

STAT 340: FALL 2009: Information sheet for Final Exam: 2:30-4:20, Wed. Dec. 16

(A Normal probability table is on the back of this sheet.)

1. Permutations and combinations

There are $n! = \prod_{i=1}^n i = 1.2.3.4.\dots n$ permutations of n objects.

There are $\binom{n}{k} = n!/(k!(n-k)!)$ ways of choosing k objects from n .

2. Joint and conditional probabilities

If C and D are any events: $P(C \cup D) = P(C) + P(D) - P(C \cap D)$.

The conditional probability of C given D is $P(C | D) = P(C \cap D) / P(D)$.

C and D are independent if $P(C \cap D) = P(C).P(D)$.

3. Laws and theorems

Suppose E_1, \dots, E_k is a partition of Ω . That is $E_i \cap E_j$ is empty, and $E_1 \cup E_2 \cup \dots \cup E_k = \Omega$.

The law of total probability states that: $P(D) = \sum_{j=1}^k P(D \cap E_j) = \sum_{j=1}^k P(D | E_j) P(E_j)$

Bayes' Theorem states that: $P(E_i | D) = P(D | E_i) P(E_i)/P(D)$

4. Random variables and distributions

	discrete (mass)	continuous (density)
Probability mass/density function	pmf: $P(X = x) = p_X(x)$ defined for all x with $p_X(x) > 0$	pdf: $f_X(x)$ $-\infty < x < \infty$
Cumulative distrib. func. cdf: $P(X \leq x)$	$F_X(x) = \sum_{y \leq x} p_X(y)$	$F_X(x) = \int_{-\infty}^x f_X(y) dy$
Expectation $E(X)$	$\sum_x x p_X(x)$	$\int_{-\infty}^{\infty} x f_X(x) dx$ provided the sum/integral converges.
Result: $E(g(X))$	$\sum_x g(x) p_X(x)$	$\int_{-\infty}^{\infty} g(x) f_X(x) dx$
Variance: $E((X - (E(X)))^2) = E(X^2) - (E(X))^2$		

5. Standard distributions: p_X or f_X

	values	mean $E(X)$	variance
(a) Binomial; $B(n, p)$; index n , parameter p ;			
$P(X = k) = \binom{n}{k} p^k (1-p)^{n-k}$	$k = 0, 1, 2, \dots, n$	np	$np(1-p)$
(b) Geometric; $G(p)$; parameter p ;			
$P(X = k) = p(1-p)^{k-1}$	$k = 1, 2, 3, 4, \dots$	$1/p$	$(1-p)/p^2$
(c) Negative Binomial; $NegB(r, p)$			
$P(X = k) = \binom{k-1}{r-1} p^r (1-p)^{k-r}$	$k = r, r+1, r+2, \dots$	r/p	$r(1-p)/p^2$
(d) Poisson; $Po(\mu)$; parameter μ ;			
$P(X = k) = \exp(-\mu) \mu^k / k!$	$k = 0, 1, 2, 3, \dots$	μ	μ
(e) Hypergeometric (N, m, n);			
$P(X = k) = \binom{m}{k} \binom{N-m}{n-k} / \binom{N}{n}$	$k = 0, \dots, n$ $k \geq m+n-N, k \leq m$	nm/N	—
(f) Normal, $N(\mu, \sigma^2)$;			
$f_X(x) = (1/\sqrt{2\pi\sigma^2}) \exp(-(x-\mu)^2/2\sigma^2)$	$-\infty < x < \infty$	μ	σ^2
(g) Uniform on interval (a, b) ; $U(a, b)$;			
$f_X(x) = 1/(b-a)$	$a < x < b$	$(a+b)/2$	$(b-a)^2/12$
(h) Exponential, $E(\lambda)$, rate parameter λ ;			
$f_X(x) = \lambda \exp(-\lambda x)$	$0 \leq x < \infty$	$1/\lambda$	$1/\lambda^2$

6. Summation of series:

(a) $\sum_{i=0}^{n-1} x^i = (1 + x + x^2 + \dots + x^{n-1}) = (1 - x^n)/(1 - x)$

(b) $\sum_{i=0}^{\infty} x^i/i! = (1 + x + x^2/2 + x^3/6 + \dots) = \exp(x)$

7. Here is a copy of the Normal distribution probability table from Kelly P.601

Tabulated is the cdf $P(X \leq x)$ for a random variable X which is Normal with mean 0 and variance 1.