Sehgal_Umang_lab03

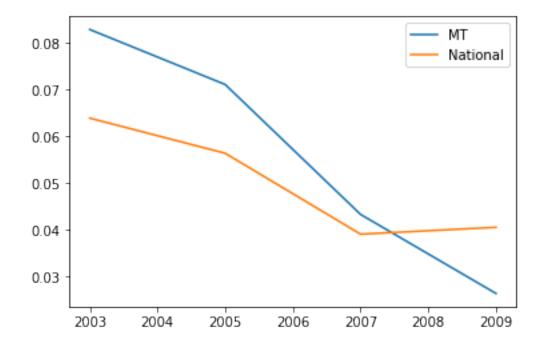
April 18, 2018

```
In [3]: import numpy as np
        import pandas as pd
        import statsmodels.formula.api as smf
        import matplotlib.pyplot as plt
        %matplotlib inline
In [4]: yrbs = pd.read_table('C:/Users/Umang/Downloads/yrbs.tsv.bz2')
        yrbs.head()
Out[4]:
           year
                 age sex
                          meth
                                tv state
        0
          2003
                  13
                       Μ
                                  1
                                       XX
        1 2003
                  13
                       Μ
                             1
                                  1
                                       XX
        2 2003
                  13
                             1
                                 1
                                      XX
                       Μ
        3 2003
                                  1
                                       XX
                  13
                       Μ
                             0
        4 2003
                  13
                             0
                                 0
                                      XX
                       М
0.0.1 1.1 load the data
In [5]: yrbs = pd.read_table('yrbs.tsv.bz2')
        print(yrbs.head())
        print(yrbs.shape)
        print("Unique years: \n",yrbs.year.unique())
        print("\nUnique Age: \n",yrbs.age.unique())
        print(yrbs.sex.unique())
        print(yrbs.meth.unique())
        print(yrbs.tv.unique())
        print(yrbs.state.unique())
                  meth
                        tv state
  year
         age sex
0 2003
          13
               Μ
                     0
                         1
                              XX
1 2003
                              XX
          13
              M
                     1
                         1
2 2003
                              XX
          13
              Μ
                     1
                         1
3 2003
          13
               Μ
                     0
                         1
                              XX
4 2003
          13
               М
                         0
                              XX
(58077, 6)
Unique years:
 [2003 2005 2007 2009]
```

```
Unique Age:
 [13 14 15 16 17]
['M' 'F']
[0 1]
[1 0]
['XX' 'MT']
In [6]: print("Dimension :", np.shape(yrbs))
Dimension: (58077, 6)
In [7]: yrbs['before'] = np.where(yrbs['year'] <= 2005, 'before', 'after')</pre>
In [8]: yrbs.head()
Out[8]:
                                             before
           year
                                  tv state
                  age sex
                           meth
           2003
                   13
                        М
                               0
                                   1
                                         XX
                                             before
        1
           2003
                   13
                                   1
                                         XX
                                             before
                        Μ
                               1
        2
                                             before
           2003
                   13
                                         XX
        3
           2003
                   13
                                   1
                                         XX
                                             before
           2003
                   13
                                   0
                                         XX
                                             before
```

0.0.2 1.2 graphical exploration

Out[9]: <matplotlib.legend.Legend at 0x20d0007f128>



0.0.3 1.3 Before-After

```
In [10]: print(yrbs[yrbs.state == 'MT'].groupby('before').meth.mean())
        m_bae = smf.ols(formula = 'meth ~ before', data = yrbs[yrbs.state == 'MT']).fit()
before
after
          0.038115
before
          0.076734
Name: meth, dtype: float64
```

There was a positive effect of camapign which reduced meth usage in Montana.

```
In [131]: m_bae.summary()
```

Out[131]: <class 'statsmodels.iolib.summary.Summary'>

OLS Regression Results

Dep. Variable:	meth	R-squared:	0.007
Model:	OLS	Adj. R-squared:	0.007
Method:	Least Squares	F-statistic:	67.66
Date:	Wed, 18 Apr 2018	Prob (F-statistic):	2.19e-16
Time:	22:41:25	Log-Likelihood:	417.76
No. Observations:	9754	AIC:	-831.5
Df Residuals:	9752	BIC:	-817.1
Df Model:	1		
Coursiance Tune:	nonrohuat		

		-
${\tt Covariance}$	Type:	nonrobust

	coef	std err	t	P> t	[0.025	0.975]
Intercept before[T.before]	0.0381 0.0386	0.003 0.005	11.484 8.225	0.000	0.032 0.029	0.045
Omnibus: Prob(Omnibus): Skew: Kurtosis:		7326.455 0.000 3.765 15.304	Durbin-Watso Jarque-Bera Prob(JB): Cond. No.		84570. 0	957 020 0.00

[1] Standard Errors assume that the covariance matrix of the errors is correctly spe-11 11 11

The Intercept of the above regression test confirms that after the proportion of meth users was 0.0381 as compared to (0.0381 + 0.0386) 0.0767 before the campaign.

```
In [11]: m_bae_full = smf.ols(formula = 'meth ~ before + age + sex + tv', data = yrbs[yrbs.sta
In [12]: m_bae_full.summary()
Out[12]: <class 'statsmodels.iolib.summary.Summary'>
                           OLS Regression Results
      ______
      Dep. Variable:
                                    R-squared:
                                                             0.009
                               meth
      Model:
                                OLS
                                    Adj. R-squared:
                                                             0.009
      Method:
                                    F-statistic:
                                                             23.20
                        Least Squares
      Date:
                      Wed, 18 Apr 2018 Prob (F-statistic):
                                                          4.11e-19
      Time:
                            23:41:07
                                    Log-Likelihood:
                                                            430.25
      No. Observations:
                                                            -850.5
                               9754
                                    AIC:
                                    BTC:
      Df Residuals:
                               9749
                                                            -814.6
      Df Model:
      Covariance Type:
                           nonrobust
      ______
                                                        [0.025
                                                                 0.975]
                                                P>|t|
                        coef
                             std err
                                          t
      ______
      Intercept
                     -0.1476
                               0.038
                                      -3.911
                                                0.000
                                                        -0.222
                                                                 -0.074
      before[T.before]
                     0.0386
                              0.005
                                      8.221
                                                0.000
                                                        0.029
                                                                 0.048
      sex[T.M]
                     -0.0031
                               0.005
                                      -0.653
                                                0.514
                                                        -0.012
                                                                 0.006
                     0.0118
                                                        0.007
      age
                               0.002
                                      4.953
                                                0.000
                                                                  0.016
                      0.0035
                               0.005
                                       0.739
                                                0.460
                                                        -0.006
                                                                  0.013
      ______
                            7302.251
                                    Durbin-Watson:
      Omnibus:
                                                             1.963
      Prob(Omnibus):
                              0.000
                                    Jarque-Bera (JB):
                                                         83719.425
      Skew:
                              3.750
                                    Prob(JB):
                                                              0.00
      Kurtosis:
                              15.237
                                    Cond. No.
                                                              255.
```

Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly spec

We observe that when multiple variables or controls are added the effect changes for both Montana(drops by 0.1476) and nationally(drops by 0.109). That is there is a reverse impact.

1 Cross section estimator

In [15]: m_cse.summary()

Out[15]: <class 'statsmodels.iolib.summary.Summary'>

 $\Pi \ \Pi \ \Pi$

OLS Regression Results

Dep. Variable:	meth	R-squared:	0.000
Model:	OLS	Adj. R-squared:	-0.000
Method:	Least Squares	F-statistic:	0.3503
Date:	Wed, 18 Apr 2018	Prob (F-statistic):	0.554
Time:	23:41:43	Log-Likelihood:	6404.0
No. Observations:	29728	AIC:	-1.280e+04
Df Residuals:	29726	BIC:	-1.279e+04
Df Model:	1		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
Intercept state[T.XX]	0.0381 0.0018	0.003 0.003	13.648 0.592	0.000 0.554	0.033 -0.004	0.044
Omnibus: Prob(Omnibus): Skew: Kurtosis:		27196.729 0.000 4.720 23.277	Jarque	•	61	1.976 9647.247 0.00 4.75

Warnings:

In [17]: m_cse_full.summary()

[1] Standard Errors assume that the covariance matrix of the errors is correctly spec """

The regression test with no controls shows that for 0.0381 percent effect in Montana the nation-wide effect of campaign for the reduction in meth use was 0.0399

```
In [16]: m_cse_full = smf.ols(formula = 'meth ~ state+ age + sex + tv ', data = yrbs[yrbs.be
```

Out[17]: <class 'statsmodels.iolib.summary.Summary'>

OLS Regression Results

=======================================			==========
Dep. Variable:	meth	R-squared:	0.002
Model:	OLS	Adj. R-squared:	0.001
Method:	Least Squares	F-statistic:	11.60
Date:	Wed, 18 Apr 2018	Prob (F-statistic):	2.07e-09
Time:	23:42:00	Log-Likelihood:	6427.0
No. Observations:	29728	AIC:	-1.284e+04
Df Residuals:	29723	BIC:	-1.280e+04

Df Model:	4
Covariance Type:	nonrobust

	coef	std err	t	P> t	[0.025	0.975]
Intercept state[T.XX] sex[T.M] age	-0.0649 0.0015 0.0081 0.0064 -0.0016	0.018 0.003 0.002 0.001 0.002	-3.615 0.494 3.563 5.648 -0.683	0.000 0.621 0.000 0.000 0.494	-0.100 -0.005 0.004 0.004 -0.006	-0.030 0.008 0.013 0.009 0.003
Omnibus: Prob(Omnibus): Skew: Kurtosis:		27150.346 0.000 4.709 23.214	Durbi			1.980 615983.002 0.00 253.

Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly spec

On adding controls the values of intercepts changed for Montana the campaign effect reversed and meth use increased by a proportion of 0.0634.

Diiferences-in-Differences Estimator

```
In [19]: yrbs[yrbs.before == 'after'].groupby('state').meth.mean()
Out[19]: state
        MT
              0.038115
              0.039923
        Name: meth, dtype: float64
In [20]: yrbs[yrbs.before == 'before'].groupby('state').meth.mean()
Out[20]: state
        MT
              0.076734
              0.060319
        Name: meth, dtype: float64
In [21]: yrbs.groupby(['state', 'before']).meth.mean()
Out[21]: state before
        MT
               after
                        0.038115
               before 0.076734
        XX
               after
                        0.039923
               before 0.060319
        Name: meth, dtype: float64
In [22]: print("Average use in Montana before and after the campaign: ", (abs(0.038115-0.076)
```

```
Average use in Montana before and after the campaign: 0.03861899999999994
```

In [111]: print("\n Average use nationwide before and after the campaign : ", (abs(0.039923-0

Average use nationwide before and after the campaign : 0.02039599999999997

In [23]: print("\n Trend difference : ", (abs(0.0386189999999999999999999999999999)))

Trend difference: 0.0182229999999999

In [25]: m_did = smf.ols(formula = 'meth ~ state*before ', data = yrbs).fit()

In [26]: m_did.summary()

Out[26]: <class 'statsmodels.iolib.summary.Summary'>

|| || ||

OLS Regression Results

Dep. Variable:	meth	R-squared:	0.003
Model:	OLS	Adj. R-squared:	0.003
Method:	Least Squares	F-statistic:	62.92
Date:	Wed, 18 Apr 2018	Prob (F-statistic):	1.31e-40
Time:	23:44:09	Log-Likelihood:	5566.9
No. Observations:	58077	AIC:	-1.113e+04
Df Residuals:	58073	BIC:	-1.109e+04
Df Model:	3		
О			

Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025
Intercept	0.0381	0.003	12.110	0.000	0.032
state[T.XX]	0.0018	0.003	0.525	0.599	-0.005
before[T.before]	0.0386	0.004	8.674	0.000	0.030
state[T.XX]:before[T.before]	-0.0182	0.005	-3.733	0.000	-0.028
				========	====
Omnibus:	46583.747	Durbin-Watso	on:	1.	.971
Proh(Omnibus).	0 000	Tarque-Rera	(TR) ·	670089	493

 Omnibus:
 40583.747
 Durbin-watson:
 1.971

 Prob(Omnibus):
 0.000
 Jarque-Bera (JB):
 670089.493

 Skew:
 4.057
 Prob(JB):
 0.00

 Kurtosis:
 17.528
 Cond. No.
 12.2

Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly spec """

Dif-in-dif regression tells the result of trend difference of simple test and confirms that the campaign had a reverse effect on meth use.

```
In [27]: m_did_full= smf.ols(formula = 'meth ~ state*before + age + sex + tv ', data = yrbs).f
In [120]: m_did_full.summary()
Out[120]: <class 'statsmodels.iolib.summary.Summary'>
```

OLS Regression Results

Dep. Variable:	meth	R-squared:	0.005
Model:	OLS	Adj. R-squared:	0.004
Method:	Least Squares	F-statistic:	44.37
Date:	Wed, 18 Apr 2018	Prob (F-statistic):	1.87e-54
Time:	20:55:51	Log-Likelihood:	5605.5
No. Observations:	58077	AIC:	-1.120e+04
Df Residuals:	58070	BIC:	-1.113e+04
Df Model:	6		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025
Intercept	-0.0727	0.015	-4.925	0.000	-0.102
state[T.XX]	0.0018	0.003	0.521	0.602	-0.005
before[T.before]	0.0389	0.004	8.740	0.000	0.030
sex[T.M]	0.0064	0.002	3.486	0.000	0.003
state[T.XX]:before[T.before]	-0.0185	0.005	-3.785	0.000	-0.028
age	0.0070	0.001	7.630	0.000	0.005
tv	-0.0041	0.002	-2.154	0.031	-0.008
=======================================					====
Omnibus:	46508.899	Durbin-Watso	on:	1	.973
<pre>Prob(Omnibus):</pre>	0.000	Jarque-Bera	(JB):	666670	.662

Warnings:

Skew: Kurtosis:

[1] Standard Errors assume that the covariance matrix of the errors is correctly spe-

Prob(JB):

Cond. No.

0.00

259.

Based on the full regression with all the controls we say that the campaign had efficacy in Montana as compared to the rest of the nation.

4.049

17.489