Stats Final Exam Cheat Sheet

Gamma function

$$\begin{split} \Gamma(z) &= \int_0^\infty t^{z-1} e^{-t} dt; \text{ for } \mathcal{R}(z) > 0 \\ \Gamma(z+1) &= z \Gamma(z); \Gamma(n+1) = n! \text{ for } n \in \mathbb{N} \end{split}$$

Gamma distribution

 $f_{Y}(y) = \frac{\beta^{\alpha}}{\Gamma(\alpha)} y^{\alpha-1} e^{-\beta y}; \ y, \alpha, \beta > 0$ $E[Y] = \alpha/\beta; \ Var(Y) = \alpha/\beta^{2}$ $MGF(t) = \left(1 - \frac{t}{\beta}\right)^{-\alpha} \text{ for } t < \beta$ $\chi^{2}(k) \sim \text{Gamma}(k/2, 1/2)$ $Y_{i} \text{ indip. } \text{Gamma}(\alpha_{i}, \beta) \Rightarrow \sum_{i=1}^{n} Y_{i} \sim \text{Gamma}(\sum_{i=1}^{n} \alpha_{i}, \beta)$ $k * \text{Gamma}(\alpha, \beta) = \text{Gamma}(\alpha, \beta/k)$

Inverse Gamma distribution

 $\begin{array}{lll} X & \sim & \operatorname{Gamma}(\alpha,\beta) & Y & = & 1/X & \sim \\ \operatorname{I-Gamma}(\alpha,\beta) & & & & \\ f_Y(y) & = & \frac{\beta^\alpha}{\Gamma(\alpha)} y^{-\alpha-1} e^{-\beta/y}; \ y,\alpha,\beta > 0 \\ E[Y] & = & \beta/(\alpha-1); \ Var(Y) = & \beta^2/[(\alpha-1)^2(\alpha-2)] \\ \operatorname{Mode} & = & \beta/(\alpha+1) \end{array}$

Inverse Gaussian distribution

$$\begin{split} &Y \sim \text{I-G}(\phi, \lambda) \\ &f_Y(y) \\ &\frac{\sqrt{\lambda}}{\sqrt{2\pi}} e^{\sqrt{\lambda \phi}} y^{-3/2} \exp\{-\frac{1}{2} \left(\phi y + \frac{\lambda}{y}\right)\}; \\ &y, \phi \geq 0, \lambda > 0 \\ &E[Y] = \left(\frac{\lambda}{\phi}\right)^{1/2}; \ Var(Y) = \sqrt{\frac{\lambda}{\phi^3}} \end{split}$$

Normal Inverse Gamma distribution

 $\begin{array}{l} \theta \sim \text{N-IGa}(\eta_0,\lambda_0,\alpha_0,\beta_0) \\ \sigma^2 \sim \text{IGa}(\alpha_0,\beta_0) \\ \mu | \sigma^2 \sim \text{N}(\eta_0,\sigma^2/\lambda_0) \\ f(\mu,\sigma^2) \propto \left(\frac{1}{\sigma^2}\right)^{\alpha_0+3/2} \exp\{\lambda_0\eta_0\frac{\mu}{\sigma^2} - \frac{1}{2\sigma^2}(\lambda_0\eta_0^2+2\beta_0) - \lambda_0\frac{\mu^2}{2\sigma^2}\} \\ \text{Posterior params (from Normal model s.r.s.):} \\ \alpha = \alpha_0 + \frac{n}{2}, \ \lambda = \lambda_0 + n \\ n = \frac{\lambda_0\eta_0 + n\bar{y}}{\sigma^2} \end{array}$

Beta distribution

 $f_Y(y) = \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} y^{\alpha-1} (1-y)^{\beta-1}; \ \alpha, \beta > 0, \ y \in [0,1]$ $E[Y] = \frac{\alpha}{\alpha+\beta}; \ Var(Y) = \frac{\alpha\beta}{(\alpha+\beta)^2(\alpha+\beta+1)}$

 $\beta = \beta_0 + \frac{n}{2}v^2 + \frac{n\lambda_0}{2(\lambda_0 + n)}(\bar{y} - \eta_0)^2$ $v = n^{-1} \sum_{i=1}^n (y_i - \bar{y})^2$

Multivariate Normal distribution

 $f_{Y}(y) = \frac{\det(A)^{1/2}e^{-\frac{1}{2}b^{T}Ab}}{(2\pi)^{d/2}} \exp\{-\frac{1}{2}(y^{T}Ay) + b^{T}y\}$ $E[Y] = A^{-1}b; \quad \text{MGF}(t) = \exp\{\mu t + \sigma^{2}t^{2}/2\}$ $Y_{1}|Y_{2} = y_{2} \sim N(\mu_{1} + \Sigma_{12}\Sigma_{22}^{-1}(y_{2} - \mu_{2}); \Sigma_{11} - \Sigma_{12}\Sigma_{22}^{-1}\Sigma_{21})$

t-Student distribution

 $f_Y(y) = \frac{\Gamma(\frac{\nu+1}{2})}{\sqrt{\nu\pi}\Gamma(\frac{\nu}{2})} \left(1 + \frac{y^2}{\nu}\right)^{-(\nu+1)/2}$ $\nu > 0$

Uniform distribution

 $Y \sim Unif(a,b)$ $f_Y(y) = \frac{1}{b-a} \text{ for } y \in [a,b] \text{ otherwise } 0$ $F_Y(y) = \frac{y-a}{b-a} \text{ for } y \in [a,b]$ $E[Y] = \frac{1}{2}(a+b); Var(Y) = \frac{1}{12}(b-a)^2$

Hypergeometric distribution

 $f_Y(y) = \frac{\binom{K}{y}\binom{N-K}{n-y}}{\binom{N}{n}}$ $E[Y] = n\frac{K}{N}; \ Var(Y) = n\frac{N-K}{N}\frac{K}{N}\frac{N-n}{N-1}$

Negative Binomial distribution

Y = "n. of failures till r-th success" $f_Y(k) = {k+r-1 \choose k} (1-p)^k p^r$ $E[Y] = \frac{r(1-p)}{p}; Var(Y) = \frac{r(1-p)}{p^2}$

Weibull distribution

$$\begin{split} Y \sim Wei(\gamma,\beta) \ y,\beta,\gamma > 0 \\ f(y) = \gamma \beta^{-\gamma} y^{\gamma-1} e^{-(y/\beta)^{\gamma}} \\ E[Y] = \beta \Gamma(1+1/\gamma) \end{split}$$

Series

 $\sum_{i=0}^{n} r^{i} = \frac{1-r^{n+1}}{1-r} \text{ for } |r| < 1$ $e^{x} = \sum_{n=0}^{\infty} x^{n}/n! = \lim_{n \to \infty} \left(1 + \frac{x}{n}\right)^{n}$ $\sum_{i=0}^{n} \frac{1}{x^{p}} \text{ for } p > 1$

Log base change

 $\log_a(x) = \frac{\log_b(x)}{\log_b(a)}$

Derivatives

 $\frac{d}{dx}\sin(x) = \cos(x); \frac{d}{dx}\cos(x) = -\sin(x)$ $\frac{d}{dx}\arcsin(x) = \frac{1}{\sqrt{1-x^2}}; \frac{d}{dx}\arccos(x) =$ $-\frac{1}{\sqrt{1-x^2}}$ $\frac{d}{dx}\tan(x) = \frac{1}{\cos^2(x)}; \frac{d}{dx}\arctan(x) =$ $\frac{1}{1+x^2}$

Integrals

 $\int x^k dx = \frac{x^{k+1}}{k+1} + C \quad k \neq 1; \quad \int \frac{1}{x} dx = \log(|x|) + C$ $\int \tan(x) dx = \log(\frac{1}{\cos(x)}) + C$ $\int \frac{1}{a^2 + x^2} dx = \frac{1}{a} \arctan(\frac{x}{a}) + C$ $\int \frac{1}{\sqrt{a^2 + x^2}} dx = \arcsin(\frac{x}{a}) + C$ $\int_0^\infty y^\alpha e^{-ky} dy < \infty \text{ if } \alpha > -1 \quad (k > 0)$

Taylor expansions

 $e^{x} = \sum_{i=0}^{k} \frac{x^{i}}{i!} + o(x^{k}) \text{ for } x \to 0$ $\log(1+x) = \sum_{i=1}^{k} (-1)^{i-1} \frac{x^{i}}{i} + o(x^{k}) \text{ for } x \to 0$ $\sin(x) = \sum_{i=0}^{n} \frac{(-1)^{i} x^{2i+1}}{(2i+1)!} + o(x^{2n})$ $\cos(x) = \sum_{i=0}^{n} \frac{(-1)^{i} x^{2i}}{(2i)!} + o(x^{2n})$

Matrix

 $X = \begin{pmatrix} A & B \\ C & D \end{pmatrix}$

If X is a 2×2 matrix:

$$X^{-1} = \frac{1}{\det X} \begin{pmatrix} D & -B \\ -C & A \end{pmatrix}$$

X is positive definite if A>0 and $\det X>0$

If X is a block Matrix X^{-1} has elements:

$$(A - BD^{-1}C)^{-1}$$

 $-A^{-1}B(D - CA^{-1}B)^{-1}$
 $-(D - CA^{-1}B)^{-1}CA^{-1}$
 $(D - CA^{-1}B)^{-1}$

 $\dot{i}^{\psi\psi} = (i_{\psi\psi} - i_{\psi\lambda}(i_{\lambda\lambda})^{-1}i_{\lambda\psi})^{-1}$ $i^{\psi\lambda} = -i^{\psi\psi}i_{\psi\lambda}i_{\lambda\lambda}^{-1}$

Order Statistics

i.i.d continuous case.

$$\begin{array}{ll} f_{Y_{(1)},\dots,Y_{(n)}}(y_1,.,y_n) & = \\ n!f(y_1)f(y_2)\dots f(y_n) \\ f_{Y(r)}(y_r) & = \\ \frac{n!}{(n-r)!(r-1)!}[F(y_r)]^{r-1}f(y_r)[1 \\ - F(y_r)]^{n-r} \\ f_{Y_r(Y_s)}(y,z) & = \\ \frac{n!}{(n-s)!(s-r-1)!(r-1)!}[F(y)]^{r-1}f(y)[F(z)-F(s)]^{s-r-1}f(z)[1-F(z)]^{n-r} \end{array}$$

Profile pivot

Likelihood profile pivot $\theta = (\psi, \lambda)$ $W_{P_e}(\psi) = (\hat{\psi} - \psi)^T j_P(\hat{\psi})(\hat{\psi} - \psi)$ $W_{P_u}(\psi) = U_P(\psi)^T i_P^{\psi\psi}(\psi, \hat{\lambda_{\psi}}) U_P(\psi)$ we can use $j^{\psi\psi}(\theta) = (j_{\psi\psi} - j_{\psi\lambda}(j_{\lambda\lambda})^{-1} j_{\lambda\phi})^{-1}$

Confidence intervals and Test

Binomial: for $n\theta$, $n(1-\theta) > 5$

 $q_2(\theta, \hat{\theta}) = \frac{\hat{\theta} - \theta}{\sqrt{\frac{\theta(1-\theta)}{n}}}$ $\frac{\hat{\theta} + (z_{1-\alpha/2})^2 / 2n}{1 + (z_{1-\alpha/2})^2 / n} \pm$ $\frac{z_{1-\alpha/2} / \sqrt{n}}{1 + (z_{1-\alpha/2})^2 / n} \sqrt{\hat{\theta}(1-\hat{\theta}) + \frac{(z_{1-\alpha/2})^2}{4n}}$ Clopper Pearson: $\{\theta \in (0,1) : \mathbb{P}_{\theta}(Y \leq y) \geq \alpha/2\}$ Exact distribution test: $n\hat{\theta} \sim Bi(n,\theta)$

 $\sup_{\theta \in \Theta_0} \{ \mathbb{P}_{\theta}(t(Y) > t(y)) + \frac{1}{2} \mathbb{P}_{\theta}(t(Y) =$

 α_{mid}^{oss}

Netwon - Raphson

Mid-pvalue:

 $x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}$

Total Variance

 $\overline{Var(Y) = E[Var(Y|X)] + Var(E[Y|X])}$

Delta Method

 $\sqrt{n}(Y_n - \theta) \xrightarrow{D} U$ $\sqrt{n}(g(Y_n) - g(\theta)) \xrightarrow{D} g'(\theta)U$

Posterior Predicitive Density

if
$$(y^*|y,\theta) = (y^*|\theta)$$

 $p(y^*|y) = \int p(y^*|\theta)p(\theta|y)d\theta$

Another topic

some stuff