Probability and Counting, Appendix C

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Binomial Coefficients

$$\binom{n}{k} = \frac{n!}{k!(n-k)!} = \binom{n}{n-k}$$
$$(x+y)^n = \sum_{k=0}^n \binom{n}{k} x^k y^{n-k}$$
$$2^n = \sum_{k=0}^n \binom{n}{k}$$

 $(n+a)^b = O(n^b)$

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$$(n+a)^b = \sum_{k=0}^b \binom{b}{k} n^k a^{b-k}$$

$$\leq \sum_{k=0}^b \binom{b}{k} n^b a^{b-k}$$

$$= n^b \sum_{k=0}^b \binom{b}{k} a^{b-k}$$

$$= n^b \sum_{k=0}^b \binom{b}{k} 1^k a^{b-k}$$

$$= n^b (1+a)^b$$

$$= O(n^b)$$

Binomial Bounds

Sample Space

- Set of all possible things that can happen.
- ► Each thing that can happen is an **elementary event**.
- Examples:
 - ► Flip a coin twice and observe which side is up: {HH, HT, TH, TT}
 - ▶ Flip a coin twice and count the heads: {0,1,2}
 - ▶ Throw a coin down the stairs and see what step it lands on: $\{1, 2, 3, ..., n\}$, where n is the number of steps.
 - ▶ Deal two cards: $\{\{A\diamondsuit, 5\clubsuit\}, \{10\heartsuit, K\clubsuit\}, \{A\spadesuit, 3\heartsuit\}, \ldots\}$
 - ► See who wins the election: { Clinton, Sanders, Trump, Cruz...}

Events

- ▶ A subset of the sample space, S.
- Examples:
 - ► {HT, TH} ⊆ {HH, HT, TH, TT}
 ► {{A♠, A♣}, {A♡, A◊}} ⊆ {{A⋄, 5♣}, {10♡, K⋄}, {A♠, 3♡},...}
 - ► The **certain event**: *S*.
 - ► The **null event**: ∅

A probability distribution on a sample space S

Pr {} is a mapping from events to real numbers such that:

- 1. $Pr\{A\} \ge 0$
- 2. $Pr\{S\} = 1$
- 3. $Pr\{A \cup B\} = Pr\{A\} + Pr\{B\}$ whenever $A \cap B = \emptyset$
 - ▶ Theorem:

$$Pr{A \cup B} = Pr{A} + Pr{B} - Pr{A \cap B}$$

$$\leq Pr{A} + Pr{B}$$

Discrete probability distribution

▶ If *S* is finite or countably infinite.

$$\Pr\left\{A\right\} = \sum_{s \in A} \Pr\left\{s\right\}$$

▶ If *S* is finite and each elementary event has the same probability, we have **uniform probability distribution**.

$$\Pr\{s\} = 1/|S|$$

Continuous uniform distribution

- ► Each real number between a and b is equally likely.
- Not all subsets have probabilities.
- Just use intervals, and countable unions of intervals.

$$\Pr\{[c,d]\} = \frac{d-c}{b-a}$$

Conditional probability

$$\Pr\{A \mid B\} = \frac{\Pr\{A \cap B\}}{\Pr\{B\}}$$

- Probability as if B were the sample space.
- Can condition a variable on events:

$$Pr\{B\} = Pr\{B \cap A\} + Pr\{B \cap \overline{A}\}$$
$$= Pr\{A\} Pr\{B \mid A\} + Pr\{\overline{A}\} Pr\{B \mid \overline{A}\}$$

Conditioning example

- Suppose we flip a coin.
- ▶ If it's heads, we roll a 6-sided die; if it's tails an 8-sided die.
- What's the probability of getting a 6?

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$$= Pr\{A\} Pr\{B \mid A\} + Pr\{\overline{A}\} Pr\{B \mid \overline{A}\}$$

$$Pr\{6\} = Pr\{6 \cap H\} + Pr\{6 \cap \overline{H}\}$$

$$= Pr\{H\} Pr\{6 \mid H\} + Pr\{\overline{H}\} Pr\{6 \mid \overline{H}\}$$

$$= (1/2)(1/6) + (1/2)(1/8)$$

$$= 7/48$$

Bayes's theorem

$$\Pr\{A \mid B\} = \frac{\Pr\{A\} \Pr\{B \mid A\}}{\Pr\{B\}}$$

This follows easily from

$$Pr\{A \cap B\} = Pr\{A\} Pr\{B \mid A\} = Pr\{B\} Pr\{A \mid B\}$$

▶ We can combine this with a conditioning of *B* getting

$$\Pr\{A \mid B\} = \frac{\Pr\{A\} \Pr\{B \mid A\}}{\Pr\{A\} \Pr\{B \mid A\} + \Pr\{\overline{A}\} \Pr\{B \mid \overline{A}\}}$$



Bayes's theorem example

- ▶ We have a fair coin and a biased coin with Pr {H} = 2/3. We choose a coin at random and flip it twice. It comes up heads both times. What is the probability we chose the biased coin?
- ▶ Let *A* be the event of choosing a biased coin, and let *B* be the event of coming up heads twice in a row.

$$\Pr\{A \mid B\} = \frac{\Pr\{A\} \Pr\{B \mid A\}}{\Pr\{A\} \Pr\{B \mid A\} + \Pr\{\overline{A}\} \Pr\{B \mid \overline{A}\}}$$

$$= \frac{(1/2)(4/9)}{(1/2)(4/9) + (1/2)(1/4)}$$

$$= \frac{(2/9)}{(2/9) + (1/8)}$$

$$= \frac{(2/9)}{(25/72)}$$

$$= 16/25$$

Joint Probability Distribution, Marginal Distributions

)			
		2		4		$\sum_{x} f(x,y)$
	Α	0.30	0.00	0.10 0.10 0.10	0.21	0.61
Y	В	0.03	0.04	0.10	0.08	0.25
	C	0.00	0.03	0.10	0.01	0.14
$\sum_{y} f(x,y)$		0.33	0.07	0.30	0.30	

Independence

$$\Pr\left\{A\cap B\right\}=\Pr\left\{A\right\}\Pr\left\{B\right\}$$

This implies

$$\Pr\{A \mid B\} = \Pr\{A\}$$

Joint Probability Distribution, Independent Distributions

)			
		2	3			$\sum_{x} f(x, y)$
	Α	0.18	0.18 0.09 0.03	0.18	0.06	0.6
Y	В	0.09	0.09	0.09	0.03	0.3
	C	0.03	0.03	0.03	0.01	0.1
$\sum_{y} f(x,y)$		0.3	0.3	0.3	0.1	

Discrete random variables

- ► Given finite or countable *S*, a **random variable** *X* is a function from *S* to the real numbers.
- ▶ The event X = x is

$$\{s \in S : X(s) = x\}$$

▶ Therefore

$$\Pr\{X = x\} = \sum_{s \in S: X(s) = x} \Pr\{s\}$$

► The probability density function:

$$f(x) = \Pr\left\{X = x\right\}$$

▶ With two random variables X and Y, the joint probability density:

$$f(x,y) = \Pr\{X = x \text{ and } Y = x\}$$

▶ What does independence imply about the joint distribution?



Expected value

▶ The expected value or expectation or mean:

$$E[X] = \sum_{x} x \cdot \Pr\left\{X = x\right\}$$

- ▶ Denoted by μ_X or μ
- Linearity of expectation:

$$E[X + Y] = E[X] + E[Y]$$

Holds even if X and Y are not independent.

Variance

$$Var[X] = E[X - E[X]]^2$$

= $E[X^2] - E^2[X]$

- The latter can be computed in one pass, but is not as numerically stable as the first.
- ▶ If *X* and *Y* are independent:

$$Var[X + Y] = Var[X] + Var[Y]$$

- ▶ The **standard deviation** is the square root of the variance.
 - ▶ It is denoted σ_X or σ .
- ▶ The variance is denoted σ^2 .

Bernoulli trials

- ▶ Repeatedly flip a biased coin with probability *p* of heads.
- ► Each flip is independent of the others.
- ▶ Instead of heads and tails we say **success** and **failure**.

Geometric distribution

▶ With Bernoulli trials with probability of success *p*, what is the probability we try *k* times to get the first success?

$$\Pr\left\{X=k\right\}=q^{k-1}p$$

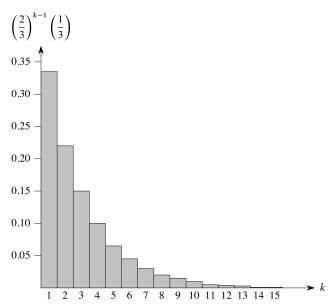
Expectation:

$$E[X] = \sum_{k=1}^{\infty} kq^{k-1}p$$
$$= 1/p$$

Variance:

$$Var[X] = q/p^3$$

Geometric distribution



Binomial distribution

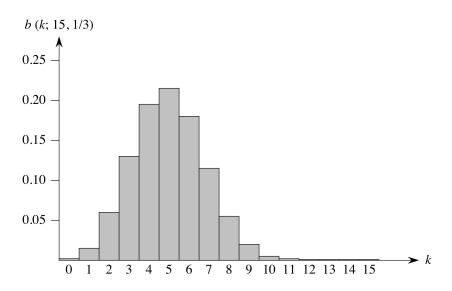
Probability of k successes occur in n Bernoulli trials with probability p:

$$\Pr\{X=k\} = \binom{n}{k} p^k (1-p)^{n-k}$$

We define a family of distributions:

$$b(k:n,p) = \binom{n}{k} p^k (1-p)^{n-k}$$

Binomial distribution



Binomial distribution expectation

$$\Pr\left\{X=k\right\} = \binom{n}{k} p^k (1-p)^{n-k}$$

Expectation:

$$E[X] = \sum_{k=0}^{n} k \cdot \Pr\{X = k\}$$
$$= \sum_{k=0}^{n} k \binom{n}{k} p^{k} q^{n-k}$$
$$= ?$$

Binomial distribution expectation

$$\Pr\left\{X=k\right\} = \binom{n}{k} p^k (1-p)^{n-k}$$

- Let X_i be an **indicator random variable** for the ith trial.
 - \triangleright X_i is 1 if the *i*th trial is a success, 0 otherwise.
- Easier math:

$$E[X] = E\left[\sum_{i=1}^{n} X_i\right]$$

$$= \sum_{i=1}^{n} E[X_i]$$
 linearity of expectation
$$= \sum_{i=1}^{n} p = np$$

Binomial distribution variance

$$\Pr\{X = k\} = \binom{n}{k} p^k (1 - p)^{n - k}$$

$$\text{Var}[X_i] = E[X_i^2] - E^2[X_i]$$

$$= E[X_i] - E^2[X_i]$$

$$= p - p^2 = pq$$

$$Var[X] = Var \left[\sum_{i=1}^{n} X_i \right]$$
$$= \sum_{i=1}^{n} Var[X_i]$$
$$= \sum_{i=1}^{n} pq = npq$$

Binomial tails

$$\Pr\{X \ge k\} = \sum_{i=k}^{n} b(i:n,p)$$

$$\le \binom{n}{k} p^k$$

$$\Pr\{X \le k\} = \sum_{i=0}^{k} b(i:n,p)$$

$$\le \binom{n}{k} (1-p)^{n-k}$$