Test a Perceptual Phenomenon

December 25, 2018

0.0.1 Analyzing the Stroop Effect

Perform the analysis in the space below. Remember to follow the instructions and review the project rubric before submitting. Once you've completed the analysis and write-up, download this file as a PDF or HTML file, upload that PDF/HTML into the workspace here (click on the orange Jupyter icon in the upper left then Upload), then use the Submit Project button at the bottom of this page. This will create a zip file containing both this .ipynb doc and the PDF/HTML doc that will be submitted for your project.

(1) What is the independent variable? What is the dependent variable?

1-independent variable: **Congruent,Incongruent**2-dependent variable: **The response time to pronounce the color**

(2) What is an appropriate set of hypotheses for this task? Specify your null and alternative hypotheses, and clearly define any notation used. Justify your choices.

1- H0 (Null Hypothesis)

There is no difference in response time between incongruent and congruent.

H0: (C - I) = 0

2- H1 (Alternate Hypothesis)

The response time of the incongruent will not be equal the the response time of the congruent. H0: (C - I)!=0

(3) Report some descriptive statistics regarding this dataset. Include at least one measure of central tendency and at least one measure of variability. The name of the data file is 'stroopdata.csv'.

```
In [1]: import matplotlib.pyplot as plt
    import pandas as pd
    import numpy as np
    import seaborn as sns
    import scipy.stats as st

    df = pd.read_csv("stroopdata.csv")
In [2]: df.head(60)
```

```
Out[2]:
             Congruent
                         Incongruent
        0
                 12.079
                               19.278
        1
                 16.791
                               18.741
        2
                               21.214
                 9.564
        3
                 8.630
                               15.687
        4
                               22.803
                 14.669
        5
                 12.238
                               20.878
        6
                 14.692
                               24.572
        7
                               17.394
                 8.987
        8
                 9.401
                               20.762
        9
                               26.282
                 14.480
                               24.524
        10
                 22.328
        11
                 15.298
                               18.644
        12
                 15.073
                               17.510
        13
                 16.929
                               20.330
        14
                 18.200
                               35.255
        15
                 12.130
                               22.158
        16
                18.495
                               25.139
        17
                 10.639
                               20.429
        18
                 11.344
                               17.425
        19
                12.369
                               34.288
        20
                12.944
                               23.894
        21
                 14.233
                               17.960
         22
                 19.710
                               22.058
         23
                16.004
                               21.157
In [3]: df.describe()
Out[3]:
                Congruent
                             Incongruent
                               24.000000
                 24.000000
        count
        mean
                 14.051125
                               22.015917
                 3.559358
                                4.797057
        std
        min
                 8.630000
                               15.687000
        25%
                11.895250
                               18.716750
```

14.356500

16.200750

22.328000

50%

75%

max

After checking the data and the observations count above, we decided to use the (t test) WHY!!

- because we want to compare two samples with the same size and figure out if the dependant variable which is the response time will be affectedor not.
- The size of sample is small, 30 observations only, so without any dought this is the perfect senario for us to use (t test).

If the size of the samples was big then we would've used the (z test)

21.017500

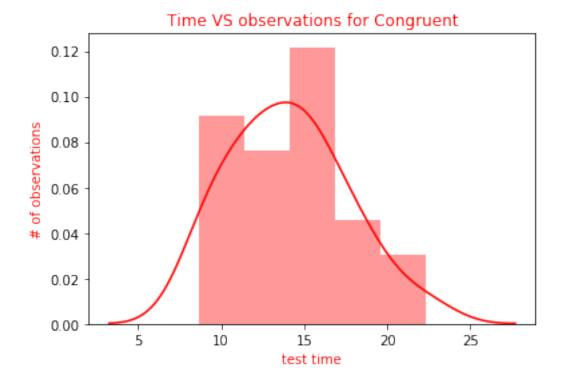
24.051500

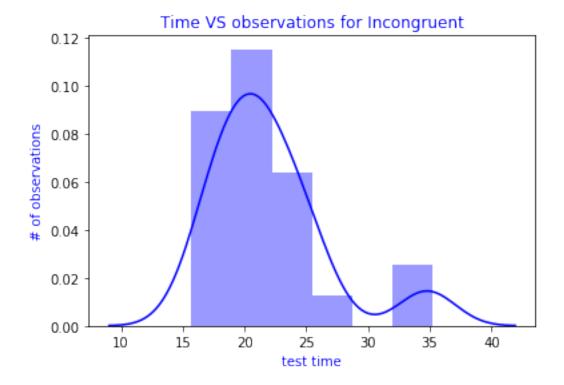
35.255000

```
In [4]: congruent_mean = df['Congruent'].mean()
        incongruent_mean = df['Incongruent'].mean()
        print("the mean of the Congruent is: ")
        print(congruent_mean)
        print("the mean of the Incongruent is: ")
        print(incongruent_mean)
the mean of the Congruent is:
14.051125
the mean of the Incongruent is:
22.0159166667
In [5]: congruent_std = df['Congruent'].std()
        incongruent_std = df['Incongruent'].std()
        print("the standard deviation of the Congruent is:")
        print(congruent_std)
        print("the standard deviation of the Incongruent is:")
        print(incongruent_std)
the standard deviation of the Congruent is:
3.55935795765
the standard deviation of the Incongruent is:
4.79705712247
In [6]: print("the standard deviation for Congruent: {0:.5f}".format(np.std(df['Congruent'].valuent')
        print("the standard deviation for Incongruent: {0:.5f}".format(np.std(df['Incongruent'].
the standard deviation for Congruent: 3.48442
the standard deviation for Incongruent: 4.69606
   --write answer here--
```

(4) 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 [7]: a=sns.distplot( df["Congruent"],color='r')
    plt.xlabel('test time',color='r')
    plt.ylabel('# of observations',color='r')
    plt.title('Time VS observations for Congruent',color='r');
```

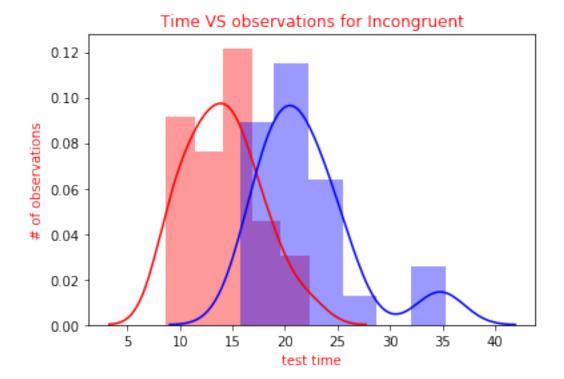


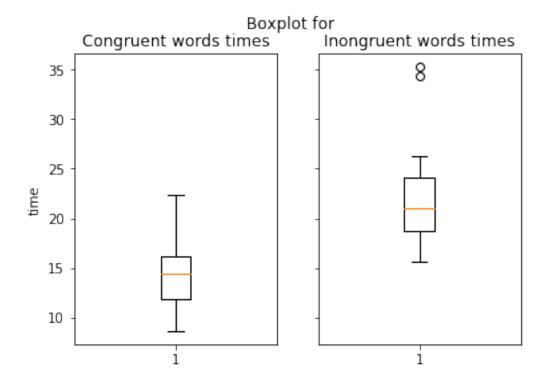


```
In [9]: sns.distplot( df["Congruent"],color='r')
    plt.xlabel('test time',color='r')
    plt.ylabel('# of observations',color='r')
    plt.title('Time VS observations for Congruent',color='r');

sns.distplot( df["Incongruent"],color='b')

plt.xlabel('test time',color='r')
    plt.ylabel('# of observations',color='r')
    plt.title('Time VS observations for Incongruent',color='r');
```





the two distributions are similar to each other but the Incongruent has an outliers, we did use the boxplot for this, to make this outliers data more visiable and clear.

(5) Now, perform the statistical test and report your results. What is your confidence level or Type I error associated with your test? What is your conclusion regarding the hypotheses you set up? Did the results match up with your expectations? **Hint:** Think about what is being measured on each individual, and what statistic best captures how an individual reacts in each environment.

calculated t-value = 8.02 p-value = 0.00000004

A paired t-test was run on a sample of 24 participants to determine whether there was a statistically significant mean difference between the reaction time between the congruent vs. incongruent condition. Typically it takes participants longer to say the ink colors (22.02 \pm 4.8 sec) than it does to read the words (14.05 \pm 3.56 sec). At the 95% confidence level (= .05) and 23 degrees of freedom, the critical statistic value for a one-tailed test is 1.714. The calculated t-value for the difference between the congruent and incongruent conditions is 8.02. The calculated t-value is greater than the critical t-value (t-table value). Additionally, the p-value is less than the alpha level: p < .05; therefore, the null hypothesis is rejected

(6) 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!

Refrences

- scipy-lectures.
- users.sussex.ac.uk.
- programiz.
- towardsdatascience.
- knowledgetack.