15. Statistics - AB testing

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Outline

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- Data types
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- Bivariate Analysis
- Null and Alternate Hypothesis
- Comparing Categorical vs Categorical
- A single Categorical variable (Goodness of fit)
- Comparing Binary vs Numeric
- Comparing Categorical (more than 2 levels) to Numeric
- Comparing Numeric to Numeric
- Review Objectives
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- Gist of the day

Objectives

- Understand hypothesis testing
- Knowing what tests to use when testing a hypothesis

Data types

- Numeric (Quantitative)
 - Continuous
 - Count (Discrete)
- Categorical (Qualitative)

- Binary
- Nominal
- Ordinal

Univariate Analysis

- Numeric
 - 5 number summary
 - $-\,$ Normally distributed Mean and Standard Deviation
 - Range and IQR
- Categorical
 - Frequencies
 - Proportions
 - Percentages
- Visualizations
 - Numeric Histograms, Boxplots
 - Categorical Barplots

Bivariate Analysis

- Numeric vs Numeric
 - Correlation
- Categorical vs Categorical
 - 2 way tables
 - Frequencies
 - Propotions
 - Percentages
- Categorical vs Numeric
 - Numeric summaries at each level of the categorical variable
- Visualizations
 - Numeric vs Numeric Scatterplots
 - Categorical vs Categorical Grouped barplots and Stacked bar plots
 - Numeric vs Categorical Grouped boxplots

Null and Alternate Hypothesis

- H_0 or the null hypothesis assumes there is no difference or things are as expected
- H_A or the alternative hypothesis is what we believe and are testing for
- p-value is the probability that we observe what we see in our data under the assumption that the H_0 is true
- The cut point used for significance is usually 0.05 (5%) and is referred to as the alpha
- Anything above 0.05 means there is no significance and we cannot reject the H_0
- Anything below 0.05 means there is significance and that we can reject the H_0
- AB Testing

Comparing Categorical vs Categorical

• Chisquare tests

```
import pandas as pd
from scipy.stats import chis2_contingency

#import dataset
nhanes = pd.read_csv("nhanes.csv")

Gender_Race = pd.crosstab(nhanes['Gender'], nhanes['Race1'])
print(Gender_Race)

chi, pvalue, dof, expected = chis2_contingency(Gender_Race)
print(chi)
print(pvalue)
print(dof)
print(expected)
```

• Interpreting a result

A single Categorical variable (Goodness of fit)

• Chisquare test

```
import pandas as pd
from scipy.stats import chisquare

Race = nhanes['Race1'].value_counts()

chi, pvalue = chisquare(Gender_Race)
print(chi)
print(pvalue)
```

• ?Interpretation

Comparing Binary vs Numeric

• T-Test

```
from scipy.stats import ttest_ind

male_age = nhanes.loc[nhanes['Gender'] == 'male', 'Age']
female_age = nhanes.loc[nhanes['Gender'] == 'female', 'Age']

print(male_age.mean(), female_age.mean())

tstat, pvalue = ttest_ind(male_age, female_age)
```

Comparing Categorical (more than 2 levels) to Numeric

• ANOVA (Analysis of variance)

```
from scipy.stats import f_oneway as anova
white_age = nhanes.loc[nhanes['Race1'] == 'White', 'Age']
black_age = nhanes.loc[nhanes['Race1'] == 'Black', 'Age']
hispanic_age = nhanes.loc[nhanes['Race1'] == 'Hispanic', 'Age']
fstat, pvalue = anova(white_age, black_age, hispanic_age)
```

Comparing Numeric to Numeric

- A regression is simply a line
- All straight lines can be defined by an equation
- Simple Linear Regression
- y = mx + c
 - where y = what we want to predict or the dependednt variable
 - -x =the predictor variable or independent variable
 - -c = intercept (the value of y when x = 0)
 - m = the slope also known as $\frac{rise}{run}$
- y = c + mx
- $y = \beta_0 + \beta_1 x$
 - where y = what we want to predict or the dependednt variable
 - x =the predictor variable or independent variable
 - $-\beta_0$ = intercept (the value of y when x = 0)
 - $-\beta_1$ = the slope also known as $\frac{rise}{run}$

Review Objectives

- Understand hypothesis testing
- Knowing what tests to use when testing a hypothesis

$\mathbf{Q\&A}$

Gist of the day

- Get the gist
- Get the pdf
- The Jupyter Notebook will be uploaded

We continue on Friday