

# Cheatography

## MAT2377 Cheat Sheet by t847222 via cheatography.com/80028/cs/19345/

Classical + Relative	Normal Distribution	PMF	Sample Variance
$P(A) = N(A)/N(S)$ $P(A) = f(A)/n$	$f(x) = 1/\sqrt{2\pi\sigma^2} e^{-(x-\mu)^2/(2\sigma^2)}$ , $-\infty < x < \infty$ $X \sim N(\mu, \sigma^2)$ $E[X] = \mu$ $V[X] = \sigma^2$	$f(x) = P(X=x)$	$s^2 = ((\sum x_i^2 - n\bar{x}^2)/(n-1))$
Conditional	Sample Mean	Variance	CLT
$P(A B) = P(A \cap B)/P(B)$ A given B	$\bar{x} = \sum x_i/n$	$\sigma^2 = V[X] = \sum x^2 f(x) - E[X]^2$ Standard deviation = $\sqrt{V[X]}$	$Z = (X-\mu)/(\sigma/\sqrt{n})$ $X \sim N(\mu, \sigma^2/n) \Rightarrow Z \sim N(0,1)$
CDF	Box Plot	Joint Properties	Confidence Level
$F(x) = P(X \leq x) = \sum f(x_i)$	<p>Describe histogram: skewness, uni/bimodal</p>	$E[g(X,Y)] = \sum x \sum y g(x,y) p(x,y)$ $E[X] = \sum x p(x)$ $E[Y] = \sum y p(y)$ $E[X+Y] = E[X] + E[Y]$ $\text{Cov}[X,Y] = (\sum x \sum y x y p(x,y)) - E[X]E[Y]$ $V[X+Y] = V[X] + V[Y] + 2\text{Cov}[X,Y]$	$\alpha = P(Z > z_{-\alpha}) = 1 - \Phi(z)$ $\mu \in [\bar{x} - E, \bar{x} + E]$ $\sigma^2 \text{ known: } E = z_{-\alpha/2} \sigma / \sqrt{n}$ $\sigma^2 \text{ unknown: } T = (\bar{X} - \mu) / (S/\sqrt{n}) \sim T(n-1)$ $P(T > t_{-\alpha/2}) = \alpha$ $z_{-\alpha} = t_{-\alpha/2}$ $E = t_{-\alpha/2} \sigma / \sqrt{n}$ $\sigma^2 \text{ unknown, } n \geq 40: (\bar{X} - \mu) / (S/\sqrt{n}) \sim N(0,1)$ $E = z_{-\alpha/2} s / \sqrt{n}$ $n \geq ((z_{-\alpha/2} \sigma) / E)^2$
Joint PMF	Constructing Confidence Interval	Poisson Distribution	Combinations
$p(x,y) = P(X=x, Y=y) = P(\{X=x\} \cap \{Y=y\})$	$P = Y/n$ $Y \sim b(n,p)$ $Z = (P-p)/\sqrt{(p(1-p)n)} \sim N(0,1)$ $E = z_{-\alpha/2} \sqrt{(p(1-p)/n)}$	$X = \# \text{ of events in time } [0,1]$ $p(x) = e^{-\mu} \mu^x / x!$ , for $x=0,1,\dots$ $X \sim P(\mu)$ $E[X] = V[X] = \mu$ $\text{Approximation: binomial } f(x) \approx p(x), \mu = np$ $\text{Process: between } [0,t], \mu = \lambda t$	$n = n_1 * \dots * n_k$ $nCr = (^n r) = n! / r!(n-r)!$ Order doesn't matter
Geometric Distribution	Sample Correlation	Continuous Uniform Distribution	Multiplication Rule
$X = \# \text{ of trials until 1st success}$ $X \sim g(p)$ $f(x) = (1-p)^{x-1} p$ , for $x=1,2,\dots$ $F(x) = 1 - (1-p)^x$ , for $x=1,2,\dots$ $E[X] = 1/p$ $V[X] = (1-p)/p^2$	$r = \text{cov}(s_x, s_y)$ $s_x$ and $s_y$ are standard dev.	$f(x) = 1/(b-a)$ , $a \leq x \leq b$ $= 0$ , elsewhere $X \sim U[a,b]$ $E[X] = (a+b)/2$ $V[X] = (b-a)^2/12$	$P(A \cap B) = P(B A)P(A) = P(A B)P(B) = P(A)P(B) \text{ if ind.}$
Continuous Variable	Permutations	Transformation	
$P(a < X < b) = \int f(x) dx = F(b) - F(a)$ $f(x) = F'(x)$ $F(x) = P(X < x) = \int f(t) dt$ $E[X] = \int x f(x) dx$ $V[X] = \int x^2 f(x) dx - E[X]^2$ $E[g(X)] = \int g(x) f(x) dx$ $V[g(X)] = \int (g(x))^2 f(x) dx - E[g(X)]^2$	$n! = n(n-1)(n-2)*\dots*1$ if $n \geq 1$ $= 1$ if $n=0$ $nPr = n! / (n-r)!$ Order matters	$E[g(X)] = \sum g(x) f(x)$ $V[g(X)] = [\sum (g(x))^2 f(x)] - (E[g(X)])^2$	

By **t847222**  
cheatography.com/t847222/

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Bernoulli Trial	Hypothesis	Addition Rules	Standard Normal Distribution
$S = \{\text{success, failure}\} = \{p, q\}$ $p = P(I=1)$ $I \sim \text{Ber}(p)$ $E[I] = p$ $V[I] = p(1-p)$	Null hyp: make no change Alternate hyp: test according to question $\Rightarrow$ Test 1: $\mu \neq \mu_0$ ; 2: $\mu > \mu_0$ ; 3: $\mu < \mu_0$ Confidence interval decision: reject $H_0$ for $H_1$ if $\mu_0$ is not in confidence interval $Z_0$ or $T_0$ decision: $\sigma^2$ known: $Z_0 = (\bar{X} - \mu_0) / (\sigma / \sqrt{n}) \sim N(0, 1)$ Test 1: reject if $ Z_0  > z_{[\alpha/2]}$ ; 2: $z_0 > z_\alpha$ ; 3: $z_0 < -z_\alpha$ $\sigma^2$ unknown: $T_0 = (\bar{X} - \mu_0) / (S / \sqrt{n}) \sim T_{[n-1]}$ Test 1: $ T_0  > t_{[\alpha/2, n-1]}$ ; 2: $t_0 > t_{[\alpha, n-1]}$ ; 3: $t_0 < -t_{[\alpha, n-1]}$ Pop. & $\sigma^2$ unknown: replace $\sigma$ with $S$ from $\sigma^2$ known p-Value decision: reject if p-value $< \alpha$ p-value = $2[1 - \Phi( Z_0 )]$ , test 1 & z-value = $1 - \Phi(z_0)$ , test 2 & z-value = $\Phi(z_0)$ , test 3 & z-value = $2P(T >  T_0 )$ , test 1 & t-value = $P(T > t_0)$ , test 2 & t-value = $P(T < t_0)$ , test 3 & t-value	$P(A \cap B') = P(A) - P(A \cap B)$ $P(A \cup B) = P(A) + P(B) - P(A \cap B)$ $P(A' \cap B') = 1 - P(A \cup B)$ $P(A \cup B \cup C) = P(A) + P(B) + P(C) - P(A \cap B) - P(A \cap C) - P(B \cap C) + P(A \cap B \cap C)$ $P(A_1 \cup \dots \cup A_n) = 1 - P(A_1^c \cap \dots \cap A_n^c)$	$Z \sim N(0, 1)$ PMF: $\phi(z) = 1 / \sqrt{(2\pi)^n} e^{-z^2/2}$ CDF: $\Phi(z) = P(Z \leq z) = \int_{-\infty}^z \phi(t) dt$ $\Phi(0) = 0.5$ $P(Z \leq -z) = P(Z \geq z)$ $\Phi(-z) = 1 - \Phi(z)$ $P(a \leq Z \leq b) = \Phi(b) - \Phi(a)$ $P(-a \leq Z \leq -b) = \Phi(a) - \Phi(b)$
Negative Binomial Distribution	Expected Value	Marginal PMF	Linear Combination
$X = \# \text{ of trials to until } r^{\text{th}}$ success $X \sim \text{Nb}(r, p)$ $f(x) = \binom{x-1}{r-1} (1-p)^{x-r} p^r$ , for $x=r, r+1, \dots$ $E[X] = r/p$ $V[X] = r(1-p)/p^2$	$\mu = E[X] = \sum x f(x)$	$p(x) = P(X=x) = \sum_y p(x,y)$ $p(y) = P(Y=y) = \sum_x p(x,y)$	$Y \sim N(\mu_Y, \sigma^2_Y)$ $E[Y] = \sum c_i E[X_i]$ $V[Y] = \sum c_i^2 V[X_i]^2$ $X = 1/n \sum X_i$ $E[X] = \mu$ $V[X] = \sigma^2/n$
Erlang Distribution	Binomial Distribution	Exponential Distribution	Sample Covariance
$T = \text{time until } r^{\text{th}}$ outcome of Poisson process $F(x) = P(T \leq x) = 1 - P(T > x) = 1 - \sum_{k=0}^{r-1} e^{-\lambda x} (\lambda x)^k / k!$ $E[T] = r/\lambda$ $V[T] = r(1-\lambda)/\lambda^2$	$X = \# \text{ of successes from } n \text{ trials}$ $X \sim b(n, p)$ $f(x) = \binom{n}{x} p^x (1-p)^{n-x}$ , for $x=0, 1, \dots, n$ $E[X] = np$ $V[X] = np(1-p)$	Waiting time $X \sim \text{Exp}(\lambda)$ $f(x) = \lambda e^{-\lambda x}, x > 0$ $F(x) = 1 - e^{-\lambda x}, x > 0$ $E[X] = 1/\lambda$ $V[X] = 1/\lambda^2$ Lack of memory: $P(X > s+t   X > s) = P(X > t)$	$Y = c_1 X_1 + \dots + c_n X_n$
Standardization Thm			Line of Best Fit
$Z = (X - E[X]) / \sqrt{V[X]}$ $F(x) = P(X \leq x) = \Phi((x - \mu) / \sigma)$ $P(a < X < b) = F(b) - F(a)$			$y = a + Bx$ $B = ((\sum x_i y_i) - (\sum x_i)(\sum y_i) / n) / ((\sum x_i^2) - (\sum x_i)^2 / n)$
Percentile			
Rank of $k^{\text{th}}$ percentile: $(n+1) * k / 100 = m+p$ , $0 \leq p < 1$ $k^{\text{th}}$ percentile = $y_m + p(y_{[m+1]} - y_m)$ IQR = $q_3 - q_1$ Median is $50^{\text{th}}$ percentile			

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