# # Test a Perceptual Phenomenon

Course Name: Machine Learning Foundation - Udacity

Project Name: Test a Perceptual Phenomenon

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## **Background Information**

In a Stroop task, participants are presented with a list of words, with each word displayed in a color of ink. The participant's task is to say out loud the color of the ink in which the word is printed. The task has two conditions: a congruent words condition, and an incongruent words condition. In the congruent words condition, the words being displayed are color words whose names match the colors in which they are printed: for example RED, BLUE. In the incongruent words condition, the words displayed are color words whose names do not match the colors in which they are printed: for example PURPLE, ORANGE. In each case, we measure the time it takes to name the ink colors in equally-sized lists. Each participant will go through and record a time from each condition.

### Questions for Investigation

As a general note, be sure to keep a record of any resources that you use or refer to in the creation of your project. You will need to report your sources as part of the project submission.

Q1. What is our independent variable? What is our dependent variable?

Independent Variable: Congruent words condition, and an incongruent words condition

Dependent Variable: Time taken to name the ink colors

Q2. What is an appropriate set of hypotheses for this task? What kind of statistical test do you expect to perform? Justify your choices

μC: Population Mean of Response Time under Congruent Condition

ul: Population Mean of Response Time under Incongruent Condition

Null Hypothesis ( $H_0$ ): There is no difference in the population means of response time under Congruent and Incogruent conditions, i.e.  $H_0$ :  $\mu$ C =  $\mu$ I

Alternative Hypothesis ( $H_1$ ): Population Mean of the response time under Incongruent Condition will be significantly larger than the response time under Congruent Condition, i.e.  $H_1$ :  $\mu$ C <  $\mu$ I

Since we have reasons to assume the directionality beforehand, this would be a one-tailed test.

Directional Hypothesis (<a href="http://support.minitab.com/en-us/minitab/17/topic-library/basic-statistics-and-graphs/hypothesis-tests/basics/directional-and-nondirectional-hypotheses/">http://support.minitab.com/en-us/minitab/directional-and-nondirectional-hypotheses/</a>
(<a href="http://support.minitab.com/en-us/minitab/17/topic-library/basic-statistics-and-graphs/hypothesis-tests/basics/directional-and-nondirectional-hypotheses/">http://support.minitab.com/en-us/minitab/17/topic-library/basic-statistics-and-graphs/hypothesis-tests/basics/directional-and-nondirectional-hypotheses/</a>)): A directional alternative hypothesis states that the null hypothesis is wrong, and also specifies whether the true value of the parameter is greater than or less than the reference value specified in null hypothesis.

#### Justification:

Objective: We need to compare the means of two "related groups" to understand and determine whether the difference between the two means is statistically significant or not.

Assumption: Normal Distributions and comparison is taking place between dependent data samples

Sample Size: 24 observations only

Population Standard Deviation is not available

Q3. Report some descriptive statistics regarding this dataset. Include at least one measure of central tendency and at least one measure of variability.

```
In [1]: import pandas as pd

df=pd.read_csv('stroopdata.csv')
    print(df.head(10))
```

	Congruent	Incongruent
0	12.079	19.278
1	16.791	18.741
2	9.564	21.214
3	8.630	15.687
4	14.669	22.803
5	12.238	20.878
6	14.692	24.572
7	8.987	17.394
8	9.401	20.762
9	14.480	26.282

```
In [2]: print("Congruent (Mean): ", df['Congruent'].mean())
         print("Congruent (Median): ", df['Congruent'].median())
print("Congruent (Mode): ", df['Congruent'].mode())
         print("Congruent (Quantile): ", df['Congruent'].quantile())
         print("Congruent (Variance): ", df['Congruent'].var())
         print("Congruent (Standard Deviation):", df['Congruent'].std())
         print("Congruent (Skewness):", df['Congruent'].skew())
         print("Congruent (Kurtosis):", df['Congruent'].kurt())
         Congruent (Mean): 14.051125000000004
         Congruent (Median): 14.3565
         Congruent (Mode): 0
                                     8.630
         1
                8.987
         2
                9.401
         3
                9.564
         4
               10.639
         5
               11.344
         6
               12.079
         7
               12.130
         8
               12.238
         9
               12.369
         10
               12.944
               14.233
         11
         12
               14.480
         13
               14.669
         14
               14.692
         15
               15.073
         16
               15.298
         17
               16.004
               16.791
         18
         19
               16.929
         20
               18.200
         21
               18,495
         22
               19.710
         23
               22.328
         dtype: float64
         Congruent (Quantile): 14.3565
         Congruent (Variance): 12.669029070652174
         Congruent (Standard Deviation): 3.559357957645195
         Congruent (Skewness): 0.41689987447903953
```

Congruent (Kurtosis): -0.20522482332339598

```
In [3]: | print("Incongruent (Mean): ", df['Incongruent'].mean())
        print("Incongruent (Median): ", df['Incongruent'].median())
        print("Incongruent (Mode): ", df['Incongruent'].mode())
        print("Incongruent (Quantile): ", df['Incongruent'].quantile())
        print("Incongruent (Variance): ", df['Incongruent'].var())
        print("Incongruent (Standard Deviation):", df['Incongruent'].std())
        print("Incongruent (Skewness):", df['Incongruent'].skew())
        print("Incongruent (Kurtosis):", df['Incongruent'].kurt())
        Incongruent (Mean): 22.01591666666667
        Incongruent (Median): 21.0175
        Incongruent (Mode): 0
                                    15.687
        1
              17.394
        2
              17.425
        3
              17.510
        4
              17.960
        5
              18.644
        6
              18.741
        7
              19.278
        8
              20.330
        9
              20.429
        10
              20.762
              20.878
        11
        12
              21.157
        13
              21.214
              22.058
        14
        15
              22.158
        16
              22.803
              23.894
        17
              24.524
        18
        19
              24.572
        20
              25.139
        21
              26.282
        22
              34.288
              35.255
        23
        dtype: float64
        Incongruent (Quantile): 21.0175
        Incongruent (Variance): 23.011757036231884
        Incongruent (Standard Deviation): 4.797057122469138
        Incongruent (Skewness): 1.547590025915552
        Incongruent (Kurtosis): 2.6889001984359964
```

Q4. Provide one or two visualizations that show the distribution of the sample data. Write one or two sentences noting what you observe about the plot or plots.

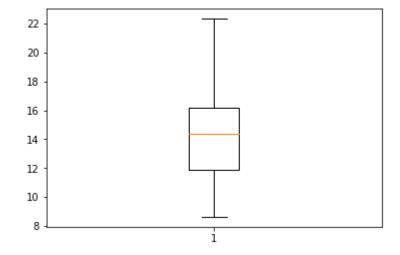
```
In [4]: import matplotlib.pyplot as plt
import numpy as np

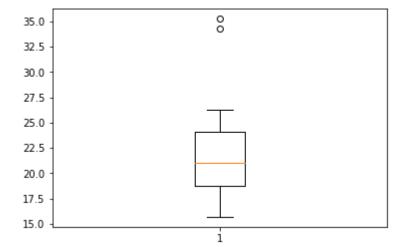
plt.plot(df.index, df['Congruent'], df.index, df['Incongruent'], linewidth=2.0)
plt.title('Time taken to identify ink color of Congruent words (blue) vs Incongruent plt.xlabel('Records');
plt.ylabel('Time (seconds)');
print(plt.show())
```

<matplotlib.figure.Figure at 0x2250de9c198>

None

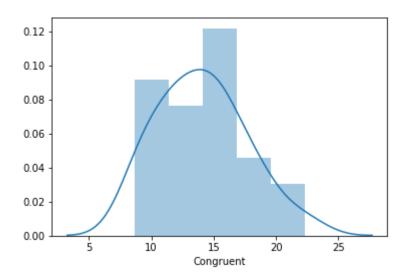
```
In [5]: plt.figure()
    plt.boxplot(df["Congruent"])
    plt.show()
    plt.figure()
    plt.boxplot(df["Incongruent"])
    plt.show()
```





```
In [6]: import seaborn as sns
from scipy.stats import t
    sns.distplot(df['Congruent'])
```

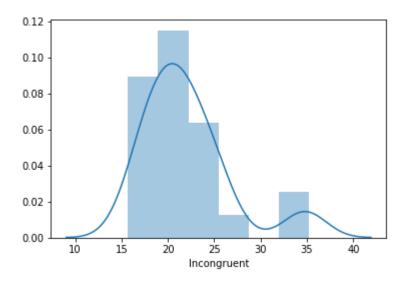
Out[6]: <matplotlib.axes.\_subplots.AxesSubplot at 0x22510633748>



(Congruent) Normal distribution is depicted in the above graph.

```
In [7]: sns.distplot(df['Incongruent'])
```

Out[7]: <matplotlib.axes.\_subplots.AxesSubplot at 0x2251064b400>



(Incongruent) Although the above graph is also Normally distributed, it is skewed towards the Right (i.e. Positively Skewed).

Q5. Now, perform the statistical test and report your results. What is your confidence level and your critical statistic value? Do you reject the null hypothesis or fail to reject it? Come to a conclusion in terms of the experiment task. Did the results match up with your expectations?

```
In [8]:
        print(df['Congruent'].size)
        print(df['Incongruent'].size)
        24
```

24

```
In [9]:
        #Critical value for 95% confidence level and 23 d.f. (size-1, i.e. 24-1)
        t.ppf(.95, 23)
```

Out[9]: 1.7138715277470473

Difference between Incongruent (Mean) & Congruent (Mean), i.e. 22.01591666666667 -14.051125000000004 = 7.96479166666667

```
In [10]:
         df['DifferenceSD'] = df['Congruent'] - df['Incongruent']
         print("Standard Deviation of the Difference: ", df['DifferenceSD'].std(axis=0))
```

Standard Deviation of the Difference: 4.864826910359056

t-statistic:

Difference of Means / (Standard Deviation of the Difference / Square Root of number of Observations)

```
In [11]: import math
         (22.01591666666667 - 14.051125000000004) / (4.864826910359056 / math.sqrt(24))
```

#### Out[11]: 8.020706944109955

Result:

T-statistic (8.020706944109955) > Critical value (1.7138715277470473)

Hence, Null Hypothesis (that there is no difference between the 2 times) can be rejected.

The result is as per my expectation because it took me less time to complete words with Congruent words condition than it took me to complete words with Incongruent words condition.

#### References:

https://en.wikipedia.org/wiki/Stroop\_effect (https://en.wikipedia.org/wiki/Stroop\_effect)

https://imotions.com/blog/the-stroop-effect/ (https://imotions.com/blog/the-stroop-effect/)

https://www.youtube.com/watch?v=EGpzftQf8ol (https://www.youtube.com/watch?v=EGpzftQf8ol)

https://en.wikipedia.org/wiki/Stroop effect#Warped words (https://en.wikipedia.org/wiki/Stroop\_effect#Warped\_words)

Q6. Optional: What do you think is responsible for the effects observed? Can you think of an

alternative or similar task that would result in a similar effect? Some research about the problem will be helpful for thinking about these two questions!

- On a daily basis, we are accustomed to reading words and not focusing on the ink of the color
  of the words. In this task, we had to say aloud the color of the ink of the word instead of the
  word, which was a change from what we have been used to.
- Our exposure to reading text printed in black color makes us accustomed to recognizing words and not colors. Hence, when confronted with printed text in any color apart from black, our brain spends additional effort in making sure that we focus on the color of the ink instead of the word itself.

As per Wikipedia, "the warped words Stroop effect produces the same findings similar to the original Stroop effect. Much like the Stroop task, the printed word's color is different from the ink color of the word; however, the words are printed in such a way that it is more difficult to read (typically curved-shaped).[36] The idea here is the way the words are printed slows down both the brain's reaction and processing time, making it harder to complete the task."

In [ ]:	