

MATH20802: Statistical Methods
Semester 2
Formulas to remember for the final exam

The moment generating function of a random variable X is $M_X(t) = E[\exp(tX)]$.

The fact that $E(X^n) = M_X^{(n)}(0)$.

The moment generating function of a $\Gamma(a, \lambda)$ random variable (where a is the shape parameter and λ is the scale parameter) is $M_X(t) = \left(\frac{\lambda}{\lambda-t}\right)^a$.

The moment generating function of an $Exp(\lambda)$ random variable is $M_X(t) = \frac{\lambda}{\lambda-t}$.

$\hat{\theta}$ is an unbiased estimator of θ if $E(\hat{\theta}) = \theta$.

$\hat{\theta}$ is an asymptotically unbiased estimator of θ if $\lim_{n \rightarrow \infty} E(\hat{\theta}) = \theta$.

The bias of $\hat{\theta}$ is $E(\hat{\theta}) - \theta$.

The mean squared error of $\hat{\theta}$ is $E\left[(\hat{\theta} - \theta)^2\right]$.

$\hat{\theta}$ is a consistent estimator of θ if $\lim_{n \rightarrow \infty} E\left[(\hat{\theta} - \theta)^2\right] = 0$.

The discrete uniform distribution on the set of integers $\{1, 2, \dots, N\}$ has the probability mass function $p(x) = \frac{1}{N}$ for $x = 1, 2, \dots, N$.

The gamma function is defined by $\Gamma(a) = \int_0^{\infty} t^{a-1} \exp(-t) dt$.

The fact that $\Gamma(n) = (n-1)!$.

The fact that $\Gamma(x+1) = x\Gamma(x)$.

The fact that $\Gamma(1/2) = \sqrt{\pi}$.

The beta function is defined by $B(a, b) = \int_0^1 t^{a-1} (1-t)^{b-1} dt$.

The fact that $B(a, b) = \frac{\Gamma(a)\Gamma(b)}{\Gamma(a+b)}$.

The probability density function of $X \sim Exp(\lambda)$ is $f_X(x) = \lambda \exp(-\lambda x)$.

The cumulative distribution function of $X \sim Exp(\lambda)$ is $F_X(x) = 1 - \exp(-\lambda x)$.

The cumulative distribution function of $X \sim N(0, 1)$ is $\Phi(x)$.

The Type I error of $H_0 : \mu = \mu_0$ versus $H_0 : \mu \neq \mu_0$ occurs if H_0 is rejected when in fact $\mu = \mu_0$.

The Type II error of $H_0 : \mu = \mu_0$ versus $H_0 : \mu \neq \mu_0$ occurs if H_0 is accepted when in fact $\mu \neq \mu_0$.

The significance level of $H_0 : \mu = \mu_0$ versus $H_0 : \mu \neq \mu_0$ is the probability of type I error.

The power function of $H_0 : \mu = \mu_0$ versus $H_0 : \mu \neq \mu_0$ is $\Pi(\mu) = \Pr(\text{Reject } H_0 \mid \mu)$.

Let X_1, X_2, \dots, X_m be a random sample from a normal population with mean μ_X and variance σ_X^2 assumed known. Let Y_1, Y_2, \dots, Y_n be a random sample from a normal population with mean μ_Y and variance σ_Y^2 assumed known. Assume independence of