Sample Size Determination for a single sample: Suppose we want to be able to detect a difference between a sample mean and a "standard" mean M, if the absolute value of the difference is at least Δ . We do not want the interval $xbar \pm c \times SE(xbar)$ to contain M if the absolute value of (xbar - M) is greater than or equal to Δ . (We say such a difference is "statistically significant".) Equivalently, we do not want the interval $(xbar - M) \pm c \times SE(xbar)$ to contain 0 if $|xbar - M| \ge \Delta$. Then we want $c \times SE(xbar) \le \Delta$. Since SE(xbar) = s/sqrt(n), the inequality we need to solve is: $n \ge c^2 s^2 / \Delta^2$.

Example: Our company is developing a new drug to lower blood pressure. Here is the drop in BP (mm Hg) we see in a sample of 5 volunteers: 8, 9, 11, 12, 14.

One sample t-test of mu = 10 vs not = 10

Р Variable Ν Mean StDev SE 95% CI Т 2.39 1.07 0.75 0.495 change in BP 5 10.80 (7.84, 13.76)

How big must the sample size be to say that the mean of the sample is "significantly" greater than 10 mm Hg? We have $\Delta = 0.8$ mm Hg. To be conservative, let c = 2.5 and s = 2.4. Then $n \ge 2.5^2 \times 2.4^2 / 0.8^2 = 56.25$. We need a sample of at least 57 volunteers. (If we let c = 2, the sample size estimate becomes n = 36.)

Another example: In the past, you have been able to "cure" 60% of patients with a given disease. You have a new treatment to test and you want to say it is better than the old treatment if it "cures" at least 70% of patients. How many patients with the disease to you need to treat with the new treatment? We will be "conservative" and let $s^2 = 0.5 \times 0.5 = 0.25$ and let c = 2.5. Then $\Delta = .7 - .6 = .1$ and $n \ge 2.5^2 \times (0.5)^2 / 0.1^2 = 156.25$ or 157 patients. (If we let c = 2, the estimate is n = 100 patients.)

[Recall that to use this large sample analysis for 0/1 data, we need at least 5 observations in each of the 2 categories, which seems reasonable here.]

<u>Sample Size Determination for comparison of 2 means</u>: Suppose we want to be able to detect a "significant" difference between 2 population means if the absolute value of the difference is greater than or equal to Δ .

We do not want the confidence interval $(xbar - ybar) \pm c \times SE(xbar - ybar)$ to contain 0 if $|xbar - ybar| \ge \Delta$ (we will say such a difference is "statistically significant").

Therefore, we want $c \times SE(xbar-ybar) \leq \Delta$.

Since $SE(xbar-ybar) = s \times sqrt(2/n)$, the inequality we need to solve is: $n \ge 2c^2 s^2 / \Delta^2$. Here s^2 is the pooled estimate of the variance of the 2 populations sampled.

Example: Maximal oxygen capacity was measured for professional women athletes in 2 sports: basketball speedskating (units not given) 42.3 46.1 42.9 52.0 49.6 N Mean StDev SE Mean basketball 3 44.93 4.05 2.3

speedskating 2 49.05 4.17 3.0

Difference = mu (basketball) - mu (speedskating)Estimate for difference: -4.1295% CI for difference: (-16.01, 7.77)

2- Sample t-test of difference = 0 (vs not =): T-Value = -1.10 P-Value = 0.351 DF = 3 Both use Pooled StDev = 4.0927

The difference between basketball players and speedskaters is not significant for mean lung capacity. What sample size do we need to see a significant difference in average lung capacity between basketball players and speedskaters? We want to be able to detect a difference of at least 49.05-44.93=4.12 units. Let $\Delta = 4$ and s = 4.1 (pooled StDev shown above). To be conservative, let c = 2.5. Then $n \ge 2c^2 s^2 / \Delta^2$ becomes: $n \ge 2 \times 2.5^2 \times 4.1^2 / 4^2 = 13.1$. So we need to plan on 14 or more athletes in each group.