## Homework 1

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#### Part 1

- 1. FRESH: annual spending (m.u.) on fresh products (Continuous and Ratio);
- 2. MILK: annual spending (m.u.) on milk products (Continuous and Ratio);
- 3. GROCERY: annual spending (m.u.) on grocery products (Continuous and Ratio);
- 4. FROZEN: annual spending (m.u.) on frozen products (Continuous and Ratio)
- 5. DETERGENTS\_PAPER: annual spending (m.u.) on detergents and paper products (Continuous and Ratio)
- 6. DELICATESSEN: annual spending (m.u.) on and delicatessen products (Continuous and Ratio);
- 7. CHANNEL: customers' Channel Horeca (Hotel/Restaurant/Café) or Retail channel (Discrete and Nominal)
- 8. REGION: customers' Region Lisbon, Oporto or Other (Discrete and Nominal)

#### Part 2

It isn't entirely clear whether total or average spending per region is desired. I assumed average, but for the sake of completeness, total spending is included as well. Most spending is from 'Other' as most data points fell into that region. Likewise, Oporto had the fewest, data points, and the least total spending (and, for the same reason, the highest standard errors).

For the average spending by region, the standard errors are too high for there to be a statistically significant difference in spending on most of the products. The only ones for which one exists is for fresh products (between Oporto and Other) and on delicatessen products (also between Oporto and Other). While not statistically significant, it seems that while Oporto spends less on fresh products, it instead purchases more frozen products than the other two regions.

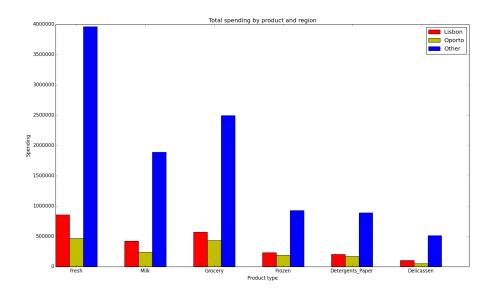


Figure 1: Total spending by region

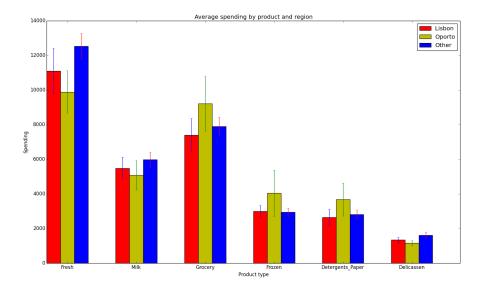


Figure 2: Average spending by region

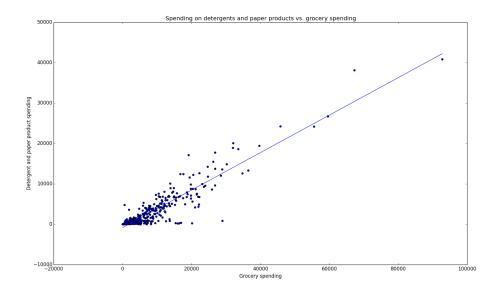


Figure 3: Relationship between spending on groceries and spending on detergents and paper, with data and best-fit line

# Part 3 There is a strong correlation between spending on groceries and spending on detergents and paper. Here is the summary:

Dep. Variable: Model: Method: Date: Time: No. Observations:		Detergents_Pape	er R-so	Adj. R-squared: F-statistic: Prob (F-statistic):		0.855 0.855 2582.	
		0:	LS Adj.				
		Least Square	es F-st				
		Wed, 14 Sep 20	16 Prob			9.56 e - 186 - 3925.7	
		17:29:3	39 Log-				
		440	40 AIC:			7855.	
Df Residuals:		4:	38 BIC:			7864.	
Df Model:			1				
Covariance Type:		nonrobu	st				
	coef	std err	t	P >   t	[95.0% C	onf. Int.]	
Intercept	-807.1336	113.050	-7.140	0.000	-1029.322	-584.945	
Grocery	0.4639	0.009	50.812	0.000	0.446	0.482	
Omnibus:		97.68	34 Durb	in-Watson:		1.932	
Prob (Omnibus):		0.00	00 Jarq	ue - Bera (JB):		1291.839	
Skew:		-0.5	11 Prob	(JB):		3.02e-281	
Kurtosis:		11.33	32 Cond	. No.		1.62e+04	

#### Part 4

```
#! /usr/bin/env python
import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
```

```
import statsmodels.formula.api as sm
# Read the data
data = pd.read_csv('Wholesale customers data.csv')
totals = []
means = []
stderrs = []
# Compute average spending by category
for i in range (1, 4):
      # Get data from region i (1, 2, or 3) for spending on Fresh, Milk, ←
             &c .
       regdata = data[data.Region == i][['Fresh', 'Milk', 'Grocery', '←
Frozen', 'Detergents_Paper', 'Delicassen']]
            Frozen', 'Detergents_Paper',
       totals += [regdata.sum().as_matrix()]
       means += [regdata.mean().as_matrix()]
       # Compute standard error of mean
       stderrs += [ regdata.std().as_matrix() / np.sqrt(len(regdata)) ]
\mathtt{fig}\;,\;\;\mathtt{ax}\;=\;\mathtt{plt}\;.\,\mathtt{subplots}\;(\,)
fig.set\_size\_inches(18.5, 10.5) \# Default size is too small
# Bars for bar graph - using total
rects = [ax.bar(ind, totals[0], width, color='r'),
ax.bar(ind + width, totals[1], width, color='y'),
ax.bar(ind + 2*width, totals[2], width, color='b')]
# Labels, titles, ticks, &c.
ax.set_xlabel('Product type')
ax.set_ylabel('Spending')
ax.set_title('Total spending by product and region')
ax.set_xticks(ind + 1.5 * width)
ax.set_xticklabels(['Fresh', 'Milk', 'Grocery', 'From Detergents_Paper', 'Delicassen'])
ax legend((rects[0][0] rects[1][0]) rects[2][0]) (
                                                            'Grocery', 'Frozen', '↔
 ax.legend((rects[0][0], rects[1][0], rects[2][0]), ('Lisbon', 'Oporto' \leftarrow
           'Other'))
# Save and clear for next image
plt.savefig('part2a.png')
plt.cla()
# Bars for bar graph - using mean
rects = [ax.bar(ind, means [0], width, color='r', yerr=stderrs [0]), ax.bar(ind + width, means [1], width, color='y', yerr=stderrs [1]), ax.bar(ind + 2*width, means [2], width, color='b', yerr=stderrs [2]) \leftarrow
             ]
# Labels, titles, ticks, &c.
# Labels, titles, ticks, &c.
ax.set_xlabel('Product type')
ax.set_ylabel('Spending')
ax.set_title('Average spending by product and region')
ax.set_xticks(ind + 1.5 * width)
ax.set_xticklabels(['Fresh', 'Milk', 'Grocery', 'Frozer
Detergents_Paper', 'Delicassen'])
                                                             'Grocery', 'Frozen', '↔
\texttt{ax.legend} \ ((\texttt{rects} \ [0] \ [0] \ , \ \texttt{rects} \ [1] \ [0] \ , \ \texttt{rects} \ [2] \ [0]) \ , \ (\texttt{'Lisbon'}, \ \texttt{'Oporto'} \ \leftarrow \ \texttt{oporto'} \ )
           'Other'))
# Save and clear for part 3
plt.savefig('part2b.png')
plt.cla()
\# Ordinary least squares fit for Grocery (independent) and \hookleftarrow
Detergents_Paper(dependent)

fit = sm.ols(formula="Detergents_Paper Grocery", data=data).fit()

ind = np.arange(min(data['Grocery']), max(data['Grocery']))
```

```
print fit.summary()
print fit.pvalues # Print separately because very close to zero

# Plot points and best-fit
ax.scatter(data['Grocery'], data['Detergents_Paper'])
ax.plot(ind, fit.params[0] + ind*fit.params[1])

# Labels and title
ax.set_xlabel('Grocery spending')
ax.set_ylabel('Detergent and paper product spending')
ax.set_title('Spending on detergents and paper products vs. grocery \( \to \) spending')

# Save
plt.savefig('part3.png')
```