

# untitled11-1

May 6, 2024

```
[1]: import pandas as pd
df= pd.read_csv('/content/archive ass.zip')
print(df)
```

	Date	Open	High	Low	Close \
0	2010-06-29	3.800000	5.000000	3.508000	4.778000
1	2010-06-30	5.158000	6.084000	4.660000	4.766000
2	2010-07-01	5.000000	5.184000	4.054000	4.392000
3	2010-07-02	4.600000	4.620000	3.742000	3.840000
4	2010-07-06	4.000000	4.000000	3.166000	3.222000
...	...	...	...	...	...
2951	2022-03-18	874.489990	907.849976	867.390015	905.390015
2952	2022-03-21	914.979980	942.849976	907.090027	921.159973
2953	2022-03-22	930.000000	997.859985	921.750000	993.979980
2954	2022-03-23	979.940002	1040.699951	976.400024	999.109985
2955	2022-03-24	1009.729980	1024.489990	988.799988	1013.919983

	Adj Close	Volume
0	4.778000	93831500
1	4.766000	85935500
2	4.392000	41094000
3	3.840000	25699000
4	3.222000	34334500
...	...	...
2951	905.390015	33408500
2952	921.159973	27327200
2953	993.979980	35289500
2954	999.109985	40225400
2955	1013.919983	22901900

[2956 rows x 7 columns]

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[5]: from sklearn.model_selection import train_test_split
from sklearn.metrics import mean_absolute_error ,mean_squared_error,
median_absolute_error,confusion_matrix,accuracy_score,r2_score
from sklearn.linear_model import LinearRegression ,ARDRegression
from sklearn.neighbors import KNeighborsRegressor
from sklearn.ensemble import RandomForestRegressor
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from sklearn.linear_model import Ridge
from sklearn.svm import SVR
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[12]: X = df.drop(columns="Close")
y = df["Close"]
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.25,
↳random_state=42)
print("X Train : ", X_train.shape)
print("X Test  : ", X_test.shape)
print("Y Train : ", y_train.shape)
print("Y Test  : ", y_test.shape)
```

```
X Train : (2217, 6)
X Test  : (739, 6)
Y Train : (2217,)
Y Test  : (739,)
```

```
[28]: from sklearn.model_selection import train_test_split
from sklearn.metrics import mean_absolute_error, mean_squared_error,
↳median_absolute_error, confusion_matrix, accuracy_score, r2_score
from sklearn.linear_model import LinearRegression, ARDRegression
from sklearn.neighbors import KNeighborsRegressor
from sklearn.ensemble import RandomForestRegressor
from sklearn.linear_model import Ridge
from sklearn.svm import SVR
X = df[['Open', 'High', 'Low', 'Adj Close', 'Volume']]
y = df['Close']
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
↳random_state=0)

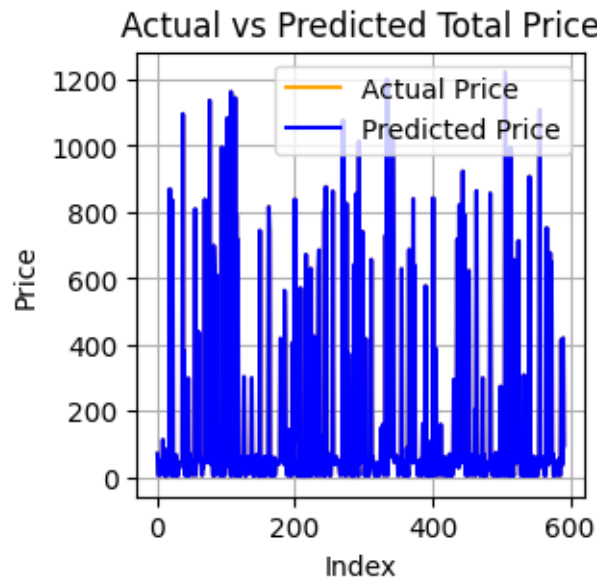
models = [
    LinearRegression(),
    RandomForestRegressor(),
    KNeighborsRegressor(n_neighbors=5),
    Ridge(alpha=1.0, random_state=42),
    SVR()
]
for model in models:
    model.fit(X_train, y_train)
    y_pred = model.predict(X_test)
    mse = mean_squared_error(y_test, y_pred)
    mae = mean_absolute_error(y_test, y_pred)
    r2 = r2_score(y_test, y_pred)
    mean_actual = y_test.mean()
    mse_percentage = (mse / mean_actual) * 100
    mape = mean_absolute_error(y_test, y_pred) / mean_actual * 100
    rmse=np.sqrt(mean_squared_error(y_test, y_pred))
```

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print(model,f"R-squared: {r2}")
print(model,f"Root Mean Squared Error (RMSE): {rmse}")
print(model,f"Mean Squared Error: {mse}")
print(model,f"Mean Absolute Error (MAE): {mae}")
print(model,f"Mean Absolute Error Percentage (MAPE): {mape:.2f}%")
print(model,f"Mean Squared Error Percentage: {mse_percentage:.2f}%")
print(model,f"R-squared: {r2}")
import matplotlib.pyplot as plt
plt.figure(figsize=(3, 3))
plt.plot(y_test.values, color='orange', label='Actual Price')
plt.plot(y_pred, color='blue', label='Predicted Price')
plt.xlabel('Index')
plt.ylabel('Price')
plt.title('Actual vs Predicted Total Price')
plt.legend()
plt.grid(True)
plt.show()

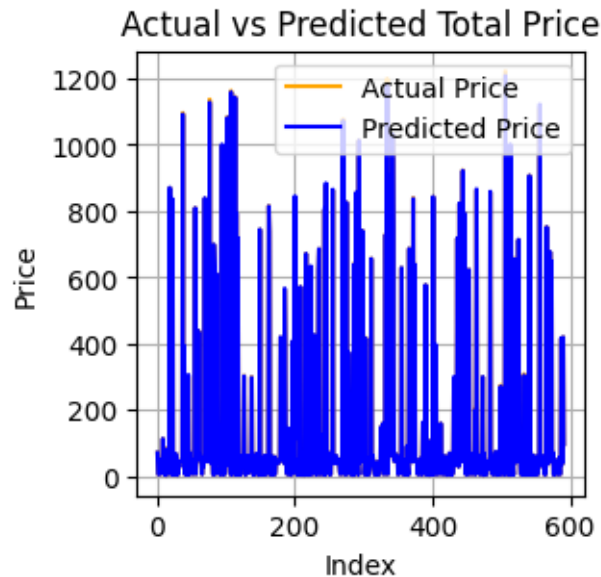
```

LinearRegression() R-squared: 1.0  
 LinearRegression() Root Mean Squared Error (RMSE): 1.9722369296243927e-12  
 LinearRegression() Mean Squared Error: 3.889718506574252e-24  
 LinearRegression() Mean Absolute Error (MAE): 1.2785860958574846e-12  
 LinearRegression() Mean Absolute Error Percentage (MAPE): 0.00%  
 LinearRegression() Mean Squared Error Percentage: 0.00%  
 LinearRegression() R-squared: 1.0

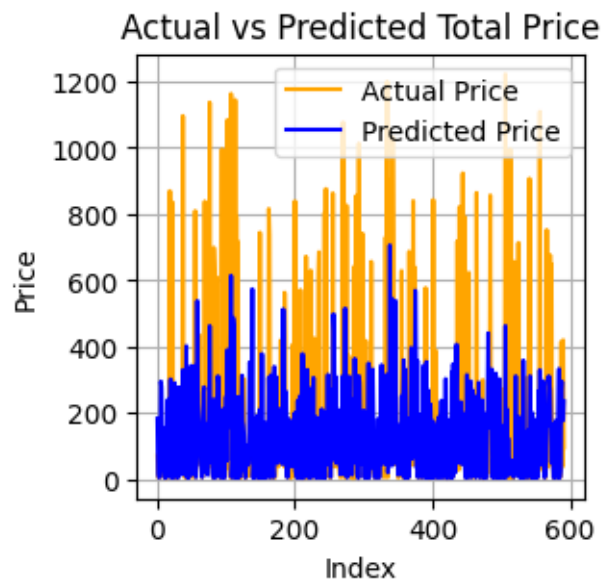


RandomForestRegressor() R-squared: 0.9999589567711437  
 RandomForestRegressor() Root Mean Squared Error (RMSE): 1.6771257192600226

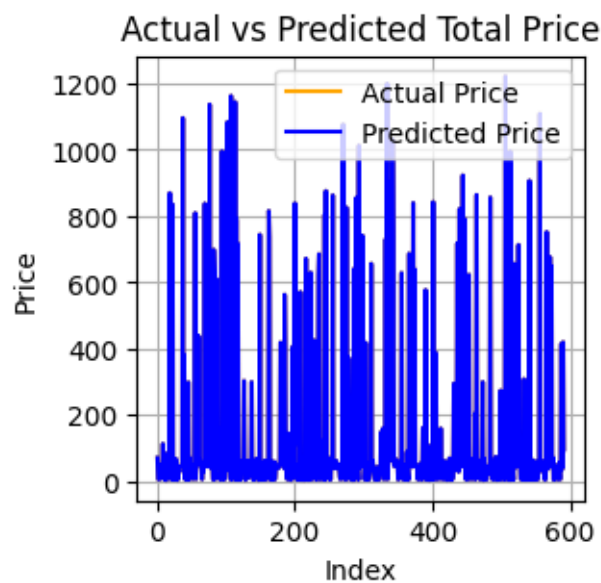
RandomForestRegressor() Mean Squared Error: 2.812750678203448  
RandomForestRegressor() Mean Absolute Error (MAE): 0.4647565066047183  
RandomForestRegressor() Mean Absolute Error Percentage (MAPE): 0.32%  
RandomForestRegressor() Mean Squared Error Percentage: 1.91%  
RandomForestRegressor() R-squared: 0.9999589567711437



KNeighborsRegressor() R-squared: 0.022284108680276082  
KNeighborsRegressor() Root Mean Squared Error (RMSE): 258.8518042462109  
KNeighborsRegressor() Mean Squared Error: 67004.25656151869  
KNeighborsRegressor() Mean Absolute Error (MAE): 154.95178142263515  
KNeighborsRegressor() Mean Absolute Error Percentage (MAPE): 105.45%  
KNeighborsRegressor() Mean Squared Error Percentage: 45599.33%  
KNeighborsRegressor() R-squared: 0.022284108680276082



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Ridge(random_state=42) R-squared: 0.999999999998596
Ridge(random_state=42) Root Mean Squared Error (RMSE): 9.810192856505214e-05
Ridge(random_state=42) Mean Squared Error: 9.623988388182592e-09
Ridge(random_state=42) Mean Absolute Error (MAE): 3.811183654131503e-05
Ridge(random_state=42) Mean Absolute Error Percentage (MAPE): 0.00%
Ridge(random_state=42) Mean Squared Error Percentage: 0.00%
Ridge(random_state=42) R-squared: 0.999999999998596
```



SVR() R-squared: -0.12378511839715034  
SVR() Root Mean Squared Error (RMSE): 277.51502272572236  
SVR() Mean Squared Error: 77014.5878384582  
SVR() Mean Absolute Error (MAE): 117.39413582086844  
SVR() Mean Absolute Error Percentage (MAPE): 79.89%  
SVR() Mean Squared Error Percentage: 52411.80%  
SVR() R-squared: -0.12378511839715034

