

Statistics Formula – Sheet 1

Mean & Standard Deviation	Counting Rules
$\bar{x} = \frac{\sum f \cdot x}{\sum f}$, $\bar{x} = \frac{\sum w \cdot x}{\sum w}$ $s_x = \sqrt{\frac{\sum f \cdot (x - \bar{x})^2}{n-1}} = \sqrt{\frac{n \sum f \cdot x^2 - (\sum f \cdot x)^2}{n(n-1)}}$ $\sigma = \sqrt{\frac{\sum f \cdot (x - \mu)^2}{N}} = \sqrt{\frac{N \sum f \cdot x^2 - (\sum f \cdot x)^2}{N^2}}$	$n! = n \cdot (n-1) \cdot (n-2) \dots 3 \cdot 2 \cdot 1$ $\binom{n}{n_1 \ n_2 \ \dots \ n_k} = \frac{n!}{n_1! n_2! \dots n_k!}$ $\binom{n}{r} = C_r^n = \frac{n!}{(n-r)! r!}$ (Combinations) $P_r^n = \frac{n!}{(n-r)!}$ (Permutations)
Probability	Probability Distribution
$0 \leq P(x) \leq 1$ & $\sum P(x) = 1$ $P(A \text{ and } B) = P(A)P(B A)$ If A and B are independent. $P(A \text{ and } B) = P(A) \cdot P(B)$ $P(A^c) = 1 - P(A)$	$\mu = \sum x \cdot p(x)$ $\sigma = \sqrt{\sum x^2 \cdot P(x) - \mu^2}$
Bayes' Formula	Chebyshev's Theorem
$P(A_i B) = \frac{P(B A_i)P(A_i)}{\sum_{j=1}^n P(B A_j)P(A_j)}$ Bayes' Formula – Two – Event Case $P(A B) = \frac{P(A \& B)}{P(B)} = \frac{P(B A)P(A)}{P(B A)P(A) + P(B A^c)P(A^c)}$	$P(x - \mu < k\sigma) \geq 1 - \frac{1}{k^2}$ where $k > 0$ or $P(x - \mu \geq k\sigma) \leq \frac{1}{k^2}$
Coefficient of Variation	Expected Value
	$CV = \frac{\sigma}{\mu} \cdot 100\%$ Variance = $Cov(X, X) = V(x)$
Expected Value	Variance
$E(x) = \sum x \cdot p(x)$ (Probability Distributions) $E(x) = \int x \cdot f(x)dx$ (Continuous Distributions)	$V(x) = \sigma^2 = E(x^2) - E(x)^2 = E(x^2) - \mu^2$
Covariance	Central Limit Theorem
$Cov(X, Y) = E(XY) - E(X)E(Y)$	$x \sim N(\mu, \sigma)$ or $n \geq 30 \Rightarrow \bar{x} \sim N\left(\mu, \frac{\sigma}{\sqrt{n}}\right)$
Linear Correlation	Test Statistics (One Population)
$r = \frac{\sum Z_x \cdot Z_y}{n-1} = \frac{SS_{xy}}{\sqrt{SS_{xx} \cdot SS_{yy}}} = \frac{\sum (x - \bar{x})(y - \bar{y})}{\sqrt{\sum (x - \bar{x})^2} \sqrt{\sum (y - \bar{y})^2}}$ $r = \frac{n \sum xy - \sum x \sum y}{\sqrt{n \sum x^2 - (\sum x)^2} \sqrt{n \sum y^2 - (\sum y)^2}}$	$t = \frac{\bar{x} - \mu}{\frac{s}{\sqrt{n}}}$ (Mean – σ – unknown) $Z = \frac{\bar{x} - \mu}{\frac{\sigma}{\sqrt{n}}}$ (Mean – σ – known) $Z = \frac{\hat{p} - p}{\sqrt{\frac{pq}{n}}}$ (Proportion) $\chi^2 = \frac{(n-1)s^2}{\sigma^2}$ (σ^2 or σ)
Regression	Skewness
$\hat{y} = b_1 x + b_0$ $b_1 = \frac{SS_{xy}}{SS_x} = \frac{n \sum xy - \sum x \sum y}{n \sum x^2 - (\sum x)^2} = r \frac{S_y}{S_x}$ $b_0 = \bar{y} - b_1 \bar{x}$	$S_k = \frac{\text{Mean} - \text{Mode}}{SD}$ (Karl Pearson) If mode doesn't exist, $S_k = \frac{3(\text{Mean} - \text{Median})}{SD}$