

1 Basic probability

1.1 Sample spaces and events

- ◆ A "sample space," is the set of all possible outcomes of an experiment. The sample points are the elements in a sample space.
- ◆ An "event" is a subset of the sample space
- ◆ Two events are mutually exclusive if ...

1.2 Assignment of probabilities

- ◆ Probability axioms
- ◆ $P(A \cup B) = \dots$
- ◆ $P(A^c) = \dots$

1.3 Simulation of events on the computer

- ◆ How do you simulate a coin tossing on a computer?

1.4 Counting techniques

- ◆ Multiplication rule
- ◆ Permutation of n distinct items taken r at a time
- ◆ Combination: an unordered arrangement of r items selected from n distinct items

1.5 Conditional probability

- ◆ $P(B|A) = \dots$
- ◆ Multiplication rule $P(A \cap B) = \dots$
- ◆ Law of total probability: if E_1, \dots, E_k are a partition of the sample space, then $P(A) = \dots$

1.6 Independent event

- ◆ Events A and B are independent if \dots
- ◆ Events E_1, E_2, \dots are mutually independent if \dots

2 Discrete random variables

2.1 Random variables

- ◆ The probability mass function $p(x)$ satisfies \dots

- ◆ CDF $F(x) = P(X \leq x)$

2.2 Joint distributions and independent random variables

- ◆ The joint probability mass function $p(x,y) = P(X=x, Y=y)$
- ◆ The marginal probability mass function $p_x(x) = \dots$
- ◆ The conditional probability mass function $P(X=x|Y=y)$
- ◆ Discrete random variables X and Y are independent if \dots
- ◆ Discrete random variables X_1, X_2, \dots, X_n are mutually independent if \dots

2.3 Expected values

- ◆ $E(X) = \dots$
- ◆ If Y is a function of a random variable X, then $E(Y) = E(Q(X)) = \dots$
- ◆ $E(a+bX) = \dots$
- ◆ If $Y = g_1(X) + \dots + g_2(X)$, then $E(Y) = \dots$

2.4 Variance and standard deviation

- ◆ $VAR(X) = \dots = \dots$
- ◆ $STD(X) = \dots$
- ◆ Chebychev's inequality
- ◆ $VAR(a+bX) = \dots$
- ◆ $STD(a+bX) = \dots$

2.5 Sampling and simulation

- ◆ Random variables X_1, \dots, X_n are a random sample if they are i.i.d.
- ◆ Empirical probability distribution; empirical probability mass function

2.6 Sample statistics

- ◆ Sample mean $\bar{X} = \dots$
- ◆ Sample variance $S^2 = \frac{1}{n} \sum_{i=1}^n (X_i - \bar{X})^2$
- ◆ Sample standard deviation $S = \dots$

2.7 Expected values of jointly distributed random variables and the law of large numbers

- ◆ $E(Q(X,Y)) = \dots$

- ◆ $E(aX+bY) = \dots$
- ◆ $E(a_1X_1 + a_2X_2 + \dots + a_nX_n) = \dots$
- ◆ If X and Y are independent, $E(XY) = \dots$
- ◆ If X and Y are independent, $VAR(aX+bY) = \dots$
- ◆ If X_1, \dots, X_n are independent, $VAR(a_1X_1 + \dots + a_nX_n) = \dots$
- ◆ If X_1, \dots, X_n iid with $E(X_i) = \mu$ and $VAR(X_i) = \sigma^2$, then $E(\bar{X}), VAR(\bar{X}), STD(\bar{X}) = \dots$
- ◆ Weak law of large number: under the i.i.d. setup as above, $\lim_{n \rightarrow \infty} P(|\bar{X} - \mu| > \epsilon) = 0$ for any $\epsilon > 0$.

2.8 Covariance and correlation

- ◆ $COV(X,Y) = \dots$
- ◆ $CORR(X,Y) = \dots$
- ◆ $VAR(aX+bY) = \dots$
- ◆ Bivariate random sample
- ◆ The sample covariance $cov(X,Y)$
- ◆ The sample correlation $corr(X,Y)$

2.9 Conditional expected values

- ◆ $E(Y|X=x) = \dots$
- ◆ $E(Y) = E(E(X|Y))$
- ◆ $VAR(Y|X=x) = \dots$
- ◆ $VAR(Y) = E[\dots] + VAR[\dots]$

3 Special discrete random variables

3.1 Binomial random variable

- ◆ A binomial random variable is \dots
- ◆ For $Y \sim \text{bin}(n,p)$, $p(y) = \dots$. Also, $E(Y) = \dots$ and $VAR(Y) = \dots$

3.2 Geometric and negative binomial random variables

- ◆ The geometric random variable is \dots . We write it as $X \sim \text{Geo}(p)$. In that case, $p(x) = \dots$, $E(X) = \dots$, $VAR(X) = \dots$
- ◆ Given an integer $k > 1$, the negative binomial random variable is \dots . We write it as $Y \sim \text{Nbinom}(k,p)$. In that case, $p(x) = \dots$, $E(X) = \dots$, $VAR(X) = \dots$

3.3 Hypergeometric random variables

- ◆ Consider a jar containing the total of N balls, r of which red, the remaining white. Randomly selecting n balls from the jar WOR, the number of red balls in the sample $X \sim \text{Hypergeo}(N,r,n)$. In that case, $p(x) = \dots$, $E(X) = \dots$, $VAR(X) = \dots$
- ◆ Approximately binomial with $p = \dots$

3.4 Multinomial random variables

- ◆ Let Y_1, \dots, Y_k denote the number of times the mutually exclusive outcomes C_1, \dots, C_k occur in n iid trials. Let $p_i = P(C_i)$ for $i=1, \dots, k$. Then we say $(Y_1, Y_2, \dots, Y_k) \sim \text{multinom}(n, p_1, \dots, p_k)$ and $P(Y_1=y_1, \dots, Y_k=y_k) = \dots$ where \dots
- ◆ $E(Y_i) = \dots$ and $VAR(Y_i) = \dots$
- ◆ $COV(Y_s, Y_t) = \dots$ if $s \neq t$.

3.5 Poisson random variables

- ◆ $X \sim \text{Poisson}(\mu)$ if $p(x) = \dots$
- ◆ $E(X) = \dots$ and $VAR(X) = \dots$
- ◆ If $X \sim \text{bin}(n,p)$, $P(X=x) \sim p(x)$ of $\text{Poisson}(\mu)$ if \dots and $\mu = \dots$

3.6 Moments and moment-generating functions (MGF)

4 Markov chains

4.1 Introduction: modeling a simple queuing system

- ◆ A queue is a waiting line
- ◆ Consider the following queuing system:
 - A company has an assigned technician to handle service for 5 computers.
 - Each computer independently fails with probability .2 during any day
 - The technician can fix one computer per day.
 - If a machine breaks down, it will be fixed the day it fails provided there is no backlog. If there is a backlog, it will join a service queue and wait until the technician fixes those ahead of it.