Basics on Probability

Coin Flips

- You flip a coin
 - Head with probability 0.5

- You flip 100 coins
 - How many heads would you expect

Coin Flips cont.

- You flip a coin
 - Head with probability p
 - Binary random variable
 - Bernoulli trial with success probability p
- You flip k coins
 - How many heads would you expect
 - Number of heads X: discrete random variable
 - Binomial distribution with parameters k and p

Discrete Random Variables

- Random variables (RVs) which may take on only a countable number of distinct values
 - E.g., the total number of heads X you get if you flip 100 coins
- X is a RV with arity k if it can take on exactly one value out of $\{x_1, \dots, x_k\}$
 - E.g., the possible values that X can take are 0, 1, 2, ..., 100

Probability of Discrete RV

- Probability mass function (pmf): $P(X = x_i)$
- Easy facts about pmf

$$- \sum_{i} P(X = x_{i}) = 1$$

$$- P(X = x_{i} \cap X = x_{j}) = 0 \text{ if } i \neq j$$

$$- P(X = x_{i} \cup X = x_{j}) = P(X = x_{i}) + P(X = x_{j}) \text{ if } i \neq j$$

$$- P(X = x_{1} \cup X = x_{2} \cup ... \cup X = x_{k}) = 1$$

Common Distributions

- Uniform $X \sim U[1,...,N]$
 - X takes values 1, 2, ..., N

$$-P(X=i)=1/N$$

- E.g., picking balls of different colors from a box
- Binomial $X \sim Bin(n, p)$
 - X takes values 0, 1, ..., n

$$-P(X=i) = \binom{n}{i} p^i (1-p)^{n-i}$$

- E.g., coin flips

Coin Flips of Two Persons

- Your friend and you both flip coins
 - Head with probability 0.5
 - You flip 50 times; your friend flip 100 times
 - How many heads will both of you get

Joint Distribution

- Given two discrete RVs X and Y, their joint distribution is the distribution of X and Y together
 - E.g., P(You get 21 heads AND you friend get 70 heads)

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$$\sum_{x} \sum_{y} P(X = x \cap Y = y) = 1$$

– E.g.,

$$\sum_{i=0}^{50} \sum_{j=0}^{100} P(\text{You get } i \text{ heads AND your friend get } j \text{ heads}) = 1$$

Conditional Probability

- P(X = x | Y = y) is the probability of X = x, given the occurrence of Y = y
 - E.g., you get 0 heads, given that your friend gets
 61 heads

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$$P(X = x | Y = y) = \frac{P(X = x \cap Y = y)}{P(Y = y)}$$

Law of Total Probability

• Given two discrete RVs X and Y, which take values in $\{x_1,\dots,x_m\}$ and $\{y_1,\dots,y_n\}$, we have

$$P(X = x_i) = \sum_{j} P(X = x_i \cap Y = y_j)$$
$$= \sum_{j} P(X = x_i | Y = y_j) P(Y = y_j)$$

Marginalization

Marginal Probability $P(X = x_i) = \sum_{j} P(X = x_i \cap Y = y_j)$ $= \sum_{j} P(X = x_i | Y = y_j) P(Y = y_j)$ Conditional Probability Marginal Probability

Bayes Rule

X and Y are discrete RVs...

$$P(X = x | Y = y) = \frac{P(X = x \cap Y = y)}{P(Y = y)}$$

$$P(X = x_i | Y = y_j) = \frac{P(Y = y_j | X = x_i)P(X = x_i)}{\sum_{k} P(Y = y_j | X = x_k)P(X = x_k)}$$

Independent RVs

- Intuition: X and Y are independent means that X = x neither makes it more or less probable that Y = y
- Definition: X and Y are independent iff

$$P(X = x \cap Y = y) = P(X = x)P(Y = y)$$

More on Independence

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$$P(X = x \cap Y = y) = P(X = x)P(Y = y)$$

 $P(X = x | Y = y) = P(X = x)$ $P(Y = y | X = x) = P(Y = y)$

 E.g., no matter how many heads you get, your friend will not be affected, and vice versa

Conditionally Independent RVs

- Intuition: X and Y are conditionally independent given Z means that once Z is known, the value of X does not add any additional information about Y
- Definition: X and Y are conditionally independent given Z iff

$$P(X = x \cap Y = y | Z = z) = P(X = x | Z = z)P(Y = y | Z = z)$$

More on Conditional Independence

$$P(X = x \cap Y = y | Z = z) = P(X = x | Z = z) P(Y = y | Z = z)$$

$$P(X = x | Y = y, Z = z) = P(X = x | Z = z)$$

$$P(Y = y | X = x, Z = z) = P(Y = y | Z = z)$$

Probability of Events

- X denotes an event that could possibly happen
 - E.g., X="you will fail in this course"
- P(X) denotes the likelihood that X happens, or X=true
 - E.g., what's the probability that you will fail in this course?
- Ω denotes the entire event set

$$-\Omega = \left\{ X, \overline{X} \right\}$$

The Axioms of Probabilities

- 0 <= P(X) <= 1
- $P(\Omega) = 1$
- $P(X_1 \cup X_2 \cup ...) = \sum_i P(X_i)$, where X_i are disjoint events
- Useful rules

$$-P(X_1 \cup X_2) = P(X_1) + P(X_2) - P(X_1 \cap X_2)$$
$$-P(\bar{X}) = 1 - P(X)$$

Interpreting the Axioms

