

Poisson $P(X = x) = \frac{(\lambda t)^x e^{-\lambda t}}{x!}, x = 0, 1, 2, \dots + \infty; \quad \mu = \lambda t \text{ & } \sigma = \sqrt{\lambda t}$

Exponential $f(x) = \lambda e^{-\lambda x}, x > 0; \text{ where } F(x) = 1 - e^{-\lambda x} \text{ and } \mu = \frac{1}{\lambda} \text{ & } \sigma = \frac{1}{\lambda}$

Weibull $f(x) = \frac{\beta}{\delta} \left(\frac{x}{\delta}\right)^{\beta-1} e^{-\left(\frac{x}{\delta}\right)^\beta}, x > 0; \text{ where } F(x) = 1 - e^{-\left(\frac{x}{\delta}\right)^\beta} \mu = \delta \Gamma\left(1 + \frac{1}{\beta}\right) \text{ &} \\ \sigma^2 = \delta^2 \Gamma\left(1 + \frac{2}{\beta}\right) - \mu^2$

Interval Estimation $\text{Point Estimate} \pm \text{Critical Value} * s.e(\text{point estimate})$

Test Statistic $\frac{\text{Point Estimate} - \text{Hypothesized Value}}{s.e(\text{point estimate})}$

$P\text{-value} = P(\text{Test Statistic is more extreme under the alternative than its observed value})$

Simple Linear Regression

$$s_{xy} = \sum (x - \bar{x})(y - \bar{y}) = \sum xy - \frac{1}{n} \sum x \sum y$$

$$\hat{\beta}_1 = \frac{s_{xy}}{s_{xx}}, \quad \hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x}$$

$$s.e(\hat{\beta}_1) = \sqrt{\frac{\sigma^2}{s_{xx}}}, \quad s.e(\hat{\beta}_0) = \sqrt{\sigma^2 \left(\frac{1}{n} + \frac{\bar{x}^2}{s_{xx}} \right)},$$

$$s.e(\widehat{\mu(x)}) = \sqrt{\sigma^2 \left(\frac{1}{n} + \frac{(x - \bar{x})^2}{s_{xx}} \right)}, \quad s.e(\widehat{y(x)}) = \sqrt{\sigma^2 \left(1 + \frac{1}{n} + \frac{(x - \bar{x})^2}{s_{xx}} \right)}$$

$$\text{SSR} = \hat{\beta}_1 s_{xy}$$

$$r = \frac{s_{xy}}{\sqrt{s_{xx}s_{yy}}}, \quad \hat{\sigma}^2 = \frac{SSE}{n-p}$$

$$R^2 = \frac{SSR}{SST}, \quad R^2_{adj} = 1 - \frac{SSE}{SST} \left(\frac{n-1}{n-p} \right), \quad F = \frac{MSR}{MSE}$$

Multiple Linear Regression

$$\hat{\beta} = (X'X)^{-1}X'Y = CX'Y$$

$$s.e(\hat{\beta}_j) = \sqrt{\hat{\sigma}^2 C_{jj}}$$

$$s.e(\hat{\mu}_{y|x_0}) = \sqrt{\hat{\sigma}^2 (x'_0 C x_0)}, \quad s.e(\hat{y}_0) = \sqrt{\hat{\sigma}^2 (1 + x'_0 C x_0)}$$

