

Probability Exam sheet by David A Wheeler

GMU, MATH 351, Spring 1986, Dr. Bolstein, for textbook  
 "Introductory Probability and Statistical Applications, 2<sup>nd</sup> edition"  
 by Paul L. Meyer

$$\int u^n du = \frac{u^{n+1}}{n+1} + C$$

$a \in B$   $\Leftrightarrow$  a is member of B

Ch!

$A \cup B$  A or B

$A \subset B \Leftrightarrow A$  is a subset of B  $\int \frac{du}{u} = \ln|u| + C$

$A \cap B$  A and B  $\int \cos u du = \sin u + C$

$$uv = \int u dv + \int v du ; \int a^u du = \frac{1}{\ln a} a^u + C ; \int \sin u du = -\cos u + C ; \int \tan u du = \ln|\sec u| + C$$

$$\int \frac{du}{\sqrt{a^2 - u^2}} = \sin^{-1} \frac{u}{a} + C$$

$$P(A) = 1 - P(\bar{A})$$

If  $A \subset B$ ,  $P(A) \leq P(B)$

$$\int \frac{du}{a^2 + u^2}$$

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

$$\int \frac{du}{u \sqrt{u^2 - a^2}}$$

$$P(A \cup B \cup C) = P(A) + P(B) + P(C) - P(A \cap B) - P(A \cap C) - P(B \cap C) + P(A \cap B \cap C)$$

$$= \frac{1}{a} \tan^{-1} \frac{u}{a} + C \text{ if } = \text{ly likely outcomes } P(A) = \frac{\# \text{ possible favorable}}{\# \text{ possible}}$$

$$\int \frac{du}{u^2 - a^2}$$

$$\text{Factorial } n! = (n)(n-1) \cdots (1)$$

$$= \frac{1}{a} \sec^{-1} \frac{u}{a} + C$$

Permutation:  $n$  different objects, choose  $r$  and permute (order important)

$$\int \sin u du = -\cos u + C$$

$$nPr = \frac{n!}{(n-r)!}$$

$$\int \cos u du = \sin u + C$$

Combination: order unimportant  $\binom{n}{r} = \frac{n!}{r!(n-r)!} = \binom{n}{n-r} = \binom{n}{r}$

$$\frac{d}{dx} U^c = cU \frac{dU}{dx}$$

Hypergeometric: Given  $N$  objects, choose  $n$  of them. Probability of

$$\frac{d}{dx} \ln V = u \frac{dv}{dx} + v \frac{du}{dx}$$

$\sum_1 r_1$  A's and  $\sum_2 r_2$  B's, where  $r_1 + r_2 = N$ ,  $\sum_1$  A's and  $\sum_2$  B's is

$$\frac{d}{dx} \frac{u}{v} = \frac{v \frac{du}{dx} - u \frac{dv}{dx}}{v^2}$$

if not all different, but

$$\frac{d}{dx} \sin u = \cos u \frac{du}{dx}$$

$$\frac{d}{dx} \tan u = \sec^2 u \frac{du}{dx}$$

$$n_1 \text{ of 1, } n_2 \text{ of 2, } n_1 + n_2 = n; \text{ Permutation} = \frac{n!}{n_1! n_2! \cdots n_k!} \quad \frac{n!}{\binom{r_1}{s_1} \binom{r_2}{n-s_1}} \quad \frac{\binom{r_1}{s_1} \binom{r_2}{n-s_1}}{\binom{N}{n}}$$

$$\frac{d}{dx} \frac{f(u)}{g(x)} = \frac{g(x) f'(u) - f(u) g'(x)}{g(x)^2}$$

$$P(B|A) = \frac{P(A \cap B)}{P(A)}$$

$$\frac{d}{dx} \frac{u}{v} = \frac{v \frac{du}{dx} - u \frac{dv}{dx}}{v^2}$$

Multiplication Theorem  
 Prob. of B, given A occurred. So  $P(A \cap B) = P(A)P(B|A)$

$$\frac{d}{dx} \frac{u}{v} = \frac{v \frac{du}{dx} - u \frac{dv}{dx}}{v^2}$$

$$\text{Baye's Theorem (find cause)} P(B_i|A) = \frac{P(A|B_i)P(B_i)}{\sum_{j=1}^k P(A|B_j)P(B_j)}$$

where one & only one  $B_i$  can occur (partitioned)

Law of Total Prob:  $P(A) = P(A|B_1)P(B_1) + P(A|B_2)P(B_2) + \cdots$  if  $B_1, B_2, \dots$  partition

Independant Events: iff  $P(A \cap B) = P(A)P(B)$

$A, B, C$  mutually independant iff  $P(A \cap B) = P(A)P(B)$ ,

$P(A \cap C) = P(A)P(C)$ ,  $P(B \cap C) = P(B)P(C)$ ,  $P(A \cap B \cap C) = P(A)P(B)P(C)$

Sphere Surface Area =  $4\pi r^2$ ; Sphere Volume =  $\frac{4}{3}\pi r^3$

$$\cos x = \sum$$

$$e^x = \sum_{n=0}^{\infty} \frac{x^n}{n!}$$

$$\sin x = \sum_{i=1}^{\infty} (-1)^{i-1} \frac{x^{2i-1}}{(2i-1)!}$$

ch 4

Geometric Series  $1+r+r^2+\dots = \frac{1}{1-r}$  if  $|r| < 1$ .

BINOMIAL distribution:  $X$  be binomial variable

Given  $n$  [independant] trials, prob  $p$  that event is true,

$x = \#$  of occurrences of event in  $n$  trials

$$P[X=x] = \binom{n}{x} p^x q^{n-x} \text{ where } q = 1-p$$

If  $x$  left as variable this is probability mass function (pmf)

pmf = largest value is at  $\begin{cases} x = (n+1)p - 1 \text{ AND } (n+1)p \text{ if } (n+1)p \text{ integer} \\ x = \text{INT}((n+1)p) \text{ if } (n+1)p \text{ not integer} \end{cases}$

pdf: probability density function (for continuous functions) = pdf of  $X$

$$f(x) \geq 0 \text{ for all } x, \int_{-\infty}^{\infty} f(x) dx = 1, P(a \leq X \leq b) = \int_a^b f(x) dx$$

cdf = cumulative distribution function =  $F(x) = P(X \leq x)$ . If  $X$  discrete,

$$F(x) = \sum p(x_j) \text{ where } x_j \leq x; \text{ continuous, } F(x) = \int_{-\infty}^x f(s) ds \text{ and pdf } f(x) = F'(x) \text{ (cdf)}$$

if  $X$  cont. RV, pdf  $f$ , H is cont function,  $\forall H(X)$ , to find pdf  $g(y)$ : 1. find cdf

of  $Y$   $G(y) = P(Y \leq y)$  by subst.  $y$  2. pdf  $g(y) = \frac{dG(y)}{dy}$  find where  $f(x) \geq 0$ .

$$\text{Ex: } f(x) = \begin{cases} 2x & 0 \leq x \leq 1 \\ 0 & \text{else} \end{cases}; H(x) = 3x + 1; G(y) = P(Y \leq y) = P(3X + 1 \leq y) = \int_0^y 2x dx; g(y) = G'(y)$$

( $X, Y$ ): discrete: prob function  $p(x_i, y_j) \geq 0, \sum \sum p(x_i, y_j) = 1$

continuous: joint prob. density  $f(x, y) \geq 0, \int \int_R f(x, y) dx dy = 1$

$$\text{cdf } F(x, y) = P(X \leq x, Y \leq y); \frac{\partial^2 F(x, y)}{\partial x \partial y} = f(x, y)$$

Marginal prob. functions:  $g(x) = \int_{-\infty}^{\infty} f(x, y) dy, h(y) = \int_{-\infty}^{\infty} f(x, y) dx$

Uniformly distributed if  $f(x, y) = \begin{cases} \frac{1}{\text{area}} & \text{if } (x, y) \in R, 0 \text{ else} \end{cases}$

Conditional pdf of  $X$  for  $Y=y$   $g(x|y) = \frac{f(x, y)}{h(y)}$  ( $h(y) \neq 0$ )

$X \& Y$  independent iff  $p(x_i, y_j) = p(x_i)p(y_j)$  always,  $f(x, y) = g(x)h(y)$ .

<sup>INDEPEND.</sup> if  $W = XY$ , pdf  $p(w) = \int_{-\infty}^{\infty} g(u) h\left(\frac{w}{u}\right) \left|\frac{1}{u}\right| du \quad \begin{cases} \text{independent} \\ \text{continuous} \end{cases}$

if  $W = XY$ , pdf  $g(z) = \int_{-\infty}^{\infty} g(vz) h(v) |v| dv$

if  $(X, Y)$  continuous, joint pdf:  $f$ , let  $Z = H_1(X, Y), W = H_2(X, Y)$ , then joint pdf

of  $(Z, W)$  is  $k(z, w) = f(G_1(z, w), G_2(z, w)) |J(z, w)|$  where  $G_1$  &  $G_2$  are

found by  $X = G_1(Z, W); Y = G_2(Z, W)$ ; and  $J(z, w) = \begin{vmatrix} \frac{\partial x}{\partial z} & \frac{\partial x}{\partial w} \\ \frac{\partial y}{\partial z} & \frac{\partial y}{\partial w} \end{vmatrix}$

Chebyshev's:  $P[X-\mu| \geq k\sigma] \leq \frac{1}{k^2}$ ;  $P[|X-\mu| \geq \epsilon] \leq \frac{1}{\epsilon^2} E[(X-\mu)^2]$

Ch 7)  $E(X) = \sum_{\text{discrete}} x_i p(x_i) \stackrel{\text{pmf}}{=} \sum H(y_i) p(y_i) \quad \text{if } X = H(Y)$   
 $\int_{-\infty}^{\infty} x f(x) dx \stackrel{\text{pdf}}{=} \int_{-\infty}^{\infty} H(y) f(y) dy$

If  $Z = H(X, Y)$ ,  $E(Z) = \sum_{i=1}^c \sum_{j=1}^c H(x_i, y_j) p(x_i, y_j)$  joint pdf  
 $\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} H(x, y) f(x, y) dx dy$

If  $X = C$  (constant),  $E(X) = C$ ,  $V(X) = 0$ ;  $E(CX) = CE(X)$ ;  $E(X+Y) = E(X) + E(Y)$

If independent,  $E(XY) = E(X)E(Y)$ ;  $V[\sum X_i] = \sum V[X_i] + 2 \sum_{i < j} \text{Cov}[X_i, X_j]$

Covariance  $\text{Cov}(X, Y) = E[(X - \mu_X)(Y - \mu_Y)]$ ; Variance  $V[X+Y] = V[X] + V[Y] + \text{Cov}[X, Y]$

$V[X] = E[X - E(X)]^2 = E[X^2] - (E[X])^2 = E[(X-a)^2] - [E(X)-a]^2$  any  $a$

$V[X+a] = V[X]$ ;  $V[CX] = C^2 V[X]$ ; if independent  $V[X+Y] = V[X] + V[Y]$

Approx: If  $\mu = E(X)$ ,  $\sigma^2 = V[X]$ ,  $Y = H[X]$ ;  $E[Y] \approx H(\mu) + \frac{1}{2} H''(\mu) \sigma^2$

and  $V[Y] \approx [H'(\mu)]^2 \sigma^2$

\* Correlation coeff  $\rho_{XY} = \frac{E\{(X-E(X))(Y-E(Y))\}}{\sqrt{V(X)V(Y)}} = \frac{E(XY) - E(X)E(Y)}{\sqrt{V(X)V(Y)}}$   
 $(-1 \leq \rho_{XY} \leq 1)$

If  $X \& Y$  independent,  $\rho_{XY} = 0$  (converse only necessarily true for Bernoulli)

If  $Y = aX + b$ ,  $\rho = \pm 1$  (converse true)

$E(X|Y) = \sum_{i=1}^c x_i p(x_i|Y)$   
 $\int_{-\infty}^{\infty} x g(x|Y) dx$

$E[E(X|Y)] = E[X]$  and  $E[E(Y|X)] = E(Y)$

If  $X \& Y$  independent,  $E[X|Y] = E[X]$ ; regression curve of  $Y$  on  $X \Rightarrow E(Y|X)$  vs.  $X$ .

If regression of  $Y$  on  $X$  is linear,  $\rho = \text{coeff}$  and  $E(Y|X) = \mu_Y + \rho \frac{\sigma_Y}{\sigma_X} (X - \mu_X)$

If regression of  $X$  on  $Y$  linear,  $E(X|Y) = \mu_X + \rho \frac{\sigma_X}{\sigma_Y} (Y - \mu_Y)$

Ch 8  
 ① Discrete Binomial:  $X \sim \text{Bin}(n, p) = \# \text{ successes in } n \text{ independent trials w prob } p$ ;  $q = 1-p$ ,  
 $P[X=k] = \binom{n}{k} p^k q^{n-k}$ ,  $E[X] = np$ ,  $V[X] = npq$ ;  $P[X] \approx \text{Pois}(\lambda = np)$

② Bernoulli: Binomial of  $n=1$ ,  $V[X] = pq$ ,  $E[X] = p$   $X \sim \{0, 1\}$ ,  $E[X^2] = \frac{2}{p^2} - 1$

③ Geometric:  $X \sim \text{Geom}(p) = \# \text{ trials until success}$ .  $p_X(k) = p q^{k-1}$ ,  $E[X] = 1/p$ ,  $V[X] = \frac{1-p}{p^2}$   
 MEMORYLESS!

(4) Pascal ("Negative Binomial")  $Y \sim \text{Pas}(r, p)$  = # trials until  $r^{\text{th}}$  success.

$$p_Y(k) = P[Y=k] = P[\text{exactly } k \text{ trials for } r^{\text{th}} \text{ success}] = \binom{k-1}{r-1} p^r q^{k-r} \quad k=r, r+1, \dots$$

$$E[Y] = \frac{r}{p}, V[Y] = r \left( \frac{1-p}{p^2} \right); P[Y \leq n] = P[X \geq r]$$

(5) Hypergeometric: simple random sample w/out replace. of  $n$  items,  $N$  population,

$$p = \text{proportion "successful" in } N: p_X(k) = \frac{\binom{Np}{k} \binom{N-Np}{n-k}}{\binom{N}{n}} \quad k=0, 1, \dots, n$$

population correction factor

$$E[X] = np; V[X] = npq \frac{\binom{N-n}{k}}{\binom{N}{n}}$$

if  $N \gg 20n$ ,  $p_X(k) \approx \binom{n}{k} p^k q^{n-k}$

(6) Multinomial Distribution (multiple outcomes), generalization of Binomial,

$$\text{pmf } p(n_1, n_2, \dots, n_k) = \frac{n!}{n_1! n_2! \dots n_k!} p_1^{n_1} p_2^{n_2} \dots p_k^{n_k} ; \sum n_i = n$$

$$\text{Poisson Distribution } X \sim \text{Pois}(\alpha) \quad p_X(k) = e^{-\alpha} \frac{\alpha^k}{k!} \quad (k=0, 1, 2, \dots)$$

$$E[X] = \alpha, V[X] = \alpha$$

(Ch 9) (1) Uniform  $X \sim \text{Unif}(a, b)$ ; pmf  $f(x) = \begin{cases} \frac{1}{b-a} & \text{if } a < x < b; \\ 0 & \text{else} \end{cases}$

Contin. F.V.  $E[X] = \frac{1}{2}(a+b), V[X] = \frac{(b-a)^2}{12}$

(2) Normal distribution  $X \sim \text{Norm}(\mu, \sigma^2)$ ;  $f(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2}\left[\frac{(x-\mu)}{\sigma}\right]^2}$

$$E[X] = \mu, V[X] = \sigma^2, P[a \leq X \leq b] = P\left[\frac{a-\mu}{\sigma} \leq Z \leq \frac{b-\mu}{\sigma}\right] = \Phi\left(\frac{b-\mu}{\sigma}\right) - \Phi\left(\frac{a-\mu}{\sigma}\right)$$

(3) Standard Normal distribution  $Z \sim \mathbf{N}(\mu=0, \sigma=1)$  has pdf  $\varphi = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^2}$

$$\text{cdf } \Phi(Z) = P[Z \leq z] = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^z e^{-\frac{1}{2}t^2} dt \quad \text{see table}$$

(Thm) If  $X \sim N(\mu, \sigma^2)$  then  $Y = aX + b \sim N(a\mu + b, a^2\sigma^2)$ ; if  $X \sim N(\mu, \sigma^2)$ ,  $Z = \frac{X-\mu}{\sigma} \sim N(0, 1)$

Time required until first occurrence of A if  $X \sim \text{Exp}(\alpha)$  if pdf  $f(x) = \begin{cases} \alpha e^{-\alpha x} & \text{if } x > 0, \\ 0 & \text{else} \end{cases}$  and  $\alpha > 0$ .

4-Pois(); No memory. Cdf  $F(x) = \begin{cases} 1 - e^{-\alpha x} & \text{if } x \geq 0, \\ 0 & \text{else} \end{cases}$ ;  $E[X] = \frac{1}{\alpha}$ ;  $V[X] = \frac{1}{\alpha^2}$

Ch 12.  $S_A = \frac{n_A}{n}$  where  $n_A = \# \text{ of actual successes, trials, } n = \# \text{ of binomial trials}$ ;  $P[|S_A - p| \leq \epsilon] \geq 1 - \frac{8(1-p)}{n\epsilon^2}$  Large  $n$ , large  $\epsilon$ .

Normal Approx:  $X \sim \text{Bin}(n, p)$ ;  $Y = \frac{X - np}{\sqrt{np(1-p)}} \sim N(0, 1)$  so  $\lim_{n \rightarrow \infty} P[Y \leq y] = \Phi(y)$  close to  $\frac{1}{2}$ .

Corrections for continuity:  $P(X=k) \approx P(k-\frac{1}{2} \leq X \leq k+\frac{1}{2})$ ;  $P[a \leq X \leq b] \approx [a-\frac{1}{2} \leq X \leq b+\frac{1}{2}]$

Central Limit:  $Z_n = \frac{X - \sum \mu_i}{\sqrt{\sum \sigma_i^2}} \sim N(0, 1)$ ; if each  $S = \sum X_i$ ,  $T_n = \frac{S - n\mu}{\sqrt{n}\sigma}$

$$P[T_n \leq t] = \Phi(t)$$