Measurements of height and weight of 22 Truman students were taken, resulting in the following data:

<table>
<thead>
<tr>
<th></th>
<th>Male</th>
<th>Female</th>
</tr>
</thead>
<tbody>
<tr>
<td>Height (in)</td>
<td>69 70 65 72 76 70 70 66 68 73</td>
<td>65 61 67 65 70 62 63 60 66 65 64</td>
</tr>
<tr>
<td>Weight (lb)</td>
<td>192 148 140 190 248 197 170 137 160 185</td>
<td>110 105 136 135 187 125 147 118 128 66 120</td>
</tr>
</tbody>
</table>

When you enter it in Minitab, it should look like this:

Say we want to determine if the average height for females at Truman is 63 inches. When running a t-test, the two-sided hypothesis would be:

\[ H_0: \mu_F = 63 \text{ inches} \quad \text{and} \quad H_A: \mu_F \neq 63 \text{ inches} \]

Since the non-parametric tests look at median, we need to rewrite these as:

\[ H_0: \theta_F = 63 \text{ inches} \quad \text{and} \quad H_A: \theta_F \neq 63 \text{ inches} \]

**Wilcoxon Signed-Rank Test (One-Sample):**

To test the height of females according to our hypothesis go to Stat: Nonparametric: 1-Sample Wilcoxon. We enter ‘FemaleHeight’ as our variable and choose the ‘test median’ option. Enter 63 in the ‘test median’ box. In this case we have a two-sided hypothesis, so we choose ‘not equal,’ but notice the ‘less than’ and ‘greater than’ options in the drop down menu for the alternative hypothesis. These will be used for one-sided hypotheses.

The output looks like this:

Our p-value is 0.100, so at the 95% level, we fail to reject the null hypothesis and conclude that it is possible that the median height of women could be 63 inches.
Minitab Tutorial for Nonparametric Statistics: Rank Tests

Sign Test (One-Sample):

The sign test is very similar and we will use the same hypothesis. This time go to Stat: Nonparametric: 1-Sample Sign.

Our output looks like this:

<table>
<thead>
<tr>
<th>Sign Test for Median: Height</th>
</tr>
</thead>
<tbody>
<tr>
<td>Sign test of median = 63.00 versus no trend = 63.00</td>
</tr>
<tr>
<td>Below</td>
</tr>
<tr>
<td>11</td>
</tr>
</tbody>
</table>

The p-value this time is 0.2266, so once again we fail to reject the null hypothesis.

Sign and Wilcoxon Tests (Matched Pairs):

We can also run the Wilcoxon and Sign Tests for matched-ordinal data. The process is very similar to what we just did, except this time we must calculate the difference of our two samples. Then we set the ‘test median’ to 0. Notice how this will affect our hypothesis. If there is no difference in our pairs, then the differences should be equal to 0, but if a difference is found then the median difference will not be equal to 0.

\[ H_0: RT - RB = 0 \quad H_A: RT - RB \neq 0 \]

The p-value for the Wilcoxon test is 0.036, so we reject the null hypothesis and conclude that the difference is not equal to 0. This means that we have evidence to conclude that the two samples are different.

The same process is used for the sign test. The p-value is 0.0313, so again we reject the null hypothesis.
**Mann-Whitney Test (Two Independent Samples):**
This test can be used to compare two samples that are independent of each other, such as height of males versus females. Our hypothesis is:

\[ H_0 : \theta_F = \theta_M \quad \& \quad H_A : \theta_F \neq \theta_M \]

This time we go to **Stat: Nonparametric: Mann-Whitney**. Enter ‘MaleHeight’ as the first sample, and ‘FemaleHeight’ as the second sample. The output looks like this:

We see that the p-value is 0.0017, and so we reject the null hypothesis and conclude that we have statistically significant evidence that the heights of males versus females are different.

**Kruskal-Wallis Test (3+ Samples):**
Next we will look at a test that will allow us to compare more than two samples. The following example with essentially look at four samples, and will compare year in school to the number of credit hours that students are taking. Consider the following data set:

Our null hypothesis would be that year in school has no effect on credit hours taken. The alternate hypothesis is that there is a difference between the four groups. After entering the data, we go to **Stat: Nonparametric: Kruskal-Wallis**. We enter year in school as the ‘factor’ and credit hours as the ‘response.’ Here is the output:

We see that the p-value is 0.161, so we cannot reject the null hypothesis. We have found no difference in number of credit hours taken based on year in school.

**Friedman Test:**
For this test we will look at scores of gymnasts from three different judges based on what gym the attend. Here is the data set:

<table>
<thead>
<tr>
<th>Judge</th>
<th>Gymnast</th>
<th>Score</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>1</td>
<td>9</td>
</tr>
<tr>
<td>2</td>
<td>1</td>
<td>8</td>
</tr>
</tbody>
</table>
We want to determine if scoring of any one judge differs from the other judges. Our null hypothesis is that there is no difference. Our alternate hypothesis is that one judge scores differently than the others.

We go to Stat: Nonparametric: Friedman.

Since score is what is being measured, this is the response variable. The treatment is the judge, and the gymnast is the block variable. Here is our output:

Since the p-value is 0.697, we fail to reject the null hypothesis, and conclude that no one judge scores differently than all of the other judges.