

ECON662D1__ass3__JSM__code

August 29, 2022

```
[1]: # ECON662D1 Assignment 3 - Jean-Sébastien Matte 260913682
```

```
[2]: # import modules
import numpy as np
import pandas as pd
import scipy as sp
import statsmodels.api as sm
import matplotlib
matplotlib.use("pgf")
matplotlib.rcParams.update({
    "pgf.texsystem": "pdflatex",
    'font.family': 'serif',
    'text.usetex': True,
    'pgf.rcfonts': False,
})

np.random.seed(123)
```

```
[3]: # load data
in_dir = '/Users/jsmatte/github/ECON662D1/Assignment3'
in_file = str(in_dir + '/e662.as3.20.dat.txt')
df = pd.read_csv(in_file, sep = '\s+', header = None)
print(df)
```

	0	1	2	3
0	-0.429193	91.52508	-7.879150	-22.82027
1	-1.719862	58.32108	-10.377350	-46.89894
2	-5.007316	71.19264	8.240685	-83.72890
3	-0.287603	85.37536	-7.939428	-98.93214
4	-5.385808	84.63815	28.309510	-80.68243
..
95	-4.539189	26.54987	-40.947840	-146.96500
96	-9.798897	69.94029	75.729620	-126.41690
97	-2.504187	89.50351	-44.335000	-76.11952
98	-2.118638	100.67600	48.674720	-16.33207
99	-0.782366	133.57730	4.789128	-10.57532

[100 rows x 4 columns]

```
[4]: df.columns = ['y', 'x1', 'x2', 'x3']
      print(df)
```

	y	x1	x2	x3
0	-0.429193	91.52508	-7.879150	-22.82027
1	-1.719862	58.32108	-10.377350	-46.89894
2	-5.007316	71.19264	8.240685	-83.72890
3	-0.287603	85.37536	-7.939428	-98.93214
4	-5.385808	84.63815	28.309510	-80.68243
..
95	-4.539189	26.54987	-40.947840	-146.96500
96	-9.798897	69.94029	75.729620	-126.41690
97	-2.504187	89.50351	-44.335000	-76.11952
98	-2.118638	100.67600	48.674720	-16.33207
99	-0.782366	133.57730	4.789128	-10.57532

[100 rows x 4 columns]

```
[5]: # create matrix X of the regressors (unrestricted model)
      X = np.zeros((len(df),4))
      for i in range(X.shape[0]):
          X[i] = [1, df.x1.values[i], df.x2.values[i], df.x3.values[i]]
      print(X.shape)

      # create vector y of observations
      y = np.array([[j] for j in df.y.values])
      print(y.shape)
```

(100, 4)

(100, 1)

```
[6]: # Create the different OLS models of different sample sizes, and collect data
      n = 17
      H1 = []
      H2 = []
      for i in range(n):
          print('MODEL_' + str(i))

          # unrestricted model & results
          globals()['model_' + str(i)] = sm.OLS(y[0:(20+5*i)], :, X[0:(20+5*i)], :).
          fit()

          temp_reg = globals()['model_' + str(i)]
          print(temp_reg.summary())
          print('')

          temp_resid = temp_reg.resid
          temp_fitval = temp_reg.fittedvalues
```

```

# Hypothesis 1:  $H_0: B_1 = 0$ 
print('Hypothesis 1:  $H_0: B_1 = 0$ ')
temp_P_H1 = []

# restricted model & results
X_rest1 = X[0:(20+5*i), [0, 2, 3]]
globals()['model_' + str(i) + '_rest1'] = sm.OLS(y[0:(20+5*i), :], X_rest1).
↪fit()
temp_reg_rest1 = globals()['model_' + str(i) + '_rest1']

temp_resid_rest1 = temp_reg_rest1.resid
temp_fitval_rest1 = temp_reg_rest1.fittedvalues

# p-value based on Student t-test
print('P-value based on Student t-test')
print(temp_reg.t_test('x1 = 0'))
temp_P_H1.append(temp_reg.pvalues[1])
print('')

# P-value from rescaled restricted residuals bootstrap
temp_nobs = temp_reg.nobs

# rescale residuals
rescale_factor = np.sqrt(temp_nobs / (temp_nobs - 3))
rescaled_temp_resid_rest1 = rescale_factor * temp_resid_rest1

boot_t = []
for b in range(999):
    # draw from rescaled restricted disturbances
    draws_rest1 = np.random.choice(rescaled_temp_resid_rest1, size = ↪
↪len(rescaled_temp_resid_rest1))

    # update observations using rescaled, restricted, resampled residuals
    temp_y_star_rest1 = temp_fitval_rest1 + draws_rest1
    temp_y_star_rest1 = np.array([[j] for j in temp_y_star_rest1])

    # Regress y* on the original covariates
    temp_reg_star = sm.OLS(temp_y_star_rest1, X[0:(20+5*i), :]).fit()

    # Bootstrap statistic
    boot_t.append(temp_reg_star.tvalues[1])

boot_cnt_right = 0

```

```

boot_cnt_left = 0
for bt in boot_t:
    if bt <= temp_reg.tvalues[1]:
        boot_cnt_right += 1
    elif bt > temp_reg.tvalues[1]:
        boot_cnt_left += 1
boot_p = 2 * min(boot_cnt_right / 999, boot_cnt_left / 999)
temp_P_H1.append(boot_p)
print('P-value from rescaled restricted residuals bootstrap')
print(boot_p)
print('')

# P-value from wild bootstrap
# define the pmf of s* as proposed by Mammen
xk = [-1, 1] # -> the rv_discrete class is only defined for xk values that
are integers, thus we define the
        # pmf as such. after random draws, we reassign the correct
values to the s* vector and carry on
        # with our computation
pk = [((np.sqrt(5)+1) / (2*np.sqrt(5))), ((np.sqrt(5)-1) / (2*np.sqrt(5)))]
s_pmf = sp.stats.rv_discrete(name = 's_pmf', values = (xk, pk))

boot_t_wild = []
for b_wild in range(999):
    # s_star
    s_star_raw = s_pmf.rvs(size = len(temp_resid_rest1))

    s_star = []
    for s in s_star_raw:
        if s == -1:
            s_star.append(-(np.sqrt(5) - 1) / 2)
        elif s == 1:
            s_star.append((np.sqrt(5) + 1) / 2)

    # update observations using wild disturbances
    resid_star_rest1 = []
    for resid_idx in range(len(temp_resid_rest1)):
        resid_star_rest1.append(s_star[resid_idx] *
temp_resid_rest1[resid_idx])

    temp_y_star_wild_rest1 = temp_fitval_rest1 + resid_star_rest1
    temp_y_star_wild_rest1 = np.array([[q] for q in temp_y_star_wild_rest1])

    # Regress y* on the original covariates
    temp_reg_wild = sm.OLS(temp_y_star_wild_rest1, X[0:(20+5*i), :]).fit()

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    # Wild Bootstrap statistic
    boot_t_wild.append(temp_reg_wild.tvalues[1])

boot_wild_cnt_right = 0
boot_wild_cnt_left = 0
for bt_wild in boot_t_wild:
    if bt_wild <= temp_reg.tvalues[1]:
        boot_wild_cnt_right += 1
    elif bt_wild > temp_reg.tvalues[1]:
        boot_wild_cnt_left += 1
boot_p_wild = 2 * min(boot_wild_cnt_right / 999, boot_wild_cnt_left / 999)
temp_P_H1.append(boot_p_wild)

print('P-value from wild bootstrap')
print(boot_p_wild)
print('')

H1.append(temp_P_H1)

# Hypothesis 2: H0: B1 = 0.1
print('Hypothesis 2: H0: B1 = 0.1')
temp_P_H2 = []

# restricted model & results
temp_X2 = X[0:(20+5*i), 1]
temp_X2 = np.array([[x] for x in temp_X2])
y_rest2 = y[0:(20+5*i), :] - 0.1 * temp_X2
X_rest2 = X_rest1
globals()['model_' + str(i) + '_rest2'] = sm.OLS(y_rest2, X_rest2).fit()
temp_reg_rest2 = globals()['model_' + str(i) + '_rest2']

temp_resid_rest2 = temp_reg_rest2.resid
temp_fitval_rest2 = temp_reg_rest2.fittedvalues

# P-value based on Student t-test
print('P-value based on Student t-test')
print(temp_reg.t_test('x1 = 0.1'))
temp_P_H2.append(temp_reg.t_test('x1 = 0.1').pvalue)
print('')

# P-value from rescaled restricted residuals bootstrap
# rescale residuals
rescaled_temp_resid_rest2 = rescale_factor * temp_resid_rest2

```

```

boot_t2 = []
for b2 in range(999):
    # draw from rescaled restricted disturbances
    draws_rest2 = np.random.choice(rescaled_temp_resid_rest2, size =
↳len(rescaled_temp_resid_rest2))

    # update observations using rescaled, restricted, resampled residuals
    temp_y_star_rest2 = temp_fitval_rest2 + draws_rest2
    temp_y_star_rest2 = np.array([[j] for j in temp_y_star_rest2])

    # Regress y* on the original covariates
    temp_reg_star2 = sm.OLS(temp_y_star_rest2, X[0:(20+5*i), :]).fit()

    # Bootstrap statistic
    boot_t2.append(temp_reg_star2.t_test('x1 = 0.1').tvalue[0][0])

boot_cnt_right2 = 0
boot_cnt_left2 = 0
for bt2 in boot_t2:
    if bt2 <= temp_reg.tvalues[1]:
        boot_cnt_right2 += 1
    elif bt2 > temp_reg.tvalues[1]:
        boot_cnt_left2 += 1
boot_p2 = 2 * min(boot_cnt_right2 / 999, boot_cnt_left2 / 999)
temp_P_H2.append(boot_p2)
print('P-value from rescaled restricted residuals bootstrap')
print(boot_p2)
print('')

# P-value from wild bootstrap
boot_t_wild2 = []
for b_wild2 in range(999):
    # s_star
    s_star_raw2 = s_pmf.rvs(size = len(temp_resid_rest2))

    s_star2 = []
    for s2 in s_star_raw2:
        if s2 == -1:
            s_star2.append(-(np.sqrt(5) - 1) / 2)
        elif s2 == 1:
            s_star2.append((np.sqrt(5) + 1) / 2)

    # update observations using wild disturbances
    resid_star_rest2 = []
    for resid_idx2 in range(len(temp_resid_rest2)):

```

```

        resid_star_rest2.append(s_star2[resid_idx2] *
↪temp_resid_rest2[resid_idx2])

temp_y_star_wild_rest2 = temp_fitval_rest2 + resid_star_rest2
temp_y_star_wild_rest2 = np.array([[q] for q in temp_y_star_wild_rest2])

# Regress y* on the original covariates
temp_reg_wild2 = sm.OLS(temp_y_star_wild_rest2, X[0:(20+5*i), :]).fit()

# Wild Bootstrap statistic
boot_t_wild2.append(temp_reg_wild2.t_test('x1 = 0.1').tvalue[0][0])

boot_wild_cnt_right2 = 0
boot_wild_cnt_left2 = 0
for bt_wild2 in boot_t_wild2:
    if bt_wild2 <= temp_reg.tvalues[1]:
        boot_wild_cnt_right2 += 1
    elif bt_wild2 > temp_reg.tvalues[1]:
        boot_wild_cnt_left2 += 1
boot_p_wild2 = 2 * min(boot_wild_cnt_right2 / 999, boot_wild_cnt_left2 /
↪999)
temp_P_H2.append(boot_p_wild2)

print('P-value from wild bootstrap')
print(boot_p_wild2)
print('')

H2.append(temp_P_H2)

print('-----')
print('')

```

MODEL_0

OLS Regression Results

```

=====
Dep. Variable:          y      R-squared:          0.544
Model:                  OLS    Adj. R-squared:       0.458
Method:                 Least Squares  F-statistic:        6.351
Date:                  Tue, 01 Dec 2020  Prob (F-statistic):    0.00485
Time:                  14:30:45  Log-Likelihood:      -34.074
No. Observations:      20      AIC:                76.15
Df Residuals:          16      BIC:                80.13
Df Model:              3
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]

const	-0.6174	1.558	-0.396	0.697	-3.920	2.685
x1	0.0056	0.015	0.366	0.719	-0.027	0.038
x2	-0.0424	0.012	-3.534	0.003	-0.068	-0.017
x3	0.0332	0.013	2.478	0.025	0.005	0.062

Omnibus:	3.991	Durbin-Watson:	2.331
Prob(Omnibus):	0.136	Jarque-Bera (JB):	2.583
Skew:	0.878	Prob(JB):	0.275
Kurtosis:	3.138	Cond. No.	479.

Notes:

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

Hypothesis 1: $H_0: B_1 = 0$

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
c0	0.0056	0.015	0.366	0.719	-0.027	0.038

P-value from rescaled restricted residuals bootstrap

0.7267267267267268

P-value from wild bootstrap

0.6486486486486487

Hypothesis 2: $H_0: B_1 = 0.1$

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
c0	0.0056	0.015	-6.202	0.000	-0.027	0.038

P-value from rescaled restricted residuals bootstrap

0.0

P-value from wild bootstrap

0.0

MODEL_1

OLS Regression Results

```

=====
Dep. Variable:          y      R-squared:          0.648
Model:                  OLS    Adj. R-squared:      0.597
Method:                 Least Squares    F-statistic:      12.86
Date:                  Tue, 01 Dec 2020    Prob (F-statistic):  5.48e-05
Time:                  14:30:51    Log-Likelihood:     -42.146
No. Observations:      25    AIC:              92.29
Df Residuals:          21    BIC:              97.17
Df Model:               3
Covariance Type:        nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0281	1.275	0.022	0.983	-2.623	2.679
x1	0.0075	0.013	0.571	0.574	-0.020	0.035
x2	-0.0402	0.011	-3.807	0.001	-0.062	-0.018
x3	0.0473	0.009	5.354	0.000	0.029	0.066

```

=====
Omnibus:                8.026    Durbin-Watson:          2.311
Prob(Omnibus):           0.018    Jarque-Bera (JB):        6.397
Skew:                    1.210    Prob(JB):                0.0408
Kurtosis:                3.533    Cond. No.:               456.
=====

```

Notes:

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

Hypothesis 1: $H_0: B_1 = 0$

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
c0	0.0075	0.013	0.571	0.574	-0.020	0.035

P-value from rescaled restricted residuals bootstrap
0.5945945945945946

P-value from wild bootstrap
0.47047047047047047

Hypothesis 2: $H_0: B_1 = 0.1$

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
c0	0.0075	0.013	-7.078	0.000	-0.020	0.035

P-value from rescaled restricted residuals bootstrap
0.0

P-value from wild bootstrap
0.0

MODEL_2

OLS Regression Results

```

=====
Dep. Variable:          y      R-squared:          0.601
Model:                  OLS    Adj. R-squared:       0.555
Method:                 Least Squares  F-statistic:      13.05
Date:                  Tue, 01 Dec 2020  Prob (F-statistic): 2.16e-05
Time:                  14:30:57  Log-Likelihood:   -52.025
No. Observations:      30      AIC:              112.1
Df Residuals:          26      BIC:              117.7
Df Model:               3
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	-0.4020	1.276	-0.315	0.755	-3.024	2.220
x1	0.0116	0.013	0.896	0.378	-0.015	0.038
x2	-0.0379	0.010	-3.686	0.001	-0.059	-0.017
x3	0.0447	0.009	5.177	0.000	0.027	0.062

```

=====
Omnibus:                 6.526  Durbin-Watson:       2.496
Prob(Omnibus):           0.038  Jarque-Bera (JB):     5.529
Skew:                    1.048  Prob(JB):             0.0630
Kurtosis:                 3.162  Cond. No.             475.
=====

```

Notes:

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

Hopthosis 1: $H_0: B_1 = 0$

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
--	------	---------	---	------	--------	--------

```
-----
c0          0.0116      0.013      0.896      0.378      -0.015      0.038
=====
```

P-value from rescaled restricted residuals bootstrap
0.36236236236236236

P-value from wild bootstrap
0.25825825825825827

Hopothesis 2: $H_0: B_1 = 0.1$

P-value based on Student t-test

Test for Constraints

```
=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
c0          0.0116      0.013      -6.798      0.000      -0.015      0.038
=====
```

P-value from rescaled restricted residuals bootstrap
0.0

P-value from wild bootstrap
0.0

MODEL_3

OLS Regression Results

```
=====
Dep. Variable:          y      R-squared:          0.603
Model:                OLS      Adj. R-squared:      0.565
Method:             Least Squares      F-statistic:      15.71
Date:                Tue, 01 Dec 2020      Prob (F-statistic):      2.15e-06
Time:                14:31:03      Log-Likelihood:      -59.019
No. Observations:      35      AIC:          126.0
Df Residuals:          31      BIC:          132.3
Df Model:              3
Covariance Type:      nonrobust
=====
```

```
=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
const        -0.1940      1.166      -0.166      0.869      -2.571      2.183
x1           0.0082      0.012      0.711      0.482      -0.015      0.032
x2          -0.0394      0.009      -4.296      0.000      -0.058      -0.021
x3           0.0455      0.008      5.657      0.000      0.029      0.062
=====
```

```
Omnibus:          11.785      Durbin-Watson:          2.394
```

Prob(Omnibus):	0.003	Jarque-Bera (JB):	11.410
Skew:	1.332	Prob(JB):	0.00333
Kurtosis:	3.854	Cond. No.	503.

Notes:

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

Hypothesis 1: $H_0: B_1 = 0$

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
c0	0.0082	0.012	0.711	0.482	-0.015	0.032

P-value from rescaled restricted residuals bootstrap

0.4964964964964965

P-value from wild bootstrap

0.3303303303303303

Hypothesis 2: $H_0: B_1 = 0.1$

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
c0	0.0082	0.012	-7.942	0.000	-0.015	0.032

P-value from rescaled restricted residuals bootstrap

0.0

P-value from wild bootstrap

0.0

MODEL_4

OLS Regression Results

Dep. Variable:	y	R-squared:	0.607
Model:	OLS	Adj. R-squared:	0.574
Method:	Least Squares	F-statistic:	18.54
Date:	Tue, 01 Dec 2020	Prob (F-statistic):	1.93e-07

Time: 14:31:09 Log-Likelihood: -66.268
 No. Observations: 40 AIC: 140.5
 Df Residuals: 36 BIC: 147.3
 Df Model: 3
 Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
const	-0.2863	1.084	-0.264	0.793	-2.484	1.911
x1	0.0080	0.010	0.776	0.443	-0.013	0.029
x2	-0.0381	0.009	-4.341	0.000	-0.056	-0.020
x3	0.0445	0.007	6.067	0.000	0.030	0.059
Omnibus:	12.970		Durbin-Watson:	2.441		
Prob(Omnibus):	0.002		Jarque-Bera (JB):	13.203		
Skew:	1.318		Prob(JB):	0.00136		
Kurtosis:	3.989		Cond. No.	515.		

Notes:

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

Hypothesis 1: $H_0: B_1 = 0$

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
c0	0.0080	0.010	0.776	0.443	-0.013	0.029

P-value from rescaled restricted residuals bootstrap

0.44644644644644643

P-value from wild bootstrap

0.3383383383383383

Hypothesis 2: $H_0: B_1 = 0.1$

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
c0	0.0080	0.010	-8.925	0.000	-0.013	0.029

P-value from rescaled restricted residuals bootstrap

0.0

P-value from wild bootstrap

0.0

MODEL_5

OLS Regression Results

```
=====
Dep. Variable:          y      R-squared:          0.631
Model:                  OLS    Adj. R-squared:      0.604
Method:                 Least Squares    F-statistic:      23.41
Date:                  Tue, 01 Dec 2020    Prob (F-statistic):    5.44e-09
Time:                  14:31:15    Log-Likelihood:      -72.622
No. Observations:      45    AIC:              153.2
Df Residuals:          41    BIC:              160.5
Df Model:               3
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	-0.0868	0.942	-0.092	0.927	-1.989	1.815
x1	0.0059	0.009	0.652	0.518	-0.012	0.024
x2	-0.0380	0.008	-4.808	0.000	-0.054	-0.022
x3	0.0447	0.007	6.838	0.000	0.032	0.058

```
=====
Omnibus:                14.700    Durbin-Watson:          2.429
Prob(Omnibus):          0.001    Jarque-Bera (JB):        15.915
Skew:                   1.327    Prob(JB):                0.000350
Kurtosis:               4.200    Cond. No.                489.
=====
```

Notes:

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

Hopthosis 1: $H_0: B_1 = 0$

P-value based on Student t-test

Test for Constraints

```
=====
          coef    std err          t      P>|t|      [0.025      0.975]
-----
c0          0.0059     0.009     0.652     0.518     -0.012     0.024
=====
```

P-value from rescaled restricted residuals bootstrap

0.5205205205205206

P-value from wild bootstrap
0.47847847847847846

Hopthesis 2: $H_0: B_1 = 0.1$

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
c0	0.0059	0.009	-10.302	0.000	-0.012	0.024

P-value from rescaled restricted residuals bootstrap
0.0

P-value from wild bootstrap
0.0

MODEL_6

OLS Regression Results

```

=====
Dep. Variable:          y      R-squared:          0.705
Model:                OLS      Adj. R-squared:       0.686
Method:             Least Squares      F-statistic:       36.70
Date:                Tue, 01 Dec 2020    Prob (F-statistic):   2.90e-12
Time:                14:31:20      Log-Likelihood:     -79.054
No. Observations:      50      AIC:              166.1
Df Residuals:          46      BIC:              173.8
Df Model:              3
Covariance Type:      nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0614	0.879	0.070	0.945	-1.708	1.831
x1	0.0060	0.008	0.705	0.484	-0.011	0.023
x2	-0.0397	0.007	-5.365	0.000	-0.055	-0.025
x3	0.0479	0.006	8.614	0.000	0.037	0.059

```

=====
Omnibus:              19.469      Durbin-Watson:       2.409
Prob(Omnibus):        0.000      Jarque-Bera (JB):    24.528
Skew:                 1.485      Prob(JB):            4.72e-06
Kurtosis:             4.718      Cond. No.            511.
=====

```

Notes:

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

Hopthosis 1: $H_0: B_1 = 0$

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
c0	0.0060	0.008	0.705	0.484	-0.011	0.023

P-value from rescaled restricted residuals bootstrap

0.4844844844844845

P-value from wild bootstrap

0.42042042042042044

Hopthosis 2: $H_0: B_1 = 0.1$

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
c0	0.0060	0.008	-11.118	0.000	-0.011	0.023

P-value from rescaled restricted residuals bootstrap

0.0

P-value from wild bootstrap

0.0

MODEL_7

OLS Regression Results

Dep. Variable:	y	R-squared:	0.712
Model:	OLS	Adj. R-squared:	0.695
Method:	Least Squares	F-statistic:	42.07
Date:	Tue, 01 Dec 2020	Prob (F-statistic):	7.89e-14
Time:	14:31:26	Log-Likelihood:	-84.668
No. Observations:	55	AIC:	177.3
Df Residuals:	51	BIC:	185.4
Df Model:	3		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
-----	-----	-----	-----	-----	-----	-----
const	0.0656	0.832	0.079	0.937	-1.604	1.735
x1	0.0054	0.008	0.683	0.498	-0.011	0.021
x2	-0.0396	0.007	-6.022	0.000	-0.053	-0.026
x3	0.0479	0.005	9.117	0.000	0.037	0.058
=====	=====	=====	=====	=====	=====	=====
Omnibus:		24.346	Durbin-Watson:			2.396
Prob(Omnibus):		0.000	Jarque-Bera (JB):			35.789
Skew:		1.617	Prob(JB):			1.69e-08
Kurtosis:		5.270	Cond. No.			530.
=====	=====	=====	=====	=====	=====	=====

Notes:

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

Hypothesis 1: $H_0: B_1 = 0$

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
-----	-----	-----	-----	-----	-----	-----
c0	0.0054	0.008	0.683	0.498	-0.011	0.021
=====	=====	=====	=====	=====	=====	=====

P-value from rescaled restricted residuals bootstrap

0.5225225225225225

P-value from wild bootstrap

0.45245245245245247

Hypothesis 2: $H_0: B_1 = 0.1$

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
-----	-----	-----	-----	-----	-----	-----
c0	0.0054	0.008	-11.844	0.000	-0.011	0.021
=====	=====	=====	=====	=====	=====	=====

P-value from rescaled restricted residuals bootstrap

0.0

P-value from wild bootstrap

0.0

MODEL_8

OLS Regression Results

```

=====
Dep. Variable:          y      R-squared:          0.695
Model:                OLS     Adj. R-squared:       0.678
Method:              Least Squares   F-statistic:       42.46
Date:                Tue, 01 Dec 2020   Prob (F-statistic): 1.91e-14
Time:                14:31:32   Log-Likelihood:    -93.707
No. Observations:      60      AIC:              195.4
Df Residuals:          56      BIC:              203.8
Df Model:              3
Covariance Type:      nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0668	0.818	0.082	0.935	-1.573	1.706
x1	0.0048	0.008	0.619	0.539	-0.011	0.020
x2	-0.0377	0.007	-5.750	0.000	-0.051	-0.025
x3	0.0475	0.005	9.080	0.000	0.037	0.058

```

=====
Omnibus:                22.942   Durbin-Watson:          2.402
Prob(Omnibus):          0.000   Jarque-Bera (JB):        31.724
Skew:                   1.519   Prob(JB):                1.29e-07
Kurtosis:               4.860   Cond. No.:               537.
=====

```

Notes:

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

Hypothesis 1: $H_0: B_1 = 0$

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
c0	0.0048	0.008	0.619	0.539	-0.011	0.020

P-value from rescaled restricted residuals bootstrap

0.4984984984984985

P-value from wild bootstrap

0.4624624624624625

Hypothesis 2: $H_0: B_1 = 0.1$

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
c0	0.0048	0.008	-12.341	0.000	-0.011	0.020

P-value from rescaled restricted residuals bootstrap
0.0

P-value from wild bootstrap
0.0

MODEL_9

OLS Regression Results

```

=====
Dep. Variable:          y      R-squared:          0.701
Model:                OLS     Adj. R-squared:       0.687
Method:             Least Squares   F-statistic:      47.78
Date:                Tue, 01 Dec 2020   Prob (F-statistic): 5.15e-16
Time:                  14:31:38   Log-Likelihood:   -99.692
No. Observations:      65      AIC:              207.4
Df Residuals:          61      BIC:              216.1
Df Model:                3
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	-0.1765	0.742	-0.238	0.813	-1.661	1.308
x1	0.0069	0.007	0.995	0.323	-0.007	0.021
x2	-0.0382	0.006	-6.221	0.000	-0.051	-0.026
x3	0.0469	0.005	9.331	0.000	0.037	0.057

```

=====
Omnibus:                26.144   Durbin-Watson:          2.358
Prob(Omnibus):           0.000   Jarque-Bera (JB):       39.406
Skew:                    1.578   Prob(JB):               2.77e-09
Kurtosis:                 5.142   Cond. No.                516.
=====

```

Notes:

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

Hopothesis 1: H0: B1 = 0

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
c0	0.0069	0.007	0.995	0.323	-0.007	0.021

P-value from rescaled restricted residuals bootstrap
0.36236236236236236

P-value from wild bootstrap
0.28428428428428426

Hopthesis 2: $H_0: B_1 = 0.1$

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
c0	0.0069	0.007	-13.376	0.000	-0.007	0.021

P-value from rescaled restricted residuals bootstrap
0.0

P-value from wild bootstrap
0.0

MODEL_10

OLS Regression Results

```

=====
Dep. Variable:          y      R-squared:          0.720
Model:                  OLS    Adj. R-squared:      0.707
Method:                 Least Squares  F-statistic:      56.43
Date:                   Tue, 01 Dec 2020  Prob (F-statistic):  3.38e-18
Time:                   14:31:44  Log-Likelihood:    -106.59
No. Observations:      70      AIC:              221.2
Df Residuals:          66      BIC:              230.2
Df Model:               3
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	-0.1127	0.708	-0.159	0.874	-1.526	1.300
x1	0.0065	0.007	0.977	0.332	-0.007	0.020
x2	-0.0394	0.006	-6.719	0.000	-0.051	-0.028
x3	0.0473	0.005	10.051	0.000	0.038	0.057

```
=====
Omnibus:                27.727    Durbin-Watson:                2.389
Prob(Omnibus):          0.000    Jarque-Bera (JB):        42.925
Skew:                   1.588    Prob(JB):                4.78e-10
Kurtosis:               5.152    Cond. No.                519.
=====
```

Notes:

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

Hypothesis 1: $H_0: B_1 = 0$

P-value based on Student t-test

Test for Constraints

```
=====
              coef    std err          t      P>|t|      [0.025    0.975]
-----
c0            0.0065    0.007      0.977      0.332      -0.007     0.020
=====
```

P-value from rescaled restricted residuals bootstrap

0.2902902902902903

P-value from wild bootstrap

0.3023023023023023

Hypothesis 2: $H_0: B_1 = 0.1$

P-value based on Student t-test

Test for Constraints

```
=====
              coef    std err          t      P>|t|      [0.025    0.975]
-----
c0            0.0065    0.007     -14.031      0.000      -0.007     0.020
=====
```

P-value from rescaled restricted residuals bootstrap

0.0

P-value from wild bootstrap

0.0

MODEL_11

OLS Regression Results

```
=====
Dep. Variable:          y    R-squared:                0.738
Model:                OLS    Adj. R-squared:            0.727
=====
```

```

Method:                Least Squares      F-statistic:                66.74
Date:                  Tue, 01 Dec 2020    Prob (F-statistic):         1.27e-20
Time:                  14:31:52           Log-Likelihood:             -112.36
No. Observations:      75                AIC:                        232.7
Df Residuals:          71                BIC:                        242.0
Df Model:              3
Covariance Type:       nonrobust

```

```

=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
const        -0.0612      0.684       -0.089      0.929      -1.426      1.303
x1             0.0058      0.006        0.911      0.366      -0.007      0.018
x2           -0.0394      0.006       -7.033      0.000      -0.051     -0.028
x3             0.0476      0.004       10.747      0.000       0.039      0.056
=====
Omnibus:                 31.414    Durbin-Watson:                2.392
Prob(Omnibus):            0.000    Jarque-Bera (JB):            53.210
Skew:                     1.654    Prob(JB):                    2.79e-12
Kurtosis:                 5.466    Cond. No.                     532.
=====

```

Notes:

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

Hypothesis 1: $H_0: B_1 = 0$

P-value based on Student t-test

Test for Constraints

```

=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
c0             0.0058      0.006        0.911      0.366      -0.007      0.018
=====

```

P-value from rescaled restricted residuals bootstrap

0.34434434434434436

P-value from wild bootstrap

0.3383383383383383

Hypothesis 2: $H_0: B_1 = 0.1$

P-value based on Student t-test

Test for Constraints

```

=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
c0             0.0058      0.006     -14.826      0.000      -0.007      0.018
=====

```

P-value from rescaled restricted residuals bootstrap
0.0

P-value from wild bootstrap
0.0

MODEL_12

OLS Regression Results

```
=====
Dep. Variable:          y      R-squared:          0.734
Model:                  OLS    Adj. R-squared:      0.724
Method:                 Least Squares  F-statistic:    70.06
Date:                  Tue, 01 Dec 2020  Prob (F-statistic): 7.95e-22
Time:                  14:32:00  Log-Likelihood:  -119.82
No. Observations:      80      AIC:            247.6
Df Residuals:          76      BIC:            257.2
Df Model:              3
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	-0.0231	0.658	-0.035	0.972	-1.333	1.287
x1	0.0056	0.006	0.939	0.351	-0.006	0.018
x2	-0.0409	0.005	-7.592	0.000	-0.052	-0.030
x3	0.0478	0.004	10.949	0.000	0.039	0.057

```
=====
Omnibus:                31.431  Durbin-Watson:          2.430
Prob(Omnibus):          0.000  Jarque-Bera (JB):        52.029
Skew:                   1.611  Prob(JB):                5.03e-12
Kurtosis:               5.286  Cond. No.                 536.
=====
```

Notes:

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

Hypothesis 1: $H_0: B_1 = 0$

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
c0	0.0056	0.006	0.939	0.351	-0.006	0.018

P-value from rescaled restricted residuals bootstrap
0.36436436436436437

P-value from wild bootstrap
0.2982982982982983

Hopothesis 2: $H_0: B_1 = 0.1$

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
c0	0.0056	0.006	-15.794	0.000	-0.006	0.018

P-value from rescaled restricted residuals bootstrap
0.0

P-value from wild bootstrap
0.0

MODEL_13

OLS Regression Results

Dep. Variable:	y	R-squared:	0.725
Model:	OLS	Adj. R-squared:	0.714
Method:	Least Squares	F-statistic:	71.05
Date:	Tue, 01 Dec 2020	Prob (F-statistic):	1.29e-22
Time:	14:32:08	Log-Likelihood:	-127.58
No. Observations:	85	AIC:	263.2
Df Residuals:	81	BIC:	272.9
Df Model:	3		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	-0.2322	0.649	-0.358	0.722	-1.524	1.059
x1	0.0074	0.006	1.240	0.218	-0.004	0.019
x2	-0.0417	0.005	-7.800	0.000	-0.052	-0.031
x3	0.0462	0.004	10.913	0.000	0.038	0.055

Omnibus:	30.066	Durbin-Watson:	2.332
Prob(Omnibus):	0.000	Jarque-Bera (JB):	46.956
Skew:	1.533	Prob(JB):	6.36e-11
Kurtosis:	4.964	Cond. No.	547.

Notes:

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

Hopothesis 1: $H_0: B_1 = 0$

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
c0	0.0074	0.006	1.240	0.218	-0.004	0.019

P-value from rescaled restricted residuals bootstrap

0.23423423423423423

P-value from wild bootstrap

0.1881881881881882

Hopothesis 2: $H_0: B_1 = 0.1$

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
c0	0.0074	0.006	-15.633	0.000	-0.004	0.019

P-value from rescaled restricted residuals bootstrap

0.0

P-value from wild bootstrap

0.0

MODEL_14

OLS Regression Results

Dep. Variable:	y	R-squared:	0.736
Model:	OLS	Adj. R-squared:	0.726
Method:	Least Squares	F-statistic:	79.78
Date:	Tue, 01 Dec 2020	Prob (F-statistic):	9.13e-25
Time:	14:32:16	Log-Likelihood:	-133.14
No. Observations:	90	AIC:	274.3
Df Residuals:	86	BIC:	284.3
Df Model:	3		

Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
const	-0.1376	0.617	-0.223	0.824	-1.365	1.090
x1	0.0060	0.006	1.068	0.288	-0.005	0.017
x2	-0.0418	0.005	-8.265	0.000	-0.052	-0.032
x3	0.0463	0.004	11.605	0.000	0.038	0.054
Omnibus:		34.542	Durbin-Watson:			2.375
Prob(Omnibus):		0.000	Jarque-Bera (JB):			59.314
Skew:		1.620	Prob(JB):			1.32e-13
Kurtosis:		5.306	Cond. No.			552.

Notes:

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

Hypothesis 1: $H_0: B_1 = 0$

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
c0	0.0060	0.006	1.068	0.288	-0.005	0.017

P-value from rescaled restricted residuals bootstrap

0.24824824824824826

P-value from wild bootstrap

0.24624624624624625

Hypothesis 2: $H_0: B_1 = 0.1$

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
c0	0.0060	0.006	-16.847	0.000	-0.005	0.017

P-value from rescaled restricted residuals bootstrap

0.0

P-value from wild bootstrap

0.0

MODEL_15

OLS Regression Results

```
=====
Dep. Variable:          y      R-squared:          0.732
Model:                  OLS    Adj. R-squared:      0.723
Method:                 Least Squares  F-statistic:    82.92
Date:                  Tue, 01 Dec 2020  Prob (F-statistic): 6.13e-26
Time:                  14:32:23  Log-Likelihood:  -139.57
No. Observations:      95      AIC:            287.1
Df Residuals:          91      BIC:            297.4
Df Model:               3
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	-0.3144	0.580	-0.542	0.589	-1.466	0.837
x1	0.0078	0.005	1.487	0.140	-0.003	0.018
x2	-0.0422	0.005	-8.529	0.000	-0.052	-0.032
x3	0.0457	0.004	11.767	0.000	0.038	0.053

```
=====
Omnibus:                34.800  Durbin-Watson:          2.361
Prob(Omnibus):           0.000  Jarque-Bera (JB):        59.065
Skew:                    1.584  Prob(JB):                1.49e-13
Kurtosis:                 5.212  Cond. No.                 535.
=====
```

Notes:

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

Hypothesis 1: $H_0: B_1 = 0$

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
c0	0.0078	0.005	1.487	0.140	-0.003	0.018

P-value from rescaled restricted residuals bootstrap

0.11411411411411411

P-value from wild bootstrap

0.08808808808808809

Hypothesis 2: $H_0: B_1 = 0.1$

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
c0	0.0078	0.005	-17.522	0.000	-0.003	0.018

P-value from rescaled restricted residuals bootstrap

0.0

P-value from wild bootstrap

0.0

MODEL_16

OLS Regression Results

Dep. Variable:	y	R-squared:	0.760
Model:	OLS	Adj. R-squared:	0.752
Method:	Least Squares	F-statistic:	101.3
Date:	Tue, 01 Dec 2020	Prob (F-statistic):	1.23e-29
Time:	14:32:30	Log-Likelihood:	-146.04
No. Observations:	100	AIC:	300.1
Df Residuals:	96	BIC:	310.5
Df Model:	3		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	-0.1631	0.567	-0.288	0.774	-1.288	0.962
x1	0.0063	0.005	1.228	0.222	-0.004	0.016
x2	-0.0425	0.004	-9.609	0.000	-0.051	-0.034
x3	0.0465	0.004	12.692	0.000	0.039	0.054

Omnibus:	37.704	Durbin-Watson:	2.361
Prob(Omnibus):	0.000	Jarque-Bera (JB):	67.499
Skew:	1.617	Prob(JB):	2.20e-15
Kurtosis:	5.396	Cond. No.	545.

Notes:

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

Hypothesis 1: $H_0: B_1 = 0$

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
c0	0.0063	0.005	1.228	0.222	-0.004	0.016

P-value from rescaled restricted residuals bootstrap

0.2082082082082082

P-value from wild bootstrap

0.18618618618618618

Hopothesis 2: $H_0: B_1 = 0.1$

P-value based on Student t-test

Test for Constraints

	coef	std err	t	P> t	[0.025	0.975]
c0	0.0063	0.005	-18.404	0.000	-0.004	0.016

P-value from rescaled restricted residuals bootstrap

0.0

P-value from wild bootstrap

0.0

```
[7]: # Plot p-value results
```

```
H11 = []
H12 = []
H13 = []
for H in H1:
    H11.append(H[0])
    H12.append(H[1])
    H13.append(H[2])

H21 = []
H22 = []
H23 = []
for HH in H2:
    H21.append(HH[0])
```

```

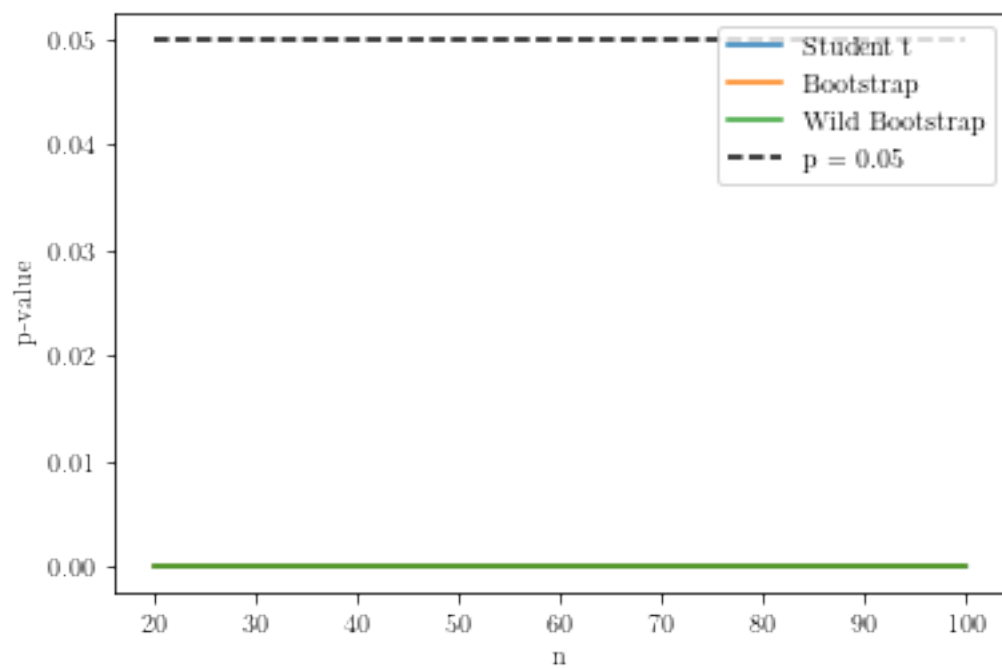
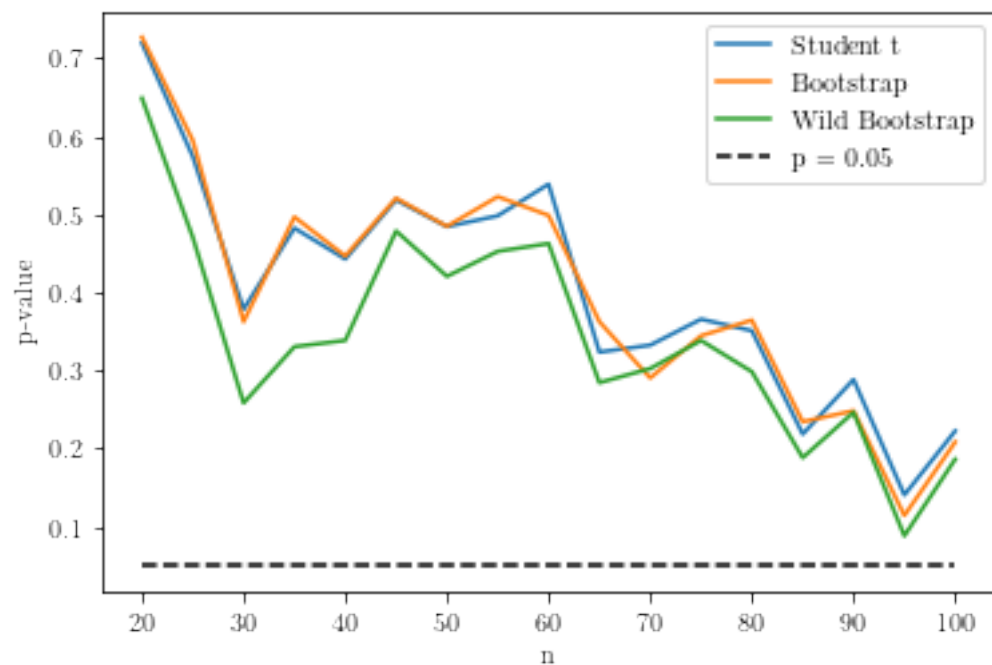
H22.append(HH[1])
H23.append(HH[2])

nobs = np.arange(20, 105, 5)

import matplotlib.pyplot as plt
%matplotlib inline
plt.plot(nobs, H11, label = 'Student t')
plt.plot(nobs, H12, label = 'Bootstrap')
plt.plot(nobs, H13, label = 'Wild Bootstrap')
plt.hlines(0.05, 20, 100, linestyle = 'dashed', label = 'p = 0.05')
plt.xlabel('n')
plt.ylabel('p-value')
plt.legend(loc = 'upper right')
plt.savefig('Varp.pgf')
plt.show()

import matplotlib.pyplot as plt
%matplotlib inline
plt.plot(nobs, H21, label = 'Student t')
plt.plot(nobs, H22, label = 'Bootstrap')
plt.plot(nobs, H23, label = 'Wild Bootstrap')
plt.hlines(0.05, 20, 100, linestyle = 'dashed', label = 'p = 0.05')
plt.xlabel('n')
plt.ylabel('p-value')
plt.legend(loc = 'upper right')
plt.savefig('Varp2.pgf')
plt.show()

```



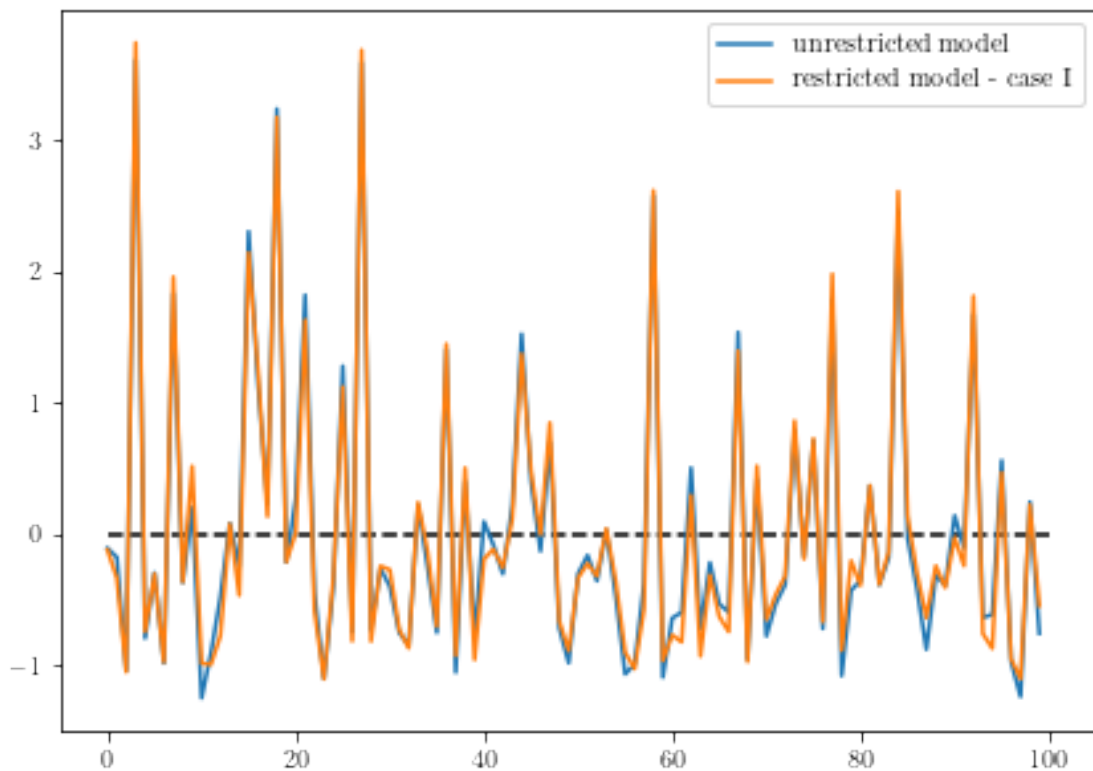
```
[8]: resid_16 = model_16.resid
     resid_rest1_16 = model_16_rest1.resid
```

```

x = np.arange(len(resid_16))

import matplotlib.pyplot as plt
%matplotlib inline
plt.figure(figsize=(7, 5))
plt.plot(resid_16, label = 'unrestricted model')
plt.plot(resid_rest1_16, label = 'restricted model - case I')
plt.hlines(0, 0, len(resid_16), color = 'k', linestyle = 'dashed')
plt.legend(loc = 'upper right')
plt.savefig('resid.pgfig', dpi = 400)
plt.show()

```



```

[9]: resid_unique, resid_counts = np.unique(resid_16, return_counts = True)
resid_rest1_unique, resid_rest1_counts = np.unique(resid_rest1_16,
↪return_counts = True)
nbins = 15
counts, bin_edges = np.histogram(resid_16, nbins)
ticks = []
for i_e in range(1, len(bin_edges)):
    ticks.append(((bin_edges[i_e] - bin_edges[i_e - 1]) / 2) + bin_edges[i_e - 1])
↪1])

```

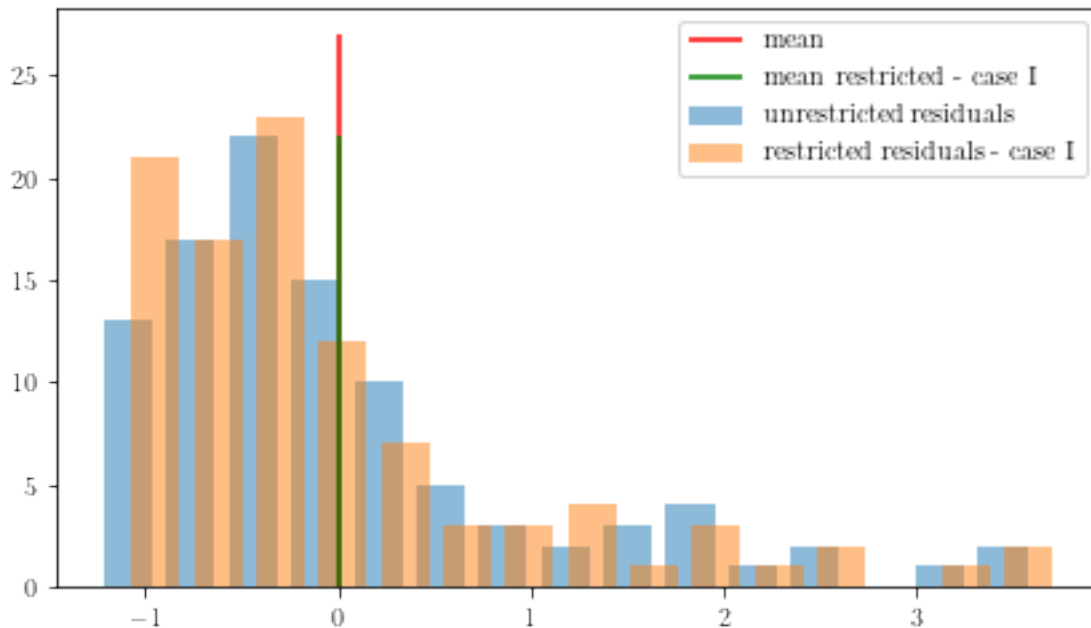


```

counts_rest1, bin_edges_rest1 = np.histogram(resid_rest1_16, nbins)
ticks_rest1 = []
for i_e_rest1 in range(1, len(bin_edges_rest1)):
    ticks_rest1.append(((bin_edges_rest1[i_e_rest1] - bin_edges_rest1[i_e_rest1_
↪ 1]) / 2) + bin_edges_rest1[i_e_rest1 - 1])

import matplotlib.pyplot as plt
%matplotlib inline
plt.figure(figsize=(7, 4))
plt.bar(ticks, counts, width = 0.25, alpha = 0.5, label = 'unrestricted_
↪ residuals')
plt.bar(ticks_rest1, counts_rest1, width = 0.25, alpha = 0.5, label = '
↪ restricted residuals - case I')
plt.vlines(np.mean(resid_16), 0, max(counts)+5, colors = 'r', label = 'mean')
plt.vlines(np.mean(resid_rest1_16), 0, max(counts), colors = 'g', label = 'mean_
↪ restricted - case I')
plt.legend(loc = 'upper right')
plt.savefig('resid_dist.pgf', dpi = 100)
plt.show()

```



```

[10]: # Results

print(H11)
print('')

```

```

print(H12)
print('')
print(H13)
print('')
print('-----')
print(H21)
print('')
print(H22)
print('')
print(H23)
print('')
print('-----')

```

[0.7191648727895876, 0.5739307657161, 0.3783781052307009, 0.4822052779834689,
0.44263342877850886, 0.5183102694510442, 0.48410435202344426,
0.4979882646508451, 0.5385403316897768, 0.3234446951074834, 0.33229488641697214,
0.36556678284711674, 0.3507738743613309, 0.21839316036920003,
0.2883284369379734, 0.14044919427501004, 0.22242857874816382]

[0.7267267267267268, 0.5945945945945946, 0.36236236236236236,
0.4964964964964965, 0.44644644644644643, 0.5205205205205206, 0.4844844844844845,
0.5225225225225225, 0.4984984984984985, 0.36236236236236236, 0.2902902902902903,
0.34434434434434436, 0.36436436436436437, 0.23423423423423423,
0.24824824824824826, 0.11411411411411411, 0.2082082082082082]

[0.6486486486486487, 0.47047047047047047, 0.25825825825825827,
0.3303303303303303, 0.3383383383383383, 0.47847847847847846,
0.42042042042042044, 0.45245245245245247, 0.4624624624624625,
0.28428428428428426, 0.3023023023023023, 0.3383383383383383, 0.2982982982982983,
0.1881881881881882, 0.24624624624624625, 0.08808808808808809,
0.18618618618618618]

[array(1.26771034e-05), array(5.53146685e-07), array(3.25099313e-07),
array(5.75686023e-09), array(1.1819935e-10), array(6.07786539e-13),
array(1.25945357e-14), array(2.95333443e-16), array(1.31067221e-17),
array(8.51666214e-20), array(1.75990338e-21), array(1.97131039e-23),
array(1.02937534e-25), array(3.52550924e-26), array(5.61418023e-29),
array(6.55491244e-31), array(3.00207401e-33)]

[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,
0.0]

[0.9409409409409409, 0.918918918918919, 0.8288288288288288, 0.7547547547547547,
0.7107107107107107, 0.7847847847847848, 0.7207207207207207, 0.7207207207207207,
0.6926926926926927, 0.5465465465465466, 0.6086086086086087, 0.5625625625625625,
0.6446446446446447, 0.46646646646646645, 0.5645645645645646,
0.37037037037037035, 0.4444444444444444]

[]: