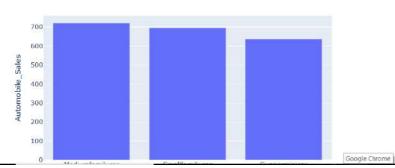
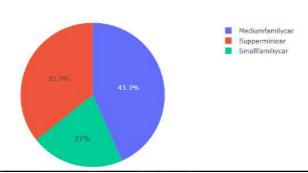
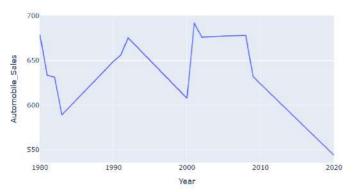


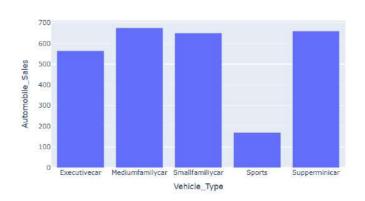
Avg. Vehicle Sales in 1980



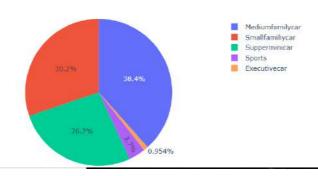
Ad Expenditure by Vehicle Type



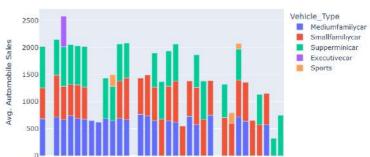




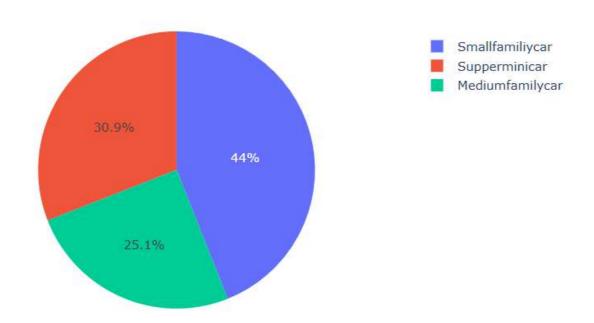
Ad Expenditure by Vehicle Type



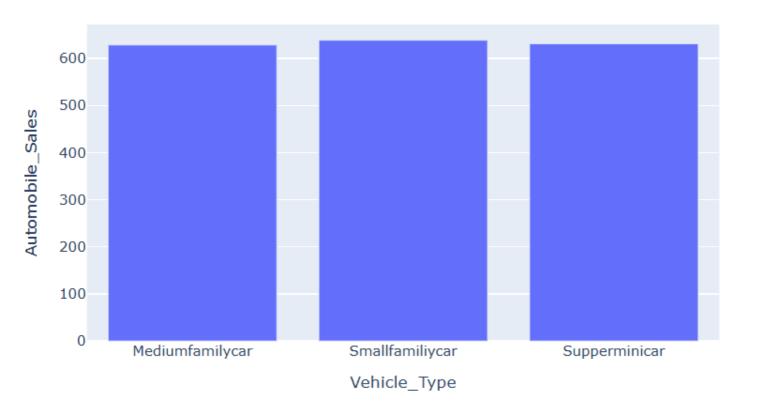
Effect of Unemployment Rate on Vehicle Type and Sales



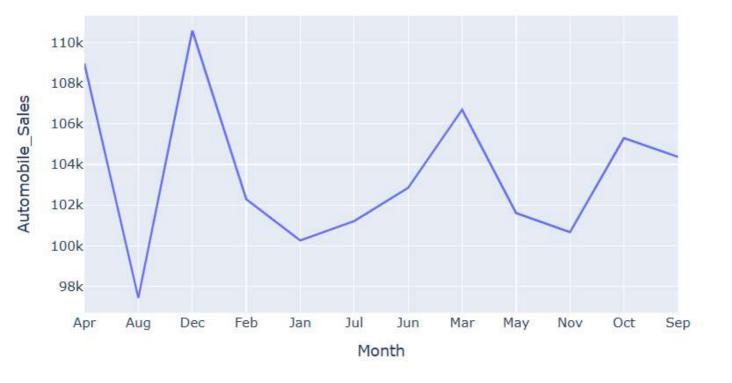
Ad Expenditure by Vehicle Type



Avg. Vehicle Sales in 1981



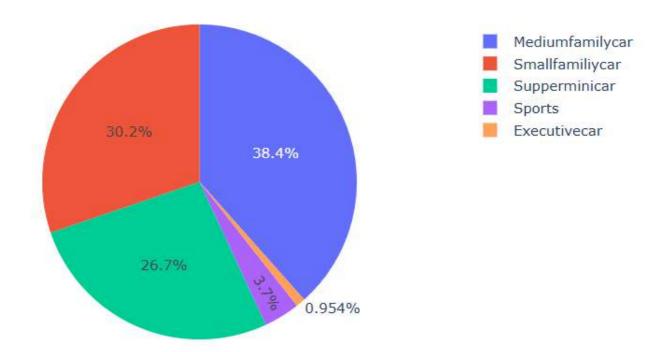
Total Monthly Automobile Sales



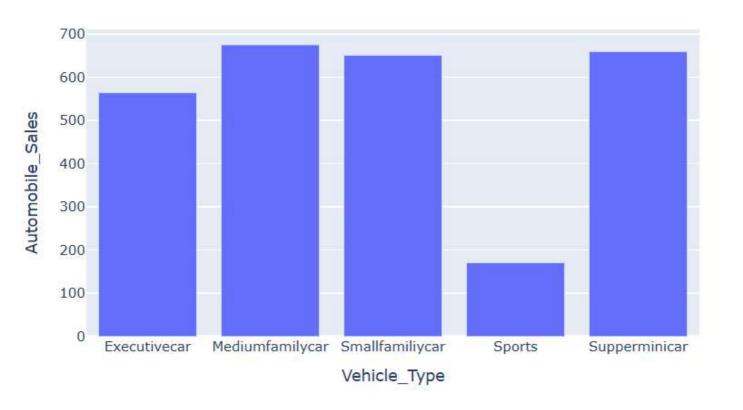
Effect of Unemployment Rate on Vehicle Type and Sales



Ad Expenditure by Vehicle Type



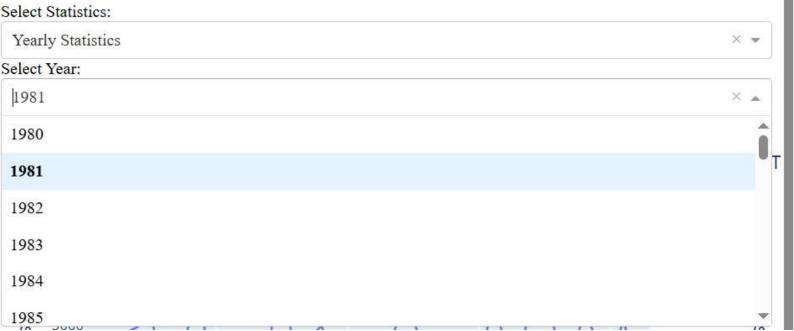
Avg. Vehicle Sales by Type During Recession

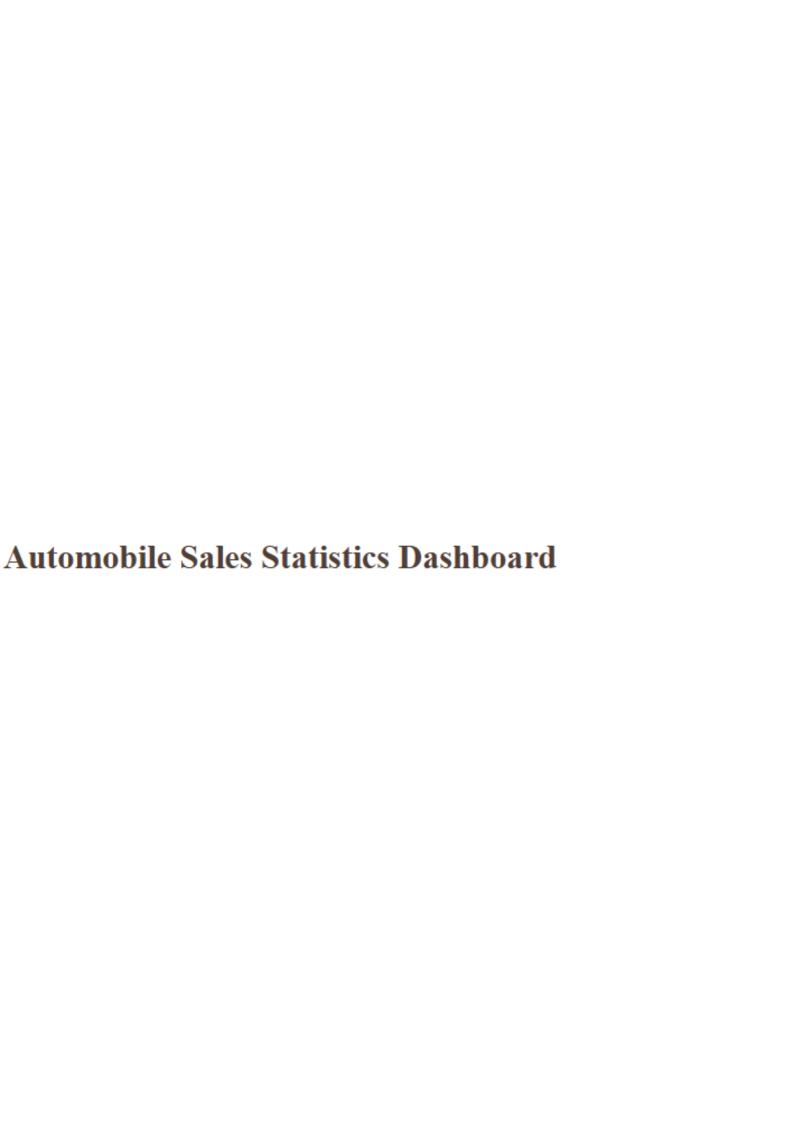


```
def update output container(selected statistics, input year):
    if selected_statistics == 'Recession Period Statistics':
        recession data = data[data['Recession'] == 1]
       # Plot 1: Yearly Average Sales
       yearly_rec = recession_data.groupby('Year')['Automobile_Sales'].mean().r
       R chart1 = dcc.Graph(figure=px.line(yearly rec, x='Year', y='Automobile
        # Plot 2: Average Sales by Vehicle Type
       average_sales = recession_data.groupby('Vehicle_Type')['Automobile_Sales
       R_chart2 = dcc.Graph(figure=px.bar(average_sales, x='Vehicle_Type', y='A
       # Plot 3: Pie chart of Ad Expenditure
        exp_rec = recession_data.groupby('Vehicle_Type')['Advertising_Expenditur
       R_chart3 = dcc.Graph(figure=px.pie(exp_rec, values='Advertising_Expendit
       # Plot 4: Effect of Unemployment Rate
       unemp_data = recession_data.groupby(['unemployment_rate', 'Vehicle_Type'
       R_chart4 = dcc.Graph(figure=px.bar(unemp_data, x='unemployment_rate', y=
                                           labels={'unemployment rate': 'Unemplo
                                           title='Effect of Unemployment Rate on
```

```
app.layout = html.Div([
    html.H1("Automobile Sales Statistics Dashboard", style={'textAlign': 'left',
    html.Div([
        html.Label("Select Statistics:"),
        dcc.Dropdown(
            id='dropdown-statistics',
            options=dropdown_options,
            value='Yearly Statistics',
            placeholder='Select report type'
    ]),
    html.Div([
        html.Label("Select Year:"),
        dcc.Dropdown(
            id='select-year',
            options=[{'label': i, 'value': i} for i in year_list],
            placeholder='Select year'
    ]),
    html.Div(id='output-container', className='chart-grid')
])
```

Automobile Sales Statistics Dashboard

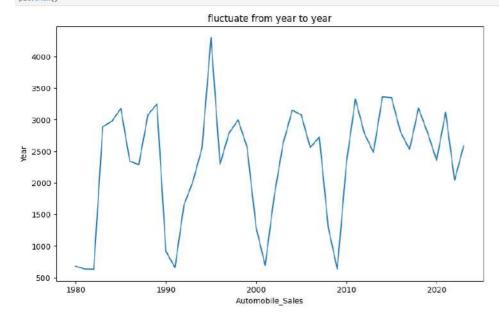




* TASK 1.1: Develop a Line chart using the functionality of pandas to show how automobile sales fluctuate from year to year

► Click here for a hint

```
df_line = df.groupby(df ['Year'])['Automobile_Sales'].mean()
plt.figure(figsize=(10,6))
df_line.plot(kind = 'line')
plt.xlabel('Automobile_Sales')
plt.ylabel('Year')
plt.title('fluctuate from year to year')
plt.show()
```



Perform a second order polynomial transform on both the training data and testing data. Create and fit a Ridge regression object using the training data, set the regularisation parameter to 0.1, and calculate the R^2 utilising the test data provided. Take a screenshot of your code and the R^2. You will need to submit it for the final project.

```
#Enter Your Code, Execute and take the Screenshot

poly = PolynomialFeatures(degree=2, include_bias=False)

X_train_poly = poly.fit_transform(X_train)

X_test_poly = poly.transform(X_test)

# Create and fit Ridge regression model

ridge_model = Ridge(alpha=0.1)

ridge_model.fit(X_train_poly, y_train)

ridge_model.score(X_train_poly, y_train)
```

16]: 0.7447768732672273

Create and fit a Ridge regression object using the training data, set the regularization parameter to 0.1, and calculate the R^2 using the test data. Take a screenshot of your code and the value of the R^2. You will need to submit it for the final project.

```
#Enter Your Code, Execute and take the Screenshot
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)

# Create and fit Ridge regression model with alpha = 0.1
ridge_model = Ridge(alpha=0.1)
ridge_model.fit(X_train, y_train)

# Predict on test data
y_pred = ridge_model.predict(X_test)

# Calculate R2 score
ridge_model.score(X_train, y_train)
```

45]: 0.6564449978833942

Use the list to create a pipeline object to predict the 'price', fit the object using the features in the list features, and calculate the R^2. Take a screenshot of your code and the value of the R^2. You will need to submit it for the final project.

```
#Enter Your Code, Execute and take the Screenshot

X = df[features]
y = df['price']

Input = [
    ('scale', StandardScaler()),
    ('polynomial', PolynomialFeatures(include_bias=False)),
    ('model', LinearRegression())
]

pipe = Pipeline(Input)

pipe.fit(X, y)
pipe.score(X, y)
```

3]: 0.7512051345272872

Fit a linear regression model to predict the 'price' using the list of features:

```
Then calculate the R^2. Take a screenshot of your code and the value of the R^2. You will need to submit it for the final project.

[35]: #Enter Your Code, Execute and take the Screenshot

X = df[features]
Y = df['price']

model = LinearRegression()
model.fit(X, Y)
```

[34]: features =["floors", "waterfront","lat" ,"bedrooms" ,"sqft_basement" ,"view" ,"bathrooms","sqft_living15","sqft_above","grade","sqft_living"]

[35]: 0.6576890354915759

Fit a linear regression model to predict the 'price' using the feature 'sqft_living' then calculate the R^2. Take a screenshot of your code and the value of the R^2. You will need to submit it for the final project.

```
[33]: #Enter Your Code, Execute and take the Screenshot
    X = df[['sqft_living']]
    Y - df['price']

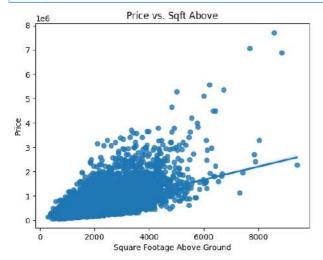
model = LinearRegression()
model.fit(X, Y)
model.score(X,Y)
```

[33]: 0.4928532179037931

Use the function regplot in the seaborn library to determine if the feature sqft_above is negatively or positively correlated with price. Take a screenshot of your code and scatterplot. You will need to submit the screenshot for the final project.

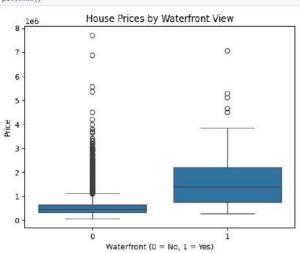
```
sns.regplot(x='sqft_obove', y='price', doto=df)

plt.title('Price vs. Sqft Above')
plt.xlabel('Square Footage Above Ground')
plt.ylobel('Price')
plt.show()
```



Use the function boxplot in the seaborn library to determine whether houses with a waterfront view or without a waterfront view have more price outliers. Take a screenshot of your code and boxplot. You will need to submit the screenshot for the final project.

```
plt.title('House Prices by Waterfront View')
plt.xlabel('Waterfront (0 = No. 1 = Yes)')
plt.ylabel('Price')
plt.show()
```



Use the method value_counts to count the number of houses with unique floor values, use the method .to_frame() to convert it to a data frame. Take a screenshot of your code and output. You will need to submit the screenshot for the final project.

```
#Enter Your Code, Execute and take the Screenshot
floor_counts = df['floors'].value_counts().to_frame()
floor_counts
```

floors 1.0 10680 2.0 8241 1.5 1910 3.0 613 2.5 161 3.5 8

df.drop(['id', 'Unnamed: 0'], axis=1, inplace=True)

Drop the columns "id" and "Unnamed: 0" from axis 1 using the method drop(), then use the method describe() to obtain a statistical summary of the data. Make sure the inplace parameter is set to True. Take a screenshot of your code and output. You will need to submit the screenshot for the final project.

	scribe()									V D +	_
	price	bedrooms	bathrooms	sqft_living	sqft_lot	floors	waterfront	view	condition	grade	sqf
count	2.161300e+04	21600.000000	21603.000000	21613.000000	2.161300e+04	21613.000000	21613.000000	21613.000000	21613.000000	21613.000000	21613
mean	5.400881e+05	3.372870	2.115736	2079.899736	1.510697e+04	1.494309	0.007542	0.234303	3.409430	7.656873	1788
std	3.671272e+05	0.926657	0.768996	918.440897	4.142051e+04	0.539989	0.086517	0.766318	0.650743	1.175459	828
min	7.500000e+04	1.000000	0.500000	290.000000	5.200000e+02	1.000000	0.000000	0.000000	1.000000	1.000000	290
25%	3.219500e+05	3.000000	1.750000	1427.000000	5.040000e+03	1.000000	0.000000	0.000000	3.000000	7.000000	1190
50%	4.500000e+05	3.000000	2.250000	1910.000000	7.618000e+03	1.500000	0.000000	0.000000	3.000000	7.000000	1560
75%	6.450000e+05	4.000000	2.500000	2550.000000	1.068800e+04	2.000000	0.000000	0.000000	4.000000	8.000000	2210
max	7.700000e+06	33.000000	8.000000	13540.000000	1.651359e+06	3.500000	1.000000	4.000000	5.000000	13.000000	9410

Display the data types of each column using the function dtypes. Take a screenshot of your code and output. You will need to submit the screenshot for the final project.

[14]: #Enter Your Code, Execute and take the Screenshot df.dtypes

[14]: Unnamed: 0 int64 int64 id object date price float64 float64 bedrooms float64 bathrooms sqft_living sqft_lot int64 floors float64 waterfront int64 int64 view condition int64 grade int64 sqft_above int64 int64 ${\sf sqft_basement}$ yr_built int64 yr_renovated int64 zipcode int64 float64 lat long float64 sqft_living15 int64 sqft_lot15 int64 dtype: object

Question 6: Plot GameStop Stock Graph

Use the make_graph function to graph the GameStop Stock Data, also provide a title for the graph. The structure to call the make_graph function is make_graph(gme_data, gme_revenue, 'GameStop'). Note the graph will only show data upto June 2021.

You just need to invoke the make_graph function with the required parameter to print the graphs. The structure to call the 'make_graph' function is 'make_graph(gme_data, gme_revenue, 'GameStop')'

make_graph(gme_data, gme_revenue, 'GameStop')

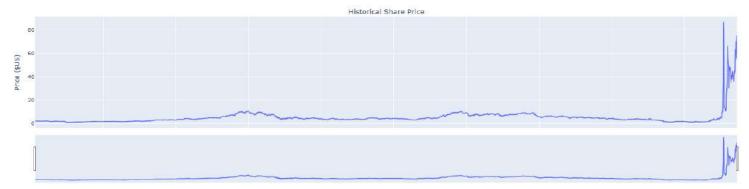
/tmp/ipykernel_781/109047474.py:5: UserWarning:

The argument 'infer_datetime_format' is deprecated and will be removed in a future version. A strict version of it is now the default, see https://pandas.pydata.org/pdeps/eee4-consistent-to-datetime-parsing.html. You can safely remove this argument.

/tmp/ipykernel_781/109047474.py:6: UserWarning:

The argument 'Infer_datetine_format' is deprecated and will be removed in a future version. A strict version of it is now the default, see https://pandas.pydata.org/pdeps/eeea-consistent-to-datetine-parsing.html. You can safely remove this argument.





Question 5: Plot Tesla Stock Graph

Use the make_graph function to graph the Tesla Stock Data, also provide a title for the graph. Note the graph will only show data upto June 2021.

▼ Hint

You just need to invoke the make_graph function with the required parameter to print the graphs. The structure to call the 'make_graph' function is 'make_graph(tesla_data, tesla_revenue, 'Tesla')'.

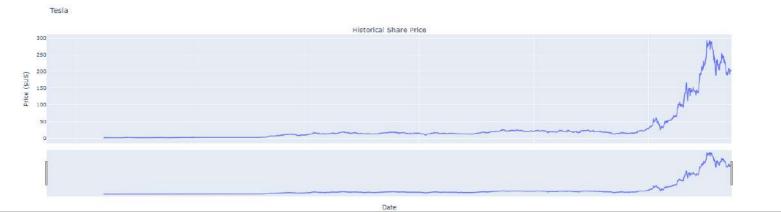
Is nake_groph(tcslo_data, tcslo_revenue, 'Tcslo')

/tmp/ipykernel_781/189047474.py:5: UserNarning:

The argument 'infer_datetime_format' is deprecated and will be removed in a future version. A strict version of it is now the default, see https://pandas.pydata.org/pdeps/0004-consistent-to-datetime-parsing.html. You can safely remove this argumen t.

/tmp/ipykernel_781/109047474.py:5: UserNarning:

The argument 'infer_datetime_format' is deprecated and will be removed in a future version. A strict version of it is now the default, see https://pandas.gydata.org/pdeps/0004-consistent-to-datetime-parsing.html. You can safely remove this argumen.



Question 3: Use yfinance to Extract Stock Data

Using the Ticker function enter the ticker symbol of the stock we want to extract data on to create a ticker object. The stock is GameStop and its ticker symbol is GNE.

|45]: gamestop = yf.Ticker("GHE")

Using the ticker object and the function in story extract stock information and save it in a dataframe named gine_data. Set the perilod parameter to "nax" so we get information for the maximum amount of time.

[45]: gme_data = gamestop.history(period="max")

Reset the index using the recet_index(inplace=True) function on the gme_data DataFrame and display the first five rows of the gme_data_dataFrame using the nead_function. Take a screenshot of the results and code from the beginning of Question 3 to the results below.

[44]: gne_data.reset_index(inplace-True)
gne_data.head()

1		Date	Open	High	Low	Close	Volume	Dividends	Stock Splits
	0	2002-02-13 00:00:00-05:00	1.620128	1.693349	1.603295	1.691666	76216000	0.0	0.0
	1	2002-02-14 00:00:00-05:00	1.712707	1.716073	1,670626	1.683250	11021600	0.0	0.0
	2	2002-02-15 00:00:00-05:00	1.683251	1.687459	7.658002	1,674834	8389600	0.0	0.0
	3	2002-02-19 00:00:00-05:00	1.666418	1.666418	1.578048	1.607505	7410400	0.0	0.0
		2022 02 20 00/00/03 05/03	1,616000	1662200	1 602006	1.662200	2002000	0.0	0.0

Question 4: Use Webscraping to Extract GME Revenue Data

Use the inequests library to download the webpage https://cf-courses-data.s3.us.doud-object-storage.appdomain.cloud/IBMDeveloper/SkillsNetwork/PV022DEN-SkillsNetwork/labs/project/stockhtml, Save the text of the response as a variable named html_data_2.

```
|25|| url = "https://cf-courses-data.s3.us.cloud-object-storage.appdomain.cloud/IBMDeveloperSkillsNetwork-PV6228EN-SkillsNetwork/labs/project/stock.html"
response = requests.get(url)
html_data = response.text
```

Parse the html data using beautiful_soup using parserile html5lib or html.parser.

```
[26]: soup - BeautifulSoup(html_data, "html.parser")
```

Using BeautifulSoup or the read_html function extract the table with GeneStop Revenue and store it into a dataframe named gene_revenue. The dataframe should have columns Date and Revenue. Make sure the commo and dollar sign is removed from the Revenue column.

Note: Use the method similar to what you did in question 2.

▶ Click here if you need help locating the table

Display the last five rows of the gne_revenue dataframe using the tail function. Take a screenshot of the results.

[30]: gmc_revenue.toil()

P31		Date	Revenue
	57	2006-01-31	\$1,567
	58	2005-10-31	\$534
	59	2005-07-31	\$416
	60	2005-04-30	\$475
	61	2005-01-31	\$709

Question 2: Use Webscraping to Extract Tesla Revenue Data

Use the requests, library to download the webpage https://df-courses-datas3.us.cloud-object-storage.appdomain.cloud//BMDeveloperSkillsNetwork-PY0220EN-SkillsNetwork/labs/project/revenue.htm Save the text of the response as a variable named html_data -

```
[18]: wrl = "https://cf.courses-data.s3.us.cloud-object-storage.appdomain.cloud/18MDeveloperSkillsNetwork-PV8228EN-SkillsNetwork/labs/project/revenue.htm" response = requests.get(url) html_data = response.text
```

Parse the html data using beautiful_soup using parserile html511b Or html.parser.

```
[11]: soup = BeautifulSoup(html_data, "html.panser")
```

Using BeautifulSoup or the read_html function extract the table with Tesia Revenue and store it into a dataframe named tesla_revenue. The dataframe should have columns Date and Revenue.

- ➤ Step-by-step instructions
- ► Click here if you need help locating the table

```
[13]: table = soup.find_all("tbody")[1]
    data = []
    for row in table.find_all("tr"):
        cols = row.find_all("td")
        if len(cols) == 2:
            date = cols[0].text.strip()
            revenue = cols[1].text.strip()
            data.append(["Date": date, "Revenue": revenue))
        tosla_revenue = pd.DataFrame(data)
```

Execute the following line to remove the comma and dollar sign from the Revenue column.

[14]: tosla_revenue("Revenue") = tosla_revenue('Revenue'),str.replace(',|\\$',"')

Execute the following lines to remove an null or empty strings in the Revenue column:

[15]: tesla_revenue.dropna(implace=True)

tesla_revenue = tesla_revenue[tesla_revenue['Revenue'] !- ""]

Display the last 5 row of the testa_revenue dataframe using the tall function, Take a screenshot of the results.

[16]: tosla_rovenue.tail()

	Date	Revenue
48	2010-09-30	\$31
49	2010-06-30	\$28
50	2010-03-31	521
52	2009-09-30	\$46
53	2009-06-20	527

Question 1: Use yfinance to Extract Stock Data

Using the Ticker function enter the ticker symbol of the stock we want to extract data on to create a ticker object. The stock is Tesla and its ticker symbol is TSLA.

[7]: tesla = yf.Ticker("TSLA")

Using the ticker object and the function history extract stock information and save it in a dataframe named tesla_data . Set the period parameter to "max" so we get information for the maximum amount of

[8]: tesla_data = tesla.history(period="max")

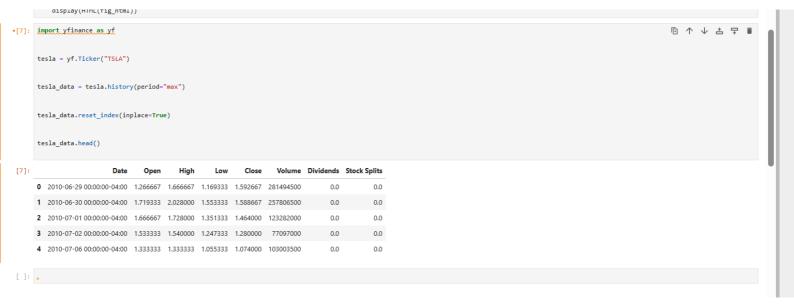
Reset the index using the reset_index(inplace=True) function on the tesla_data DataFrame and display the first five rows of the tesla_data dataframe using the head function. Take a screenshot of the results and code from the beginning of Question 1 to the results below.

[9]: tesla_data.reset_index(inplace=True)
 tesla_data.head()

9]:		Date	Open	High	Low	Close	Volume	Dividends	Stock Splits
	0	2010-06-29 00:00:00-04:00	1.266667	1.666667	1.169333	1.592667	281494500	0.0	0.0
	1	2010-06-30 00:00:00-04:00	1.719333	2.028000	1.553333	1.588667	257806500	0.0	0.0
	2	2010-07-01 00:00:00-04:00	1.666667	1.728000	1.351333	1.464000	123282000	0.0	0.0
	3	2010-07-02 00:00:00-04:00	1.533333	1.540000	1.247333	1.280000	77097000	0.0	0.0
	4	2010-07-06 00:00:00-04:00	1.333333	1.333333	1.055333	1.074000	103003500	0.0	0.0

[38]:

	Date	Revenue
48	2010-09-30	\$31
49	2010-06-30	\$28
50	2010-03-31	\$21
52	2009-09-30	\$46
53	2009-06-30	\$27

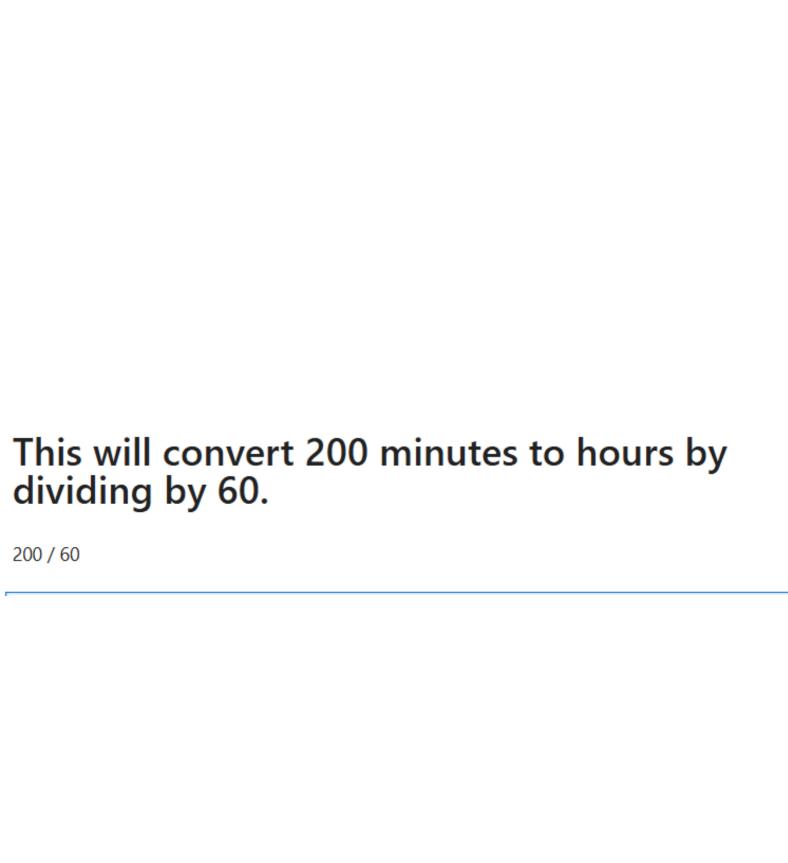


Author

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Objectives:

- List popular languages for Data Science
- List commonly used libraries in Data Science
- Create tables using markdown
- Perform arithmetic operations in code cells
- Share Jupyter Notebooks through GitHub



This is a simple arithmetic expression to multiply then add integers.

(3*4)+5

Below are a few examples of evaluating arithmetic express Python.	ions in

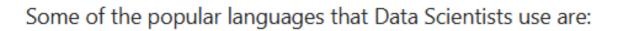
Data Science Tools

Jupyter Notebooks

RStudio

Apache Zeppelin

Some of the commonly used libraries used by Data Scientists include:
1. Pandas
NumPy Matplotlib
4. Scikit-learn



- 1. Python
- 2. R
- 3. SQL
- 4. Julia



