

SEC1 3 - Homework

January 24, 2024

```
[1]: import pandas as pd
```

1 0.) Clean the Apple Data to get a quarterly series of EPS.

```
[2]: y = pd.read_csv("AAPL_quarterly_financials.csv")
y
```

```
[2]:
```

	name	ttm	\
0	TotalRevenue	383,285,000,000	
1	\tOperatingRevenue	383,285,000,000	
2	CostOfRevenue	214,137,000,000	
3	GrossProfit	169,148,000,000	
4	OperatingExpense	54,847,000,000	
5	\tSellingGeneralAndAdministration	24,932,000,000	
6	\tResearchAndDevelopment	29,915,000,000	
7	\tDepreciationAmortizationDepletionIncomeState...	NaN	
8	\t\tDepreciationAndAmortizationInIncomeStatement	NaN	
9	\t\t\tOtherOperatingExpenses	NaN	
10	OperatingIncome	114,301,000,000	
11	NetNonOperatingInterestIncomeExpense	-183,000,000	
12	\tInterestIncomeNonOperating	3,750,000,000	
13	\tInterestExpenseNonOperating	3,933,000,000	
14	\tTotalOtherFinanceCost	NaN	
15	OtherIncomeExpense	-382,000,000	
16	\tGainOnSaleOfSecurity	NaN	
17	\tSpecialIncomeCharges	NaN	
18	\t\tRestructuringAndMergernAcquisition	NaN	
19	\t\t\tOtherSpecialCharges	NaN	
20	\t\t\tOtherNonOperatingIncomeExpenses	-382,000,000	
21	PretaxIncome	113,736,000,000	
22	TaxProvision	16,741,000,000	
23	NetIncomeCommonStockholders	96,995,000,000	
24	\tNetIncome	96,995,000,000	
25	\t\tNetIncomeIncludingNoncontrollingInterests	96,995,000,000	
26	\t\t\tNetIncomeContinuousOperations	96,995,000,000	
27	\t\t\tNetIncomeExtraordinary	NaN	
28	\t\t\tNetIncomeFromTaxLossCarryforward	NaN	

29	DilutedNI AvailtoComStockholders	96,995,000,000
30	BasicEPS	6.16
31	DilutedEPS	6.13
32	BasicAverageShares	15,744,231,000
33	DilutedAverageShares	15,812,547,000
34	TotalOperatingIncomeAsReported	114,301,000,000
35	TotalExpenses	268,984,000,000
36	NetIncomeFromContinuingAndDiscontinuedOperation	96,995,000,000
37	NormalizedIncome	96,995,000,000
38	InterestIncome	3,750,000,000
39	InterestExpense	3,933,000,000
40	NetInterestIncome	-183,000,000
41	EBIT	117,669,000,000
42	EBITDA	129,188,000,000
43	ReconciledCostOfRevenue	214,137,000,000
44	ReconciledDepreciation	11,519,000,000
45	NetIncomeFromContinuingOperationNetMinorityInt...	96,995,000,000
46	TotalUnusualItemsExcludingGoodwill	NaN
47	TotalUnusualItems	NaN
48	NormalizedEBITDA	129,188,000,000
49	TaxRateForCalcs	0.147
50	TaxEffectOfUnusualItems	0

	09/30/2023	06/30/2023	03/31/2023	12/31/2022	\
0	89,498,000,000	81,797,000,000	94,836,000,000	117,154,000,000	
1	89,498,000,000	81,797,000,000	94,836,000,000	117,154,000,000	
2	49,071,000,000	45,384,000,000	52,860,000,000	66,822,000,000	
3	40,427,000,000	36,413,000,000	41,976,000,000	50,332,000,000	
4	13,458,000,000	13,415,000,000	13,658,000,000	14,316,000,000	
5	6,151,000,000	5,973,000,000	6,201,000,000	6,607,000,000	
6	7,307,000,000	7,442,000,000	7,457,000,000	7,709,000,000	
7	NaN	NaN	NaN	NaN	
8	NaN	NaN	NaN	NaN	
9	NaN	NaN	NaN	NaN	
10	26,969,000,000	22,998,000,000	28,318,000,000	36,016,000,000	
11	-18,000,000	-18,000,000	-12,000,000	-135,000,000	
12	984,000,000	980,000,000	918,000,000	868,000,000	
13	1,002,000,000	998,000,000	930,000,000	1,003,000,000	
14	NaN	NaN	NaN	NaN	
15	47,000,000	-247,000,000	76,000,000	-258,000,000	
16	NaN	NaN	NaN	NaN	
17	NaN	NaN	NaN	NaN	
18	NaN	NaN	NaN	NaN	
19	NaN	NaN	NaN	NaN	
20	47,000,000	-247,000,000	76,000,000	-258,000,000	
21	26,998,000,000	22,733,000,000	28,382,000,000	35,623,000,000	
22	4,042,000,000	2,852,000,000	4,222,000,000	5,625,000,000	

23	22,956,000,000	19,881,000,000	24,160,000,000	29,998,000,000
24	22,956,000,000	19,881,000,000	24,160,000,000	29,998,000,000
25	22,956,000,000	19,881,000,000	24,160,000,000	29,998,000,000
26	22,956,000,000	19,881,000,000	24,160,000,000	29,998,000,000
27	NaN	NaN	NaN	NaN
28	NaN	NaN	NaN	NaN
29	22,956,000,000	19,881,000,000	24,160,000,000	29,998,000,000
30	1.47	1.27	1.53	1.89
31	1.46	1.26	1.52	1.88
32	15,599,434,000	15,697,614,000	15,787,154,000	15,892,723,000
33	15,672,400,000	15,775,021,000	15,847,050,000	15,955,718,000
34	26,969,000,000	22,998,000,000	28,318,000,000	36,016,000,000
35	62,529,000,000	58,799,000,000	66,518,000,000	81,138,000,000
36	22,956,000,000	19,881,000,000	24,160,000,000	29,998,000,000
37	22,956,000,000	19,881,000,000	24,160,000,000	29,998,000,000
38	984,000,000	980,000,000	918,000,000	868,000,000
39	1,002,000,000	998,000,000	930,000,000	1,003,000,000
40	-18,000,000	-18,000,000	-12,000,000	-135,000,000
41	28,000,000,000	23,731,000,000	29,312,000,000	36,626,000,000
42	30,653,000,000	26,783,000,000	32,210,000,000	39,542,000,000
43	49,071,000,000	45,384,000,000	52,860,000,000	66,822,000,000
44	2,653,000,000	3,052,000,000	2,898,000,000	2,916,000,000
45	22,956,000,000	19,881,000,000	24,160,000,000	29,998,000,000
46	NaN	NaN	NaN	NaN
47	NaN	NaN	NaN	NaN
48	30,653,000,000	26,783,000,000	32,210,000,000	39,542,000,000
49	0.15	0.125	0.149	0.158
50	0	0	0	0

	09/30/2022	06/30/2022	03/31/2022	12/31/2021	...	\
0	90,146,000,000	82,959,000,000	97,278,000,000	123,945,000,000	...	
1	90,146,000,000	82,959,000,000	97,278,000,000	123,945,000,000	...	
2	52,051,000,000	47,074,000,000	54,719,000,000	69,702,000,000	...	
3	38,095,000,000	35,885,000,000	42,559,000,000	54,243,000,000	...	
4	13,201,000,000	12,809,000,000	12,580,000,000	12,755,000,000	...	
5	6,440,000,000	6,012,000,000	6,193,000,000	6,449,000,000	...	
6	6,761,000,000	6,797,000,000	6,387,000,000	6,306,000,000	...	
7	NaN	NaN	NaN	NaN	...	
8	NaN	NaN	NaN	NaN	...	
9	NaN	NaN	NaN	NaN	...	
10	24,894,000,000	23,076,000,000	29,979,000,000	41,488,000,000	...	
11	-74,000,000	3,000,000	9,000,000	-44,000,000	...	
12	753,000,000	722,000,000	700,000,000	650,000,000	...	
13	827,000,000	719,000,000	691,000,000	694,000,000	...	
14	NaN	NaN	NaN	NaN	...	
15	-163,000,000	-13,000,000	151,000,000	-203,000,000	...	
16	NaN	NaN	NaN	NaN	...	

17	NaN	NaN	NaN	NaN	...
18	NaN	NaN	NaN	NaN	...
19	NaN	NaN	NaN	NaN	...
20	-163,000,000	-13,000,000	151,000,000	-203,000,000	...
21	24,657,000,000	23,066,000,000	30,139,000,000	41,241,000,000	...
22	3,936,000,000	3,624,000,000	5,129,000,000	6,611,000,000	...
23	20,721,000,000	19,442,000,000	25,010,000,000	34,630,000,000	...
24	20,721,000,000	19,442,000,000	25,010,000,000	34,630,000,000	...
25	20,721,000,000	19,442,000,000	25,010,000,000	34,630,000,000	...
26	20,721,000,000	19,442,000,000	25,010,000,000	34,630,000,000	...
27	NaN	NaN	NaN	NaN	...
28	NaN	NaN	NaN	NaN	...
29	20,721,000,000	19,442,000,000	25,010,000,000	34,630,000,000	...
30	1.29	1.2	1.54	2.11	...
31	1.29	1.2	1.52	2.1	...
32	16,030,382,000	16,162,945,000	16,278,802,000	16,391,724,000	...
33	16,118,465,000	16,262,203,000	16,403,316,000	16,519,291,000	...
34	24,894,000,000	23,076,000,000	29,979,000,000	41,488,000,000	...
35	65,252,000,000	59,883,000,000	67,299,000,000	82,457,000,000	...
36	20,721,000,000	19,442,000,000	25,010,000,000	34,630,000,000	...
37	20,721,000,000	19,442,000,000	25,010,000,000	34,630,000,000	...
38	753,000,000	722,000,000	700,000,000	650,000,000	...
39	827,000,000	719,000,000	691,000,000	694,000,000	...
40	-74,000,000	3,000,000	9,000,000	-44,000,000	...
41	25,484,000,000	23,785,000,000	30,830,000,000	41,935,000,000	...
42	28,349,000,000	26,590,000,000	33,567,000,000	44,632,000,000	...
43	52,051,000,000	47,074,000,000	54,719,000,000	69,702,000,000	...
44	2,865,000,000	2,805,000,000	2,737,000,000	2,697,000,000	...
45	20,721,000,000	19,442,000,000	25,010,000,000	34,630,000,000	...
46	NaN	NaN	NaN	NaN	...
47	NaN	NaN	NaN	NaN	...
48	28,349,000,000	26,590,000,000	33,567,000,000	44,632,000,000	...
49	0.16	0.157	0.17	0.16	...
50	0	0	0	0	...

	12/31/1987	09/30/1987	06/30/1987	03/31/1987	\
0	1,042,400,000	786,500,000	637,100,000	575,300,000	
1	1,042,400,000	786,500,000	637,100,000	575,300,000	
2	NaN	NaN	NaN	NaN	
3	NaN	NaN	NaN	NaN	
4	NaN	NaN	NaN	NaN	
5	NaN	NaN	NaN	NaN	
6	NaN	NaN	NaN	NaN	
7	NaN	NaN	NaN	NaN	
8	NaN	NaN	NaN	NaN	
9	NaN	NaN	NaN	NaN	
10	1,042,400,000	-1,503,100,000	637,100,000	575,300,000	

11	NaN	NaN	NaN	NaN
12	NaN	NaN	NaN	NaN
13	NaN	NaN	NaN	NaN
14	NaN	NaN	NaN	NaN
15	NaN	NaN	NaN	NaN
16	NaN	NaN	NaN	NaN
17	NaN	NaN	NaN	NaN
18	NaN	NaN	NaN	NaN
19	NaN	NaN	NaN	NaN
20	NaN	NaN	NaN	NaN
21	NaN	NaN	NaN	NaN
22	NaN	NaN	NaN	NaN
23	NaN	NaN	NaN	NaN
24	NaN	NaN	NaN	NaN
25	121,400,000	71,700,000	53,500,000	33,900,000
26	121,400,000	71,700,000	53,500,000	33,900,000
27	NaN	NaN	NaN	NaN
28	NaN	NaN	NaN	NaN
29	NaN	NaN	NaN	NaN
30	0.008	NaN	0.004	0.002
31	0.008	NaN	0.004	0.002
32	14,063,280,000	NaN	14,028,672,000	14,028,672,000
33	14,063,280,000	NaN	14,028,672,000	14,028,672,000
34	NaN	NaN	NaN	NaN
35	NaN	NaN	NaN	NaN
36	NaN	NaN	NaN	NaN
37	NaN	NaN	NaN	NaN
38	NaN	NaN	NaN	NaN
39	NaN	NaN	NaN	NaN
40	NaN	NaN	NaN	NaN
41	1,042,400,000	-1,503,100,000	637,100,000	575,300,000
42	1,042,400,000	-1,503,100,000	637,100,000	575,300,000
43	NaN	NaN	NaN	NaN
44	NaN	NaN	NaN	NaN
45	NaN	NaN	NaN	NaN
46	NaN	NaN	NaN	NaN
47	NaN	NaN	NaN	NaN
48	1,042,400,000	-1,503,100,000	637,100,000	575,300,000
49	NaN	NaN	NaN	NaN
50	NaN	NaN	NaN	NaN

	12/31/1986	09/30/1986	06/30/1986	03/31/1986	\
0	662,300,000	510,800,000	448,300,000	408,900,000	
1	662,300,000	510,800,000	448,300,000	408,900,000	
2	NaN	NaN	NaN	NaN	
3	NaN	NaN	NaN	NaN	
4	NaN	NaN	NaN	NaN	

5	NaN	NaN	NaN	NaN
6	NaN	NaN	NaN	NaN
7	NaN	NaN	NaN	NaN
8	NaN	NaN	NaN	NaN
9	NaN	NaN	NaN	NaN
10	662,300,000	-1,117,600,000	448,300,000	408,900,000
11	NaN	NaN	NaN	NaN
12	NaN	NaN	NaN	NaN
13	NaN	NaN	NaN	NaN
14	NaN	NaN	NaN	NaN
15	NaN	NaN	NaN	NaN
16	NaN	NaN	NaN	NaN
17	NaN	NaN	NaN	NaN
18	NaN	NaN	NaN	NaN
19	NaN	NaN	NaN	NaN
20	NaN	NaN	NaN	NaN
21	NaN	NaN	NaN	NaN
22	NaN	NaN	NaN	NaN
23	NaN	NaN	NaN	NaN
24	NaN	NaN	NaN	NaN
25	58,500,000	32,900,000	32,300,000	31,800,000
26	58,500,000	32,900,000	32,300,000	31,800,000
27	NaN	NaN	NaN	NaN
28	NaN	NaN	NaN	NaN
29	NaN	NaN	NaN	NaN
30	0.004	NaN	0.002	0.002
31	0.004	NaN	0.002	0.002
32	14,260,064,000	NaN	14,260,064,000	14,033,600,000
33	14,260,064,000	NaN	14,260,064,000	14,033,600,000
34	NaN	NaN	NaN	NaN
35	NaN	NaN	NaN	NaN
36	NaN	NaN	NaN	NaN
37	NaN	NaN	NaN	NaN
38	NaN	NaN	NaN	NaN
39	NaN	NaN	NaN	NaN
40	NaN	NaN	NaN	NaN
41	662,300,000	-1,117,600,000	448,300,000	408,900,000
42	662,300,000	-1,117,600,000	448,300,000	408,900,000
43	NaN	NaN	NaN	NaN
44	NaN	NaN	NaN	NaN
45	NaN	NaN	NaN	NaN
46	NaN	NaN	NaN	NaN
47	NaN	NaN	NaN	NaN
48	662,300,000	-1,117,600,000	448,300,000	408,900,000
49	NaN	NaN	NaN	NaN
50	NaN	NaN	NaN	NaN

	12/31/1985	09/30/1985
0	533,900,000	409,700,000
1	533,900,000	409,700,000
2	NaN	NaN
3	NaN	NaN
4	NaN	NaN
5	NaN	NaN
6	NaN	NaN
7	NaN	NaN
8	NaN	NaN
9	NaN	NaN
10	533,900,000	-1,361,300,000
11	NaN	NaN
12	NaN	NaN
13	NaN	NaN
14	NaN	NaN
15	NaN	NaN
16	NaN	NaN
17	NaN	NaN
18	NaN	NaN
19	NaN	NaN
20	NaN	NaN
21	NaN	NaN
22	NaN	NaN
23	NaN	NaN
24	NaN	NaN
25	56,900,000	22,300,000
26	56,900,000	22,300,000
27	NaN	NaN
28	NaN	NaN
29	NaN	NaN
30	0.004	NaN
31	0.004	NaN
32	13,837,824,000	NaN
33	13,837,824,000	NaN
34	NaN	NaN
35	NaN	NaN
36	NaN	NaN
37	NaN	NaN
38	NaN	NaN
39	NaN	NaN
40	NaN	NaN
41	533,900,000	-1,361,300,000
42	533,900,000	-1,361,300,000
43	NaN	NaN
44	NaN	NaN
45	NaN	NaN

46	NaN	NaN
47	NaN	NaN
48	533,900,000	-1,361,300,000
49	NaN	NaN
50	NaN	NaN

[51 rows x 155 columns]

```
[3]: y.index = y.name
```

```
[4]: y = pd.DataFrame(y.loc['BasicEPS', :]).iloc[2:,:]
y.index = pd.to_datetime(y.index)
y = y.sort_index().fillna(0.) #TIMES where earnings where not issued
```

```
[5]: y
```

```
[5]:
```

	BasicEPS
1985-09-30	0.0
1985-12-31	0.004
1986-03-31	0.002
1986-06-30	0.002
1986-09-30	0.0
...	...
2022-09-30	1.29
2022-12-31	1.89
2023-03-31	1.53
2023-06-30	1.27
2023-09-30	1.47

[153 rows x 1 columns]

```
[6]: y = y.loc["2004-03-31":]
```

```
[7]: y
```

```
[7]:
```

	BasicEPS
2004-03-31	0.002
2004-06-30	0.003
2004-09-30	0.0
2004-12-31	0.013
2005-03-31	0.013
...	...
2022-09-30	1.29
2022-12-31	1.89
2023-03-31	1.53
2023-06-30	1.27
2023-09-30	1.47

[79 rows x 1 columns]

- 2 1.) Come up with 6 search terms you think could nowcast earnings. (Different than the ones I used) Add in 3 terms that you think will not Nowcast earnings. Pull in the gtrends data. Clean it to have a quarterly average.

```
[8]: from pytrends.request import TrendReq
```

```
[9]: # Create pytrends object
pytrends = TrendReq(hl='en-US', tz=360)

# Set up the keywords and the timeframe
keywords = ['Iphone', 'Apple Event', 'New Phone', 'Cheap Phone', 'Tim Cook',
            'Used Phone', 'Vacations', 'Greece', 'Broken Screen'] # Add your keywords
            # here
start_date = '2004-01-01'
end_date = '2023-09-30'

# Create an empty DataFrame to store the results
df = pd.DataFrame()

# Iterate through keywords and fetch data
for keyword in keywords:
    pytrends.build_payload([keyword], cat=0, timeframe=f'{start_date}'
                           f'{end_date}', geo='', gprop='')
    interest_over_time_df = pytrends.interest_over_time()
    df[keyword] = interest_over_time_df[keyword]
```

```
[10]: df
```

```
[10]:
```

	Iphone	Apple Event	New Phone	Cheap Phone	Tim Cook	Used Phone	\
date							
2004-01-01	0	1	51	100	4	35	
2004-02-01	0	0	49	96	2	31	
2004-03-01	0	1	45	86	2	36	
2004-04-01	0	2	50	95	1	37	
2004-05-01	0	2	48	86	2	25	
...	
2023-05-01	45	4	72	23	12	53	
2023-06-01	47	11	74	24	17	51	
2023-07-01	48	4	79	24	9	49	
2023-08-01	47	18	82	26	8	54	
2023-09-01	66	82	90	24	23	58	

	Vacations	Greece	Broken Screen
date			
2004-01-01	96	68	6
2004-02-01	85	69	10
2004-03-01	74	69	9
2004-04-01	70	74	11
2004-05-01	63	82	9
...
2023-05-01	12	38	51
2023-06-01	13	46	50
2023-07-01	14	48	46
2023-08-01	12	41	47
2023-09-01	13	44	52

[237 rows x 9 columns]

```
[11]: df = df.resample("Q").mean()
```

3 3.) Normalize all the X data

```
[12]: from sklearn.preprocessing import StandardScaler
from sklearn.linear_model import Lasso
from sklearn.model_selection import train_test_split
```

```
[19]: scaler = StandardScaler()
X_scaled = scaler.fit_transform(df)
```

4 2.) Import data. Train, Test, Holdout (80%,15%,5%)

```
[20]: X_train, X_temp, y_train, y_temp = train_test_split(df, y, test_size=0.4,
↳random_state =42)

X_test, X_holdout, y_test, y_holdout = train_test_split(X_temp, y_temp,
↳test_size=0.4, random_state =42)
```

```
[ ]:
```

5 4.) Run a Lasso with lambda of .5. Plot a bar chart.

```
[21]: import matplotlib.pyplot as plt
```

```
[22]: lasso =Lasso(alpha=0.2)
```

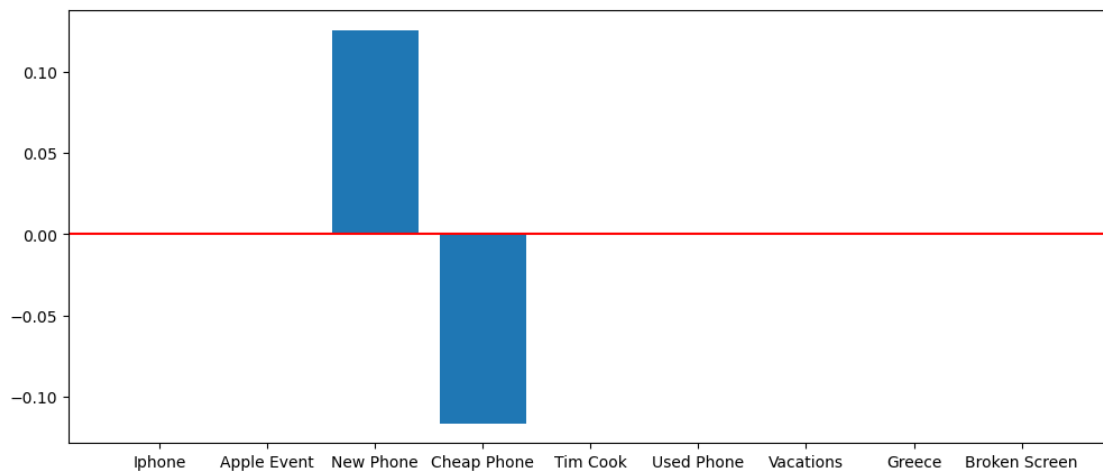
```
[23]: lasso.fit(X_scaled,y)
```

```
[23]: Lasso(alpha=0.2)
```

```
[24]: coefficients = lasso.coef_  
coefficients
```

```
[24]: array([ 0.          ,  0.          ,  0.12548522, -0.11635773,  0.          ,  
         0.          , -0.          , -0.          ,  0.          ])
```

```
[25]: plt.figure( figsize =(12,5))  
# plt.bar(range(len(coefficients)), coefficients)  
plt.bar(df.columns, coefficients)  
plt.axhline(0, color = 'red')  
plt.show()
```



```
[ ]:
```

6 5.) Do these coefficient magnitudes make sense?

```
[ ]:
```

7 6.) Run a for loop looking at 10 different Lambdas and plot the coefficient magnitude for each.

```
[ ]:
```

8 7.) Run a cross validation. What is your ideal lambda?

[]:

[]: