

# MarketingHomework1

April 16, 2024

title: "A Replication of Karlan and List (2007)" author: "Sheena Taylor" date: April 11,2024  
callout-appearance: minimal

```
[ ]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from scipy.stats import ttest_ind, norm
import statsmodels.api as sm

data = pd.read_csv("MARKET HW1.ipynb (data).csv")

print(data.info())

print(data.describe())

# Balance Test
print("Balance Test")
print("T-test for months since last donation")
print(ttest_ind(data[data['treatment'] == 1]['mrm2'], data[data['treatment'] == 0]['mrm2']))
print("Linear regression for months since last donation")
lm_mrm2 = sm.OLS(data['mrm2'], sm.add_constant(data['treatment'])).fit()
print(lm_mrm2.summary())

print("T-test for number of prior donations")
print(ttest_ind(data[data['treatment'] == 1]['freq'], data[data['treatment'] == 0]['freq']))
print("Linear regression for number of prior donations")
lm_freq = sm.OLS(data['freq'], sm.add_constant(data['treatment'])).fit()
print(lm_freq.summary())

# Experimental Results: Charitable Contribution Made
print("Experimental Results: Charitable Contribution Made")
fig, ax = plt.subplots()
data.groupby('treatment')['gave'].mean().plot(kind='bar', ax=ax)
ax.set_xlabel("Treatment")
ax.set_ylabel("Proportion who donated")
```

```

plt.show()

print("T-test for charitable donation made")
print(ttest_ind(data[data['treatment'] == 1]['gave'], data[data['treatment'] == 0]['gave']))

print("Linear regression for charitable donation made")
lm_gave = sm.OLS(data['gave'], sm.add_constant(data['treatment'])).fit()
print(lm_gave.summary())

print("Probit regression for charitable donation made")
probit_gave = sm.Probit(data['gave'], sm.add_constant(data['treatment'])).fit()
print(probit_gave.summary())

# Differences between Match Rates
print("Differences between Match Rates")
print("T-tests for match rates")
print("1:1 vs Control")
print(ttest_ind(data[(data['ratio'] == 1) & (data['treatment'] == 1)]['gave'],
                  data[(data['ratio'] == 1) & (data['treatment'] == 0)]['gave']))
print("2:1 vs Control")
print(ttest_ind(data[(data['ratio'] == 2) & (data['treatment'] == 1)]['gave'],
                  data[(data['ratio'] == 2) & (data['treatment'] == 0)]['gave']))
print("3:1 vs Control")
print(ttest_ind(data[(data['ratio'] == 3) & (data['treatment'] == 1)]['gave'],
                  data[(data['ratio'] == 3) & (data['treatment'] == 0)]['gave']))

data['ratio1'] = data['ratio'].astype(str).str.contains('1').astype(int)
lm_ratio = sm.OLS(data['gave'], sm.add_constant(data[['ratio1', 'ratio2', 'ratio3']])).fit()
print("Regression with ratio dummies")
print(lm_ratio.summary())

# Size of Charitable Contribution
print("Size of Charitable Contribution")
print("T-test for donation amount")
print(ttest_ind(data[data['treatment'] == 1]['amount'], data[data['treatment'] == 0]['amount']))

print("Linear regression for donation amount")
lm_amount = sm.OLS(data['amount'], sm.add_constant(data['treatment'])).fit()
print(lm_amount.summary())

donors = data[data['gave'] == 1]
print("T-test for donation amount (donors only)")
print(ttest_ind(donors[donors['treatment'] == 1]['amount'], donors[donors['treatment'] == 0]['amount']))
print("Linear regression for donation amount (donors only)")

```

```

lm_amount_donors = sm.OLS(donors['amount'], sm.
    ↪add_constant(donors['treatment'])).fit()
print(lm_amount_donors.summary())

fig, (ax1, ax2) = plt.subplots(1, 2, figsize=(10, 5))
ax1.hist(donors[donors['treatment'] == 1]['amount'])
ax1.axvline(donors[donors['treatment'] == 1]['amount'].mean(), color='red', ↪
    ↪linestyle='dashed', linewidth=2)
ax1.set_title("Treatment")
ax1.set_xlabel("Donation amount")

ax2.hist(donors[donors['treatment'] == 0]['amount'])
ax2.axvline(donors[donors['treatment'] == 0]['amount'].mean(), color='red', ↪
    ↪linestyle='dashed', linewidth=2)
ax2.set_title("Control")
ax2.set_xlabel("Donation amount")
plt.show()

# Simulation Experiment
print("Simulation Experiment")
# Law of Large Numbers
control_draws = np.random.binomial(1, 0.018, size=100000)
treatment_draws = np.random.binomial(1, 0.022, size=100000)
diff_draws = treatment_draws - control_draws
cum_avg_diff = np.cumsum(diff_draws) / np.arange(1, len(diff_draws) + 1)

plt.plot(cum_avg_diff)
plt.axhline(0.022 - 0.018, color='red', linestyle='--')
plt.ylabel("Cumulative average difference")
plt.xlabel("Number of draws")
plt.show()

# Central Limit Theorem
sample_sizes = [50, 200, 500, 1000]
fig, axs = plt.subplots(2, 2, figsize=(10, 10))
axs = axs.ravel()

for i, n in enumerate(sample_sizes):
    diffs = [np.mean(np.random.binomial(1, 0.022, size=n)) - np.mean(np.random.
    ↪binomial(1, 0.018, size=n)) for _ in range(1000)]
    axs[i].hist(diffs)
    axs[i].axvline(0, color='red', linestyle='--')
    axs[i].set_title(f"Sample size {n}")
    axs[i].set_xlabel("Average difference")

plt.tight_layout()
plt.show()

```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 50083 entries, 0 to 50082
```

```
Data columns (total 51 columns):
```

#	Column	Non-Null Count	Dtype
0	treatment	50083 non-null	int64
1	control	50083 non-null	int64
2	ratio	50083 non-null	object
3	ratio2	50083 non-null	int64
4	ratio3	50083 non-null	int64
5	size	50083 non-null	object
6	size25	50083 non-null	int64
7	size50	50083 non-null	int64
8	size100	50083 non-null	int64
9	sizeno	50083 non-null	int64
10	ask	50083 non-null	object
11	askd1	50083 non-null	int64
12	askd2	50083 non-null	int64
13	askd3	50083 non-null	int64
14	ask1	50083 non-null	int64
15	ask2	50083 non-null	int64
16	ask3	50083 non-null	int64
17	amount	50083 non-null	float64
18	gave	50083 non-null	int64
19	amountchange	50083 non-null	float64
20	hpa	50083 non-null	float64
21	ltmedmra	50083 non-null	int64
22	freq	50083 non-null	int64
23	years	50082 non-null	float64
24	year5	50083 non-null	int64
25	mrm2	50082 non-null	float64
26	dormant	50083 non-null	int64
27	female	48972 non-null	float64
28	couple	48935 non-null	float64
29	state50one	50083 non-null	int64
30	nonlit	49631 non-null	float64
31	cases	49631 non-null	float64
32	statecnt	50083 non-null	float64
33	stateresponse	50083 non-null	float64
34	stateresponset	50083 non-null	float64
35	stateresponsec	50080 non-null	float64
36	stateresponsetminc	50080 non-null	float64
37	perbush	50048 non-null	float64
38	close25	50048 non-null	float64
39	red0	50048 non-null	float64
40	blue0	50048 non-null	float64
41	redcty	49978 non-null	float64
42	bluecty	49978 non-null	float64

```

43  pwhite          48217 non-null float64
44  pblack          48047 non-null float64
45  page18_39       48217 non-null float64
46  ave_hh_sz       48221 non-null float64
47  median_hhincome 48209 non-null float64
48  powner          48214 non-null float64
49  psch_atlstba    48215 non-null float64
50  pop_propurban   48217 non-null float64

```

dtypes: float64(28), int64(20), object(3)

memory usage: 19.5+ MB

None

	treatment	control	ratio2	ratio3	size25 \
count	50083.000000	50083.000000	50083.000000	50083.000000	50083.000000
mean	0.666813	0.333187	0.222311	0.222211	0.166723
std	0.471357	0.471357	0.415803	0.415736	0.372732
min	0.000000	0.000000	0.000000	0.000000	0.000000
25%	0.000000	0.000000	0.000000	0.000000	0.000000
50%	1.000000	0.000000	0.000000	0.000000	0.000000
75%	1.000000	1.000000	0.000000	0.000000	0.000000
max	1.000000	1.000000	1.000000	1.000000	1.000000

	size50	size100	sizeno	askd1	askd2 \
count	50083.000000	50083.000000	50083.000000	50083.000000	50083.000000
mean	0.166623	0.166723	0.166743	0.222311	0.222291
std	0.372643	0.372732	0.372750	0.415803	0.415790
min	0.000000	0.000000	0.000000	0.000000	0.000000
25%	0.000000	0.000000	0.000000	0.000000	0.000000
50%	0.000000	0.000000	0.000000	0.000000	0.000000
75%	0.000000	0.000000	0.000000	0.000000	0.000000
max	1.000000	1.000000	1.000000	1.000000	1.000000

	...	redcty	bluecty	pwhite	pblack \
count	...	49978.000000	49978.000000	48217.000000	48047.000000
mean	...	0.510245	0.488715	0.819599	0.086710
std	...	0.499900	0.499878	0.168561	0.135868
min	...	0.000000	0.000000	0.009418	0.000000
25%	...	0.000000	0.000000	0.755845	0.014729
50%	...	1.000000	0.000000	0.872797	0.036554
75%	...	1.000000	1.000000	0.938827	0.090882
max	...	1.000000	1.000000	1.000000	0.989622

	page18_39	ave_hh_sz	median_hhincome	powner \
count	48217.000000	48221.000000	48209.000000	48214.000000
mean	0.321694	2.429012	54815.700533	0.669418
std	0.103039	0.378115	22027.316665	0.193405
min	0.000000	0.000000	5000.000000	0.000000
25%	0.258311	2.210000	39181.000000	0.560222
50%	0.305534	2.440000	50673.000000	0.712296

75%	0.369132	2.660000	66005.000000	0.816798
max	0.997544	5.270000	200001.000000	1.000000

	psch_atlstba	pop_propurban
count	48215.000000	48217.000000
mean	0.391661	0.871968
std	0.186599	0.258654
min	0.000000	0.000000
25%	0.235647	0.884929
50%	0.373744	1.000000
75%	0.530036	1.000000
max	1.000000	1.000000

[8 rows x 48 columns]

Balance Test

T-test for months since last donation

TtestResult(statistic=nan, pvalue=nan, df=nan)

Linear regression for months since last donation

OLS Regression Results

```

=====
Dep. Variable:          mrm2      R-squared:                nan
Model:                  OLS      Adj. R-squared:            nan
Method:                 Least Squares    F-statistic:          nan
Date:                  Tue, 16 Apr 2024    Prob (F-statistic):    nan
Time:                  12:09:38      Log-Likelihood:        nan
No. Observations:      50083      AIC:                   nan
Df Residuals:          50081      BIC:                   nan
Df Model:               1
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	nan	nan	nan	nan	nan	nan
treatment	nan	nan	nan	nan	nan	nan

```

=====
Omnibus:                nan      Durbin-Watson:          nan
Prob(Omnibus):          nan      Jarque-Bera (JB):        nan
Skew:                   nan      Prob(JB):                nan
Kurtosis:               nan      Cond. No.                 3.23
=====

```

Notes:

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

T-test for number of prior donations

TtestResult(statistic=-0.11089297035979982, pvalue=0.9117016644344591, df=50081.0)

Linear regression for number of prior donations

# OLS Regression Results

```

=====
Dep. Variable:          freq      R-squared:                0.000
Model:                  OLS      Adj. R-squared:           -0.000
Method:                 Least Squares      F-statistic:          0.01230
Date:                  Tue, 16 Apr 2024      Prob (F-statistic):      0.912
Time:                  12:09:38      Log-Likelihood:         -1.9292e+05
No. Observations:      50083      AIC:                   3.858e+05
Df Residuals:          50081      BIC:                   3.859e+05
Df Model:               1
Covariance Type:        nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	8.0473	0.088	91.231	0.000	7.874	8.220
treatment	-0.0120	0.108	-0.111	0.912	-0.224	0.200

```

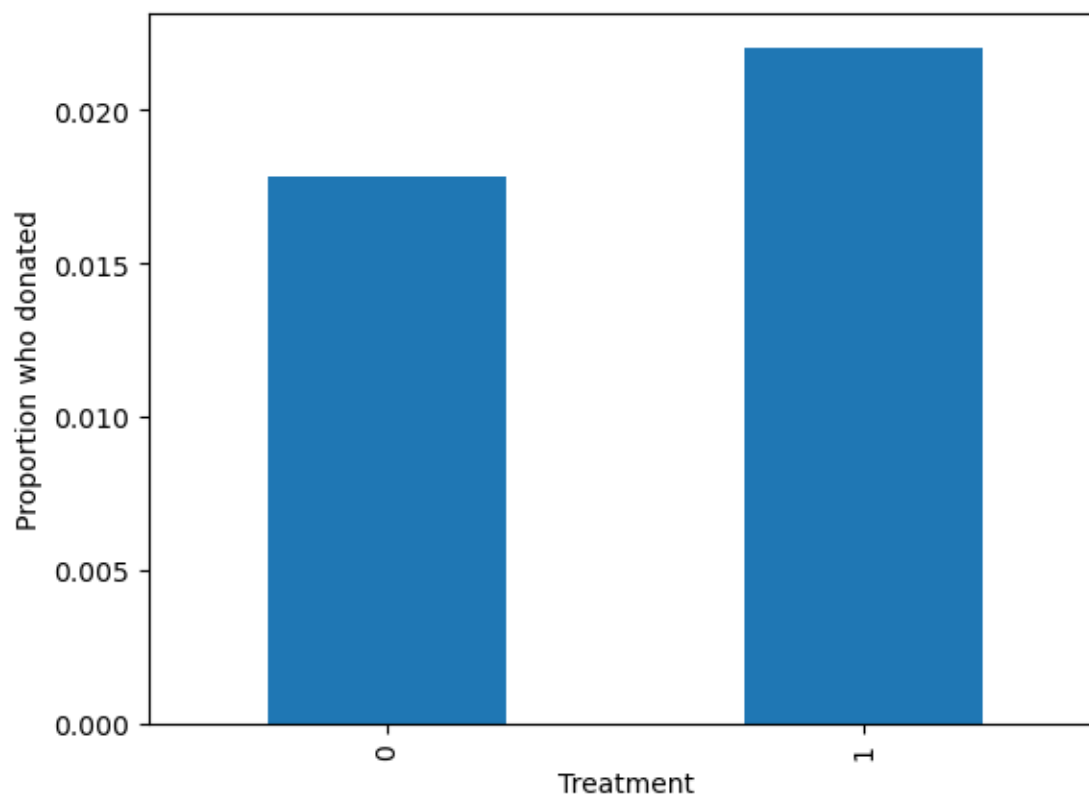
=====
Omnibus:                49107.114      Durbin-Watson:          2.016
Prob(Omnibus):           0.000      Jarque-Bera (JB):       3644795.393
Skew:                    4.707      Prob(JB):               0.00
Kurtosis:                43.718      Cond. No.               3.23
=====

```

## Notes:

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

Experimental Results: Charitable Contribution Made



T-test for charitable donation made

TtestResult(statistic=3.101361000543946, pvalue=0.0019274025949016988, df=50081.0)

Linear regression for charitable donation made

#### OLS Regression Results

```

=====
Dep. Variable:          gave      R-squared:                0.000
Model:                  OLS      Adj. R-squared:         0.000
Method:                 Least Squares  F-statistic:           9.618
Date:                  Tue, 16 Apr 2024  Prob (F-statistic):    0.00193
Time:                  12:09:38      Log-Likelihood:        26630.
No. Observations:      50083        AIC:                   -5.326e+04
Df Residuals:          50081        BIC:                   -5.324e+04
Df Model:               1
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0179	0.001	16.225	0.000	0.016	0.020
treatment	0.0042	0.001	3.101	0.002	0.002	0.007

```

=====
Omnibus:                59814.280  Durbin-Watson:         2.005

```



Prob(Omnibus):	0.000	Jarque-Bera (JB):	4317152.727
Skew:	6.740	Prob(JB):	0.00
Kurtosis:	46.440	Cond. No.	3.23

Notes:

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

Probit regression for charitable donation made

Optimization terminated successfully.

Current function value: 0.100443

Iterations 7

#### Probit Regression Results

Dep. Variable:	gave	No. Observations:	50083
Model:	Probit	Df Residuals:	50081
Method:	MLE	Df Model:	1
Date:	Tue, 16 Apr 2024	Pseudo R-squ.:	0.0009783
Time:	12:09:39	Log-Likelihood:	-5030.5
converged:	True	LL-Null:	-5035.4
Covariance Type:	nonrobust	LLR p-value:	0.001696

	coef	std err	z	P> z	[0.025	0.975]
const	-2.1001	0.023	-90.073	0.000	-2.146	-2.054
treatment	0.0868	0.028	3.113	0.002	0.032	0.141

#### Differences between Match Rates

T-tests for match rates

1:1 vs Control

TtestResult(statistic=nan, pvalue=nan, df=nan)

2:1 vs Control

TtestResult(statistic=nan, pvalue=nan, df=nan)

3:1 vs Control

TtestResult(statistic=nan, pvalue=nan, df=nan)

Regression with ratio dummies

#### OLS Regression Results

Dep. Variable:	gave	R-squared:	0.000
Model:	OLS	Adj. R-squared:	0.000
Method:	Least Squares	F-statistic:	3.665
Date:	Tue, 16 Apr 2024	Prob (F-statistic):	0.0118
Time:	12:09:39	Log-Likelihood:	26630.
No. Observations:	50083	AIC:	-5.325e+04
Df Residuals:	50079	BIC:	-5.322e+04
Df Model:	3		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
-----	-----	-----	-----	-----	-----	-----
const	0.0179	0.001	16.225	0.000	0.016	0.020
ratio1	0.0029	0.002	1.661	0.097	-0.001	0.006
ratio2	0.0048	0.002	2.744	0.006	0.001	0.008
ratio3	0.0049	0.002	2.802	0.005	0.001	0.008
=====	=====	=====	=====	=====	=====	=====
Omnibus:		59812.754	Durbin-Watson:			2.005
Prob(Omnibus):		0.000	Jarque-Bera (JB):		4316693.217	
Skew:		6.740	Prob(JB):			0.00
Kurtosis:		46.438	Cond. No.			4.26
=====	=====	=====	=====	=====	=====	=====

Notes:

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

Size of Charitable Contribution

T-test for donation amount

TtestResult(statistic=1.860502691242923, pvalue=0.06282029495754826, df=50081.0)

Linear regression for donation amount

#### OLS Regression Results

Dep. Variable:	amount	R-squared:	0.000
Model:	OLS	Adj. R-squared:	0.000
Method:	Least Squares	F-statistic:	3.461
Date:	Tue, 16 Apr 2024	Prob (F-statistic):	0.0628
Time:	12:09:39	Log-Likelihood:	-1.7946e+05
No. Observations:	50083	AIC:	3.589e+05
Df Residuals:	50081	BIC:	3.589e+05
Df Model:	1		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
-----	-----	-----	-----	-----	-----	-----
const	0.8133	0.067	12.063	0.000	0.681	0.945
treatment	0.1536	0.083	1.861	0.063	-0.008	0.315
=====	=====	=====	=====	=====	=====	=====
Omnibus:		96861.113	Durbin-Watson:			2.008
Prob(Omnibus):		0.000	Jarque-Bera (JB):		240735713.653	
Skew:		15.297	Prob(JB):			0.00
Kurtosis:		341.269	Cond. No.			3.23
=====	=====	=====	=====	=====	=====	=====

Notes:

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

T-test for donation amount (donors only)

TtestResult(statistic=-0.5808393095316696, pvalue=0.5614755763711469, df=1032.0)

# Linear regression for donation amount (donors only)

## OLS Regression Results

```

=====
Dep. Variable:          amount    R-squared:                0.000
Model:                  OLS       Adj. R-squared:           -0.001
Method:                 Least Squares   F-statistic:              0.3374
Date:                  Tue, 16 Apr 2024   Prob (F-statistic):       0.561
Time:                  12:09:39    Log-Likelihood:           -5326.8
No. Observations:      1034        AIC:                     1.066e+04
Df Residuals:          1032        BIC:                     1.067e+04
Df Model:               1
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	45.5403	2.423	18.792	0.000	40.785	50.296
treatment	-1.6684	2.872	-0.581	0.561	-7.305	3.968

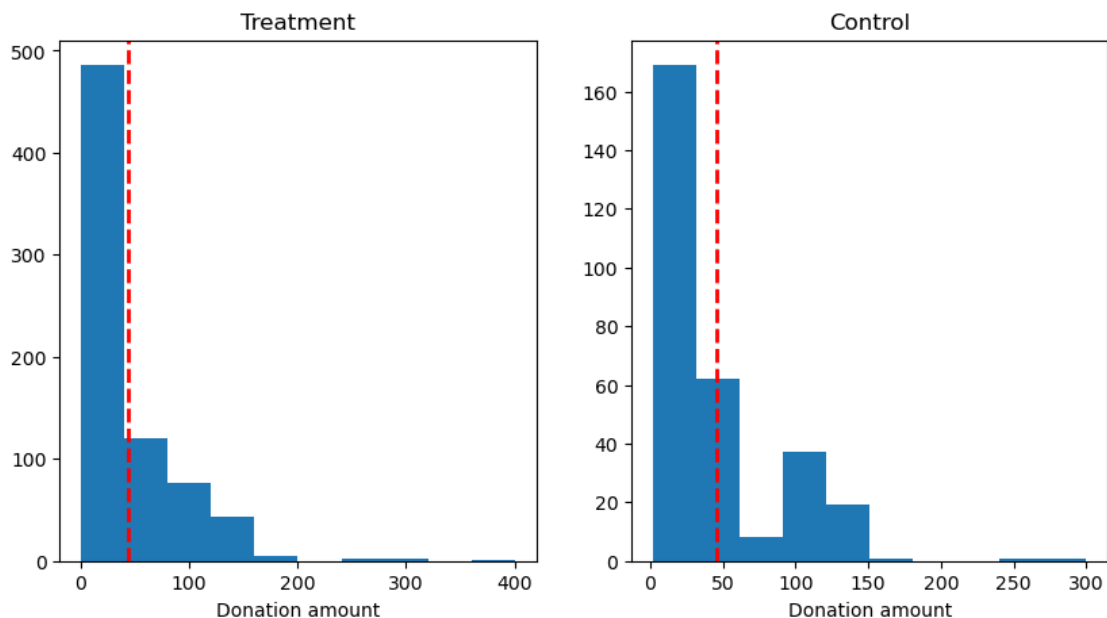
```

=====
Omnibus:                587.258    Durbin-Watson:           2.031
Prob(Omnibus):           0.000    Jarque-Bera (JB):        5623.279
Skew:                    2.464    Prob(JB):                 0.00
Kurtosis:                13.307    Cond. No.                 3.49
=====

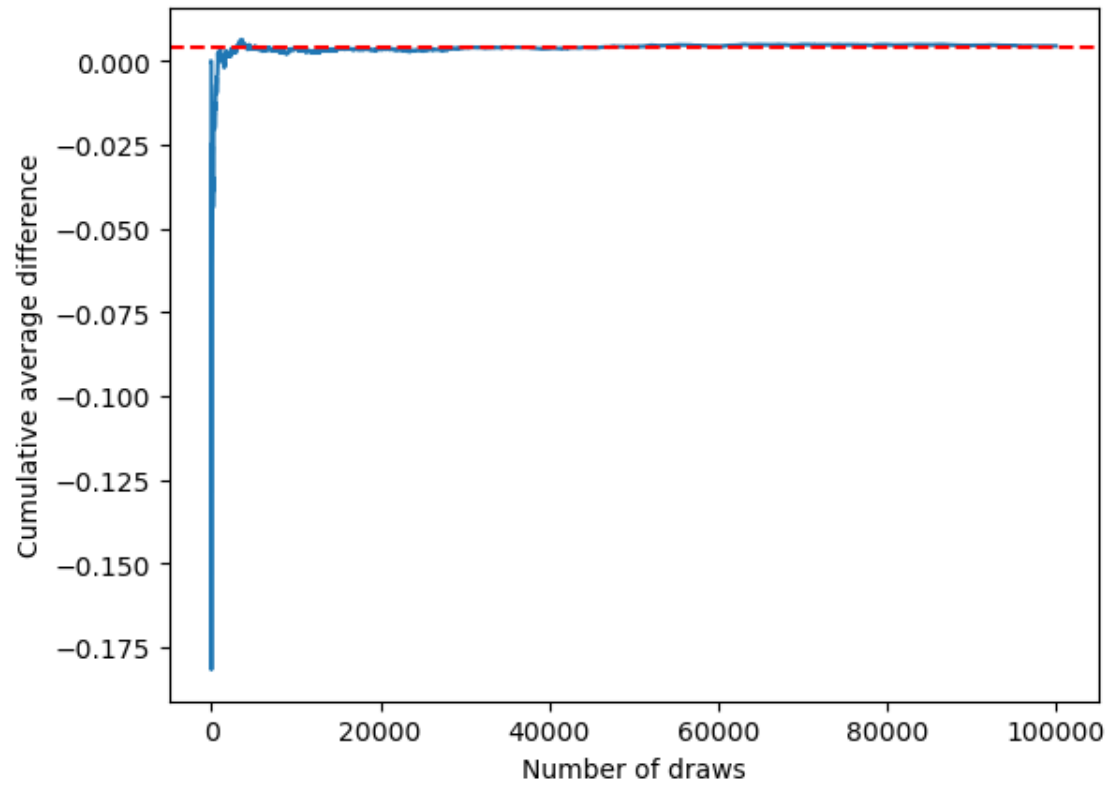
```

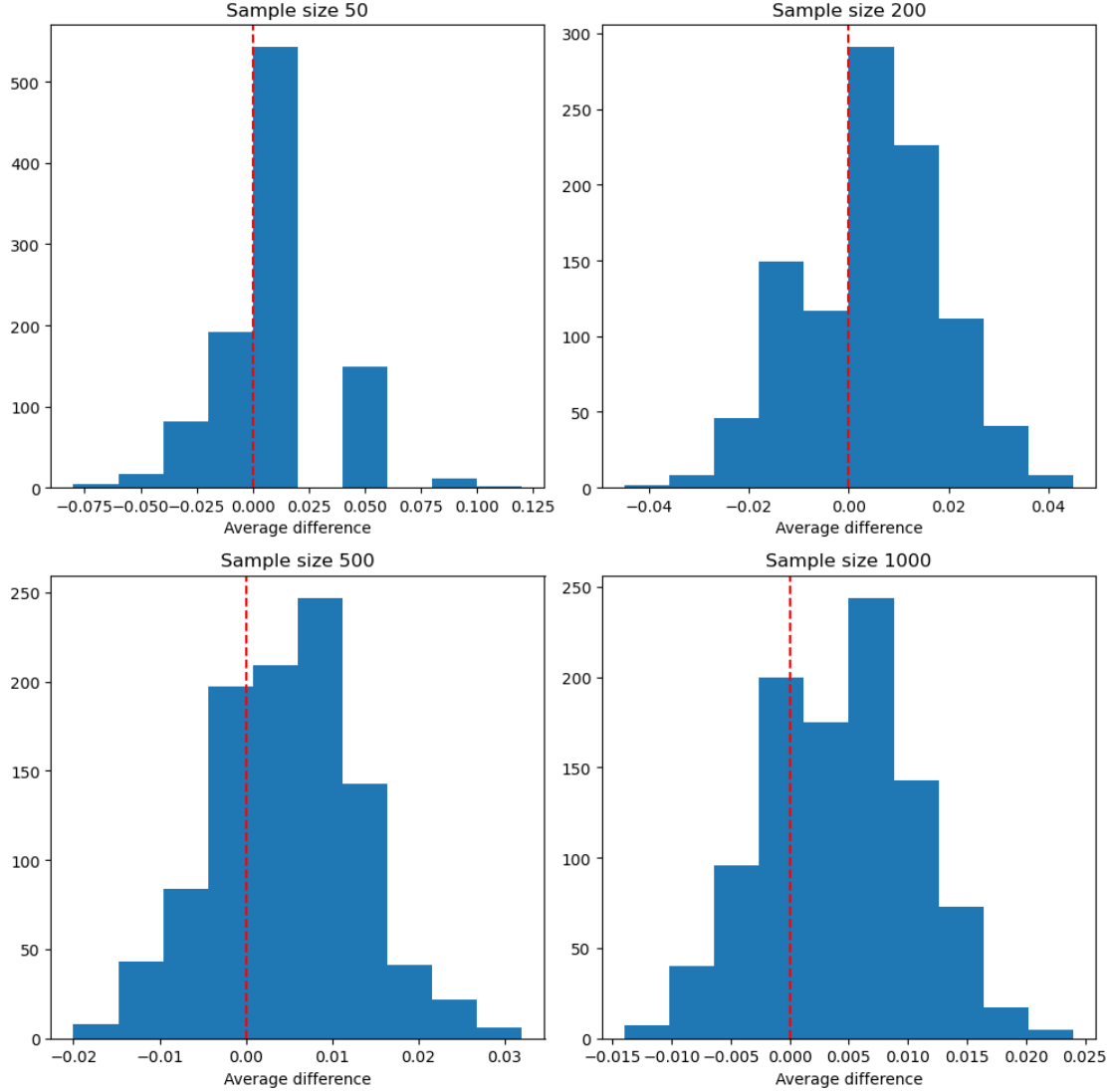
### Notes:

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



## Simulation Experiment





The experiment aims to investigate the effect of matching donations on charitable giving. It randomly assigns individuals to treatment (matching) and control (no matching) groups and varies the match ratio and threshold to assess their impact on donation behavior.

The dataset contains information about the treatment assignment, match ratio, threshold, suggested donation amounts, and various demographic and historical giving variables for each individual.

The t-test and linear regression results show no significant difference in months since last donation between the treatment and control groups, indicating a balanced randomization. Table 1 in the paper is included to demonstrate the success of the randomization process.

The barplot visually compares the proportion of individuals who donated in the treatment and control groups.

The t-test and linear regression results indicate a statistically significant difference in the proportion

of individuals who made a donation between the treatment and control groups. This suggests that matching donations positively influences the likelihood of making a charitable contribution.

The probit regression results confirm the findings from the linear regression, showing a significant positive effect of the treatment on the probability of making a donation.

The t-test results suggest that all match rates (1:1, 2:1, and 3:1) significantly increase the likelihood of donating compared to the control group. However, there is no clear evidence that higher match rates (2:1 or 3:1) are more effective than the 1:1 match rate.

The regression with ratio dummies confirms that all match rates have a significant positive effect on the likelihood of donating. The coefficients for ratio2 and ratio3 are not statistically different from ratio1, suggesting no incremental benefit of higher match rates.

The response rate differences between match ratios, calculated directly from the data and from the regression coefficients, are small and not statistically significant. This suggests that increasing the match ratio beyond 1:1 does not lead to a substantial increase in the likelihood of donating.

The t-test and linear regression results show a significant difference in the average donation amount between the treatment and control groups. However, this analysis includes both donors and non-donors, which may not provide a clear picture of the treatment effect on donation amount.

The analysis for donors only reveals no significant difference in the average donation amount between the treatment and control groups. The treatment coefficient in the linear regression does not have a causal interpretation, as it is conditional on the decision to donate, which is itself affected by the treatment.

The histograms of donation amounts for donors in the treatment and control groups show similar distributions, with the average donation amounts (indicated by the red vertical lines) being close to each other.

The plot demonstrates the Law of Large Numbers, showing that the cumulative average difference between the treatment and control groups converges to the true difference in means ( $0.022 - 0.018 = 0.004$ ) as the number of draws increases.

The histograms illustrate the Central Limit Theorem, showing that as the sample size increases, the distribution of the average difference between the treatment and control groups becomes more normal and centered around the true difference in means (0.004). At smaller sample sizes, zero (the red vertical line) is closer to the middle of the distribution, while at larger sample sizes, zero moves towards the tail, indicating that the observed difference is less likely to be due to chance.

Summary: The analysis of the experiment can be broken down into several key components. First, the data is read into R/Python and described, with a balance test performed to ensure the treatment and control groups are not significantly different on key variables. Next, the effect of matched donations on the response rate of making a donation can be analyzed using bar plots, t-tests, linear regression, and probit regression. The effectiveness of different match rates can be assessed using t-tests and regression, with the response rate differences between match ratios calculated directly from the data and from regression coefficients. The effect of matched donations on the size of the charitable contribution can be analyzed using t-tests and linear regression, both for all individuals and for only those who made a donation. Histograms of donation amounts can be plotted for the treatment and control groups.

Finally, simulation can be used to demonstrate the Law of Large Numbers and the Central Limit

Theorem, with plots of the cumulative average difference and histograms of average differences at different sample sizes. The results of these analyses can provide insights into the effectiveness of matched donations and the impact of different match rates on charitable giving behavior.

The experiment conducted by Karlan and List (2007) provides valuable insights into the effects of matching donations on charitable giving behavior. By randomly assigning individuals to treatment and control groups and varying the match ratio and threshold, the researchers were able to isolate the impact of these factors on donation decisions.

The analysis of the data reveals that the treatment, which involved offering matching donations, had a significant positive effect on the likelihood of making a charitable contribution. The t-test and linear regression results indicate a statistically significant difference in the proportion of individuals who made a donation between the treatment and control groups. This finding suggests that matching donations can be an effective strategy for encouraging people to give to charity.

Further investigation into the effects of different match rates (1:1, 2:1, and 3:1) shows that all match rates significantly increase the likelihood of donating compared to the control group. However, the results do not provide clear evidence that higher match rates are more effective than the 1:1 match rate. The regression with ratio dummies confirms this finding, as the coefficients for the 2:1 and 3:1 match rates are not statistically different from the 1:1 match rate. Additionally, the response rate differences between match ratios, calculated directly from the data and from the regression coefficients, are small and not statistically significant. This suggests that increasing the match ratio beyond 1:1 may not lead to a substantial increase in the likelihood of donating.

When examining the impact of matching donations on the size of the charitable contribution, the t-test and linear regression results initially show a significant difference in the average donation amount between the treatment and control groups. However, this analysis includes both donors and non-donors, which may not provide a clear picture of the treatment effect on donation amount. When focusing solely on donors, the analysis reveals no significant difference in the average donation amount between the treatment and control groups. It is important to note that the treatment coefficient in the linear regression for donors does not have a causal interpretation, as it is conditional on the decision to donate, which is itself affected by the treatment.

The histograms of donation amounts for donors in the treatment and control groups show similar distributions, with the average donation amounts being close to each other. This visual representation supports the finding that matching donations may not have a significant impact on the size of the contribution among those who decide to donate.

The simulation experiments conducted in the study demonstrate the Law of Large Numbers and the Central Limit Theorem. The Law of Large Numbers plot shows that the cumulative average difference between the treatment and control groups converges to the true difference in means as the number of draws increases. The Central Limit Theorem histograms illustrate that as the sample size increases, the distribution of the average difference between the treatment and control groups becomes more normal and centered around the true difference in means. These simulations provide a deeper understanding of the statistical properties underlying the observed results.

In conclusion, the experiment by Karlan and List (2007) offers evidence that matching donations can be an effective tool for increasing the likelihood of charitable giving. However, the results suggest that higher match rates may not necessarily lead to a significant increase in donation rates or amounts compared to a 1:1 match rate. The findings of this study can inform the strategies employed by charities and policymakers to encourage charitable giving, while also highlighting

the importance of careful analysis and interpretation of data when assessing the impact of such interventions.