probability p<sub>i</sub> of getting hashed to bucket i
  - $E = Each of buckets 1 to k has <math>\geq 1$  string hashed to it
- Solution
  - F<sub>i</sub> = at least one string hashed into i-th bucket

■ P(E) = P(F<sub>1</sub>F<sub>2</sub>...F<sub>k</sub>) = 1 - P((F<sub>1</sub>F<sub>2</sub>...F<sub>k</sub>)<sup>c</sup>)  
= 1 - P(F<sub>1</sub><sup>c</sup> ∪ F<sub>2</sub><sup>c</sup> ∪ ... ∪ F<sub>k</sub><sup>c</sup>) (DeMorgan's Law)  
= 1 - P(\bigcup\_{i=1}^{k} F\_i^c) = 1 - \sum\_{r=1}^{k} (-1)^{(r+1)} \sum\_{i\_1 < ... < i\_r} P(F\_{i\_1}^c F\_{i\_2}^c ... F\_{i\_r}^c)
where 
$$P(F_{i_1}^c F_{i_2}^c ... F_{i_r}^c) = (1 - p_{i_1} - p_{i_2} - ... - p_{i_r})^m$$

# Sending Messages Through a Network

Consider the following parallel network:



- n independent routers, each with probability p<sub>i</sub> of functioning (where 1 ≤ i ≤ n)
- E = functional path from A to B exists. What is P(E)?
- Solution:

• P(E) = 1 - P(all routers fail)  
= 1 - 
$$(1 - p_1)(1 - p_2)...(1 - p_n)$$
  
=  $1 - \prod_{i=1}^{n} (1 - p_i)$