Binomial Probability Distribution

$$P(x=k) = C_k^n p^k q^{n-k} = \frac{n!}{k!(n-k)!} p^k q^{n-k}$$

Mean: $\mu = np$

Variance: $\sigma^2 = npq$

Standard deviation: $\sigma = \sqrt{npq}$

Hypergeometric Probability Distribution

$$P(x = k) = \frac{C_k^M C_{n-k}^{N-M}}{C_n^N}$$

 $M \rightarrow$ successes, $N-M \rightarrow$ failures, $n \rightarrow$ size of the random sample space

$$Mean: \mu = n \left(\frac{M}{N}\right)$$

Variance:
$$\sigma^2 = n \left(\frac{M}{N} \right) \left(\frac{N-M}{N} \right) \left(\frac{N-n}{N-1} \right)$$

Poisson Probability Distribution

$$P(x=k) = \frac{\mu^k e^{-\mu}}{k!}$$

Mean: $E(x) = \mu$

Variance : $\sigma^2 = \mu$

Standard deviation: $\sigma = \sqrt{\mu}$

Variance of a Sample

$$s^{2} = \frac{\sum (x_{i} - \overline{x})^{2}}{n - 1} = \frac{\sum x_{i}^{2} - \frac{(\sum x_{i})^{2}}{n}}{n - 1}$$

Variance of Population: $\sigma^2 = \frac{\sum (x_i - \mu)^2}{N}$

Correlation coefficient, $r = \frac{S_{xy}}{S_x S_y}$

Covariance, $s_{xy} = \frac{\sum (x_i - \overline{x})(y_i - \overline{y})}{n-1}$, or

$$s_{xy} = \frac{\sum x_i y_i - \frac{(\sum x_i)(\sum y_i)}{n}}{n-1}$$

Regression line, y = a + bx

$$b = r \frac{s_y}{s_x} \qquad , \qquad a = \overline{y} - b\overline{x}$$

Standardizing the value of x:

$$z = \frac{x - \mu}{\sigma}$$
, or in a sample, $z = \frac{x - x}{s}$