ps_4

February 28, 2019

1 Problem Set 4 (python stuff)

Formatting enviornment:

```
In [1]: import numpy as np
        import pandas as pd
        import matplotlib.pyplot as plt
        import seaborn as sns
        from pathlib import Path
        import requests
        import statsmodels.api as sm
        import statsmodels.formula.api as smf
        from stargazer.stargazer import Stargazer
        from math import isclose
        from scipy import stats
        from IPython.display import display
  Loading Data:
In [2]: def fetch_and_cache(data_url, file, data_dir="data", force = False):
            (Credit: John DeNero)
            Download and cache a url and return the file object.
            Dependent: from pathlib import Path
                        import requests
            data_url: the web address to download
            file: the file in which to save the results.
            data_dir: (default="data") the location to save the data
            force: if true the file is always re-downloaded
            return: The pathlib. Path to the file.
            data_dir = Path(data_dir)
            data_dir.mkdir(exist_ok=True)
            file_path = data_dir/Path(file)
```

```
if force and file_path.exists():
        file_path.unlink()
    if force or not file_path.exists():
        print('Downloading...', end = ' ')
        resp = requests.get(data url)
        with file_path.open('wb') as f:
            f.write(resp.content)
        print('Done!')
    else:
        import time
        created = time.ctime(file_path.stat().st_ctime)
        print("Using cached version downloaded at", created)
    return file_path
data_url = 'https://bcourses.berkeley.edu/files/74496429/download?download_frd=1'
file = 'ovb.csv'
ovb_path = fetch_and_cache(data_url, file)
ovb_df = pd.read_csv(ovb_path)
```

Using cached version downloaded at Wed Feb 27 12:57:56 2019

1.1 Question 2:

```
In [3]: models = ["logwage ~ 1 + educ", "educ ~ 1 + logwage"]
        f_df = ovb_df.loc[ovb_df['female'] == 1]
       rho = np.corrcoef(f_df['logwage'], f_df['educ'])[0,1]
        rho_squared = rho**2
       r_squared = []
        for mod in models:
            res = smf.ols(mod, data = f_df).fit()
            r_squared.append(res.rsquared)
        assert(isclose(r_squared[0], r_squared[1], abs_tol = 1e-6))
        assert(isclose(r_squared[0], rho_squared, abs_tol = 1e-6))
        print("correlation coefficient = {}".format(rho_squared))
        print("r^2 of logwage on educ = {}".format(r_squared[0]))
        print("r^2 of educ on logwage = {}".format(r_squared[1]))
correlation coefficient = 0.22388700299289657
r^2 of logwage on educ = 0.22388700299289477
r^2 of educ on logwage = 0.2238870029928961
```

1.2 Question 3:

1.2.1 a)

```
In [4]: q3_a = pd.DataFrame(
            Γ
                list((np.mean(s), np.std(s), len(s))) for s in
                    [f_df.loc[f_df['imm'] == i]['logwage'] for i in [0,1]]
                ]
            , columns = ['mean', 'std', "n"]
        mean\_diff = (q3\_a['mean'][1] - q3\_a['mean'][0])
        se = np.sqrt(sum(q3_a['std']**2 / q3_a['n']))
        t_test = mean_diff/se
        q3_a.rename({0: "non-immigrants", 1: "immigrants"}, axis = 'index',
                    inplace = True)
        display(q3_a)
        print("the t-score for the difference in means test is {}".format(t_test))
        print("the difference in means is: {} \n with se {}".format(mean_diff,se))
                               std
                    mean
                                       n
non-immigrants
               2.886378 0.651487
                                    8616
                2.706393 0.715594 1985
immigrants
the t-score for the difference in means test is -10.268432855577307
the difference in means is: -0.17998575423169516
 with se 0.017528064580364447
  The t-test implies the means are different.
1.2.2 b)
In [5]: models = "logwage ~ 1 + imm"
        res = smf.ols(models, data = f_df).fit()
        beta = res.params
        assert(isclose(mean_diff, beta['imm'], abs_tol = 1e-6))
        print("the OLS estimate for logwage on imm is: {} \n is equal to difference in means:
        sigma = res.bse['imm']
        new_t = res.tvalues['imm']
        print("however, the new std error is {} \n which leads to t-score of: {}".format(sigma
the OLS estimate for logwage on imm is: -0.1799857542317315
```

is equal to difference in means: -0.17998575423169516

```
however, the new std error is 0.016531953532059624 which leads to t-score of: -10.887143729426397
```

which leads to t-score of: -10.268432855579304

This difference is because the original t-test was used heteroskedastic std error, but the regression model used a homoskedastic estimate.

1.2.3 c)

```
In [6]: #refitting data with White's heteroskedastic robust std. error
    het_res = smf.ols(models, data = f_df).fit(cov_type = 'HCO')

het_beta = het_res.params
    assert(isclose(mean_diff, het_beta['imm'], abs_tol = 1e-6))
    print("the OLS estimate for logwage on imm is: {} \n is equal to difference in means:
    het_sigma = het_res.bse['imm']
    het_new_t = het_res.tvalues['imm']
    print("the heteroskedastic robust std error is {} \n which leads to t-score of: {}".for
the OLS estimate for logwage on imm is: -0.1799857542317315
    is equal to difference in means: -0.17998575423169516
```

The heteroskedastic robust std. error is equal to the original std. error which gives rise to the same t-score. The heteroskedastic robust std. error is slightly larger than the homoskedastic std. error.

In []: