Distributions

Uniform:
$$f(x) = \frac{1}{b-a}$$
; $a \le x \le b$; $\bar{X} = \frac{a+b}{2}$; $Var(x) = \frac{(b-a)^2}{12}$ $mgf = \frac{e^{tb} - e^{ta}}{t(b-a)}$

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Normal:
$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} \exp \frac{-1}{2\sigma^2} (x - \mu)^2;$$

 $\bar{X} = \mu; Var(x) = \frac{1}{\sigma^2};$

$$\bar{X} = \mu$$
; $Var(x) = \sigma^2$

$$mgf = \exp \mu t + \frac{t^2 \sigma^2}{2}$$

Exponential:
$$f(x) = \frac{1}{\beta}e^{\frac{-x}{\beta}}$$
; $\bar{X} = \beta$; $Var(x) = \beta^2$; $mgf = (1 = \beta t)^{-1}$;

$$\bar{X} = \beta$$
; $Var(x) = \beta^2$;

$$mgf = (1 = \beta t)^{-1};$$

Gamma:
$$f(x) = \frac{1}{\Gamma(\alpha)\beta^{\alpha}} x^{\alpha-1} e^{\frac{-x}{\beta}};$$

$$\bar{X} = \alpha \beta; Var(x) = \alpha \beta^2$$

$$\bar{X} = \alpha \beta; Var(x) = \alpha \beta^2;$$

 $mgf = (1 = \beta t)^{-\alpha}; \Gamma(\alpha) = \int_0^\infty y^{\alpha - 1} e^{-y} dy$

Beta:
$$f(x) = \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} (1-x)^{\beta-1};$$

$$\bar{X} = \frac{\alpha}{\alpha + \beta};$$

$$Var(x) = \frac{\alpha\beta}{(\alpha+\beta)^2(\alpha+\beta+1)};$$

Misc

Moment Generating Function:

$$m_{X_1,\dots X_n} = \mathbb{E}[e^{tX_1+\dots+tX_n}] \stackrel{iid}{=} \prod_{i=1}^n \mathbb{E}[e^{tX_i}]$$

Jacobian Method: Y = g(X) with g(X) monotone:

$$f_Y(y) = f_X(g^{-1}) \left| \frac{d}{dy} g^{-1}(y) \right|.$$

If g(X) not monotone, then f_Y is sum of piecewise monotone $g_i(X)$.

3 Order Statistics

Maximum:
$$F_{X_{(n)}}(x) = F_X(x)^n$$
,

$$f_{X_{(n)}} = n(F_X(x))^{n-1} f_X(x)$$

Minimum:
$$F_{X_{(1)}}(x) = 1 - (1 - F_X(x))^n$$
,

$$f_{X_{(1)}}(x) = n(1 - F_X(x))^{n-1} f_X(x)$$

General:

$$f_{X_{(j)}} = \frac{n!}{(j-1)!(n-j)!} F_X(x)^{j-1} f_X(x) (1 - F_X(x))^{n-j}$$

Normal, Chi-Square Distributions

Mean:
$$\bar{X} \sim N(\mu, \frac{\sigma^2}{n})$$
.

Given
$$X \sim N(\mu, \sigma^2), Z_i = \frac{X_i - \mu}{\sigma}$$
. Then: $\sum_{i=1}^n Z_i^2 \sim \chi_n^2$

$$\chi_{\nu}^2 = Gamma(\frac{\nu}{2}, \beta = 2), \ \mathbb{E}[X] = \nu, Var(x) = 2\nu.$$

Theorem 7.3: \bar{x}, S^2 are independent, and:

$$\frac{(n-1)S^2}{\sigma^2} \sim \chi_{n-1}^2 \qquad S^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$$

Student T Distribution: if $Z \sim N(0,1)$ and $W \sim \chi^2_{\nu}$, and Z, W independent. Then RV: $T = \frac{Z}{\sqrt{W/\nu}}$ follows the student T distribution. Where:

$$f_T(x) = \frac{\Gamma(\frac{\nu+1}{2})}{\Gamma(\frac{\nu}{2})\sqrt{\pi\nu}} (1 + \frac{t^2}{\nu})^{-(\nu+1)/2}$$