

```
!pip install yfinance seaborn scikit-learn matplotlib
```

```

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```

```

import yfinance as yf
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt

```

```

from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_absolute_error, mean_squared_error, r2_score
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import MinMaxScaler

```

```
plt.style.use('seaborn-v08-whitegrid')
```



```
price, volume = session.get_data(
```

```
df = yf.download("AAPL", start="2018-01-01", end="2023-12-31")
df.head()
```



YF.download() has changed argument auto\_adjust default to True

[\*\*\*\*\*100%\*\*\*\*\*] 1 of 1 completed

Price	Close	High	Low	Open	Volume	
Ticker	AAPL	AAPL	AAPL	AAPL	AAPL	
Date						
2018-01-02	40.426819	40.436208	39.722764	39.933983	102223600	
2018-01-03	40.419773	40.964244	40.356411	40.490179	118071600	
2018-01-04	40.607533	40.710794	40.384583	40.492536	89738400	
2018-01-05	41.069866	41.156698	40.612231	40.703758	94640000	
2018-01-08	40.917316	41.213018	40.818746	40.917316	82271200	

Next steps:

[Generate code with df](#)
[View recommended plots](#)
[New interactive sheet](#)

```
# baresi maghadir gom shode
print(df.isnull().sum())

# hazf radif hayi k maghadire gom shode darand
df.dropna(inplace=True)

# tabdil tarikh be datetime index
df.index = pd.to_datetime(df.index)
```



```
Price    Ticker
Close    AAPL      0
High     AAPL      0
Low      AAPL      0
Open     AAPL      0
Volume   AAPL      0
dtype: int64
```

```
scaler = MinMaxScaler()
scaled_data = scaler.fit_transform(df[['Open', 'High', 'Low', 'Close', 'Volume']])
scaled_df = pd.DataFrame(scaled_data, columns=['Open', 'High', 'Low', 'Close', 'Volume'],
```

```
# 1. ravand gheimat payani
df['Close'].plot(figsize=(12, 4), title='ravand gheimat payani')
```

```

df['Close'].plot(figsize=(12,4), title='ravand gheimat payani')

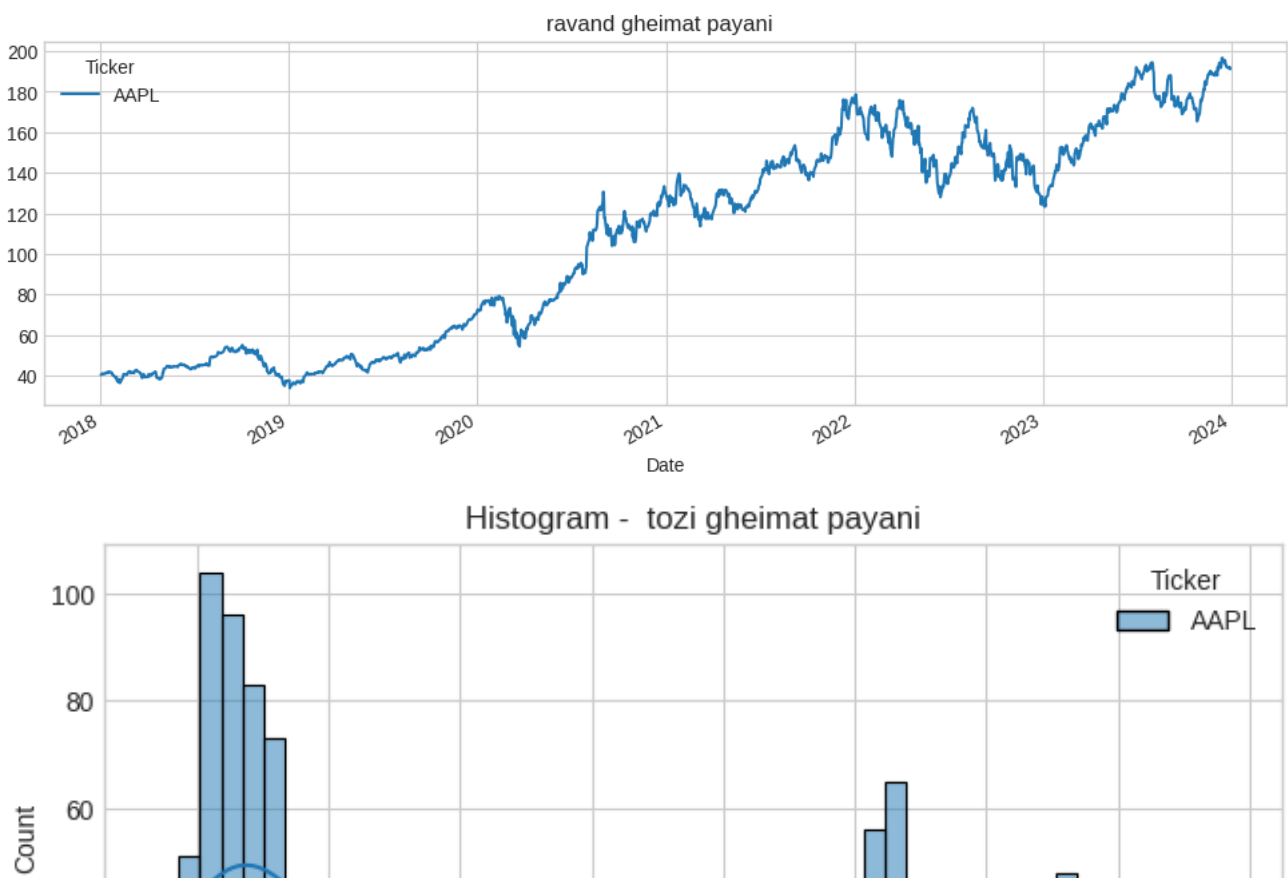
# 2. tozi gheimat payani
plt.figure(figsize=(8,4))
sns.histplot(df['Close'], bins=50, kde=True)
plt.title("Histogram - tozi gheimat payani")
plt.show()

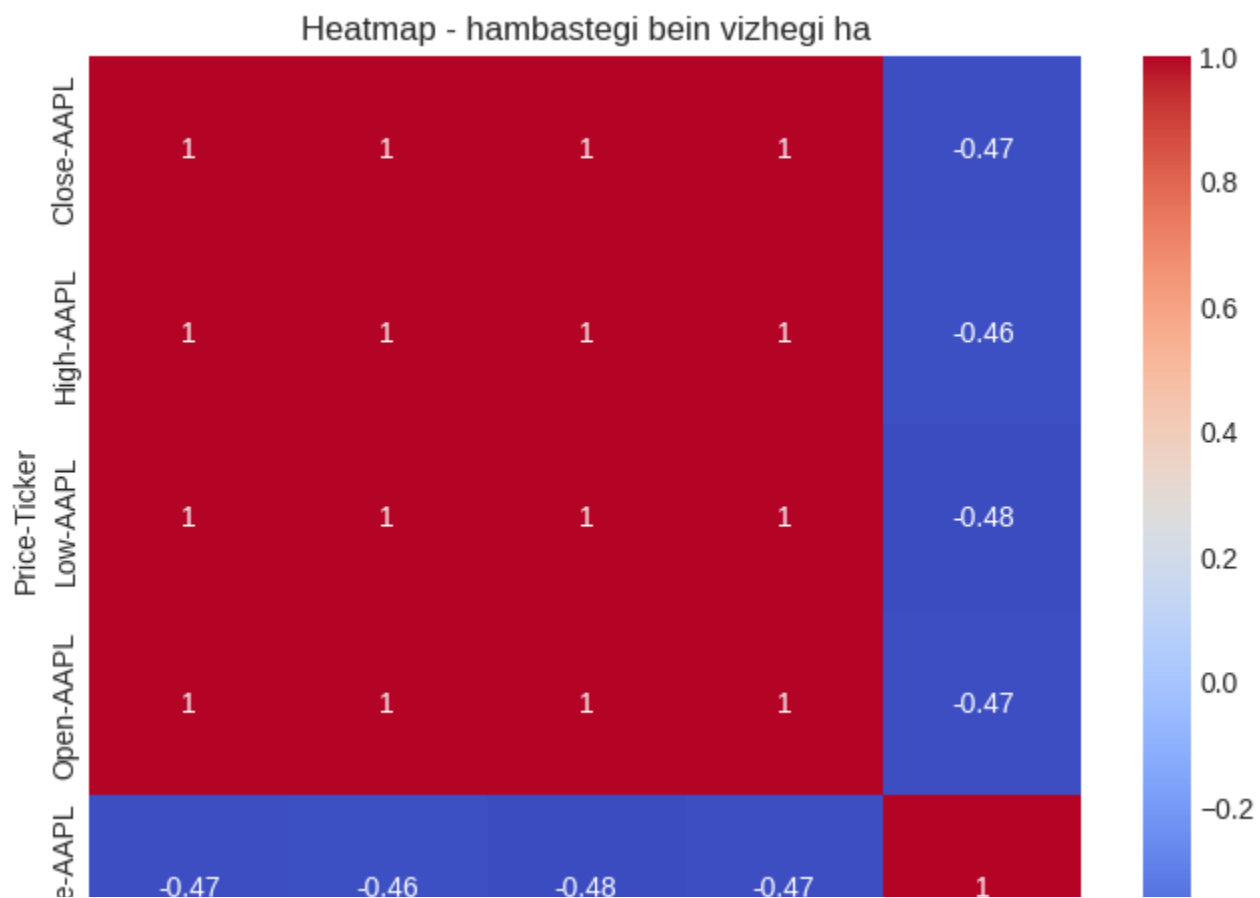
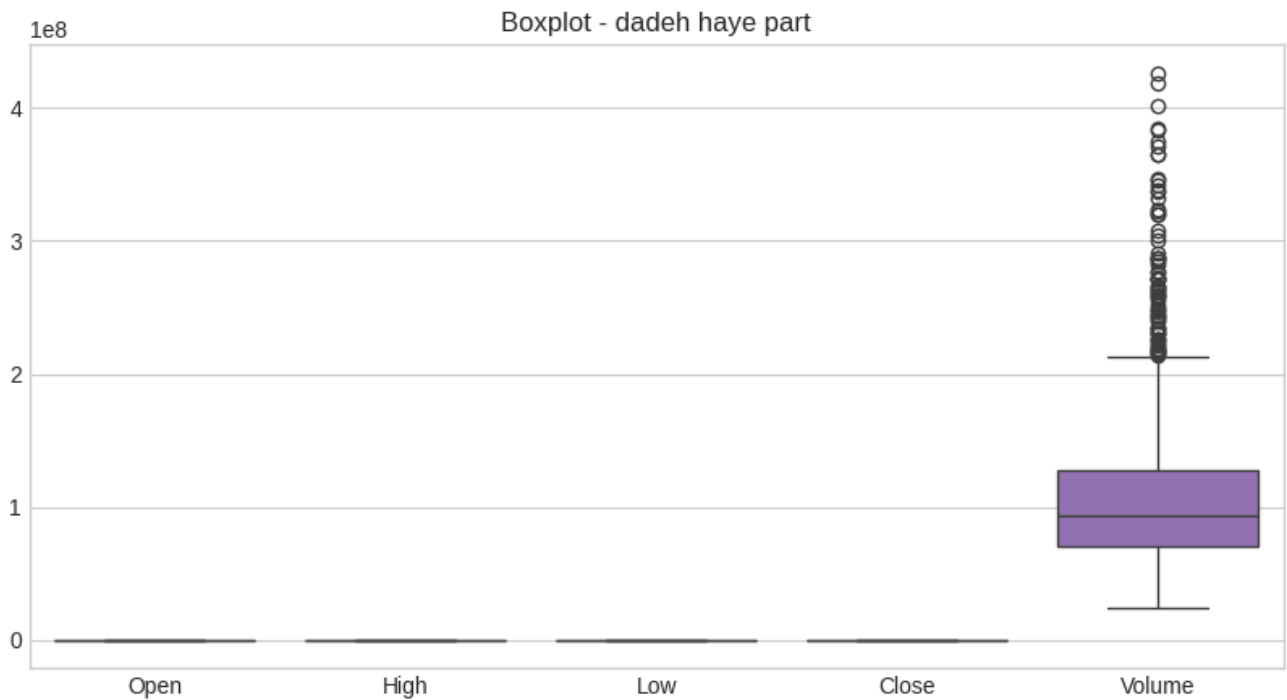
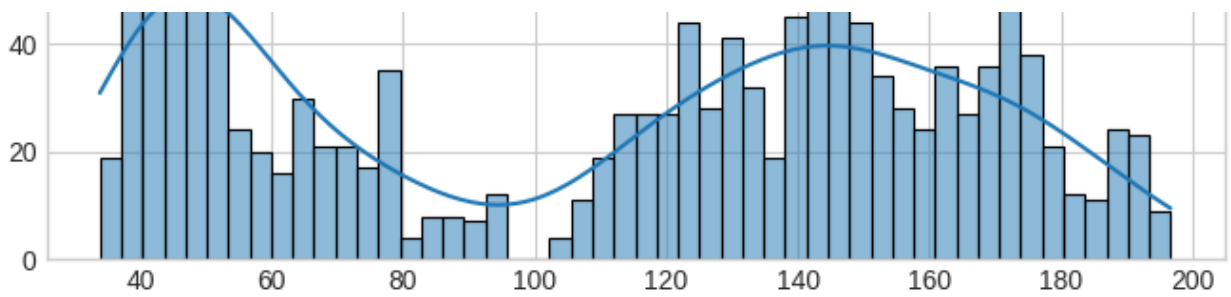
# 3. Boxplot baraye outliers
plt.figure(figsize=(10,5))
sns.boxplot(data=df[['Open', 'High', 'Low', 'Close', 'Volume']])
plt.title("Boxplot - dadeh haye part")
plt.show()

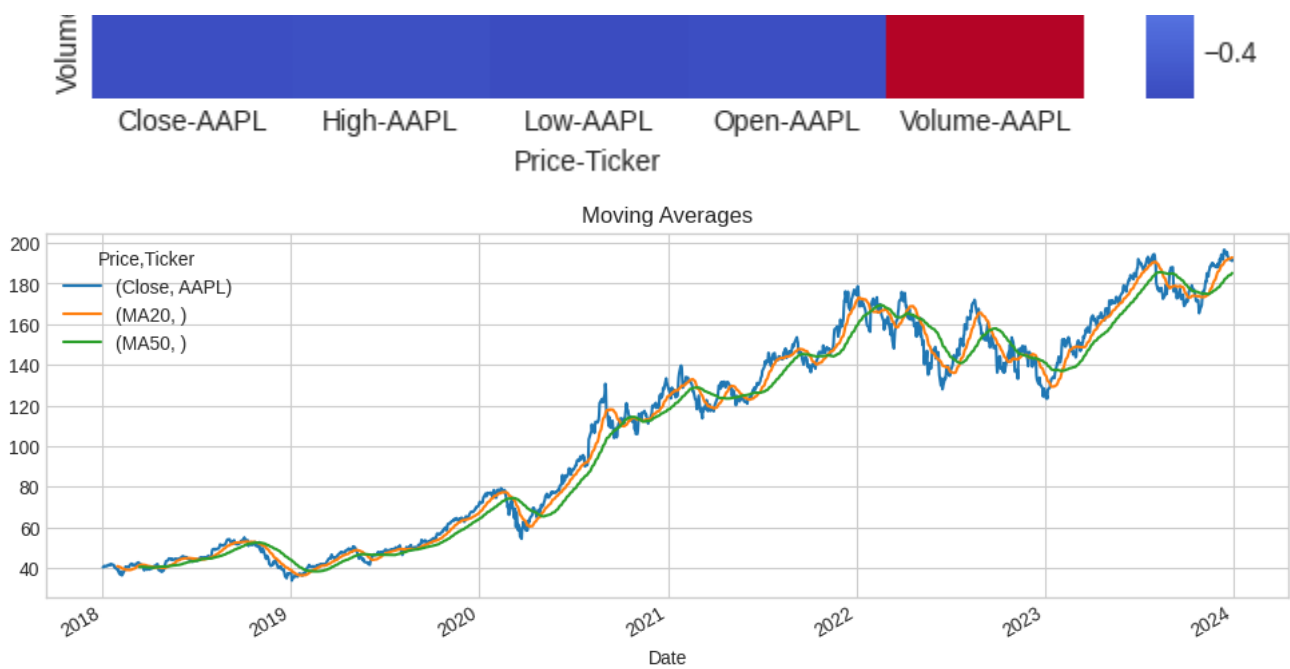
# 4. Heatmap hambastegui
plt.figure(figsize=(8,6))
sns.heatmap(df.corr(), annot=True, cmap='coolwarm')
plt.title("Heatmap - hambastegi bein vizhegi ha")
plt.show()

# 5. miangin moteharek
df['MA20'] = df['Close'].rolling(window=20).mean()
df['MA50'] = df['Close'].rolling(window=50).mean()
df[['Close', 'MA20', 'MA50']].plot(figsize=(12,4), title='Moving Averages')
plt.show()

```







```
X = scaled_df[['Open', 'High', 'Low', 'Volume']]
y = scaled_df['Close']
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, shuffle=False)
```

```
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_absolute_error, mean_squared_error, r2_score
import numpy as np
```

```
1. model = LinearRegression()
```

```
lr_model = LinearRegression()
lr_model.fit(X_train, y_train)
lr_pred = lr_model.predict(X_test)
lr_mae = mean_absolute_error(y_test, lr_pred)
lr_rmse = np.sqrt(mean_squared_error(y_test, lr_pred))
lr_r2 = r2_score(y_test, lr_pred)

print("Linear Regression")
print(f"MAE: {lr_mae:.4f}")
print(f"RMSE: {lr_rmse:.4f}")
print(f"R2 Score: {lr_r2:.4f}")

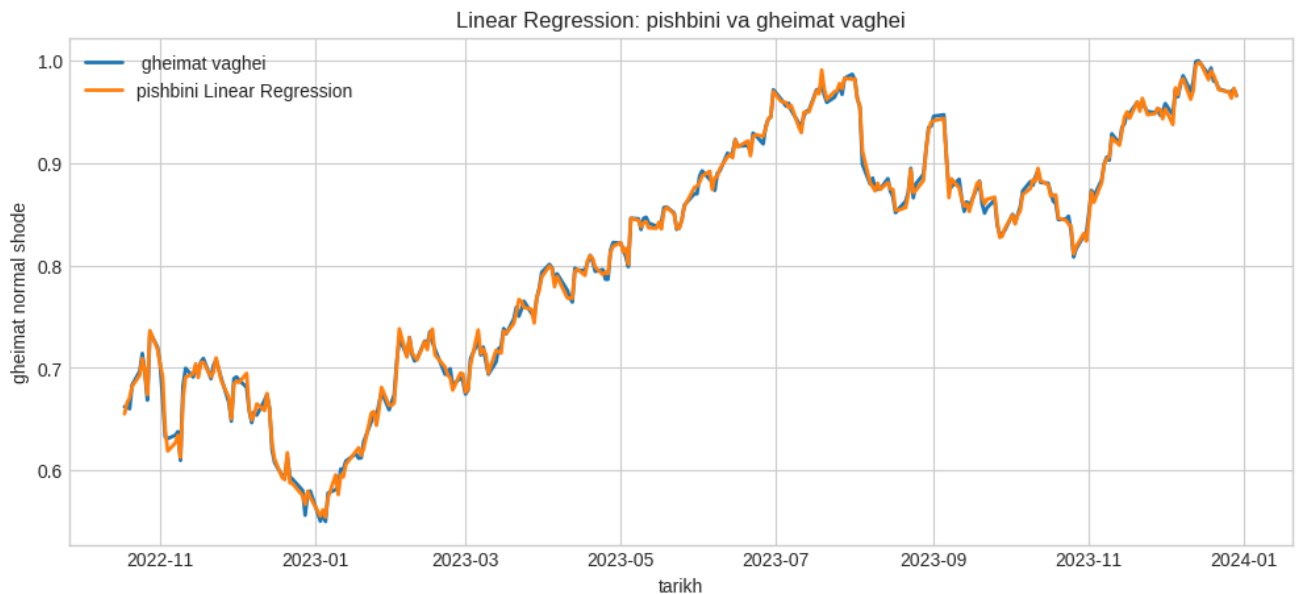
# nemodare moghayese
plt.figure(figsize=(12,5))
plt.plot(y_test.index, y_test, label=' gheimat vaghei', linewidth=2)
plt.plot(y_test.index, lr_pred, label='pishbini Linear Regression', linewidth=2)
plt.title("Linear Regression: pishbini va gheimat vaghei")
plt.xlabel("tarikh")
plt.ylabel("gheimat normal shode")
plt.legend()
plt.grid(True)
plt.show()
```

Linear Regression

MAE: 0.0041

RMSE: 0.0052

R<sup>2</sup> Score: 0.9981



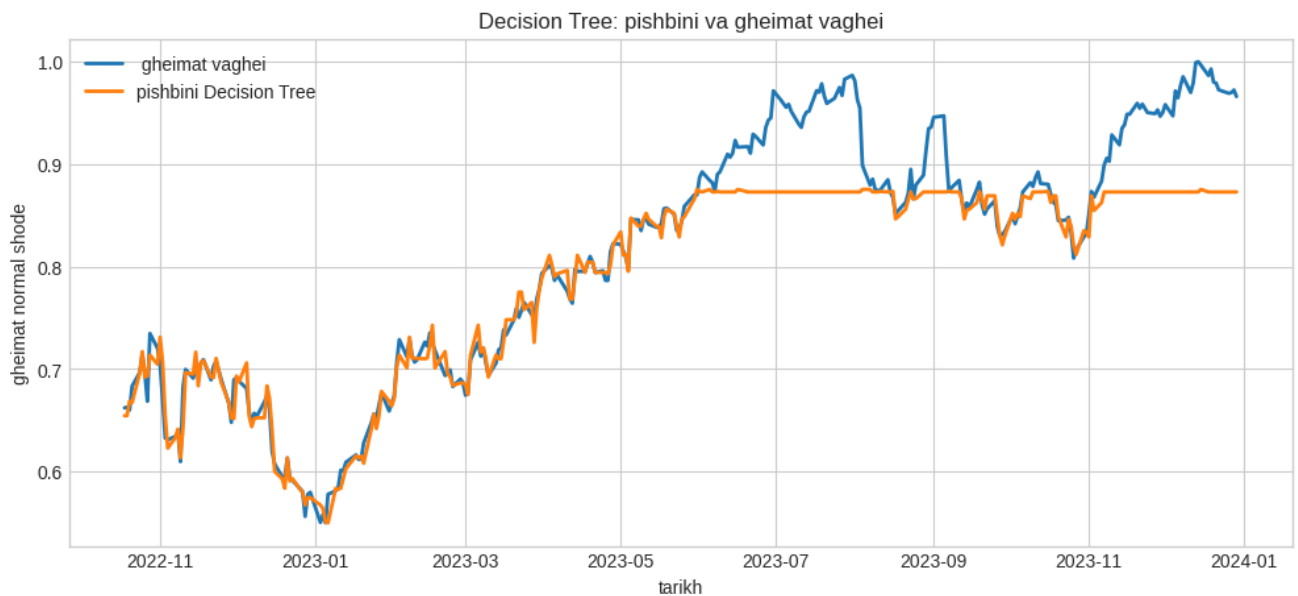
```
from sklearn.tree import DecisionTreeRegressor

dt_model = DecisionTreeRegressor()
dt_model.fit(X_train, y_train)
dt_pred = dt_model.predict(X_test)
dt_mae = mean_absolute_error(y_test, dt_pred)
dt_rmse = np.sqrt(mean_squared_error(y_test, dt_pred))
dt_r2 = r2_score(y_test, dt_pred)

print("Decision Tree")
print(f"MAE: {dt_mae:.4f}")
print(f"RMSE: {dt_rmse:.4f}")
print(f"R2 Score: {dt_r2:.4f}")

plt.figure(figsize=(12,5))
plt.plot(y_test.index, y_test, label=' gheimat vaghei', linewidth=2)
plt.plot(y_test.index, dt_pred, label='pishbini Decision Tree', linewidth=2)
plt.title("Decision Tree: pishbini va gheimat vaghei")
plt.xlabel("tarikh")
plt.ylabel("gheimat normal shode")
plt.legend()
plt.grid(True)
plt.show()
```

Decision Tree  
MAE: 0.0269  
RMSE: 0.0438  
R<sup>2</sup> Score: 0.8674



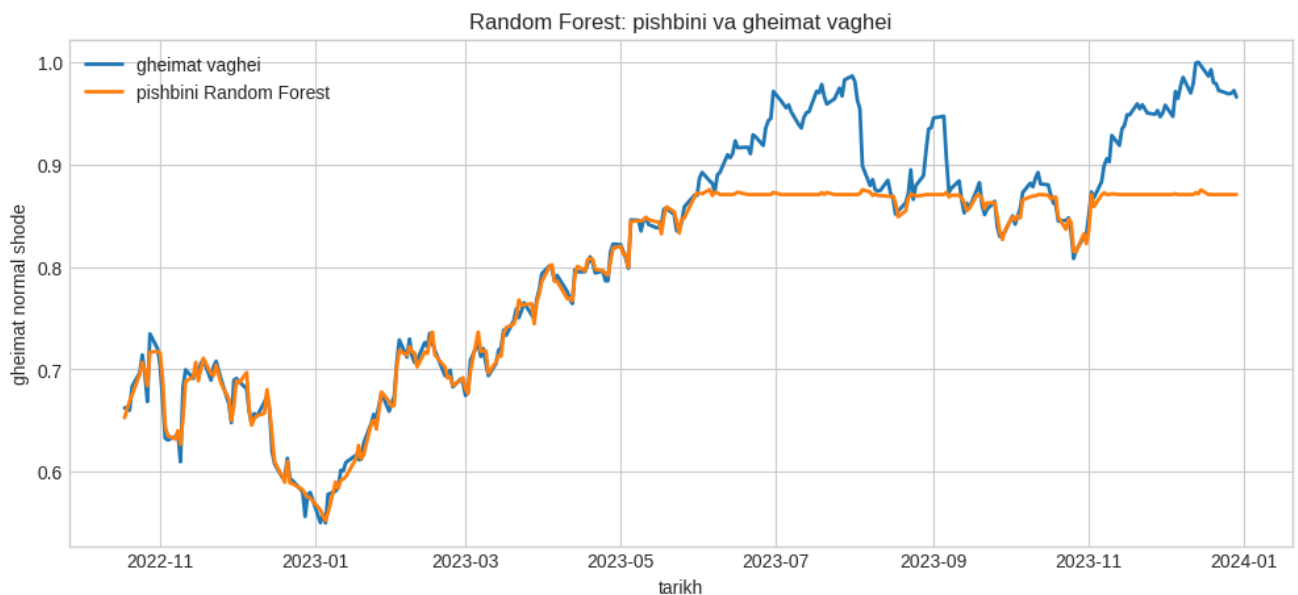
```
from sklearn.ensemble import RandomForestRegressor

rf_model = RandomForestRegressor(n_estimators=100)
rf_model.fit(X_train, y_train)
rf_pred = rf_model.predict(X_test)
rf_mae = mean_absolute_error(y_test, rf_pred)
rf_rmse = np.sqrt(mean_squared_error(y_test, rf_pred))
rf_r2 = r2_score(y_test, rf_pred)

print("Random Forest")
print(f"MAE: {rf_mae:.4f}")
print(f"RMSE: {rf_rmse:.4f}")
print(f"R2 Score: {rf_r2:.4f}")

plt.figure(figsize=(12,5))
plt.plot(y_test.index, y_test, label='gheimat vaghei ', linewidth=2)
plt.plot(y_test.index, rf_pred, label='pishbini Random Forest', linewidth=2)
plt.title("Random Forest: pishbini va gheimat vaghei")
plt.xlabel("tarikh")
plt.ylabel("gheimat normal shode")
plt.legend()
plt.grid(True)
plt.show()
```

Random Forest  
MAE: 0.0265  
RMSE: 0.0443  
R<sup>2</sup> Score: 0.8641



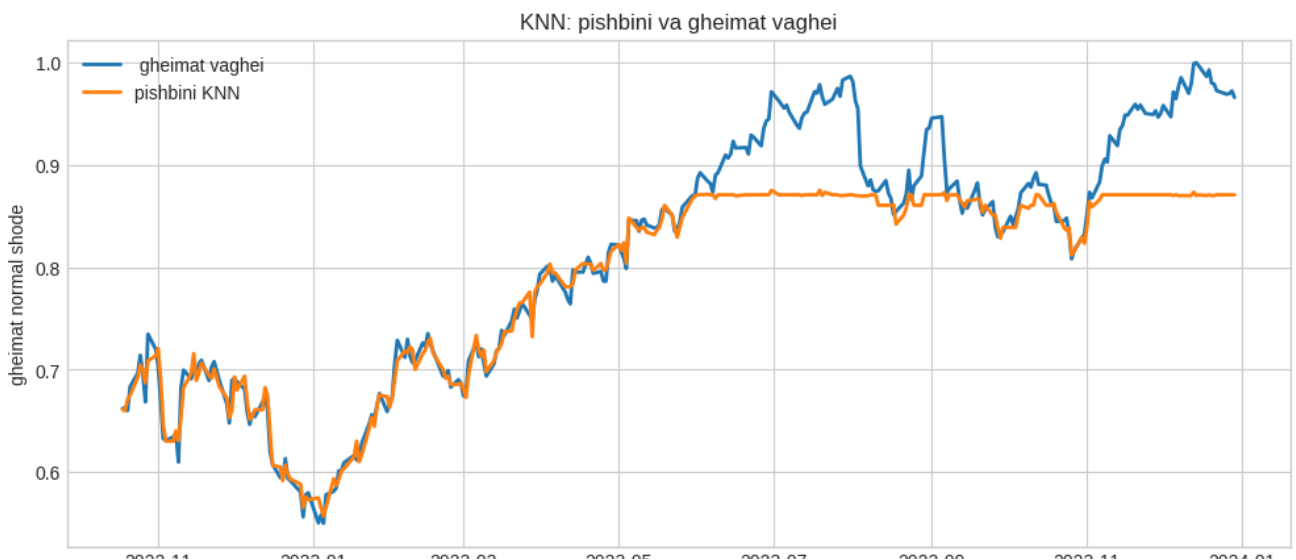


```
from sklearn.neighbors import KNeighborsRegressor
knn_model = KNeighborsRegressor()
knn_model.fit(X_train, y_train)
knn_pred = knn_model.predict(X_test)
knn_mae = mean_absolute_error(y_test, knn_pred)
knn_rmse = np.sqrt(mean_squared_error(y_test, knn_pred))
knn_r2 = r2_score(y_test, knn_pred)

print("KNN")
print(f"MAE: {knn_mae:.4f}")
print(f"RMSE: {knn_rmse:.4f}")
print(f"R2 Score: {knn_r2:.4f}")

plt.figure(figsize=(12,5))
plt.plot(y_test.index, y_test, label=' gheimat vaghei', linewidth=2)
plt.plot(y_test.index, knn_pred, label='pishbini KNN', linewidth=2)
plt.title("KNN: pishbini va gheimat vaghei")
plt.xlabel("tarikh")
plt.ylabel("gheimat normal shode")
plt.legend()
plt.grid(True)
plt.show()
```

KNN  
MAE: 0.0279  
RMSE: 0.0447  
R<sup>2</sup> Score: 0.8614



2022-11

2023-01

2023-03

2023-05

2023-07

2023-09

2023-11

2024-01

tarikh

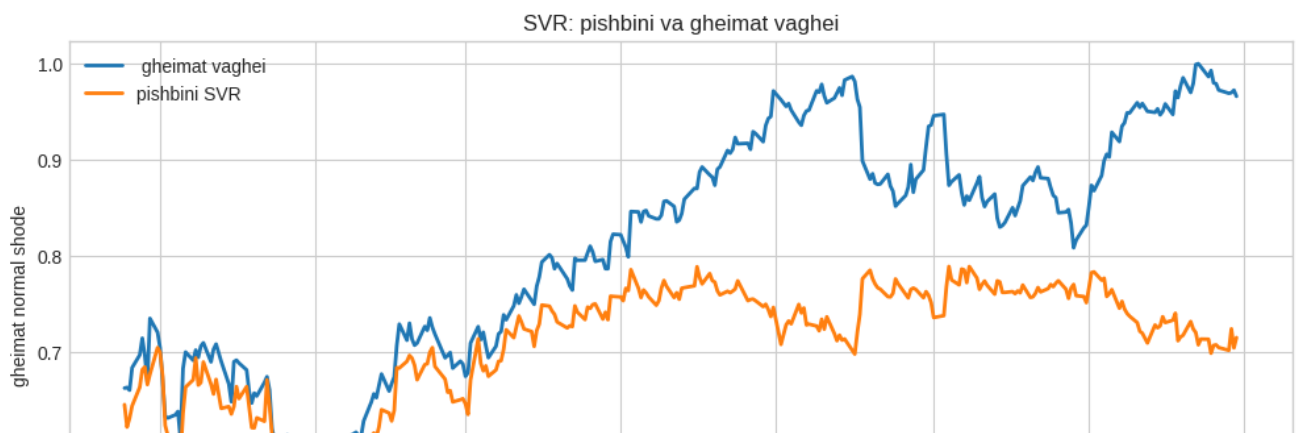
```
from sklearn.svm import SVR

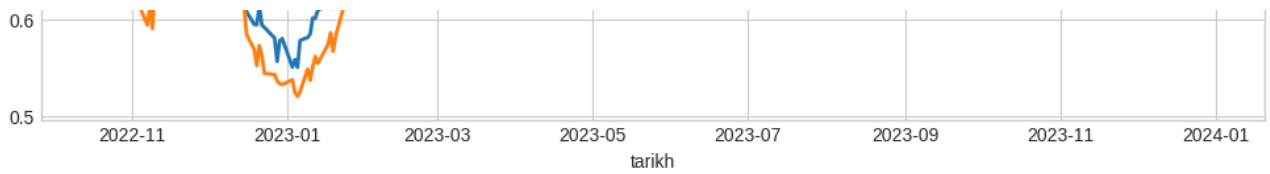
svr_model = SVR()
svr_model.fit(X_train, y_train)
svr_pred = svr_model.predict(X_test)
svr_mae = mean_absolute_error(y_test, svr_pred)
svr_rmse = np.sqrt(mean_squared_error(y_test, svr_pred))
svr_r2 = r2_score(y_test, svr_pred)

print("SVR")
print(f"MAE: {svr_mae:.4f}")
print(f"RMSE: {svr_rmse:.4f}")
print(f"R2 Score: {svr_r2:.4f}")

plt.figure(figsize=(12,5))
plt.plot(y_test.index, y_test, label=' gheimat vaghei', linewidth=2)
plt.plot(y_test.index, svr_pred, label='pishbini SVR', linewidth=2)
plt.title("SVR: pishbini va gheimat vaghei")
plt.xlabel("tarikh")
plt.ylabel("gheimat normal shode")
plt.legend()
plt.grid(True)
plt.show()
```

SVR  
MAE: 0.1004  
RMSE: 0.1284  
R<sup>2</sup> Score: -0.1406





```
comparison_df = pd.DataFrame({ 'Model': ['Linear Regression', 'Decision Tree', 'Random Fo
                                'MAE': [lr_mae, dt_mae, rf_mae, knn_mae, svr_mae],
                                'RMSE': [lr_rmse, dt_rmse, rf_rmse, knn_rmse, svr_rmse],
                                'R² Score': [lr_r2, dt_r2, rf_r2, knn_r2, svr_r2]
                                })
```

```
comparison_df = comparison_df.sort_values(by='R² Score', ascending=False)
print(" moghayese amalkarde model ha:")
print(comparison_df)
```

```
moghayese amalkarde model ha:
      Model      MAE      RMSE  R² Score
0  Linear Regression  0.004136  0.005223  0.998111
1    Decision Tree   0.026852  0.043766  0.867408
2    Random Forest   0.026480  0.044304  0.864124
3              KNN   0.027933  0.044749  0.861379
4              SVR   0.100394  0.128364 -0.140621
```

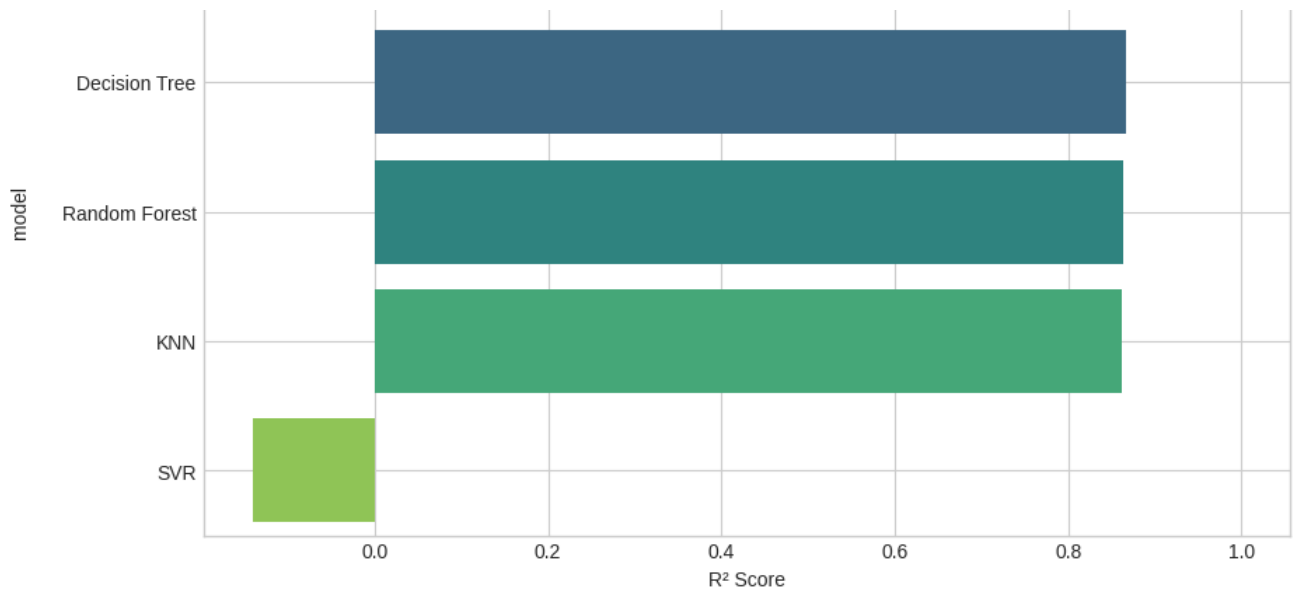
```
plt.figure(figsize=(10,6))
sns.barplot(data=comparison_df, x='R² Score', y='Model', palette='viridis')
plt.title("moghayese deghat model (R² Score)")
plt.xlabel("R² Score")
plt.ylabel("model")
plt.grid(True)
plt.show()
```

<ipython-input-16-8a7236082465>:2: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.

```
sns.barplot(data=comparison_df, x='R² Score', y='Model', palette='viridis')
moghayese deghat model (R² Score)
```





```
import seaborn as sns
sns.pairplot(df[['Open', 'High', 'Low', 'Close', 'Volume']])
```

<seaborn.axisgrid.PairGrid at 0x7d6511203910>

