Confidence Intervals

This tutorial is going to demonstrate how to load data, clean/manipulate a dataset, and construct a confidence interval for the difference between two population proportions and means.

We will use the 2015-2016 wave of the NHANES data for our analysis.

*Note: We have provided a notebook that includes more analysis, with examples of confidence intervals for one population proportions and means, in addition to the analysis I will show you in this tutorial. I highly recommend checking it out!

For our population proportions, we will analyze the difference of proportion between female and male smokers. The column that specifies smoker and non-smoker is "SMQ020" in our dataset.

For our population means, we will analyze the difference of mean of body mass index within our female and male populations. The column that includes the body mass index value is "BMXBMI".

Additionally, the gender is specified in the column "RIAGENDR".

In [1]: import pandas as pd import numpy as np import matplotlib matplotlib.use('Agg') import seaborn as sns %matplotlib inline import matplotlib.pyplot as plt import statsmodels.api as sm

In [2]: url = "nhanes_2015_2016.csv" da = pd.read_csv(url)

Investigating and Cleaning Data

```
In [3]:
       # Recode SMQ020 from 1/2 to Yes/No into new variable SMQ020x
       da["SMQ020x"] = da.SMQ020.replace({1: "Yes", 2: "No", 7: np.nan, 9: np.nan})
       da["SMQ020x"]
Out[3]: 0
            Yes
       1
            Yes
       2
            Yes
       3
             No
       4
             No
       5
             No
       6
            Yes
       7
             No
       8
             No
       9
             No
       10
             Yes
       11
             Yes
       12
             Yes
       13
             No
       14
             No
       15
             No
       16
             No
       17
             No
       18
             Yes
       19
             No
       20
             No
       21
             No
       22
             Yes
       23
             No
       24
             No
       25
             No
       26
             Yes
       27
             Yes
       28
             No
       29
             No
       5705 Yes
       5706
             Yes
       5707
              No
       5708
              No
       5709
              Yes
       5710
              No
       5711
             Yes
       5712
              No
       5713
              No
       5714
              No
       5715
              No
       5716
              Yes
       5717
              Yes
       5718
              No
       5719
              Yes
       5720
              No
       5721
              No
       5722
              No
       5723
              Yes
       5724
              No
       5725
              No
       5726
             Yes
       5727
              No
       5728
              No
       5729
              No
       5730
              Yes
```

5731

5732

5733

5734

No

Yes

Yes

Name: SMQ020x, Length: 5735, dtype: object

```
In [4]:
       # Recode RIAGENDR from 1/2 to Male/Female into new variable RIAGENDRX
       da["RIAGENDRx"] = da.RIAGENDR.replace({1: "Male", 2: "Female"})
        da["RIAGENDRx"]
Out[4]: 0
              Male
              Male
       1
       2
              Male
       3
             Female
       4
             Female
       5
             Female
       6
              Male
       7
             Female
       8
              Male
       9
              Male
       10
              Male
       11
              Male
       12
             Female
       13
             Female
       14
              Male
       15
             Female
       16
             Female
       17
             Female
       18
             Female
       19
             Female
       20
              Male
       21
             Female
       22
             Female
       23
             Female
       24
              Male
       25
             Female
       26
              Male
       27
             Female
       28
              Male
       29
             Female
       5705
               Male
       5706
               Male
       5707
              Female
       5708
              Female
       5709
               Male
       5710
              Female
       5711
               Male
       5712
              Female
       5713
               Male
       5714
               Male
       5715
              Female
       5716
              Female
       5717
               Male
       5718
               Male
       5719
              Female
       5720
               Male
       5721
              Female
       5722
              Female
       5723
              Female
       5724
              Female
       5725
               Male
       5726
               Male
       5727
              Female
       5728
               Male
       5729
               Male
       5730
              Female
       5731
               Male
              Female
       5732
       5733
               Male
       5734
              Female
```

In [5]: dx = da[["SMQ020x", "RIAGENDRx"]].dropna() pd.crosstab(dx.SMQ020x, dx.RIAGENDRx)

Name: RIAGENDRx, Length: 5735, dtype: object

Out[5]:

RIAGENDRX	Female	Male
SMQ020x		
No	2066	1340
Yes	906	1413

In [6]: # Recode SMQ020x from Yes/No to 1/0 into existing variable SMQ020x dx["SMQ020x"] = dx.SMQ020x.replace({"Yes": 1, "No": 0})

```
In [7]: dz = dx.groupby("RIAGENDRx").agg({"SMQ020x": [np.mean, np.size]}) dz.columns = ["Proportion", "Total n"] dz
```

Out[7]:

	Proportion	Total n
RIAGENDRx		
Female	0.304845	2972
Male	0.513258	2753

Constructing Confidence Intervals

Now that we have the population proportions of male and female smokers, we can begin to calculate confidence intervals. From lecture, we know that the equation is as follows:

$$Best\ Estimate \pm Margin\ of\ Error$$

Where the Best Estimate is the observed population proportion or mean from the sample and the Margin of Error is the t-multiplier.

The equation to create a 95% confidence interval can also be shown as:

$$Population\ Proportion\ or\ Mean\ \pm (t-multiplier*\ Standard\ Error)$$

The Standard Error (SE) is calculated differenly for population proportion and mean:

$$Standard\ Error\ for\ Population\ Proportion = \sqrt{rac{Population\ Proportion*(1-Population\ Proportion)}{Number\ Of\ Observations}}$$
 $Standard\ Error\ for\ Mean = rac{Standard\ Deviation}{\sqrt{Number\ Of\ Observations}}$

Lastly, the standard error for difference of population proportions and means is:

$$Standard\ Error\ for\ Difference\ of\ Two\ Population\ Proportions\ Or\ Means = \sqrt{(SE_1)^2 + (SE_2)^2}$$

Difference of Two Population Proportions

```
In [8]: p = .304845

n = 2972

se_female = np.sqrt(p * (1 - p)/n)

se_female
```

Out[8]: 0.00844415041930423

```
In [9]: p = .513258
n = 2753
se_male = np.sqrt(p * (1 - p)/ n)
se_male
```

Out[9]: 0.009526078787008965

```
In [10]: se_diff = np.sqrt(se_female**2 + se_male**2) se_diff
```

Out[10]: 0.012729880335656654

```
In [11]: d = .304845 - .513258

lcb = d - 1.96 * se_diff

ucb = d + 1.96 * se_diff

(lcb, ucb)
```

Out[11]: (-0.23336356545788706, -0.18346243454211297)

Difference of Two Population Means

In [13]: da.groupby("RIAGENDRx").agg({"BMXBMI": [np.mean, np.std, np.size]})

Out[13]:

	вмхвмі		
	mean	std	size
RIAGENDRX			
Female	29.939946	7.753319	2976.0
Male	28.778072	6.252568	2759.0

In [14]: sem_female = 7.753319 / np.sqrt(2976) sem_male = 6.252568 / np.sqrt(2759) (sem_female, sem_male)

Out[14]: (0.14212523289878048, 0.11903716451870151)

In [15]: sem_diff = np.sqrt(sem_female**2 + sem_male**2) sem_diff

Out[15]: 0.18538993598139303

In [16]: d = 29.939946 - 28.778072

In [17]: $| lcb = d - 1.96 * sem_diff | ucb = d + 1.96 * sem_diff | (lcb, ucb)$

Out[17]: (0.798509725476467, 1.5252382745235278)