# In [1]: import numpy as np import pandas as pd

## In [2]: pip install lightgbm

Requirement already satisfied: lightgbm in c:\users\archa\anaconda3\lib\site-packages (4.1.0)

Requirement already satisfied: numpy in c:\users\archa\anaconda3\lib\site-packages (from lightgbm) (1.23.5)

Requirement already satisfied: scipy in c:\users\archa\anaconda3\lib\site-packages (from lightgbm) (1.10.0)

Note: you may need to restart the kernel to use updated packages.

## In [3]: pip install catboost

Requirement already satisfied: catboost in c:\users\archa\anaconda3\lib\si te-packages (1.2.2)

Requirement already satisfied: numpy>=1.16.0 in c:\users\archa\anaconda3\l
ib\site-packages (from catboost) (1.23.5)

Requirement already satisfied: plotly in c:\users\archa\anaconda3\lib\site -packages (from catboost) (5.9.0)

Requirement already satisfied: six in c:\users\archa\anaconda3\lib\site-pa ckages (from catboost) (1.16.0)

Requirement already satisfied: graphviz in c:\users\archa\anaconda3\lib\si te-packages (from catboost) (0.20.1)

Requirement already satisfied: matplotlib in c:\users\archa\anaconda3\lib \site-packages (from catboost) (3.7.0)

Requirement already satisfied: pandas>=0.24 in c:\users\archa\anaconda3\li b\site-packages (from catboost) (1.5.3)

Requirement already satisfied: scipy in c:\users\archa\anaconda3\lib\site-packages (from catboost) (1.10.0)

Requirement already satisfied: python-dateutil>=2.8.1 in c:\users\archa\an aconda3\lib\site-packages (from pandas>=0.24->catboost) (2.8.2)

Requirement already satisfied: pytz>=2020.1 in c:\users\archa\anaconda3\lib\site-packages (from pandas>=0.24->catboost) (2022.7)

Requirement already satisfied: pillow>=6.2.0 in c:\users\archa\anaconda3\l ib\site-packages (from matplotlib->catboost) (9.4.0)

Requirement already satisfied: contourpy>=1.0.1 in c:\users\archa\anaconda 3\lib\site-packages (from matplotlib->catboost) (1.0.5)

Requirement already satisfied: pyparsing>=2.3.1 in c:\users\archa\anaconda 3\lib\site-packages (from matplotlib->catboost) (3.0.9)

Requirement already satisfied: fonttools>=4.22.0 in c:\users\archa\anacond a3\lib\site-packages (from matplotlib->catboost) (4.25.0)

Requirement already satisfied: kiwisolver>=1.0.1 in c:\users\archa\anacond a3\lib\site-packages (from matplotlib->catboost) (1.4.4)

Requirement already satisfied: packaging>=20.0 in c:\users\archa\anaconda3 \lib\site-packages (from matplotlib->catboost) (22.0)

Requirement already satisfied: cycler>=0.10 in c:\users\archa\anaconda3\li b\site-packages (from matplotlib->catboost) (0.11.0)

Requirement already satisfied: tenacity>=6.2.0 in c:\users\archa\anaconda3 \lib\site-packages (from plotly->catboost) (8.0.1)

Note: you may need to restart the kernel to use updated packages.

```
import numpy as np
In [4]:
        import pandas as pd
        from sklearn.model_selection import train_test_split
        from sklearn.linear model import LinearRegression
        from sklearn.preprocessing import StandardScaler
        from sklearn.preprocessing import LabelEncoder
        from sklearn.linear_model import LinearRegression, Ridge, Lasso
        from sklearn.neighbors import KNeighborsRegressor
        from sklearn.neural_network import MLPRegressor
        from sklearn.svm import LinearSVR, SVR
        from sklearn.tree import DecisionTreeRegressor
        from sklearn.ensemble import RandomForestRegressor, GradientBoostingRegress
        from xgboost import XGBRegressor
        from lightgbm import LGBMRegressor
        from catboost import CatBoostRegressor
        import seaborn as sns
        from sklearn. linear_model import Lasso
        import warnings
        warnings.filterwarnings(action='ignore')
```

# In [5]: import pandas as pd data=pd.read\_csv('C:/Users/archa/Downloads/dataset Tomato (Price) new.csv') data

#### Out[5]:

_		Date	Incoming Quintel	Lowest Price	Highest Price	Avg_Price	Price	Temperature in CC	Month	Tomat Qualit
	0	1/1/2021	106.0	400.0	600.0	500.0	5300.0	17.8	Jan	Goo
	1	1/2/2021	130.0	500.0	700.0	600.0	7800.0	18.7	Jan	Goo
	2	1/3/2021	152.0	400.0	600.0	400.0	60800.0	19.7	Jan	Goo
	3	1/4/2021	104.0	300.0	640.0	475.0	49400.0	20.2	Jan	Goo
	4	1/5/2021	95.0	300.0	700.0	400.0	38000.0	20.8	Jan	Averag
	816	10/2/2023	115.0	500.0	1000.0	750.0	65500.0	29.5	Oct	Goo
	817	10/3/2023	126.0	500.0	1000.0	750.0	94500.0	28.0	Oct	Goo
	818	10/4/2023	119.0	600.0	900.0	750.0	89200.0	28.0	Oct	Goo
	819	10/5/2023	193.0	300.0	900.0	600.0	115400.0	27.0	Oct	Goo
	820	10/7/2023	216.0	250.0	750.0	500.0	10800.0	26.0	Oct	Goo

821 rows × 10 columns

**◆** 

# In [6]: data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 821 entries, 0 to 820
Data columns (total 10 columns):

#	Column	Non-Null Count	Dtype
0	Date	821 non-null	object
1	Incoming Quintel	820 non-null	float64
2	Lowest Price	820 non-null	float64
3	Highest Price	820 non-null	float64
4	Avg_Price	820 non-null	float64
5	Price	820 non-null	float64
6	Temperature in CC	821 non-null	float64
7	Month	821 non-null	object
8	Tomato Quality	821 non-null	object
9	Season	821 non-null	object

dtypes: float64(6), object(4)

memory usage: 64.3+ KB

```
data.replace({'Season':{'Cloudy':1,'Sunny':2,'Rainy':3}},inplace=True)
data.replace({'Tomato Quality':{'Good': 1,'Bad': 2,'Average': 3}},inplace=T
In [7]:
           data.replace({'Month':{
                      'Jan': 1,
                      'Feb': 2,
                      'Mar': 3,
                      'Apr': 4,
                      'May': 5,
                      'Jun': 6,
                      'Jul': 7,
                      'Aug': 8,
                      'Sep': 9,
                      'Oct': 10,
                      'Nov': 11,
                      'Dec': 12}},inplace=True)
           data
```

# Out[7]:

	Date	Incoming Quintel	Lowest Price	Highest Price	Avg_Price	Price	Temperature in CC	Month	Tomat Qualit
0	1/1/2021	106.0	400.0	600.0	500.0	5300.0	17.8	1	
1	1/2/2021	130.0	500.0	700.0	600.0	7800.0	18.7	1	
2	1/3/2021	152.0	400.0	600.0	400.0	60800.0	19.7	1	
3	1/4/2021	104.0	300.0	640.0	475.0	49400.0	20.2	1	
4	1/5/2021	95.0	300.0	700.0	400.0	38000.0	20.8	1	
816	10/2/2023	115.0	500.0	1000.0	750.0	65500.0	29.5	10	
817	10/3/2023	126.0	500.0	1000.0	750.0	94500.0	28.0	10	
818	10/4/2023	119.0	600.0	900.0	750.0	89200.0	28.0	10	
819	10/5/2023	193.0	300.0	900.0	600.0	115400.0	27.0	10	
820	10/7/2023	216.0	250.0	750.0	500.0	10800.0	26.0	10	

821 rows × 10 columns

In [8]: data.replace({'Tomato Quality':{'Good': 1,'Bad': 2,'Average': 3}},inplace=T
data

Out[8]:

		Date	Incoming Quintel	Lowest Price	Highest Price	Avg_Price	Price	Temperature in CC	Month	Tomat Qualit
· ·	0	1/1/2021	106.0	400.0	600.0	500.0	5300.0	17.8	1	
	1	1/2/2021	130.0	500.0	700.0	600.0	7800.0	18.7	1	
	2	1/3/2021	152.0	400.0	600.0	400.0	60800.0	19.7	1	
	3	1/4/2021	104.0	300.0	640.0	475.0	49400.0	20.2	1	
	4	1/5/2021	95.0	300.0	700.0	400.0	38000.0	20.8	1	
8	16	10/2/2023	115.0	500.0	1000.0	750.0	65500.0	29.5	10	
8	17	10/3/2023	126.0	500.0	1000.0	750.0	94500.0	28.0	10	
8	18	10/4/2023	119.0	600.0	900.0	750.0	89200.0	28.0	10	
8	19	10/5/2023	193.0	300.0	900.0	600.0	115400.0	27.0	10	
8	20	10/7/2023	216.0	250.0	750.0	500.0	10800.0	26.0	10	

821 rows × 10 columns

In [9]: index\_to\_delete = 236

data = data.drop(index\_to\_delete)

In [10]: data

Out[10]:

	Date	Incoming Quintel	Lowest Price	Highest Price	Avg_Price	Price	Temperature in CC	Month	Tomat Qualit
0	1/1/2021	106.0	400.0	600.0	500.0	5300.0	17.8	1	
1	1/2/2021	130.0	500.0	700.0	600.0	7800.0	18.7	1	
2	1/3/2021	152.0	400.0	600.0	400.0	60800.0	19.7	1	
3	1/4/2021	104.0	300.0	640.0	475.0	49400.0	20.2	1	
4	1/5/2021	95.0	300.0	700.0	400.0	38000.0	20.8	1	
816	10/2/2023	115.0	500.0	1000.0	750.0	65500.0	29.5	10	
817	10/3/2023	126.0	500.0	1000.0	750.0	94500.0	28.0	10	
818	10/4/2023	119.0	600.0	900.0	750.0	89200.0	28.0	10	
819	10/5/2023	193.0	300.0	900.0	600.0	115400.0	27.0	10	
820	10/7/2023	216.0	250.0	750.0	500.0	10800.0	26.0	10	

820 rows × 10 columns

4

```
In [11]:
         print(data.columns)
         data = data.dropna()
         Index(['Date', 'Incoming Quintel', 'Lowest Price', 'Highest Price',
                 'Avg_Price', 'Price', 'Temperature in CC', 'Month', 'Tomato Qualit
         у',
                 'Season'],
                dtype='object')
In [12]: import pandas as pd
         y = data['Price']
         X = data.drop(['Price', "Date"], axis=1)
         # #for demand
         z = data['Avg_Price']
         w = data.drop(['Lowest Price', 'Highest Price', 'Date', 'Incoming Quintel', 'Av
In [13]: y
Out[13]: 0
                   5300.0
         1
                   7800.0
         2
                  60800.0
         3
                  49400.0
                  38000.0
         816
                  65500.0
         817
                  94500.0
         818
                 89200.0
         819
                 115400.0
         820
                  10800.0
         Name: Price, Length: 819, dtype: float64
```

In [14]:

#### Out[14]:

	Incoming Quintel	Lowest Price	Highest Price	Avg_Price	Temperature in CC	Month	Tomato Quality	Season
0	106.0	400.0	600.0	500.0	17.8	1	1	1
1	130.0	500.0	700.0	600.0	18.7	1	1	1
2	152.0	400.0	600.0	400.0	19.7	1	1	1
3	104.0	300.0	640.0	475.0	20.2	1	1	1
4	95.0	300.0	700.0	400.0	20.8	1	3	1
816	115.0	500.0	1000.0	750.0	29.5	10	1	1
817	126.0	500.0	1000.0	750.0	28.0	10	1	1
818	119.0	600.0	900.0	750.0	28.0	10	1	1
819	193.0	300.0	900.0	600.0	27.0	10	1	1
820	216.0	250.0	750.0	500.0	26.0	10	1	1

819 rows × 8 columns

```
In [15]:
         print(data.columns)
         Index(['Date', 'Incoming Quintel', 'Lowest Price', 'Highest Price',
                 'Avg_Price', 'Price', 'Temperature in CC', 'Month', 'Tomato Qualit
         у',
                 'Season'],
               dtype='object')
In [16]: X_train,X_test,y_train,y_test=train_test_split(X,y,train_size=0.8,shuffle=T
         X_train,X_test,y_train,y_test=train_test_split(X,y,train_size=0.8,shuffle=T
         w_train,w_test,z_train,z_test=train_test_split(w,z,train_size=0.8,shuffle=T
         w_train,w_test,z_train,z_test=train_test_split(w,z,train_size=0.8,shuffle=T
In [17]: X_test.shape
Out[17]: (164, 8)
In [18]: # Check for missing values
         print(X train.isnull().sum())
         print(y_train.isnull().sum())
         # Check data types
         print(X_train.dtypes)
         print(y_train.dtypes)
         Incoming Quintel
                               0
         Lowest Price
                               0
         Highest Price
                               0
         Avg_Price
         Temperature in CC
                               0
                               0
         Month
                               0
         Tomato Quality
         Season
                               0
         dtype: int64
         Incoming Quintel
                               float64
         Lowest Price
                               float64
         Highest Price
                               float64
         Avg Price
                               float64
         Temperature in CC
                               float64
         Month
                                 int64
         Tomato Quality
                                 int64
         Season
                                 int64
         dtype: object
         float64
In [19]:
         np.random.seed(0)
         X=np.random.rand(821,10)
         y=np.random.rand(821,1)
In [20]: |print("X_train shape:", X_train.shape)
         print("y_train shape:", y_train.shape)
         X train shape: (655, 8)
         y_train shape: (655,)
```

```
print("NaN values in X_train:", X_train.isnull().sum().sum())
In [21]:
         print("NaN values in y_train:", y_train.isnull().sum().sum())
         NaN values in X_train: 0
         NaN values in y train: 0
In [22]: print("Data types in X_train:", X_train.dtypes.unique())
         print("Data type in y_train:", y_train.dtypes)
         Data types in X_train: [dtype('float64') dtype('int64')]
         Data type in y_train: float64
In [23]: X_train = X_train.values
         y_train = y_train.values
In [24]: # Create a linear regression model
         model = LinearRegression()
         # Train the model
         model.fit(X_train, y_train)
Out[24]: LinearRegression()
```

In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

```
In [54]: # model.predict(X_test, y_test)
In [25]: models = {
                                   Linear Regression": LinearRegression(),
             " Linear Regression (L2 Regularization)": Ridge(),
             " Linear Regression (L1 Regularization)": Lasso(),
                                 K-Nearest Neighbors": KNeighborsRegressor(),
                                      Neural Network": MLPRegressor(),
             "Support Vector Machine (Linear Kernel)": LinearSVR(),
                 Support Vector Machine (RBF Kernel)": SVR(),
                                       Decision Tree": DecisionTreeRegressor(),
                                       Random Forest": RandomForestRegressor(),
                                   Gradient Boosting": GradientBoostingRegressor(),
                                             XGBoost": XGBRegressor(),
                                             LightGBM": LGBMRegressor(),
                                            CatBoost": CatBoostRegressor(verbose=0)
         }
```

```
In [26]:
         for name, model in models.items():
             model.fit(X_train, y_train)
             print(name + " trained.")
         for name, model in models.items():
             print(name + " R^2 Score: {:.5f}".format(model.score(X_test, y_test)))
                               Linear Regression trained.
          Linear Regression (L2 Regularization) trained.
          Linear Regression (L1 Regularization) trained.
                            K-Nearest Neighbors trained.
                                  Neural Network trained.
         Support Vector Machine (Linear Kernel) trained.
            Support Vector Machine (RBF Kernel) trained.
                                  Decision Tree trained.
                                   Random Forest trained.
                               Gradient Boosting trained.
                                         XGBoost trained.
         [LightGBM] [Info] Auto-choosing col-wise multi-threading, the overhead
         of testing was 0.000065 seconds.
         You can set `force_col_wise=true` to remove the overhead.
         [LightGBM] [Info] Total Bins 346
         [LightGBM] [Info] Number of data points in the train set: 655, number o
         f used features: 8
         [LightGBM] [Info] Start training from score 141367.983206
         [LightGBM] [Warning] No further splits with positive gain, best gain: -
In [43]: #model selection linear regression
         price_reg_model= CatBoostRegressor(verbose=0)
         price_reg_model.fit(X_train,y_train)
```

Out[43]: <catboost.core.CatBoostRegressor at 0x12a0d21f820>

```
In [48]: #visualize the actual prices and Predicted prices
    from matplotlib import pyplot as plt
    #plt.scatter(X_train,y_train, color = 'blue', label = 'Actual')
    plt.scatter(y_train,price_data_pred)
    plt.xlabel("Actual Price")
    plt.ylabel("Predicted Price")
    plt.title("Actual Price Vs Predicted Price")
    plt.legend()
    plt.grid(True)
    plt.show()
```

No artists with labels found to put in legend. Note that artists whose label start with an underscore are ignored when legend() is called with no a rgument.



In [46]: data

Out[46]:

	Date	Incoming Quintel	Lowest Price	Highest Price	Avg_Price	Price	Temperature in CC	Month	Tomat Qualit
0	1/1/2021	106.0	400.0	600.0	500.0	5300.0	17.8	1	
1	1/2/2021	130.0	500.0	700.0	600.0	7800.0	18.7	1	
2	1/3/2021	152.0	400.0	600.0	400.0	60800.0	19.7	1	
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816	10/2/2023	115.0	500.0	1000.0	750.0	65500.0	29.5	10	
817	10/3/2023	126.0	500.0	1000.0	750.0	94500.0	28.0	10	
818	10/4/2023	119.0	600.0	900.0	750.0	89200.0	28.0	10	
819	10/5/2023	193.0	300.0	900.0	600.0	115400.0	27.0	10	
820	10/7/2023	216.0	250.0	750.0	500.0	10800.0	26.0	10	

819 rows × 10 columns

In [ ]: