

# Homework #3: Causality

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## Overview & instructions

For homework, you will first analyze a panel dataset from a new product category. Then you will evaluate the effectiveness of an online display ad campaign.

## Instructions

1. Your task is to fill in all R code blocks that currently contain “#TBD” comments. Similarly, insert text responses wherever you see \*TBD\* in the markdown file.
2. PLEASE ADD YOUR NAME TO THE AUTHOR LINE ABOVE

## Task 1 description

Your task is to construct a linear model of demand for a common packaged good (a laundry detergent). The effect of interest is the average effect of **price** on **sales** (units sold).

You have access to scanner data across a set of stores of a retail chain in the Chicago metro region. The data are in the file **detergent\_data.csv** here. The variables in the data set are:

<b>store</b>	Store id number
<b>week</b>	Week number
<b>promoflag</b>	= 1 if any product in the category was on promotion
<b>sales</b>	Tide 128oz laundry detergent: unit sales
<b>price</b>	Tide 128oz laundry detergent: price (\$)

## 1) Data Description

### 1.1) Read in the data

Read in the data from the csv file, and store to dataframe **DF1**.

Also, report (print):

- 1) the number of unique stores in the dataset, and
- 2) number of unique time periods in the dataset

*Hint: the functions **unique()** and **length()** can be useful in calculating the number of unique observations*

```
DF1<-read.csv("/Users/raveena/Desktop/Classroom - R/Marketing Analytics/data/detergent_data.csv")
length(unique(DF1$store))
```

```
## [1] 86
```

```
length(unique(DF1$week))
```

```
## [1] 224
```

What is the total number of observations? 14744

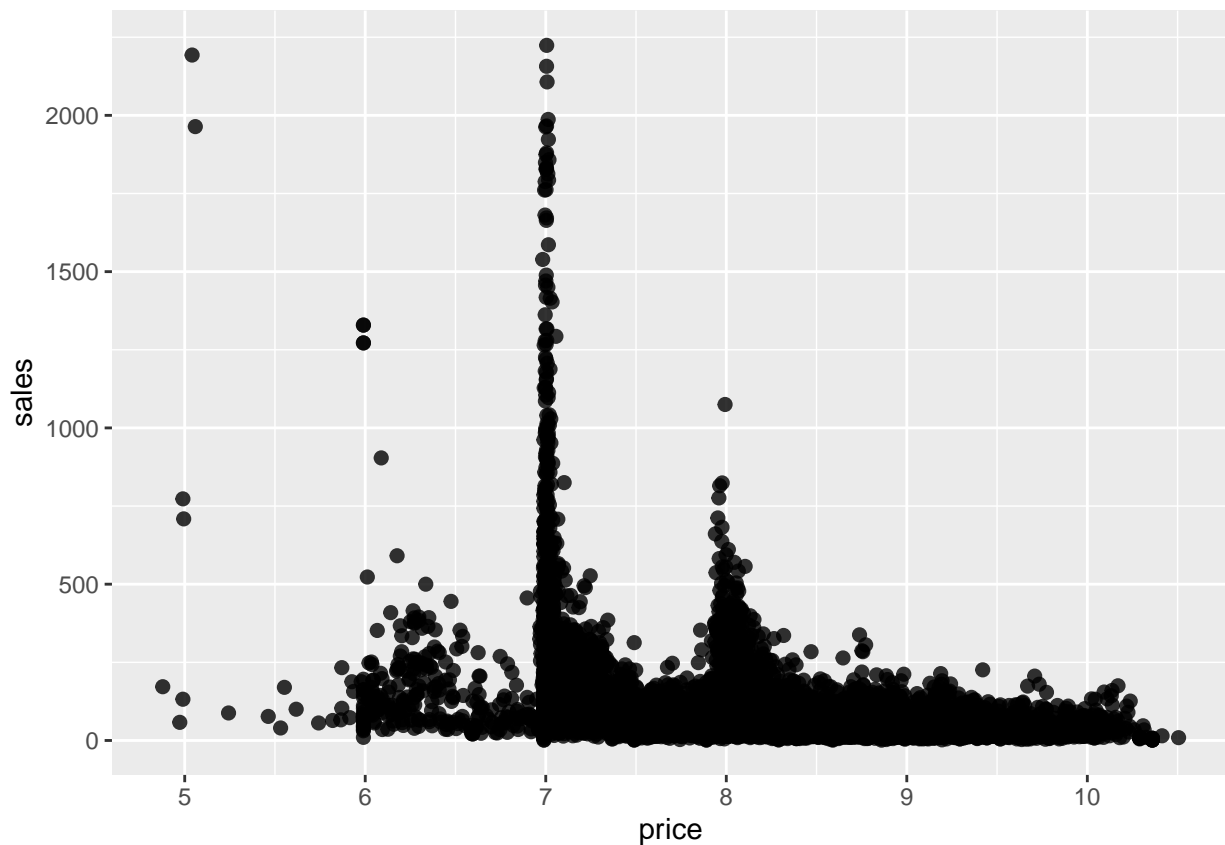
**Is the data a balanced or unbalanced panel? Why?** The data is unbalanced panel. This is because each store should have same number of time periods. If we group the data by store and count the weeks for each store it should be same for all the stores for the data to be a balanced panel.

**Interpret the mean value of the variable `promoflag`** Mean is 0.81 which means that 81% of the panel dataset have used promotions.

## 1.2) Scatterplot of sales vs. price

Generate a scatterplot of sales vs. price. Add a linear regression line.

```
library(ggplot2)
ggplot(data = DF1, aes(x = price, y = sales)) +
  geom_point(alpha = .8, size = 2)
```



**Comment on the distribution of sales and prices. What patterns do you notice?** The sales is high for some price points like 7 and then falls. It shows a fluctuation where sales is increasing with increase in price (between \$7-\$8) and then decreasing as price further increases.

**Is the (sign of the) fitted regression line slope as expected? Why or why not?** Yes. The sign is -ve which is expected from the model. It does show that as price increases unit sold decreases.

## 2) Regression models where DV = sales

### 2.1) regressors: price, promoflag

Use `lm()` to estimate a model of sales with regressors: price, promoflag. Use `summary()` to summarize the results.

```
summary( lm(sales ~ price + promoflag, data=DF1))

##
## Call:
## lm(formula = sales ~ price + promoflag, data = DF1)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -259.94  -47.25  -15.51   19.36  2045.53
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  636.795     11.573   55.024 <2e-16 ***
## price       -68.622      1.336  -51.347 <2e-16 ***
## promoflag     22.379      2.635    8.492 <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 123 on 14741 degrees of freedom
## Multiple R-squared:  0.1588, Adjusted R-squared:  0.1587
## F-statistic: 1391 on 2 and 14741 DF, p-value: < 2.2e-16
```

### 2.2) regressors: price, promoflag, time (week) fixed effects

Use `lm()` to estimate a model of sales with regressors: price, promoflag and time (week) fixed effects. Use `summary()` to summarize the results.

```
summary( lm(sales ~ price + promoflag + week, data=DF1))

##
## Call:
## lm(formula = sales ~ price + promoflag + week, data = DF1)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -262.39  -47.39  -15.66   19.46  2044.02
```

```
##
## Coefficients:
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept) 633.94646   11.73102  54.040 < 2e-16 ***
## price       -68.54153    1.33747 -51.247 < 2e-16 ***
## promoflag    21.59894    2.68730   8.037 9.87e-16 ***
## week         0.02843     0.01918   1.482  0.138
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 123 on 14740 degrees of freedom
## Multiple R-squared:  0.1589, Adjusted R-squared:  0.1587
## F-statistic: 928.2 on 3 and 14740 DF, p-value: < 2.2e-16
```

```
summary( lm(sales ~ price + promoflag + factor(week), data=DF1))
```

```
##
## Call:
## lm(formula = sales ~ price + promoflag + factor(week), data = DF1)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1191.94   -18.41    -3.96    12.65   1878.07
##
## Coefficients:
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept)    179.2997    12.7714   14.039 < 2e-16 ***
## price          -17.3808     1.2545  -13.854 < 2e-16 ***
## promoflag       -7.2057     3.0677   -2.349 0.018845 *
## factor(week)2    -4.2591    10.8031   -0.394 0.693406
## factor(week)3     7.1346    10.7676    0.663 0.507599
## factor(week)4   105.6222    10.6652    9.903 < 2e-16 ***
## factor(week)5    67.2293    10.2984    6.528 6.88e-11 ***
## factor(week)6    21.1420    10.3973    2.033 0.042030 *
## factor(week)7    30.3143    10.4385    2.904 0.003689 **
## factor(week)8    31.5211    10.2880    3.064 0.002189 **
## factor(week)9    22.2312    10.1885    2.182 0.029127 *
## factor(week)10   24.1935    10.7717    2.246 0.024718 *
## factor(week)11     5.3013    10.9161    0.486 0.627227
## factor(week)12   23.7954    10.6671    2.231 0.025715 *
## factor(week)13   33.6070    10.4728    3.209 0.001335 **
## factor(week)14   52.7934    10.2992    5.126 3.00e-07 ***
## factor(week)15   48.3027    10.2661    4.705 2.56e-06 ***
## factor(week)16   63.8106    10.3656    6.156 7.66e-10 ***
## factor(week)17  368.5436    10.2499   35.956 < 2e-16 ***
## factor(week)18   21.7282    10.7363    2.024 0.043010 *
## factor(week)19   48.4430    10.7443    4.509 6.57e-06 ***
## factor(week)20  333.7572    10.5487   31.640 < 2e-16 ***
## factor(week)21   44.5825    10.8911    4.093 4.27e-05 ***
## factor(week)22   18.5942    10.8013    1.721 0.085185 .
## factor(week)23   22.5426    10.8458    2.078 0.037685 *
## factor(week)24   16.7553    11.1806    1.499 0.134000
## factor(week)25   20.5327    11.3084    1.816 0.069437 .
## factor(week)26    7.4717    11.3090    0.661 0.508823
```

## factor(week)27	13.3017	11.6735	1.139	0.254522	
## factor(week)28	126.9142	11.3049	11.226	< 2e-16	***
## factor(week)29	57.6310	11.5619	4.985	6.28e-07	***
## factor(week)30	25.4194	11.8501	2.145	0.031962	*
## factor(week)31	27.3902	11.5488	2.372	0.017720	*
## factor(week)32	32.7177	11.3413	2.885	0.003922	**
## factor(week)33	21.5342	11.1378	1.933	0.053201	.
## factor(week)34	19.9710	11.2802	1.770	0.076673	.
## factor(week)35	23.5177	11.2427	2.092	0.036473	*
## factor(week)36	42.7294	11.0479	3.868	0.000110	***
## factor(week)37	44.6859	10.2889	4.343	1.41e-05	***
## factor(week)38	44.1526	10.2205	4.320	1.57e-05	***
## factor(week)39	35.9987	10.2099	3.526	0.000423	***
## factor(week)40	40.3781	10.3262	3.910	9.26e-05	***
## factor(week)41	37.9819	10.2181	3.717	0.000202	***
## factor(week)42	36.8931	10.1085	3.650	0.000263	***
## factor(week)43	31.6829	10.0770	3.144	0.001669	**
## factor(week)44	12.7436	10.5854	1.204	0.228653	
## factor(week)45	21.4324	10.1278	2.116	0.034346	*
## factor(week)46	20.7015	10.1558	2.038	0.041529	*
## factor(week)47	20.5786	10.2146	2.015	0.043962	*
## factor(week)48	5.5088	10.6186	0.519	0.603914	
## factor(week)49	8.8089	10.6693	0.826	0.409030	
## factor(week)50	14.7092	10.0570	1.463	0.143602	
## factor(week)51	19.0112	10.0468	1.892	0.058476	.
## factor(week)52	14.0450	10.0528	1.397	0.162398	
## factor(week)53	24.2244	10.0297	2.415	0.015736	*
## factor(week)54	188.1124	10.0515	18.715	< 2e-16	***
## factor(week)55	38.3523	10.0777	3.806	0.000142	***
## factor(week)56	-1.1565	10.0017	-0.116	0.907944	
## factor(week)57	-4.7928	10.4569	-0.458	0.646718	
## factor(week)58	2.4062	10.1506	0.237	0.812619	
## factor(week)59	3.9848	10.3925	0.383	0.701408	
## factor(week)60	23.2274	10.0277	2.316	0.020554	*
## factor(week)61	121.1052	10.0801	12.014	< 2e-16	***
## factor(week)62	44.5783	10.0692	4.427	9.62e-06	***
## factor(week)63	5.3834	9.9838	0.539	0.589745	
## factor(week)64	5.8605	10.4604	0.560	0.575315	
## factor(week)65	1.2656	9.9762	0.127	0.899050	
## factor(week)66	13.1524	9.9120	1.327	0.184556	
## factor(week)67	31.8729	9.9468	3.204	0.001357	**
## factor(week)68	44.3713	10.0841	4.400	1.09e-05	***
## factor(week)69	45.7043	10.0262	4.559	5.19e-06	***
## factor(week)70	41.1927	10.0887	4.083	4.47e-05	***
## factor(week)71	41.4639	9.9849	4.153	3.30e-05	***
## factor(week)72	58.9130	10.0208	5.879	4.22e-09	***
## factor(week)73	303.8825	10.0267	30.307	< 2e-16	***
## factor(week)74	40.2134	9.9938	4.024	5.76e-05	***
## factor(week)75	33.8421	10.0237	3.376	0.000737	***
## factor(week)76	38.3706	10.0230	3.828	0.000130	***
## factor(week)77	35.4360	10.0218	3.536	0.000408	***
## factor(week)78	54.3113	10.0846	5.386	7.33e-08	***
## factor(week)79	171.1303	10.1337	16.887	< 2e-16	***
## factor(week)80	49.6456	10.1263	4.903	9.56e-07	***

## factor(week)81	55.1670	10.0225	5.504	3.77e-08	***
## factor(week)82	50.9635	10.0985	5.047	4.55e-07	***
## factor(week)83	48.9224	10.0128	4.886	1.04e-06	***
## factor(week)84	293.9316	10.0249	29.320	< 2e-16	***
## factor(week)85	46.1738	10.0083	4.614	3.99e-06	***
## factor(week)86	53.7184	9.9987	5.373	7.88e-08	***
## factor(week)87	62.5244	10.0427	6.226	4.92e-10	***
## factor(week)88	103.7413	10.0139	10.360	< 2e-16	***
## factor(week)89	48.6748	10.1971	4.773	1.83e-06	***
## factor(week)90	56.5608	10.0537	5.626	1.88e-08	***
## factor(week)91	31.4695	10.0184	3.141	0.001686	**
## factor(week)92	37.5773	9.9300	3.784	0.000155	***
## factor(week)93	34.4123	9.9792	3.448	0.000566	***
## factor(week)94	36.4961	9.9851	3.655	0.000258	***
## factor(week)95	24.3166	10.0267	2.425	0.015312	*
## factor(week)96	20.3438	10.0312	2.028	0.042574	*
## factor(week)97	20.4977	10.0303	2.044	0.041014	*
## factor(week)98	15.7109	10.4708	1.500	0.133519	
## factor(week)99	15.9032	10.4724	1.519	0.128890	
## factor(week)100	16.7480	9.9992	1.675	0.093968	.
## factor(week)101	15.4032	10.5299	1.463	0.143542	
## factor(week)102	21.1013	10.0264	2.105	0.035345	*
## factor(week)103	17.9604	10.5010	1.710	0.087222	.
## factor(week)104	23.0040	10.0535	2.288	0.022143	*
## factor(week)105	236.1183	9.9571	23.714	< 2e-16	***
## factor(week)106	163.0363	9.9672	16.357	< 2e-16	***
## factor(week)107	32.9097	9.8963	3.325	0.000885	***
## factor(week)108	47.8498	10.2052	4.689	2.77e-06	***
## factor(week)109	23.7042	9.9779	2.376	0.017530	*
## factor(week)110	24.2984	10.3804	2.341	0.019257	*
## factor(week)111	39.7323	9.9097	4.009	6.12e-05	***
## factor(week)112	25.5935	9.9247	2.579	0.009925	**
## factor(week)113	221.0864	9.9092	22.311	< 2e-16	***
## factor(week)114	119.2756	9.9143	12.031	< 2e-16	***
## factor(week)115	39.7831	9.9662	3.992	6.59e-05	***
## factor(week)116	24.4513	9.9390	2.460	0.013900	*
## factor(week)117	26.6132	9.9392	2.678	0.007424	**
## factor(week)118	26.5014	9.9095	2.674	0.007496	**
## factor(week)119	27.9500	9.9323	2.814	0.004899	**
## factor(week)120	36.6042	9.9109	3.693	0.000222	***
## factor(week)121	35.0496	9.9317	3.529	0.000418	***
## factor(week)122	125.9602	10.0031	12.592	< 2e-16	***
## factor(week)123	25.5963	9.9056	2.584	0.009775	**
## factor(week)124	17.5425	9.9489	1.763	0.077876	.
## factor(week)125	20.1312	9.9533	2.023	0.043137	*
## factor(week)126	27.9608	9.9188	2.819	0.004824	**
## factor(week)127	201.9281	10.0769	20.039	< 2e-16	***
## factor(week)128	13.9993	9.9048	1.413	0.157563	
## factor(week)129	28.6200	9.9021	2.890	0.003855	**
## factor(week)130	16.1986	9.9103	1.635	0.102170	
## factor(week)131	18.3100	9.9058	1.848	0.064565	.
## factor(week)132	17.8786	9.9652	1.794	0.072816	.
## factor(week)133	20.0554	9.9378	2.018	0.043599	*
## factor(week)134	19.4797	9.9176	1.964	0.049532	*

## factor(week)135	26.7934	9.9058	2.705	0.006842	**
## factor(week)136	94.8138	9.8966	9.580	< 2e-16	***
## factor(week)137	17.2997	9.8579	1.755	0.079296	.
## factor(week)138	21.0836	9.8960	2.131	0.033145	*
## factor(week)139	42.8556	9.9132	4.323	1.55e-05	***
## factor(week)140	32.7314	9.8574	3.320	0.000901	***
## factor(week)141	23.3636	9.8689	2.367	0.017927	*
## factor(week)142	28.2204	10.0108	2.819	0.004824	**
## factor(week)143	95.4180	10.0084	9.534	< 2e-16	***
## factor(week)144	17.5950	9.9636	1.766	0.077430	.
## factor(week)145	20.7300	10.0328	2.066	0.038825	*
## factor(week)146	20.4586	9.9451	2.057	0.039689	*
## factor(week)147	25.1984	9.9487	2.533	0.011324	*
## factor(week)148	88.0258	9.9367	8.859	< 2e-16	***
## factor(week)149	32.6844	9.8952	3.303	0.000959	***
## factor(week)150	13.7384	10.0233	1.371	0.170505	.
## factor(week)151	18.8133	9.9274	1.895	0.058101	.
## factor(week)152	19.8200	9.8978	2.002	0.045253	*
## factor(week)153	11.9356	10.4270	1.145	0.252362	.
## factor(week)154	57.4455	9.8579	5.827	5.75e-09	***
## factor(week)155	18.8106	10.0070	1.880	0.060163	.
## factor(week)156	16.2379	9.9254	1.636	0.101862	.
## factor(week)157	230.4539	9.8825	23.319	< 2e-16	***
## factor(week)158	120.2192	9.8559	12.198	< 2e-16	***
## factor(week)159	17.9834	9.9079	1.815	0.069537	.
## factor(week)160	18.8139	9.8843	1.903	0.057008	.
## factor(week)161	1174.6969	9.9166	118.458	< 2e-16	***
## factor(week)162	26.6810	9.8638	2.705	0.006840	**
## factor(week)163	10.3208	9.9157	1.041	0.297961	.
## factor(week)164	13.5380	9.8832	1.370	0.170772	.
## factor(week)165	8.2425	10.0769	0.818	0.413395	.
## factor(week)166	9.2891	9.9148	0.937	0.348828	.
## factor(week)167	13.9235	9.9152	1.404	0.160262	.
## factor(week)168	0.3635	9.8640	0.037	0.970605	.
## factor(week)169	21.5975	9.8607	2.190	0.028521	*
## factor(week)170	2.1990	9.8916	0.222	0.824079	.
## factor(week)171	2.6709	9.9760	0.268	0.788912	.
## factor(week)172	1.2012	9.8916	0.121	0.903349	.
## factor(week)173	6.4766	9.8905	0.655	0.512590	.
## factor(week)174	81.0736	10.2697	7.894	3.12e-15	***
## factor(week)175	36.1773	10.2702	3.523	0.000429	***
## factor(week)176	-1.6349	9.8931	-0.165	0.868743	.
## factor(week)177	-0.2837	9.8906	-0.029	0.977117	.
## factor(week)178	611.0807	9.9709	61.286	< 2e-16	***
## factor(week)179	92.3112	9.9782	9.251	< 2e-16	***
## factor(week)180	-10.2435	10.3341	-0.991	0.321591	.
## factor(week)181	-2.6761	9.9183	-0.270	0.787308	.
## factor(week)182	-3.9632	9.8910	-0.401	0.688654	.
## factor(week)183	-3.3458	9.8892	-0.338	0.735117	.
## factor(week)184	-6.3216	10.0039	-0.632	0.527457	.
## factor(week)185	-7.9301	9.9212	-0.799	0.424124	.
## factor(week)186	39.1561	10.1801	3.846	0.000120	***
## factor(week)187	-1.7573	9.9199	-0.177	0.859392	.
## factor(week)188	-14.6410	11.9202	-1.228	0.219373	.

```

## factor(week)189 -11.0933 24.3997 -0.455 0.649367
## factor(week)190 579.7828 31.4204 18.452 < 2e-16 ***
## factor(week)191 -20.0315 43.7986 -0.457 0.647423
## factor(week)192 -13.1714 43.7910 -0.301 0.763588
## factor(week)193 -21.7162 61.5765 -0.353 0.724341
## factor(week)194 -28.7832 61.4996 -0.468 0.639776
## factor(week)195 -34.5086 61.4989 -0.561 0.574721
## factor(week)196 -32.7168 61.5009 -0.532 0.594753
## factor(week)197 -12.6022 61.5125 -0.205 0.837675
## factor(week)198 -29.9069 61.5114 -0.486 0.626832
## factor(week)199 -35.8126 61.5777 -0.582 0.560855
## factor(week)200 -27.6635 61.6200 -0.449 0.653484
## factor(week)204 31.9950 61.5148 0.520 0.602989
## factor(week)205 -3.1236 61.5444 -0.051 0.959523
## factor(week)207 18.8311 61.5445 0.306 0.759628
## factor(week)210 -10.9277 61.5496 -0.178 0.859084
## factor(week)212 -30.2741 61.5208 -0.492 0.622659
## factor(week)229 24.3212 61.5159 0.395 0.692580
## factor(week)255 -42.7601 61.5990 -0.694 0.487587
## factor(week)256 -38.7601 61.5990 -0.629 0.529206
## factor(week)277 16.2798 43.8825 0.371 0.710653
## factor(week)278 1232.5170 43.8680 28.096 < 2e-16 ***
## factor(week)279 -28.5465 43.8297 -0.651 0.514860
## factor(week)280 27.8266 43.9098 0.634 0.526272
## factor(week)281 -19.5654 36.1053 -0.542 0.587897
## factor(week)282 1232.5170 43.8680 28.096 < 2e-16 ***
## factor(week)283 -28.5465 43.8297 -0.651 0.514860
## factor(week)284 -17.7229 43.8128 -0.405 0.685840
## factor(week)285 -11.2128 43.8128 -0.256 0.798011
## factor(week)286 8.1879 43.8130 0.187 0.851755
## factor(week)287 655.6741 43.9602 14.915 < 2e-16 ***
## factor(week)288 -30.8935 43.8234 -0.705 0.480850
## factor(week)289 -18.6022 61.5125 -0.302 0.762342
## factor(week)298 -36.9830 61.5528 -0.601 0.547960
## factor(week)299 -4.9830 36.0965 -0.138 0.890206
## factor(week)300 -18.9830 61.5528 -0.308 0.757781
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 61.06 on 14518 degrees of freedom
## Multiple R-squared: 0.7959, Adjusted R-squared: 0.7928
## F-statistic: 251.7 on 225 and 14518 DF, p-value: < 2.2e-16

```

### 2.3) regressors: price, promoflag, store fixed effects

Use `feols()` to estimate a store fixed-effects model of sales with regressors: `price`, `promoflag`. Use `summary()` to summarize the results.

```

library(fixest)
summary(feols(sales ~ price + promoflag | store , data = DF1))

```

```

## OLS estimation, Dep. Var.: sales
## Observations: 14,744

```



```
## Fixed-effects: store: 86
## Standard-errors: Clustered (store)
##           Estimate Std. Error  t value  Pr(>|t|)
## price      -76.4248    3.30935 -23.0936 < 2.2e-16 ***
## promoflag   21.8063    1.39980  15.5782 < 2.2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## RMSE: 118.8      Adj. R2: 0.211508
##                Within R2: 0.179565
```

## 2.4) regressors: price, promoflag, time (week) fixed effects, store fixed effects

Use `feols()` to estimate a store fixed-effects model of sales with regressors: `price`, `promoflag`, and time (week) fixed effects. Use `summary()` to summarize the results.

```
summary(feols(sales ~ price + promoflag | store + week , data = DF1))
```

```
## OLS estimation, Dep. Var.: sales
## Observations: 14,744
## Fixed-effects: store: 86, week: 224
## Standard-errors: Clustered (store)
##           Estimate Std. Error  t value  Pr(>|t|)
## price      -22.499122    6.39858 -3.516269 0.00070476 ***
## promoflag    0.219081    1.65204  0.132613 0.89481311
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## RMSE: 54.0      Adj. R2: 0.834627
##                Within R2: 0.015512
```

## Discussion questions

**What patterns do you notice? Adding which controls (time trends, time fixed effects, store fixed effects) leads to the greatest change in the price and promotion parameter estimates?** Variety of patterns are noticed. On adding time fixed effects the value of price and promotion coefficient estimate changes and standard error also changes a little bit nit much though whereas on just adding time trend we can see that the effect of “week” is not significant.

However on adding store fixed effects through `feols()` we observe the greatest change in price estimate and for the promotion estimate it’s the same as we get for other models (except when we consider fixed effects of both week and store).

When we take away the fixed effect of both week and store together then we see the lowest change in promotion estimate, however since the p value of the promotion store is high this indicates that relationship is not statistically significant and is out of randomness. Additionally the standard error has increased for price estimate when both the fixed effects are removed compared to when only store fixed effect is considered.

**What does this suggest about sources of omitted variable bias? E.g., are omitted factors more likely associated with cross-sectional units (stores) or time periods?** Omitted factors are likely associated with stores because when the fixed effect of stores was taken away we observed a larger decrease in sales of units when price increases by 1 unit.

**Which estimate would you report as your “best” estimate of the demand response to price? Why?** store. Because it shows greatest decrease in units with price increase.

## Task 2

Your second task is to evaluate the effectiveness of an online display ad campaign.

You have data from an experiment designed to measure the effectiveness of an online display advertising campaign. The experiment involves randomly assigning Internet users to a test or a control group based on cookies that uniquely identify each user visiting a site where the ad exchange (Rocket Fuel) can place an ad. Users in the test group see an ad for a newly released handbag by TaskaBella, Rocket Fuel's client. Users in the control group are shown a public service announcement that is unrelated to the advertised product. Based on the unique IDs, Rocket Fuel is able to track which users eventually purchased a handbag from TaskaBella, allowing the analyst to discern the effectiveness of the campaign.

Each row in the CSV file data set `rocketfuel_data.csv` here represents a uniquely identified user in the ad campaign. For each user, the following six variables are provided:

<code>user_id</code>	Unique identifier of the user
<code>test</code>	1 if the user was exposed to the real ad 0 if the user was in the control group and was shown a PSA
<code>converted</code>	1 if the user made a purchase, 0 otherwise
<code>tot_impr</code>	Total number of ad impressions the user encountered (treat=ad, control=PSA)
<code>mode_impr_day</code>	Day of the week on which the user encountered the most impressions (1=Mon,...,7=Sun)
<code>mode_impr_hour</code>	Hour of the day (0-23) in which the user encountered the most impressions

For these data, `converted` is the outcome, and `test` is the treatment indicator. `user_id` uniquely identifies users (and rows). The remaining variables provide additional information observed during the experiment.

The client firm TaskaBella estimates that a conversion generates approximately \$40 in incremental profit for the firm. The cost to serve ads in the experiment was \$9 CPM (\$9 per 1000 impressions).

### 3) Exploratory analysis

Read the data into R and perform some exploratory analysis. Show your work in the R chunks below, and provide text answers following the R chunk.

```
RFD <- read.csv("~/Desktop/Classroom - R/Marketing Analytics/data/rocketfuel_data.csv")
sum(RFD$test)
```

#### 3.1) How many users are in the test and control conditions?

```
## [1] 564577
```

```
sum(RFD$test == 0)
```

```
## [1] 23524
```

Users in test = 564577

Users in control = 23524

**3.2) Conversion rates** The conversion rate for a group is the fraction of users that purchase (`converted==1`) in that group.

What is the conversion rate (in percentage) for the – a) test group and b) control group?

```
mean(RFD[RFD$test == 0, ]$converted)
```

```
## [1] 0.01785411
```

```
mean(RFD[RFD$test == 1, ]$converted)
```

```
## [1] 0.02554656
```

a) test group = 2.55%

b) control group = 1.8%

#### 4) Randomization checks

Verify that Rocketfuel implemented the randomization correctly by examining whether the distributions of the variable `tot_impr` for the test and control groups are the same. If the average number of impressions that users see in each group is different, then the differences in their response rate can be (potentially) attributed to this instead of the ads that they see. We can examine the distribution of `tot_impr` for the two groups in three ways: (simple) mean comparison, distribution (histogram) comparison, and formal difference in means t-tests.

**4.1) Mean comparison** Using the `describe` command, summarize `tot_impr` for two the groups of users (in test and control conditions). What is the mean of this variable each of these groups? Are the means similar?

```
summary(RFD$tot_impr[RFD$test == 1], na.rm = T)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      1.00    4.00   13.00   24.82   27.00  2065.00
```

```
summary(RFD$tot_impr[RFD$test == 0], na.rm = T)
```

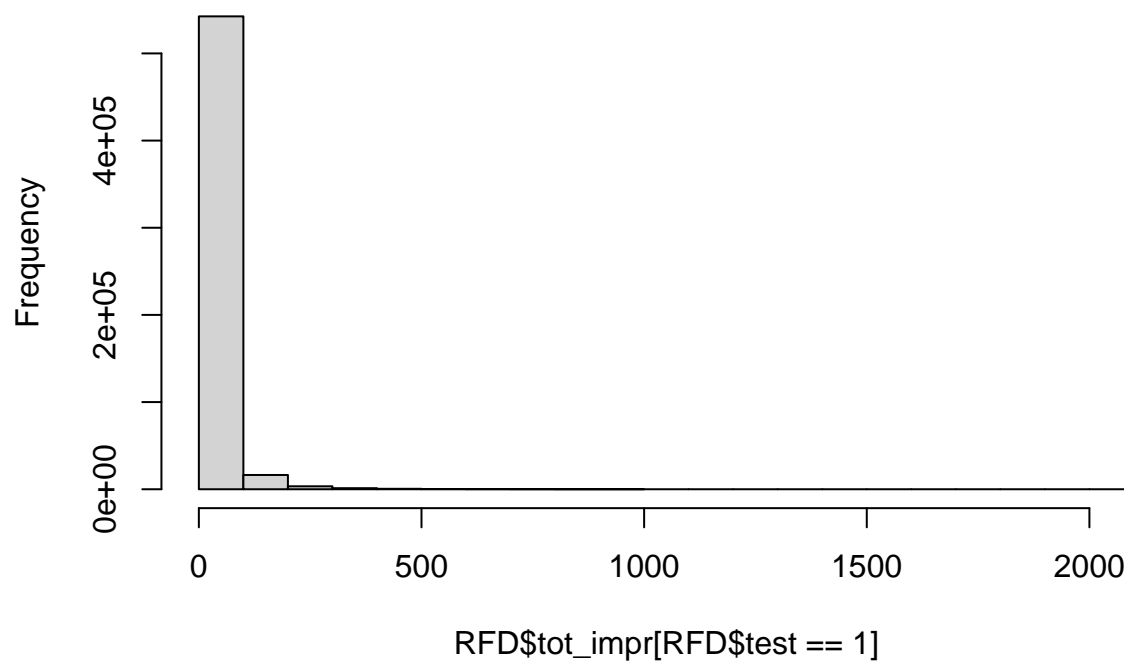
```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      1.00    4.00   12.00   24.76   26.00   907.00
```

*Yes, the means are similar*

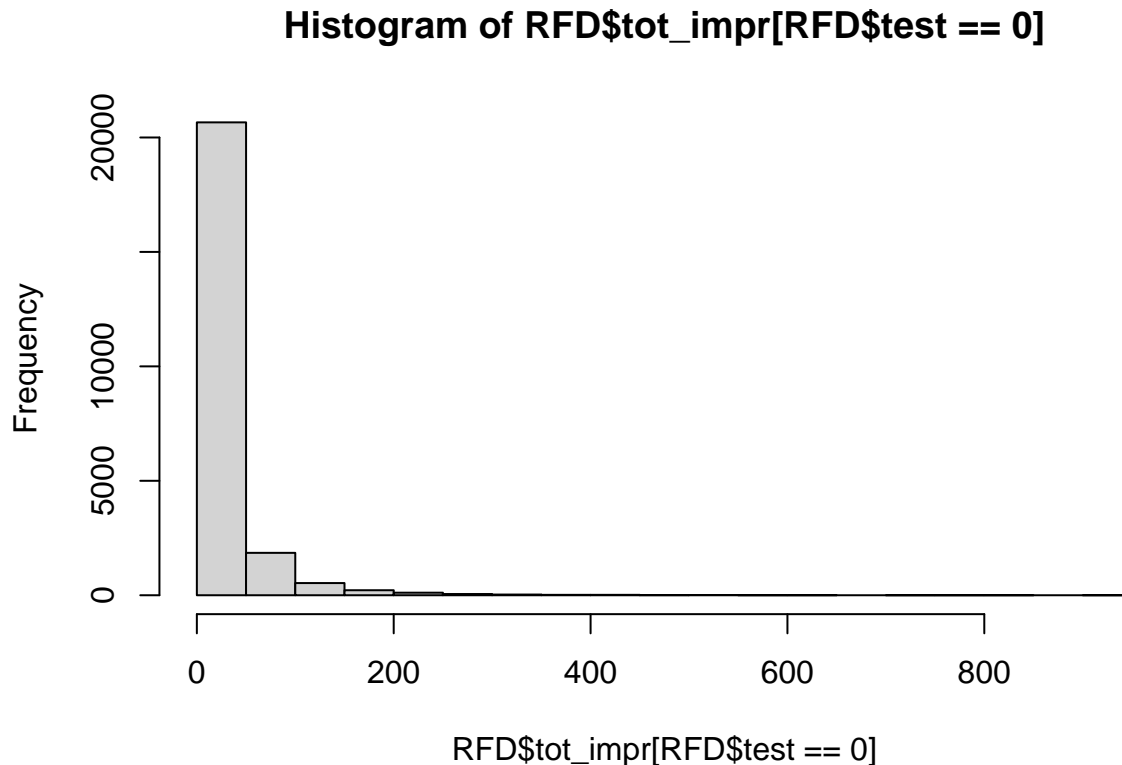
**4.2) Distribution (histogram) comparison** To further understand how the distribution of `tot_impr` looks for the two groups, plot the histograms of `tot_impr` for each of the two groups (test and control). Do the two histograms look similar?

```
hist(RFD$tot_impr[RFD$test == 1])
```

**Histogram of RFD\$tot\_impr[RFD\$test == 1]**



```
hist(RFD$tot_impr[RFD$test == 0])
```



*No the histograms don't look similar. The frequency of tot\_impr shown to some of test users is way higher than shown to some of control group*

**4.3) Formal difference in means t-test** Finally, conduct a t-test to examine whether the differences (if any) in tot\_impr across the two groups is statistically significant?

```
t.test( tot_impr ~ test, data = RFD)
```

```
##
## Welch Two Sample t-test
##
## data: tot_impr by test
## t = -0.218, df = 25608, p-value = 0.8274
## alternative hypothesis: true difference in means between group 0 and group 1 is not equal to 0
## 95 percent confidence interval:
## -0.6217286 0.4972735
## sample estimates:
## mean in group 0 mean in group 1
## 24.76114 24.82337
```

Since P value is very high at 82% hence the relationship is not statistically significant.

Based on the above analyses, can you conclude that the randomization was done correctly?

The mean and the p-value suggestions that the difference between control group and test group is not statistically significant and that the randomization was done correctly.

## 5) Average treatment effect (ATE) estimation & application

**5.2) Compute the treatment effect “by hand”** Calculate the ATE as the difference in mean outcomes across the treatment and control conditions. Report your ATE estimate as a percentage. Was the campaign effective?

```
ATE<-mean(RFD[RFD$test==1, "converted"]) -  
mean(RFD[RFD$test==0, "converted"])
```

0.77%

**5.3) Compute the treatment effect by regression** Use a regression to estimate the treatment effect (ATE). Does your estimate match the “by hand” calculation? What is the standard error of the ATE?

```
summary(lm(converted ~ test, data = RFD))
```

```
##  
## Call:  
## lm(formula = converted ~ test, data = RFD)  
##  
## Residuals:  
##      Min       1Q   Median       3Q      Max   
## -0.02555 -0.02555 -0.02555 -0.02555  0.98215   
##  
## Coefficients:  
##              Estimate Std. Error t value Pr(>|t|)      
## (Intercept)  0.017854   0.001023   17.46 < 2e-16 ***  
## test         0.007692   0.001044    7.37 1.7e-13 ***  
## ---  
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
##  
## Residual standard error: 0.1568 on 588099 degrees of freedom  
## Multiple R-squared:  9.236e-05, Adjusted R-squared:  9.066e-05   
## F-statistic: 54.32 on 1 and 588099 DF, p-value: 1.703e-13
```

*Yes, the estimate matches the one by hand. The standard error is 0.0010, very low.*

## 6) Return on investment (ROI)

We did not do this in class. But for extra points, give it a try.

**6.1) Campaign incremental conversions** For the users in the test group, how many extra conversions can be attributed to the ad campaign? In other words, what is the incremental number of conversions from the ad campaign?

Hint: the ATE is incremental (causal) effect of the campaign on conversion for each user. The total effect of the campaign on conversion is the number of users in the treatment condition times the ATE.

```
incremental_conv<-(sum(RFD$test == 1) * ATE)
```

4342

**6.2) Campaign incremental profit** Recall from the overview above that TaskaBella gets on average \$40 for each conversion.

How much more money did TaskaBella make by running the campaign (excluding advertising costs)? In other words, what is the incremental profit from the ad campaign?

```
incremental_prof<-incremental_conv* c(40)
```

\$173719

**6.3) Campaign cost** What was the cost of the campaign?

Hint: the relevant number of impressions is contained in the `tot_impr` variable.

```
total_impressions<-sum(RFD$tot_impr)
campaign_cost<- total_impressions*c(0.009)
```

\$131375

**6.4) ROI calculation** Calculate the ROI of the campaign. Percentage ROI is defined as:  $100 * (\text{incremental\_profit} - \text{campaign\_cost}) / \text{campaign\_cost}$ .

```
result <- incremental_prof - campaign_cost
ROI<- result/campaign_cost
print(ROI)
```

```
## [1] 0.3223198
```

32% is the ROI

**6.5) Control group opportunity cost** If the ad campaign had been shown to the control group as well, how much additional profit would have been generated? Explain your answer.

```
incremental_conv_control<-(sum(RFD$test == 0) * ATE)
incremental_prof_control<-incremental_conv_control*c(40)
print(incremental_prof_control)
```

```
## [1] 7238.291
```

Incremental profit generated would be \$7238.

If the control group is given the same treatment the effect would be same as that observed for test group which is because of Average treatment effect. Hence we can apply the same ATE for the control group and find incremental profit by times \$40.