

# Final Project: Semiconductor Company Sales Analysis by Kaelan Yim

for COMP 122 - Spring 2023

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## Introduction

I will be using a chart of semiconductor company market revenue worldwide from 2009 to 2022 from Statista. Some companies I expect to see on there are well-known chip manufacturers like Intel, Samsung, and Micron.

The questions I will analyze are:

1. Which company grew the most from 2009 to 2022?
2. How this number compares to the average of the company's year-over-year gains?
3. Which company produced the most revenue cumulatively?

**Import Required Libraries:** The first step of coding is to import the required libraries such as Pandas, NumPy, and Matplotlib. These libraries provide useful functions for data analysis and visualization.

```
In [ ]: import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
```

**Load the Dataset:** Use the Pandas library to load the dataset into a Pandas DataFrame object.

```
In [ ]: df = pd.read_csv(r'.\semicon.csv')
```

**Explore the Dataset:** Use various Pandas functions to explore the dataset, including:

**head()** and **tail()** functions to view the first and last few rows of the dataset.

**info()** function to get information about the data types of each column and the number of non-null values in each column.

**describe()** function to get a summary of the dataset's statistical measures such as mean, standard deviation, minimum and maximum values, etc.

**value\_counts()** function to get the frequency of each unique value in a particular column.

```
In [ ]: print(df.info())
        print(df.head(5))
        print(df.describe())
        # tail and value_counts aren't relevant here
        # the output is severely truncated, the print for describe doesn't even show up bel
```

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

```
RangeIndex: 14 entries, 0 to 13
```

```
Data columns (total 19 columns):
```

#	Column	Non-Null Count	Dtype
0	Year	14 non-null	int64
1	Samsung Electronics	14 non-null	float64
2	Intel	14 non-null	float64
3	SK Hynix	14 non-null	float64
4	Qualcomm	14 non-null	float64
5	Micron Technology	14 non-null	float64
6	Broadcom	14 non-null	float64
7	AMD	3 non-null	float64
8	Texas Instruments	14 non-null	float64
9	MediaTek	5 non-null	float64
10	Apple	2 non-null	float64
11	STMicroelectronics	9 non-null	float64
12	Western Digital	2 non-null	float64
13	Renesas Technology	6 non-null	float64
14	Infineon	2 non-null	float64
15	NXP	4 non-null	float64
16	Kioxia (Toshiba Memory)	10 non-null	float64
17	Nvidia	3 non-null	float64
18	Others	14 non-null	float64

```
dtypes: float64(18), int64(1)
```

```
memory usage: 2.2 KB
```

```
None
```

	Year	Samsung Electronics	Intel	SK Hynix	Qualcomm	Micron Technology
0	2009	17.75	33.43	6.04	6.41	4.17 \
1	2010	28.10	42.00	9.90	7.20	8.20
2	2011	27.76	50.67	9.39	10.00	7.64
3	2012	28.62	49.09	8.97	13.18	6.92
4	2013	30.64	48.59	12.63	17.21	11.92

	Broadcom	AMD	Texas Instruments	MediaTek	Apple	STMicroelectronics
0	4.32	NaN	9.14	NaN	NaN	8.46 \
1	6.60	NaN	11.90	NaN	NaN	10.30
2	7.16	NaN	11.75	NaN	NaN	9.64
3	7.85	NaN	11.11	NaN	NaN	8.42
4	8.20	NaN	10.59	NaN	NaN	8.08

	Western Digital	Renesas Technology	Infineon	NXP
0	NaN	4.54	NaN	NaN \
1	NaN	10.20	NaN	NaN
2	NaN	10.65	NaN	NaN
3	NaN	9.15	NaN	NaN
4	NaN	7.98	NaN	NaN

	Kioxia (Toshiba Memory)	Nvidia	Others
0	9.60	NaN	124.82
1	12.40	NaN	152.58
2	11.77	NaN	151.34
3	10.61	NaN	145.99
4	11.28	NaN	148.32

	Year	Samsung Electronics	Intel	SK Hynix	Qualcomm
count	14.0000	14.000000	14.000000	14.000000	14.000000 \

mean	2015.5000	44.913571	55.595000	19.807857	16.384286
std	4.1833	18.579894	11.372546	10.802587	7.376349
min	2009.0000	17.750000	33.430000	6.040000	6.410000
25%	2012.2500	29.125000	49.485000	10.582500	13.287500
50%	2015.5000	38.975000	53.210000	16.180000	15.750000
75%	2018.7500	60.165000	64.400000	26.240000	17.547500
max	2022.0000	73.710000	72.760000	36.350000	34.750000

	Micron Technology	Broadcom	AMD	Texas Instruments	MediaTek
count	14.000000	14.000000	3.000000	14.000000	5.000000 \
mean	16.655714	12.107857	16.420000	12.911429	12.300000
std	8.577648	5.621209	6.810793	2.600275	5.370731
min	4.170000	4.320000	9.670000	9.140000	6.700000
25%	9.130000	7.937500	12.985000	11.532500	7.960000
50%	15.045000	10.825000	16.300000	11.900000	10.990000
75%	22.620000	15.665000	19.795000	13.642500	17.620000
max	29.740000	23.810000	23.290000	18.810000	18.230000

	Apple	STMicroelectronics	Western Digital	Renesas Technology
count	2.000000	9.000000	2.000000	6.000000 \
mean	12.600000	8.520000	6.665000	8.300000
std	7.000357	1.132563	3.528463	2.242115
min	7.650000	6.800000	4.170000	4.540000
25%	10.125000	8.020000	5.417500	7.455000
50%	12.600000	8.420000	6.665000	8.565000
75%	15.075000	9.580000	7.912500	9.937500
max	17.550000	10.300000	9.160000	10.650000

	Infineon	NXP	Kioxia (Toshiba Memory)	Nvidia	Others
count	2.00000	4.000000	10.000000	3.000000	14.000000
mean	6.25000	8.405000	10.361000	11.596667	179.997143
std	0.79196	1.264186	1.323585	4.816787	46.785746
min	5.69000	6.540000	7.830000	7.330000	124.820000
25%	5.97000	8.197500	9.680000	8.985000	149.075000
50%	6.25000	8.885000	10.490000	10.640000	158.825000
75%	6.53000	9.092500	11.127500	13.730000	196.832500
max	6.81000	9.310000	12.400000	16.820000	277.500000

## Data Wrangling

**Clean the Dataset:** This step involves cleaning the dataset by handling missing or duplicate values, fixing data types, and removing irrelevant columns.

Some of the common data cleaning techniques include:

Dropping duplicates using **drop\_duplicates()** function.

Handling missing values using **fillna()** function or by removing rows with missing values using **dropna()**function.

Converting data types of columns using **astype()** function.

Renaming columns using **rename()** function.

Removing irrelevant columns using **drop()** function.

```
In [ ]: # no things I want to do here
        # I would do fillna but it makes it harder to find the yearly average earnings when
```

## Exploratory Data Analysis

**Analyze the Dataset:** Once the dataset is cleaned, it's time to perform some analysis to gain insights. This can involve creating visualizations using Matplotlib and performing statistical analysis to answer your research questions.

### Research Question 1: Which company grew the most from 2009 to 2022?

```
In [ ]: amogus = df.columns.tolist()
        amogus = amogus[1:-1]
        ax = df.plot.bar(x="Year", y=amogus, rot=0)
        # not sure how to reposition the legend

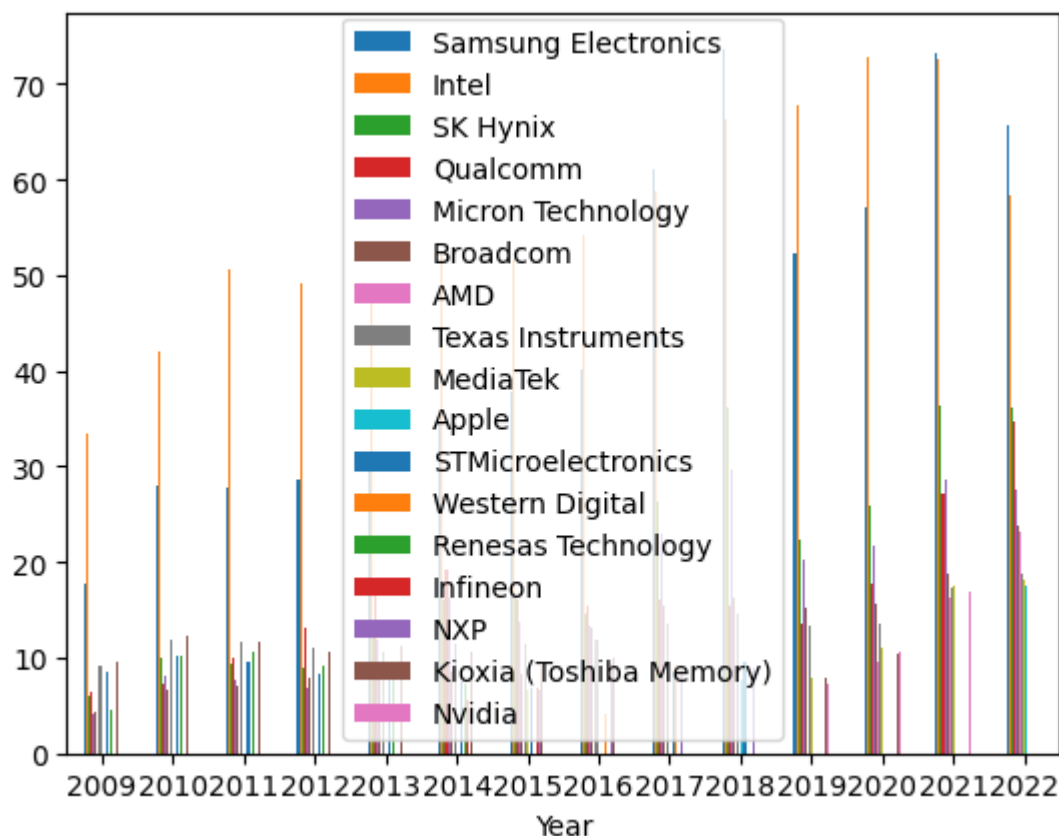
        sus1 = 0
        sus2 = 0
        sus3 = ""
        sus4 = ""
        sus5 = 0
        sus6 = 0

        for column in df.columns[1:-1]:
            begval = df[column].iloc[0]
            endval = df[column].iloc[-1]
            sus1 = endval - begval
            sus3 = column
            if sus1 > sus2:
                sus2 = sus1
                sus4 = sus3
                sus5 = begval
                sus6 = endval
            # yes I excluded "Year" and "Others" since neither are really applicable here

        sus7 = ((sus6 - sus5) / sus5) * 100
        print("The company with the largest growth was", sus4, "with", sus2, "billion dollars more revenue in the final reported year than the first reported year.")
        print("This was a", sus7, "percent increase in revenue.")
```

The company with the largest growth was Samsung Electronics with 47.84 billion dollars more revenue in the final reported year than the first reported year.

This was a 269.5211267605634 percent increase in revenue.



Research Question 2: How does this number compares to the average of the company's year-over-year gains?

```
In [ ]: sus8 = 0
sus9 = 0
sus10 = ""
sus11 = ""

for column in df.columns[1:-1]:
    amo1 = df[column].diff()
    avgchg = amo1.mean()
    sus8 = avgchg
    sus10 = column
    if sus8 > sus9:
        sus9 = sus8
        sus11 = sus10

amo2 = df[sus3].diff()
amo3 = amo2.mean()
sus12 = amo3

print("The company with the highest year-over-year gain was", sus11, "at", sus9, "p
print("The previous company mentioned above (", sus4, ") had a", sus12, "annual rev
```

The company with the highest year-over-year gain was AMD at 6.81 percent per year. The previous company mentioned above ( Samsung Electronics ) had a 4.745 annual revenue increase.

### Research Question 3: Which company produced the most revenue cumulatively?

```
In [ ]: sus13 = 0
        sus14 = ""

        for column in df.columns[1:-1]:
            coltot = df[column].sum()
            if coltot > sus13:
                sus13 = coltot
                sus14 = column

        print(sus14, "had the highest total cumulative revenue at", sus13, "billion dollars")
```

Intel had the highest total cumulative revenue at 778.33 billion dollars over the reported time period.

## Conclusions

Draw Conclusion: Finally, summarize your findings and draw conclusions based on your analysis.

If I had to invest in only one of the companies on the datasheet as a consumer, I would likely invest in **Samsung** because of the large growth the company has experienced. **AMD** would also be a good choice from the data results above but because it only has entries for three years of data the numbers could be unreliable. If I were a hedge fund manager, I would invest in **Intel** because they have produced earnings reliably for a long time.