

# Statistics 430

## Summary, Spring 2003

### Probability

- Counting methods, sample space  $S$ , events;  $P(E) = \#E/\#S$
- Unions  $P(A \cup B) = P(A) + P(B) - P(A \cap B)$
- Conditional probability  $P(A|B) = P(A \cap B)/P(B)$
- Intersections  $P(A \cap B) = P(A|B)P(B) = P(B|A)P(A)$
- Complements  $P(A^c) = 1 - P(A)$
- Law of total probability  $P(A) = \sum_j P(A|B_j)P(B_j)$ ,  $\cup B_j = B$
- Bayes  $P(A|B) = P(B|A)P(A)/P(B)$
- Independent vs. dependent events

### Random variables

- Properties of PDF and CDF
- Mean, variance, and standard deviation of random variable
- Marginal, conditional, joint distributions
- Relationships that connect the different types  
(*e.g.*, Poisson process and exponential r.v.)
- Independence of random variables; sums of random variables
- Markov chain (sequence of random variable)

		Mean	Variance
Bernoulli	$p(x) = p^x(1-p)^{1-x}$ , $x = 0, 1$	$p$	$p(1-p)$
Binomial	$p(x) = \binom{n}{x}p^x(1-p)^{n-x}$ , $x = 0, \dots, n$	$np$	$np(1-p)$
Geometric	$p(x) = (1-p)^{x-1}p$ , $x = 1, 2, \dots$	$1/p$	$(1-p)/p^2$
Poisson	$p(x) = e^{-\lambda}\lambda^x/x!$ , $x = 0, 1, \dots$	$\lambda$	$\lambda$
Uniform	$f(x) = 1$ , $0 \leq x \leq 1$	$1/2$	$1/12$
Exponential	$f(x) = \lambda e^{-\lambda x}$ , $x \geq 0$	$1/\lambda$	$1/\lambda^2$
Normal	$f(x) = e^{-(x-\mu)^2/(2\sigma^2)}/\sqrt{2\pi\sigma^2}$ , $-\infty < x < \infty$	$\mu$	$\sigma^2$

## Expected value

- Definition of  $E X$  as a weighted sum, with weights given by  $p(x)$ .
- Expected value of a sum = sum of expected values  
$$E(a + bX + cY) = a + b E(X) + c E(Y)$$
- Covariance, correlation and variance of sums; portfolios  
$$\text{Cov}(X, Y) = E((X - EX)(Y - EY)) = E(XY) - (EX)(EY)$$
- Variance of a sum = sum of variances *if* all covariances are zero, otherwise  
$$\text{Var}(aX + bY) = a^2 \text{Var}(X) + b^2 \text{Var}(Y) + 2ab \text{Cov}(X, Y)$$
- Expected value of indicator variable,  $E I = P(I = 1)$
- Conditional expected value,  $E(X|Y)$ ; use in prediction
- Marginal from conditional  $E X = \sum_y E(X|Y = y)P(Y = y)$

## Inequalities and asymptotics

- Bonferroni  $P(\cup_i A_i) \leq \sum_i P(A_i)$
- Jensen  $f(E(X)) \leq E(f(x))$  if  $f(x)$  is convex (*e.g.*,  $f(x) = x^2$ )
- Markov  $P(X \geq k) \leq EX/k$  if  $X$  is a non-negative r.v.
- Chebyshev  $P(|X - EX|/SD(X) \geq k) \leq 1/k^2$
- Central limit theorem  $\sum_i X_i \rightarrow \text{Normal}$
- Weak law of large numbers; averages get close their expected value