

# **MORE STATISTICS WITH GAUSSIANS**

10.24.2018

# PROBLEM SET 3

\* how's it going?



# RECAP

- \* functions for dealing with a gaussian distribution (pdf, cdf, sf, isf)
- \* standard error of the mean (another gaussian distribution!)
- \* gaussian confidence intervals
- \* gaussian z-test

# T-TEST



W. S. Gosset  
aka "Student"

- \* the Z-test assumes that you know exactly what the variance is
- \* what if you don't, and have to estimate it from your sample?
- \* then the Z-test doesn't apply, and you must use the t-test!

# T-TEST

- \* the t-test uses the **t statistic**, which is defined as:

$$t = \frac{Z}{s} = \frac{\bar{X} - \mu}{s/\sqrt{n}}$$

# T-TEST

- \* under the null hypothesis, the  $t$  statistic follows the  $t$  distribution, which has one parameter (the number of degrees of freedom)

# T-TEST

- \* the t-test is implemented in `scipy.stats`
- \* **`scipy.stats.ttest_1samp`** : tests whether the mean of your sample is different from a given value (assuming your sample has a gaussian distribution)

# T-TEST

- \* `scipy.stats` also includes other flavors of t-test which are useful
- \* **`scipy.stats.ttest_ind`** : tests whether two different (independent) samples have the same mean



# PAIRED T-TEST

- \* `scipy.stats` also includes other flavors of t-test which are useful
- \* **`scipy.stats.ttest_rel`** : aka the paired t-test, tests whether two *related* sets of samples have different means

# PAIRED T-TEST

- \* **fMRI example:** suppose you measure the response of each voxel in a given brain area under condition A, and then again under condition B. you now have 2 different samples, but they are related
- \* testing for the difference of the two populations could obscure a difference between the conditions

# PAIRED T-TEST

- \* **the solution:** compute the difference between  $\text{resp}(A)$  and  $\text{resp}(B)$ , then do a t-test on whether that difference is different from zero

# POWER OF STATISTICAL TESTS

- \* **power** is defined as the probability of rejecting the null hypothesis if the alternative hypothesis is actually true
- \* thus power is related to the rate of **false negative** results: how often does the test say “there is an effect” when there is actually an effect

# POWER OF STATISTICAL TESTS

- \* power is related to the p-value threshold that you choose for a test
- \* smaller threshold = lower power
- \* power is also related to whether the assumptions of the test are valid and whether the test is mis-specified

# POWER OF STATISTICAL TESTS

- \* e.g. if you have paired samples but use an un-paired t-test, then that could reduce your power to find a real effect

**END**