

EconQuizFinal

November 11, 2025

```
[1]: import pandas as pd # <--- You need this import
import statsmodels.formula.api as smf
import matplotlib.pyplot as plt
```

```
[2]: # This is the missing step. You must load your CSV into a DataFrame.
# We'll use the mock data example:
data = pd.read_csv("/Users/utkarshtyagi/Downloads/Q2_data_did.csv")
```

```
[4]: model=smf.ols('hours_worked ~ treated * post + age + female +_
sector',data=data).fit(cov_type='HC1')
```

```
[5]: print(model.summary())
```

OLS Regression Results

Dep. Variable:	hours_worked	R-squared:	0.313		
Model:	OLS	Adj. R-squared:	0.306		
Method:	Least Squares	F-statistic:	50.55		
Date:	Tue, 11 Nov 2025	Prob (F-statistic):	3.31e-50		
Time:	18:31:24	Log-Likelihood:	-1275.1		
No. Observations:	600	AIC:	2564.		
Df Residuals:	593	BIC:	2595.		
Df Model:	6				
Covariance Type:	HC1				
=====					
====					
	coef	std err	z	P> z	[0.025
0.975]					

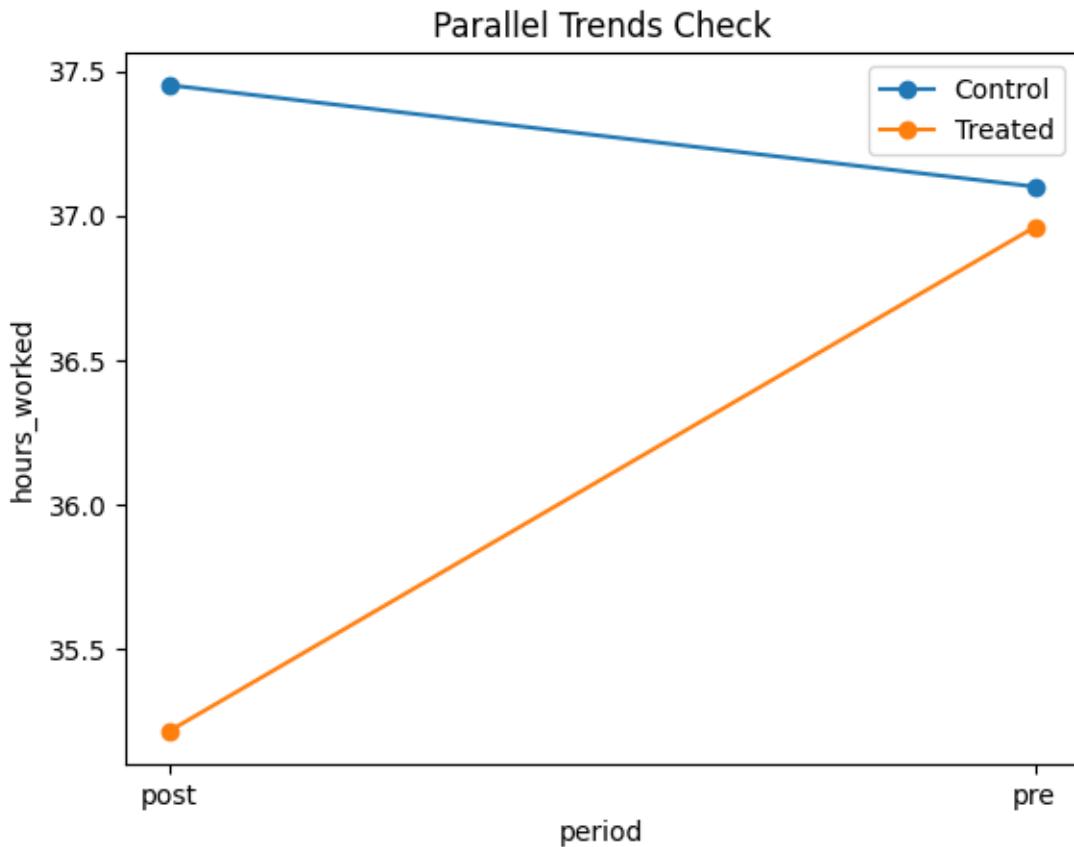
Intercept	40.4576	0.323	125.450	0.000	39.826
41.090					
sector[T.Retail]	-0.4509	0.166	-2.724	0.006	-0.775
-0.126					
treated	-0.0574	0.244	-0.236	0.814	-0.535
0.420					
post	0.3516	0.231	1.520	0.128	-0.102
0.805					

treated:post	-2.0971	0.333	-6.300	0.000	-2.749
-1.445					
age	-0.0656	0.007	-9.493	0.000	-0.079
-0.052					
female	-1.2183	0.169	-7.219	0.000	-1.549
-0.888					
<hr/>					
Omnibus:	0.617	Durbin-Watson:			2.054
Prob(Omnibus):	0.735	Jarque-Bera (JB):			0.713
Skew:	0.027	Prob(JB):			0.700
Kurtosis:	2.840	Cond. No.			220.
<hr/>					

Notes:

[1] Standard Errors are heteroscedasticity robust (HC1)

```
[7]: # Parallel trends plot
avg_trends = data.groupby(['period', 'treated'])['hours_worked'].mean().
    reset_index()
plt.plot(avg_trends[avg_trends['treated']==0]['period'], □
    avg_trends[avg_trends['treated']==0]['hours_worked'], marker='o', □
    label='Control')
plt.plot(avg_trends[avg_trends['treated']==1]['period'], □
    avg_trends[avg_trends['treated']==1]['hours_worked'], marker='o', □
    label='Treated')
plt.legend()
plt.xlabel("period")
plt.ylabel("hours_worked")
plt.title("Parallel Trends Check")
plt.show()
```



```
[11]: # --- 5. ROBUSTNESS CHECK (Add controls) ---
controls = ['age', 'female', 'sector']      # replace with your controls
robust_formula = f" hours_worked ~ treated* post + {' + '.join(controls)}"
did_robust = smf.ols(robust_formula, data=data).fit(cov_type='HC1')
print("\n===== ROBUSTNESS (DiD + CONTROLS) =====")
print(did_robust.summary())
```

```
=====
ROBUSTNESS (DiD + CONTROLS)
=====
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=====
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