

Confidence_Intervals_Differences_Population_Parameters

February 10, 2022

1 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)
```

1.0.1 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    No
5732    Yes
5733    Yes
5734    No
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
        1      Male
        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
5732    Female
5733      Male
5734    Female
Name: RIAGENDRx, Length: 5735, dtype: object

```

```

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

```

```

Out[5]:
SMQ020x RIAGENDRx
0      Yes      Male
1      Yes      Male
2      Yes      Male
3       No    Female
4       No    Female
5       No    Female
6      Yes      Male
7       No    Female
8       No      Male
9       No      Male
10     Yes      Male
11     Yes      Male
12     Yes    Female
13     No    Female

```

14	No	Male
15	No	Female
16	No	Female
17	No	Female
18	Yes	Female
19	No	Female
20	No	Male
21	No	Female
22	Yes	Female
23	No	Female
24	No	Male
25	No	Female
26	Yes	Male
27	Yes	Female
28	No	Male
29	No	Female
...
5705	Yes	Male
5706	Yes	Male
5707	No	Female
5708	No	Female
5709	Yes	Male
5710	No	Female
5711	Yes	Male
5712	No	Female
5713	No	Male
5714	No	Male
5715	No	Female
5716	Yes	Female
5717	Yes	Male
5718	No	Male
5719	Yes	Female
5720	No	Male
5721	No	Female
5722	No	Female
5723	Yes	Female
5724	No	Female
5725	No	Male
5726	Yes	Male
5727	No	Female
5728	No	Male
5729	No	Male
5730	Yes	Female
5731	No	Male
5732	Yes	Female
5733	Yes	Male
5734	No	Female

[5725 rows x 2 columns]

```
In [6]: pd.crosstab(dx.SMQ020x, dx.RIAGENDRx)
```

```
Out[6]: RIAGENDRx  Female  Male
SMQ020x
No              2066   1340
Yes              906   1413
```

```
In [9]: # Recode SMQ020x from Yes/No to 1/0 into existing variable SMQ020x
dx["SMQ020x"] = dx.SMQ020x.replace({"Yes": 1, "No": 0})
```

TypeError Traceback (most recent call last)

```
<ipython-input-9-0a796a728ba0> in <module>()
    1 # Recode SMQ020x from Yes/No to 1/0 into existing variable SMQ020x
----> 2 dx["SMQ020x"] = dx.SMQ020x.replace({"Yes": 1, "No": 0})
```

```
/opt/conda/lib/python3.6/site-packages/pandas/core/series.py in replace(self, to_replac
3839         return super(Series, self).replace(to_replace=to_replace, value=value,
3840                                             inplace=inplace, limit=limit,
-> 3841                                             regex=regex, method=method)
3842
3843     @Appender(generic._shared_docs['shift'] % _shared_doc_kwargs)
```

```
/opt/conda/lib/python3.6/site-packages/pandas/core/generic.py in replace(self, to_repl
6496
6497         return self.replace(to_replace, value, inplace=inplace,
-> 6498                           limit=limit, regex=regex)
6499     else:
6500
```

```
/opt/conda/lib/python3.6/site-packages/pandas/core/series.py in replace(self, to_replac
3839         return super(Series, self).replace(to_replace=to_replace, value=value,
3840                                             inplace=inplace, limit=limit,
-> 3841                                             regex=regex, method=method)
3842
3843     @Appender(generic._shared_docs['shift'] % _shared_doc_kwargs)
```

```
/opt/conda/lib/python3.6/site-packages/pandas/core/generic.py in replace(self, to_repl
6545                                     dest_list=value,
6546                                     inplace=inplace,
```

```

-> 6547                                     regex=regex)
6548
6549             else: # [NA, ''] -> 0

/opt/conda/lib/python3.6/site-packages/pandas/core/internals/managers.py in replace_li
557         return _compare_or_regex_match(values, s, regex)
558
--> 559     masks = [comp(s, regex) for i, s in enumerate(src_list)]
560
561     result_blocks = []

/opt/conda/lib/python3.6/site-packages/pandas/core/internals/managers.py in <listcomp>
557         return _compare_or_regex_match(values, s, regex)
558
--> 559     masks = [comp(s, regex) for i, s in enumerate(src_list)]
560
561     result_blocks = []

/opt/conda/lib/python3.6/site-packages/pandas/core/internals/managers.py in comp(s, re
555         return _compare_or_regex_match(maybe_convert_objects(values),
556                                         getattr(s, 'asm8'), regex)
--> 557         return _compare_or_regex_match(values, s, regex)
558
559     masks = [comp(s, regex) for i, s in enumerate(src_list)]

/opt/conda/lib/python3.6/site-packages/pandas/core/internals/managers.py in _compare_or
1949     raise TypeError(
1950         "Cannot compare types {a!r} and {b!r}".format(a=type_names[0],
-> 1951                                                         b=type_names[1]))
1952     return result
1953

```

TypeError: Cannot compare types 'ndarray(dtype=int64)' and 'str'

In [10]: dx

```

Out[10]:      SMQ020x  RIAGENDRx
0          1      Male
1          1      Male
2          1      Male
3          0    Female
4          0    Female

```

5	0	Female
6	1	Male
7	0	Female
8	0	Male
9	0	Male
10	1	Male
11	1	Male
12	1	Female
13	0	Female
14	0	Male
15	0	Female
16	0	Female
17	0	Female
18	1	Female
19	0	Female
20	0	Male
21	0	Female
22	1	Female
23	0	Female
24	0	Male
25	0	Female
26	1	Male
27	1	Female
28	0	Male
29	0	Female
...
5705	1	Male
5706	1	Male
5707	0	Female
5708	0	Female
5709	1	Male
5710	0	Female
5711	1	Male
5712	0	Female
5713	0	Male
5714	0	Male
5715	0	Female
5716	1	Female
5717	1	Male
5718	0	Male
5719	1	Female
5720	0	Male
5721	0	Female
5722	0	Female
5723	1	Female
5724	0	Female
5725	0	Male
5726	1	Male

5727	0	Female
5728	0	Male
5729	0	Male
5730	1	Female
5731	0	Male
5732	1	Female
5733	1	Male
5734	0	Female

[5725 rows x 2 columns]

```
In [11]: dz = dx.groupby("RIAGENDRx").agg({"SMQ020x": [np.mean, np.size]})
dz
```

```
Out[11]:
```

	SMQ020x	
	mean	size
RIAGENDRx		
Female	0.304845	2972
Male	0.513258	2753

```
In [12]: dz.columns = ["Proportion", "Total n"]
dz
```

```
Out[12]:
```

	Proportion	Total n
RIAGENDRx		
Female	0.304845	2972
Male	0.513258	2753

1.0.2 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:

$$\text{Best Estimate} \pm \text{Margin of Error}$$

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:

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

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

$$\text{Standard Error for Population Proportion} = \sqrt{\frac{\text{Population Proportion} * (1 - \text{Population Proportion})}{\text{Number Of Observations}}}$$

$$\text{Standard Error for Mean} = \frac{\text{Standard Deviation}}{\sqrt{\text{Number Of Observations}}}$$

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

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

Difference of Two Population Proportions

```
In [13]: p = .304845
         n = 2972
         se_female = np.sqrt(p * (1 - p)/n)
         se_female
```

```
Out[13]: 0.00844415041930423
```

```
In [14]: p = .513258
         n = 2753
         se_male = np.sqrt(p * (1 - p)/ n)
         se_male
```

```
Out[14]: 0.009526078787008965
```

```
In [15]: se_diff = np.sqrt(se_female**2 + se_male**2)
         se_diff
```

```
Out[15]: 0.012729880335656654
```

```
In [16]: # no calculo do intervalo de confianca, usar o standard erro da diferenca (se_diff) e
         d = .304845 - .513258
         lcb = d - 1.96 * se_diff
         ucb = d + 1.96 * se_diff
         (lcb, ucb)
```

```
Out[16]: (-0.23336356545788706, -0.18346243454211297)
```

Difference of Two Population Means

```
In [18]: da["BMXBMI"].head()
```

```
Out[18]: 0    27.8
         1    30.8
         2    28.8
         3    42.4
         4    20.3
         Name: BMXBMI, dtype: float64
```

```
In [19]: da.groupby("RIAGENDRx").agg({"BMXBMI": [np.mean, np.std, np.size]})
```

```
Out[19]:
```

	BMXBMI		
	mean	std	size
RIAGENDRx			
Female	29.939946	7.753319	2976.0
Male	28.778072	6.252568	2759.0

```
In [20]: sem_female = 7.753319 / np.sqrt(2976)
         sem_male = 6.252568 / np.sqrt(2759)
         (sem_female, sem_male)
```

```
Out[20]: (0.14212523289878048, 0.11903716451870151)
```

```
In [21]: sem_diff = np.sqrt(sem_female**2 + sem_male**2)
         sem_diff
```

```
Out[21]: 0.18538993598139303
```

```
In [24]: d = 29.939946 - 28.778072
         d
```

```
Out[24]: 1.1618739999999974
```

```
In [25]: lcb = d - 1.96 * sem_diff
         ucb = d + 1.96 * sem_diff
         (lcb, ucb)
```

```
Out[25]: (0.798509725476467, 1.5252382745235278)
```