

### 1 Sample T-Testing

For **numerical** data.

Compares a sample mean to a hypothetical population mean.

**from scipy.stats import ttest\_1samp**

*ttest\_1samp* requires two inputs, a distribution of values and an expected mean.

**tstat, pval = ttest\_1samp(example\_distribution, expected\_mean)**

### 2 Sample T-Test

For **numerical** data.

Compares two sets of data, which are both approximately normally distributed.

The *null hypothesis*, in this case, is that the two distributions have the same mean.

**from scipy.stats import ttest\_ind**

It takes the two distributions as inputs and returns the t-statistic and a p-value.

**t, pval = ttest\_ind(dataset1, dataset2)**

### ANOVA

For **numerical** data.

Compares more than two numerical datasets.

ANOVA (Analysis of Variance) tests the null hypothesis that all of the datasets have the same mean.

**from scipy.stats import f\_oneway**

It takes in each dataset as a different input and returns the t-statistic and the p-value.

**t, pval = f\_oneway(a, b, c)**

### Tukey's Range Test

For **numerical** data.

We can perform a **Tukey's Range Test** to determine the difference between datasets.

**from statsmodels.stats.multicomp import pairwise\_tukeyhsd**

We have to provide the function with *onelist of all of the data* and *a list of labels* that tell the function which elements of the list are from which set.

We also provide the *significance level* we want, which is usually 0.05.

**values = np.concatenate([a, b, c])**

**labels = ['a'] \* len(a) + ['b'] \* len(b) + ['c'] \* len(c)**

**tukey\_results = pairwise\_tukeyhsd(values, labels, 0.05)**

### Binomial Test

For **categorical** data.

To analyze a dataset with two different possibilities for entries.

The **null hypothesis**, in this case, would be that there is no difference between the observed behavior and the expected behavior.

**from scipy.stats import binom\_test**

*binom\_test* requires three inputs, the number of observed successes, the number of total trials, and an expected probability of success.

**pval = binom\_test(525, n=1000, p=0.5)**

### Chi Square Test

For **categorical** data.

To compare two or more categorical datasets.

**from scipy.stats import chi2\_contingency**

The input to *chi2\_contingency* is a **contingency table** where:

- **The columns** represent different outcomes, like "Survey Response A" vs. "Survey Response B" or "Clicked a Link" vs. "Didn't Click"

- **The rows** are each a different condition, such as men vs. women or Interface A vs. Interface B

**X = [[30, 10],**

**[35, 5],**

**[28, 12],**

**[20, 20]]**

**\_, pval, \_, \_ = chi2\_contingency(X)**