## Econ 205 - Cheat Sheet Statistics for Business and Economics

#### Descriptive statistics:

Mean:  $\bar{x}$  =average(DATA), Median =median(DATA), Mode =mode(DATA)

Variance:  $\sigma^2 = \text{var.p}\left(\text{DATA}\right)$ ,  $s^2 = \text{var.s}\left(\text{DATA}\right)$ ,  $s^2 = \sigma^2\left(\frac{N}{n-1}\right)$  or  $\sigma^2 = s^2\left(\frac{n-1}{N}\right)$ 

Standard deviation:  $s = \sqrt{\frac{\sum_{i=1}^{n}(x_i - \bar{x})^2}{n-1}} = \text{stdev.s}(\text{DATA}) = \sigma\sqrt{\frac{N}{n-1}}$ 

 $\sigma = \sqrt{\frac{\sum_{i=1}^N (x_i - \mu)^2}{N}} = \mathtt{stdev.p}(\mathrm{DATA}) = s\sqrt{\frac{n-1}{N}}$  Coefficient of variation:  $CV = \frac{s}{\bar{x}}$  or  $CV = \frac{\sigma}{\mu}$ 

P<sup>th</sup>-percentile =percentile.exc(data,P/100) or location formula  $L_P = (n+1) \frac{P}{100}$ 

Covariance:  $\sigma_{XY}$  covariance.s(data) or Data-Analysis-Toolpak Covariance,  $\sigma_{XY} = \frac{100}{N} \frac{n-1}{N}$ 

Correlation coefficient:  $\rho_{XY} = \mathtt{correl}(\mathtt{data})$  or Data-Analysis-Toolpak  $\sim$  Correlation

Regression model:  $y = \beta_0 + \beta_1 x + \varepsilon$ 

Regression line estimated:  $\hat{y} = b_0 + b_1 x$  in Scatterplot Add Trendline or Data-Analysis-Toolpak

Regression

# **Probability:**

Rule of complements:  $P(A^c) = 1 - P(A)$ 

Multiplication formulat:  $P(A \text{ and } B) = P(A|B) \times P(B)$ 

Addition formulat: P(A or B) = P(A) + P(B) - P(A and B)Conditional probability:  $P(A|B) = \frac{P(A \text{ and } B)}{P(B)} = \frac{P(A|B) \times P(B)}{P(B)}$ 

Independence: A and B are independent if  $P(A \text{ and } B) = P(A) \times P(B)$ 

or when P(A|B) = P(A)

Mean (aka expected value) of a discrete distribution:  $\mu = \sum_{i=1}^{n} x_i P(x_i)$ 

E[c] = c, Var[c] = 0;

E[X + c] = E[X] + c, Var[X + c] = Var[X];

 $E[cX] = cE[X], Var[cX] = c^2Var[X]$ 

#### Distributions:

- Binomial distribution:  $P(X = x) = \text{binom.dist}(x, n, \pi, 0) \text{ and } P(X \le x) = \text{binom.dist}(x, n, \pi, 1)$
- Mean of binomial distribution:  $\mu = E[X] = n \times \pi$  and  $\sigma^2 = Var[X] = n \times \pi \times (1 \pi)$
- Uniform distribution:  $X \sim U[a,b]$ , then density is  $f = \frac{1}{b-a}$  and  $E[X] = \frac{a+b}{2}$  and  $V[X] = \frac{a+b}{2}$  $\frac{1}{12}(b-a)^2$
- Normal distribution:

$$X \sim N\left(\mu,\sigma\right): P\left(X < x\right) = P\left(\frac{X-\mu}{\sigma} < \frac{x-\mu}{\sigma}\right) \text{ where } Z = \frac{X-\mu}{\sigma} \sim N\left(0,1\right) \text{ then:}$$

to-the-left : P(Z < z) = norm.s.dist(z), or

to-the-right : P(Z > z) = 1 - norm.s.dist(z), and

the  $\pi^{th}$  percentile  $P_{\pi} = \text{norm.s.inv}(\pi)$ 

#### • Student T-distribution:

-if  $\sigma$  unknown then  $T = \frac{X-\mu}{s} \sim T(n-1)$ :

to-the-left : P(T < t) = t.dist(t, n - 1, 1), or

to-the-right : P(T > t) = 1 - t.dist(t, n - 1, 1), and

the  $\pi^{th}$  percentile  $P_{\pi} = \text{t.inv}(\pi, n-1)$ 

## Central limit theorem:

If  $X \sim N(\mu, \sigma)$  then  $\bar{X} \sim N\left(\mu, \frac{\sigma}{\sqrt{n}}\right)$  and  $\sigma_{\bar{X}} = \frac{\sigma}{\sqrt{n}}$ 

If  $X \sim ?(\mu, \sigma)$  then  $\bar{X} \sim N\left(\mu, \frac{\sigma}{\sqrt{n}}\right)$  and  $\sigma_{\bar{X}} = \frac{\sigma}{\sqrt{n}}$  if  $n \geq 30$ 

If proportion  $\pi \sim ?$ , then sample proportion  $\hat{p} \sim N\left(\pi, \sqrt{\frac{\pi(1-\pi)}{n}}\right)$  if  $n\pi \geq 5$  and  $n(1-\pi) \geq 5$ 

# Confidence intervals:

Confidence intervals.  $CI_{\alpha} \text{ (z): } \left[ \bar{x} \pm \overbrace{z_{\alpha/2} \times \frac{\sigma}{\sqrt{n}}} \right] \text{ and conf. int. (t): } \left[ \bar{x} \pm \overbrace{t_{\alpha/2} \times \frac{s}{\sqrt{n}}} \right]$ 

Sample size formula:  $n = (z \times \frac{\sigma}{W})^2$ 

# Hypothesis testing:

 $z_{obs}$  formula:  $z_{obs} = \frac{\bar{x} - \mu_0}{\sigma / \sqrt{n}}$  or  $z_{obs} = \frac{\hat{p} - \pi}{\sqrt{\frac{\pi(1 - \pi)}{2}}}$ 

 $t_{obs}$  formula:  $t_{obs} = \frac{\bar{x} - \mu_0}{s / \sqrt{n}}$ .

# Two sample tests:

Case 1: equal variances  $\sigma_1^2 = \sigma_2^2$ :  $t = \frac{(\bar{x}_1 - \bar{x}_2) - (\mu_1 - \mu_2)}{\sqrt{s_P^2 \left(\frac{1}{n_1} + \frac{1}{n_2}\right)}}$ , where  $s_P^2 = \frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2}{n_1 + n_2 - 2}$  and

 $v = n_1 + n_2 - 2$ 

Case 2: unequal variances  $\sigma_1^2 \neq \sigma_2^2$ :  $t = \frac{(\bar{x}_1 - \bar{x}_2) - (\mu_1 - \mu_2)}{\sqrt{\left(\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}\right)}}$ , where  $v = \frac{\left(\frac{s_1^2/n_1 + s_2^2/n_2}{2}\right)^2}{\frac{\left(\frac{s_1^2/n_1}{n_1} + \frac{s_2^2/n_2}{n_2}\right)^2}{n_1 - 1} + \frac{\left(\frac{s_2^2/n_2}{n_2} + \frac{s_2^2}{n_2}\right)^2}{n_2 - 1}}$ 

## Regression analysis:

SST = SSR + SSE

Coefficient of determination:  $R^2 = \frac{SSR}{SST}$