

# data-madrid

May 2, 2024

ztc\_ail kısmı bu kodlar aracılığı ile yapıldı.

```
[1]: import pandas as pd
import numpy as np
import tensorflow as tf
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import MinMaxScaler
from sklearn.metrics import mean_squared_error, mean_absolute_error

veri = pd.read_csv(r"C:/Users/ASUS/Desktop/AI CLUB DATATHON/CSV/ztc_ail.csv")

veri['open'] = veri['open'].str.replace('\"', '').astype(float)
veri['high'] = veri['high'].str.replace('\"', '').astype(float)
veri['low'] = veri['low'].str.replace('\"', '').astype(float)
veri['close'] = veri['close'].str.replace('\"', '').astype(float)

veri['open'].fillna(veri['open'].mean(), inplace=True)
veri['high'].fillna(veri['high'].mean(), inplace=True)
veri['low'].fillna(veri['low'].mean(), inplace=True)
veri['close'].fillna(veri['close'].mean(), inplace=True)

veri = veri[['open', 'low', 'high', 'close']]

X = veri[['open', 'high', 'close']].values
y = veri['open'].values

scaler = MinMaxScaler()
X_scaled = scaler.fit_transform(X)
y_scaled = scaler.fit_transform(y.reshape(-1,1)).reshape(-1)

X_train, X_test, y_train, y_test = train_test_split(X_scaled, y_scaled,
↳test_size=0.2, random_state=42)

model = tf.keras.models.Sequential([
    tf.keras.layers.Dense(64, activation='relu', input_shape=(X_train.
↳shape[1],)),
    tf.keras.layers.Dense(32, activation='relu'),
```

```

        tf.keras.layers.Dense(1)
    ])

model.compile(optimizer='adam', loss='mean_squared_error')

model.fit(X_scaled, y_scaled, epochs=50, batch_size=32, verbose=1)

gelecek_veri = X_scaled[-201:]
tahminler = model.predict(gelecek_veri)

tahminler = scaler.inverse_transform(tahminler)

print("Gelecekteki Fiyat Tahminleri:")
print(tahminler)

y_pred = model.predict(X_test)
mse = mean_squared_error(y_test, y_pred)
rmse = np.sqrt(mse)
mae = mean_absolute_error(y_test, y_pred)
print("RMSE:", rmse)
print("MAE:", mae)

```

Epoch 1/50

C:\Users\ASUS\anaconda3\Lib\site-packages\keras\src\layers\core\dense.py:87:  
 UserWarning: Do not pass an `input\_shape`/`input\_dim` argument to a layer. When  
 using Sequential models, prefer using an `Input(shape)` object as the first  
 layer in the model instead.

```
super().__init__(activity_regularizer=activity_regularizer, **kwargs)
```

360/360 1s 819us/step -

loss: 0.0122

Epoch 2/50

360/360 0s 743us/step -

loss: 1.7933e-06

Epoch 3/50

360/360 0s 820us/step -

loss: 7.6221e-07

Epoch 4/50

360/360 1s 2ms/step -

loss: 4.4084e-07

Epoch 5/50

360/360 1s 2ms/step -

loss: 4.5971e-07

Epoch 6/50

360/360 1s 2ms/step -

loss: 2.7161e-07

Epoch 7/50

```

360/360          1s 1ms/step -
loss: 4.7412e-07
Epoch 8/50
360/360          1s 1ms/step -
loss: 2.4705e-07
Epoch 9/50
360/360          1s 2ms/step -
loss: 4.2063e-07
Epoch 10/50
360/360          1s 2ms/step -
loss: 3.2261e-07
Epoch 11/50
360/360          1s 2ms/step -
loss: 2.9180e-07
Epoch 12/50
360/360          1s 2ms/step -
loss: 5.1377e-07
Epoch 13/50
360/360          1s 2ms/step -
loss: 4.9525e-07
Epoch 14/50
360/360          1s 2ms/step -
loss: 2.5193e-06
Epoch 15/50
360/360          1s 2ms/step -
loss: 7.7375e-07
Epoch 16/50
360/360          1s 2ms/step -
loss: 2.8358e-07
Epoch 17/50
360/360          1s 2ms/step -
loss: 2.9692e-07
Epoch 18/50
360/360          1s 2ms/step -
loss: 8.0565e-07
Epoch 19/50
360/360          1s 2ms/step -
loss: 2.9873e-06
Epoch 20/50
360/360          1s 2ms/step -
loss: 1.8617e-07
Epoch 21/50
360/360          1s 2ms/step -
loss: 2.6134e-06
Epoch 22/50
360/360          1s 2ms/step -
loss: 3.4818e-07
Epoch 23/50

```

```

360/360          1s 2ms/step -
loss: 2.2442e-06
Epoch 24/50
360/360          1s 2ms/step -
loss: 1.1519e-06
Epoch 25/50
360/360          1s 2ms/step -
loss: 4.1131e-07
Epoch 26/50
360/360          1s 2ms/step -
loss: 8.1102e-07
Epoch 27/50
360/360          1s 2ms/step -
loss: 2.1929e-06
Epoch 28/50
360/360          1s 2ms/step -
loss: 5.6680e-07
Epoch 29/50
360/360          1s 2ms/step -
loss: 2.8018e-06
Epoch 30/50
360/360          1s 2ms/step -
loss: 8.3587e-07
Epoch 31/50
360/360          1s 2ms/step -
loss: 2.1951e-07
Epoch 32/50
360/360          1s 1ms/step -
loss: 3.1545e-06
Epoch 33/50
360/360          1s 2ms/step -
loss: 1.5138e-06
Epoch 34/50
360/360          1s 2ms/step -
loss: 3.4526e-06
Epoch 35/50
360/360          1s 2ms/step -
loss: 8.7778e-08
Epoch 36/50
360/360          1s 2ms/step -
loss: 2.1773e-07
Epoch 37/50
360/360          1s 2ms/step -
loss: 2.2318e-06
Epoch 38/50
360/360          1s 2ms/step -
loss: 2.7593e-07
Epoch 39/50

```

```

360/360          1s 2ms/step -
loss: 1.0940e-06
Epoch 40/50
360/360          1s 2ms/step -
loss: 5.7537e-07
Epoch 41/50
360/360          1s 2ms/step -
loss: 1.4486e-06
Epoch 42/50
360/360          1s 2ms/step -
loss: 1.0347e-07
Epoch 43/50
360/360          1s 2ms/step -
loss: 4.5939e-07
Epoch 44/50
360/360          1s 2ms/step -
loss: 3.2728e-06
Epoch 45/50
360/360          1s 2ms/step -
loss: 2.0950e-07
Epoch 46/50
360/360          1s 2ms/step -
loss: 2.0090e-07
Epoch 47/50
360/360          1s 2ms/step -
loss: 7.6189e-08
Epoch 48/50
360/360          1s 2ms/step -
loss: 2.1330e-07
Epoch 49/50
360/360          1s 2ms/step -
loss: 1.7747e-06
Epoch 50/50
360/360          1s 2ms/step -
loss: 5.7129e-07
7/7              0s 11ms/step
Gelecekteki Fiyat Tahminleri:
[[23871.344]
 [23753.27 ]
 [23651.28 ]
 [23809.049]
 [24991.963]
 [24733.021]
 [25079.967]
 [25172.107]
 [24871.201]
 [25082.748]
 [25315.697]

```

[24837.883]  
[24953.238]  
[24795.195]  
[25015.973]  
[25256.98 ]  
[25344.06 ]  
[25273.875]  
[25130.021]  
[25210.855]  
[24994.365]  
[24805.578]  
[24851.637]  
[24910.877]  
[24994.154]  
[24876.48 ]  
[24778.95 ]  
[24772.316]  
[24765.885]  
[24780.455]  
[24981.574]  
[25088.145]  
[25113.566]  
[25178.053]  
[25179.555]  
[25111.068]  
[25181.521]  
[25269.988]  
[25319.484]  
[25346.95 ]  
[25448.535]  
[25455.105]  
[25419.377]  
[25483.74 ]  
[25296.252]  
[25149.588]  
[25217.81 ]  
[25265.855]  
[25287.059]  
[25258.988]  
[25180.844]  
[25448.63 ]  
[25341.527]  
[25289.861]  
[25263.48 ]  
[25356.574]  
[25507.584]  
[25406.2 ]  
[25531.955]

[25786.533]  
[25455.904]  
[25550.541]  
[25732.371]  
[25820.217]  
[25955.623]  
[26046.025]  
[25948.787]  
[25701.227]  
[25928.812]  
[25626.074]  
[25701.766]  
[25703.623]  
[25788.688]  
[25792.299]  
[26085.506]  
[26746.512]  
[26781.87 ]  
[26995.96 ]  
[26897.016]  
[26909.578]  
[27074.371]  
[26918.666]  
[27158.908]  
[27095.088]  
[27291.162]  
[27277.44 ]  
[27276.32 ]  
[27088.95 ]  
[27096.646]  
[27274.752]  
[27222.357]  
[27109.379]  
[27080.688]  
[27150.447]  
[26930.646]  
[27004.494]  
[27179.621]  
[26969.404]  
[27055.475]  
[27022.166]  
[27035.557]  
[27094.559]  
[27126.682]  
[26936.21 ]  
[26818.732]  
[26398.543]  
[26557.182]

[26682.508]  
[26697.25 ]  
[26567.094]  
[26506.072]  
[26643.422]  
[26786.342]  
[26857.982]  
[27110.803]  
[27050.838]  
[27100.887]  
[27163.854]  
[27147.258]  
[27356.756]  
[27415.822]  
[27635.863]  
[27582.988]  
[27555.629]  
[27493.754]  
[27401.088]  
[27156.578]  
[27231.318]  
[27343.23 ]  
[27479.996]  
[27532.678]  
[27622.541]  
[27463.475]  
[27576.043]  
[27448.834]  
[27516.258]  
[27406.219]  
[27442.238]  
[27480.22 ]  
[27567.787]  
[27475.6 ]  
[27525.951]  
[27689.45 ]  
[27558.957]  
[27605.277]  
[27461.03 ]  
[27431.576]  
[27420.547]  
[27033.71 ]  
[26952.512]  
[26626.805]  
[26800.555]  
[26704.191]  
[26351.137]  
[26058.736]



[26559.285]  
[26720.146]  
[26894.479]  
[26839.03 ]  
[27030. ]  
[26920.787]  
[25949.826]  
[25640.41 ]  
[25342.227]  
[25803.71 ]  
[25716.193]  
[25203.31 ]  
[25473.576]  
[25278.988]  
[25456.709]  
[25514.707]  
[25771.562]  
[25797.584]  
[25621.287]  
[25529.914]  
[25649.312]  
[26395.55 ]  
[25934.72 ]  
[25607.486]  
[25655.39 ]  
[25669.967]  
[25722.031]  
[25288.902]  
[25472.633]  
[25542.414]  
[25826.389]  
[25746.607]  
[25637.082]  
[25567.078]  
[25500.55 ]  
[25597.553]  
[25764.291]  
[25396.95 ]  
[25179.342]  
[25373.682]  
[25241.191]  
[25340.871]  
[25195.248]  
[25180.865]  
[25420.088]  
[25459.646]]

72/72                      0s 2ms/step  
RMSE: 0.00048756292988456283

MAE: 0.00035567596396486546

```
[2]: import pandas as pd
import numpy as np
import tensorflow as tf
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import MinMaxScaler
from sklearn.metrics import mean_squared_error, mean_absolute_error

veri = pd.read_csv(r"C:/Users/ASUS/Desktop/AI CLUB DATATHON/CSV/zzc_ail.csv")

veri['open'] = veri['open'].str.replace('"', '').astype(float)
veri['high'] = veri['high'].str.replace('"', '').astype(float)
veri['low'] = veri['low'].str.replace('"', '').astype(float)
veri['close'] = veri['close'].str.replace('"', '').astype(float)

veri['open'].fillna(veri['open'].mean(), inplace=True)
veri['high'].fillna(veri['high'].mean(), inplace=True)
veri['low'].fillna(veri['low'].mean(), inplace=True)
veri['close'].fillna(veri['close'].mean(), inplace=True)

veri = veri[['open', 'low', 'high', 'close']]

X = veri[['open', 'high', 'low']].values
y = veri['close'].values

scaler = MinMaxScaler()
X_scaled = scaler.fit_transform(X)
y_scaled = scaler.fit_transform(y.reshape(-1,1)).reshape(-1)

X_train, X_test, y_train, y_test = train_test_split(X_scaled, y_scaled,
    ↪test_size=0.2, random_state=42)

model = tf.keras.models.Sequential([
    tf.keras.layers.Dense(64, activation='relu', input_shape=(X_train.
    ↪shape[1],)),
    tf.keras.layers.Dense(32, activation='relu'),
    tf.keras.layers.Dense(1)
])

model.compile(optimizer='adam', loss='mean_squared_error')
```

```

model.fit(X_scaled, y_scaled, epochs=25, batch_size=32, verbose=1)

gelecek_veri = X_scaled[-201:]
tahminler = model.predict(gelecek_veri)

tahminler = scaler.inverse_transform(tahminler)

print("Gelecekteki Fiyat Tahminleri:")
print(tahminler)

y_pred = model.predict(X_test)
mse = mean_squared_error(y_test, y_pred)
rmse = np.sqrt(mse)
mae = mean_absolute_error(y_test, y_pred)
print("RMSE:", rmse)
print("MAE:", mae)

```

Epoch 1/25

C:\Users\ASUS\anaconda3\Lib\site-packages\keras\src\layers\core\dense.py:87:  
UserWarning: Do not pass an `input\_shape`/`input\_dim` argument to a layer. When  
using Sequential models, prefer using an `Input(shape)` object as the first  
layer in the model instead.

```
super().__init__(activity_regularizer=activity_regularizer, **kwargs)
```

360/360 3s 2ms/step -

loss: 0.0126

Epoch 2/25

360/360 1s 2ms/step -

loss: 3.0413e-06

Epoch 3/25

360/360 1s 2ms/step -

loss: 2.1935e-06

Epoch 4/25

360/360 1s 2ms/step -

loss: 1.6653e-06

Epoch 5/25

360/360 1s 2ms/step -

loss: 3.1726e-06

Epoch 6/25

360/360 1s 2ms/step -

loss: 1.7359e-06

Epoch 7/25

```

360/360          1s 2ms/step -
loss: 2.0367e-06
Epoch 8/25
360/360          1s 2ms/step -
loss: 1.5843e-06
Epoch 9/25
360/360          1s 2ms/step -
loss: 1.7682e-06
Epoch 10/25
360/360          1s 2ms/step -
loss: 1.0920e-06
Epoch 11/25
360/360          1s 2ms/step -
loss: 1.2651e-06
Epoch 12/25
360/360          1s 2ms/step -
loss: 1.3778e-06
Epoch 13/25
360/360          1s 2ms/step -
loss: 1.3147e-06
Epoch 14/25
360/360          1s 2ms/step -
loss: 6.6282e-07
Epoch 15/25
360/360          1s 2ms/step -
loss: 1.8652e-06
Epoch 16/25
360/360          1s 2ms/step -
loss: 9.7717e-07
Epoch 17/25
360/360          1s 2ms/step -
loss: 1.9848e-06
Epoch 18/25
360/360          1s 2ms/step -
loss: 6.4578e-07
Epoch 19/25
360/360          1s 2ms/step -
loss: 6.3957e-07
Epoch 20/25
360/360          1s 2ms/step -
loss: 5.3102e-06
Epoch 21/25
360/360          1s 2ms/step -
loss: 7.9311e-07
Epoch 22/25
360/360          1s 2ms/step -
loss: 1.0619e-06
Epoch 23/25

```

```
360/360          1s 2ms/step -
loss: 8.0312e-07
Epoch 24/25
360/360          1s 2ms/step -
loss: 1.0719e-06
Epoch 25/25
360/360          1s 2ms/step -
loss: 1.7251e-07
7/7              0s 12ms/step
Gelecekteki Fiyat Tahminleri:
[[24046.928]
 [23932.057]
 [23889.87 ]
 [24349.926]
 [25069.434]
 [25149.174]
 [25482.625]
 [25373.389]
 [25160.959]
 [25360.232]
 [25418.852]
 [25101.629]
 [25337.133]
 [25122.69 ]
 [25308.984]
 [25388.154]
 [25438.852]
 [25361.744]
 [25426.988]
 [25249.803]
 [25059.016]
 [24967.352]
 [24968.21 ]
 [25018.826]
 [25049.662]
 [25016.379]
 [24931.635]
 [24924.873]
 [24947.703]
 [25048.5  ]
 [25261.576]
 [25212.977]
 [25237.479]
 [25350.605]
 [25298.229]
 [25263.283]
 [25609.81 ]
 [25480.148]
```

[25453.895]  
[25648.027]  
[25575.006]  
[25596.135]  
[25643.037]  
[25502.871]  
[25453.936]  
[25318.05 ]  
[25311.266]  
[25361.947]  
[25358.121]  
[25308.246]  
[25494.352]  
[25514.73 ]  
[25434.848]  
[25436.016]  
[25429.379]  
[11132.418]  
[25557.479]  
[11130.018]  
[25899.273]  
[26421.729]  
[25608.994]  
[25961.127]  
[25962.99 ]  
[25991.516]  
[26191.492]  
[26097.484]  
[25979.248]  
[25980.521]  
[25974.105]  
[25772.209]  
[11133.313]  
[25864.662]  
[25928.475]  
[26129.412]  
[26854.605]  
[27003.566]  
[27066.006]  
[27063.89 ]  
[27229.164]  
[27205.473]  
[27272.914]  
[27275.754]  
[27403.023]  
[27368.988]  
[27355.873]  
[27375.992]

[27443.08 ]  
[27241.29 ]  
[27356.72 ]  
[27422.012]  
[27296.762]  
[27266.268]  
[27307.887]  
[27211.633]  
[27140.678]  
[27153.809]  
[27262.057]  
[27169.418]  
[27129.303]  
[27222.326]  
[27140.877]  
[27175.143]  
[27382.564]  
[27040.72 ]  
[26890.078]  
[26538.996]  
[11128.871]  
[26836.416]  
[26738.725]  
[26665.18 ]  
[26722.549]  
[26841.928]  
[27033.307]  
[27166.58 ]  
[27160.111]  
[27216.842]  
[27212.84 ]  
[27211.838]  
[27412.395]  
[27589.861]  
[27765.088]  
[27745.398]  
[27664.535]  
[27688.316]  
[27582.492]  
[27515.088]  
[27485.168]  
[27559.348]  
[27532.494]  
[27597.326]  
[27679.152]  
[27694.021]  
[27674.209]  
[27624.672]

[27573.824]  
[27584.559]  
[27499.799]  
[27600.336]  
[27635.666]  
[27760.541]  
[27583.164]  
[27749.43 ]  
[27814.35 ]  
[27724.78 ]  
[27710.96 ]  
[27523.668]  
[27530.555]  
[27463.398]  
[27310.78 ]  
[27130.41 ]  
[26876.334]  
[11126.783]  
[26723.309]  
[26503.502]  
[26777.01 ]  
[26768.312]  
[26983.33 ]  
[27006.377]  
[27277.068]  
[27134.006]  
[26956.426]  
[26122.82 ]  
[25763.47 ]  
[25686.469]  
[25913.932]  
[25795.492]  
[25598.705]  
[25619.457]  
[25539.871]  
[11129.792]  
[25851.457]  
[25842.572]  
[11126.219]  
[25762.377]  
[11129.691]  
[26517.143]  
[26558.844]  
[25954.229]  
[25777.502]  
[25951.066]  
[25869.885]  
[25746.61 ]



```
[25515.67 ]
[25658.748]
[25801.068]
[25868.94 ]
[25785.473]
[25666.576]
[25687.86 ]
[25666.125]
[25782.975]
[25811.164]
[25514.508]
[25413.113]
[25627.424]
[25407.688]
[25402.021]
[25400.379]
[25380.805]
[25506.182]
[25610.861]]
72/72          0s 2ms/step
RMSE: 0.000330938398256279
MAE: 0.0002776021253427656
```

```
[3]: import pandas as pd
import numpy as np
import tensorflow as tf
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import MinMaxScaler
from sklearn.metrics import mean_squared_error, mean_absolute_error

veri = pd.read_csv(r"C:/Users/ASUS/Desktop/AI CLUB DATATHON/CSV/zzc_ail.csv")

veri['open'] = veri['open'].str.replace('"', '').astype(float)
veri['high'] = veri['high'].str.replace('"', '').astype(float)
veri['low'] = veri['low'].str.replace('"', '').astype(float)
veri['close'] = veri['close'].str.replace('"', '').astype(float)

veri['open'].fillna(veri['open'].mean(), inplace=True)
veri['high'].fillna(veri['high'].mean(), inplace=True)
veri['low'].fillna(veri['low'].mean(), inplace=True)
veri['close'].fillna(veri['close'].mean(), inplace=True)

veri = veri[['open', 'low', 'high', 'close']]

X = veri[['open', 'high', 'low']].values
y = veri['low'].values
```

```

scaler = MinMaxScaler()
X_scaled = scaler.fit_transform(X)
y_scaled = scaler.fit_transform(y.reshape(-1,1)).reshape(-1)

X_train, X_test, y_train, y_test = train_test_split(X_scaled, y_scaled,
↳test_size=0.2, random_state=42)

model = tf.keras.models.Sequential([
    tf.keras.layers.Dense(64, activation='relu', input_shape=(X_train.
↳shape[1],)),
    tf.keras.layers.Dense(32, activation='relu'),
    tf.keras.layers.Dense(1)
])

model.compile(optimizer='adam', loss='mean_squared_error')

model.fit(X_scaled, y_scaled, epochs=25, batch_size=32, verbose=1)

gelecek_veri = X_scaled[-201:]
tahminler = model.predict(gelecek_veri)

tahminler = scaler.inverse_transform(tahminler)

print("Gelecekteki Fiyat Tahminleri:")
print(tahminler)

y_pred = model.predict(X_test)
mse = mean_squared_error(y_test, y_pred)
rmse = np.sqrt(mse)
mae = mean_absolute_error(y_test, y_pred)
print("RMSE:", rmse)
print("MAE:", mae)

```

Epoch 1/25

C:\Users\ASUS\anaconda3\Lib\site-packages\keras\src\layers\core\dense.py:87:  
UserWarning: Do not pass an `input\_shape`/`input\_dim` argument to a layer. When  
using Sequential models, prefer using an `Input(shape)` object as the first  
layer in the model instead.

```
super().__init__(activity_regularizer=activity_regularizer, **kwargs)
```

360/360                      3s 2ms/step -

loss: 0.0185

Epoch 2/25

360/360                      1s 2ms/step -

```

loss: 3.2344e-06
Epoch 3/25
360/360          1s 2ms/step -
loss: 2.7664e-06
Epoch 4/25
360/360          1s 2ms/step -
loss: 2.3211e-06
Epoch 5/25
360/360          1s 2ms/step -
loss: 2.0229e-06
Epoch 6/25
360/360          1s 2ms/step -
loss: 1.4939e-06
Epoch 7/25
360/360          1s 2ms/step -
loss: 2.2801e-06
Epoch 8/25
360/360          1s 2ms/step -
loss: 1.3525e-06
Epoch 9/25
360/360          1s 2ms/step -
loss: 1.0585e-06
Epoch 10/25
360/360          1s 2ms/step -
loss: 1.0616e-06
Epoch 11/25
360/360          1s 2ms/step -
loss: 1.2644e-06
Epoch 12/25
360/360          1s 2ms/step -
loss: 1.0310e-06
Epoch 13/25
360/360          1s 2ms/step -
loss: 7.1882e-07
Epoch 14/25
360/360          1s 2ms/step -
loss: 4.0542e-07
Epoch 15/25
360/360          1s 2ms/step -
loss: 5.7388e-07
Epoch 16/25
360/360          1s 2ms/step -
loss: 1.8575e-06
Epoch 17/25
360/360          1s 2ms/step -
loss: 3.8119e-06
Epoch 18/25
360/360          1s 2ms/step -

```

```

loss: 5.0710e-07
Epoch 19/25
360/360          1s 2ms/step -
loss: 4.2924e-07
Epoch 20/25
360/360          1s 2ms/step -
loss: 6.8676e-07
Epoch 21/25
360/360          1s 2ms/step -
loss: 1.7266e-06
Epoch 22/25
360/360          1s 2ms/step -
loss: 2.9850e-07
Epoch 23/25
360/360          1s 2ms/step -
loss: 1.0493e-06
Epoch 24/25
360/360          1s 2ms/step -
loss: 1.3268e-06
Epoch 25/25
360/360          1s 2ms/step -
loss: 1.9646e-06
7/7              0s 11ms/step
Gelecekteki Fiyat Tahminleri:
[[23852.3  ]
 [23744.932]
 [23738.182]
 [23896.307]
 [24853.064]
 [24757.57  ]
 [25148.582]
 [24977.754]
 [24979.156]
 [25093.547]
 [24524.24  ]
 [24704.115]
 [24904.135]
 [24678.873]
 [25042.2   ]
 [25219.197]
 [25221.748]
 [25215.26  ]
 [25220.307]
 [25080.385]
 [24923.906]
 [24840.404]
 [24837.154]
 [24791.48  ]

```

[24913.256]  
[24885.273]  
[24814.521]  
[24829.98 ]  
[24827.701]  
[24863.86 ]  
[25076.365]  
[25153.531]  
[25195.695]  
[25140.914]  
[25180.78 ]  
[25173.85 ]  
[25170.715]  
[25234.309]  
[25355.041]  
[25432.236]  
[25506.81 ]  
[25441.064]  
[25517.6 ]  
[25366.98 ]  
[25309.502]  
[25223.88 ]  
[25276.1 ]  
[25308.258]  
[25336.234]  
[25263.42 ]  
[25255.16 ]  
[25425.701]  
[25361.271]  
[25356.592]  
[25370.578]  
[25386.643]  
[25472.42 ]  
[25385.94 ]  
[25568.303]  
[25477.12 ]  
[25021.66 ]  
[25613.8 ]  
[25770.752]  
[25800.895]  
[26039.697]  
[26076.186]  
[25750.828]  
[25805.418]  
[25390.418]  
[25568.922]  
[25748.037]  
[25798.354]

[25880.229]  
[25859.67 ]  
[26162.637]  
[26824.83 ]  
[26889.35 ]  
[27008.766]  
[26954.225]  
[26980.77 ]  
[26966.834]  
[27033.477]  
[27185.467]  
[27136.303]  
[27270.166]  
[27278.307]  
[27155.6 ]  
[27165.457]  
[27163.21 ]  
[27356.703]  
[27219.496]  
[27188.908]  
[27160.67 ]  
[26998.777]  
[26936.592]  
[27078.424]  
[27203.605]  
[27081.412]  
[27062.139]  
[27122.324]  
[26998.562]  
[27069.516]  
[27044.459]  
[26840.61 ]  
[26466.205]  
[25986.719]  
[26473.17 ]  
[26743.309]  
[26633.576]  
[26600.562]  
[26598.799]  
[26726.592]  
[26909.178]  
[26931.576]  
[27140.646]  
[27151.105]  
[27173.822]  
[27217.637]  
[27245.576]  
[27449.127]

[27512.898]  
[27650.084]  
[27630.16 ]  
[27600.916]  
[27401.508]  
[27111.668]  
[27226.494]  
[27320.1 ]  
[27337.36 ]  
[27481.152]  
[27537.234]  
[27571.566]  
[27563.652]  
[27522.127]  
[27543.326]  
[27515.865]  
[27416.678]  
[27530.135]  
[27603.197]  
[27497.723]  
[27514.297]  
[27638.768]  
[27649.46 ]  
[27678.531]  
[27666.834]  
[27469.65 ]  
[27385.79 ]  
[27029.871]  
[26910.807]  
[26672.645]  
[26399.113]  
[26635.459]  
[26220.137]  
[25919.23 ]  
[26121.115]  
[26529.426]  
[26839.955]  
[26967.764]  
[26900.709]  
[27046.533]  
[25736.566]  
[25252.803]  
[25282.57 ]  
[25343.936]  
[25739.049]  
[25326.615]  
[24790.73 ]  
[25394. ]

```
[25204.459]
[25432.205]
[25464.102]
[25639.758]
[25697.875]
[25512.102]
[25501.35 ]
[25727.537]
[25926.783]
[25514.77 ]
[25723.678]
[25768.541]
[25742.22 ]
[25423.459]
[25228.459]
[25581.133]
[25558.52 ]
[25837.69 ]
[25732.902]
[25433.098]
[25550.713]
[25566.084]
[25673.744]
[25397.979]
[25186.219]
[25166.088]
[25304.727]
[25153.596]
[25299.186]
[25263.076]
[25293.35 ]
[25420.727]
[25513.229]]
```

72/72                      0s 1ms/step

RMSE: 0.0011752951749601454

MAE: 0.0008784334568683738

```
[4]: import pandas as pd
import numpy as np
import tensorflow as tf
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import MinMaxScaler
from sklearn.metrics import mean_squared_error, mean_absolute_error

veri = pd.read_csv(r"C:/Users/ASUS/Desktop/AI CLUB DATATHON/CSV/zzc_ail.csv")

veri['open'] = veri['open'].str.replace('\"', '').astype(float)
```



```

veri['high'] = veri['high'].str.replace(' ', '').astype(float)
veri['low'] = veri['low'].str.replace(' ', '').astype(float)
veri['close'] = veri['close'].str.replace(' ', '').astype(float)

veri['open'].fillna(veri['open'].mean(), inplace=True)
veri['high'].fillna(veri['high'].mean(), inplace=True)
veri['low'].fillna(veri['low'].mean(), inplace=True)
veri['close'].fillna(veri['close'].mean(), inplace=True)

veri = veri[['open', 'low', 'high', 'close']]

X = veri[['open', 'high', 'low']].values
y = veri['close'].values

scaler = MinMaxScaler()
X_scaled = scaler.fit_transform(X)
y_scaled = scaler.fit_transform(y.reshape(-1,1)).reshape(-1)

X_train, X_test, y_train, y_test = train_test_split(X_scaled, y_scaled,
    ↪test_size=0.2, random_state=42)

model = tf.keras.models.Sequential([
    tf.keras.layers.Dense(64, activation='relu', input_shape=(X_train.
    ↪shape[1],)),
    tf.keras.layers.Dense(32, activation='relu'),
    tf.keras.layers.Dense(1)
])

model.compile(optimizer='adam', loss='mean_squared_error')

model.fit(X_scaled, y_scaled, epochs=25, batch_size=32, verbose=1)

gelecek_veri = X_scaled[-201:] # Tüm verilerin son 89 verisini kullanıyoruz
tahminler = model.predict(gelecek_veri)

tahminler = scaler.inverse_transform(tahminler)

print("Gelecekteki Fiyat Tahminleri:")
print(tahminler)

y_pred = model.predict(X_test)
mse = mean_squared_error(y_test, y_pred)
rmse = np.sqrt(mse)
mae = mean_absolute_error(y_test, y_pred)

```

```
print("RMSE:", rmse)
print("MAE:", mae)
```

Epoch 1/25

C:\Users\ASUS\anaconda3\Lib\site-packages\keras\src\layers\core\dense.py:87:  
UserWarning: Do not pass an `input\_shape`/`input\_dim` argument to a layer. When  
using Sequential models, prefer using an `Input(shape)` object as the first  
layer in the model instead.

```
super().__init__(activity_regularizer=activity_regularizer, **kwargs)
```

360/360 3s 2ms/step -

loss: 0.0055

Epoch 2/25

360/360 1s 2ms/step -

loss: 0.0021

Epoch 3/25

360/360 1s 2ms/step -

loss: 0.0018

Epoch 4/25

360/360 1s 2ms/step -

loss: 0.0020

Epoch 5/25

360/360 1s 2ms/step -

loss: 0.0021

Epoch 6/25

360/360 1s 2ms/step -

loss: 0.0020

Epoch 7/25

360/360 1s 2ms/step -

loss: 0.0021

Epoch 8/25

360/360 1s 2ms/step -

loss: 0.0019

Epoch 9/25

360/360 1s 2ms/step -

loss: 0.0022

Epoch 10/25

360/360 1s 2ms/step -

loss: 0.0018

Epoch 11/25

360/360 1s 2ms/step -

loss: 0.0017

Epoch 12/25

360/360 1s 1ms/step -

loss: 0.0020

Epoch 13/25

360/360 1s 2ms/step -

```

loss: 0.0022
Epoch 14/25
360/360          1s 1ms/step -
loss: 0.0019
Epoch 15/25
360/360          1s 1ms/step -
loss: 0.0017
Epoch 16/25
360/360          1s 1ms/step -
loss: 0.0021
Epoch 17/25
360/360          1s 1ms/step -
loss: 0.0021
Epoch 18/25
360/360          1s 1ms/step -
loss: 0.0020
Epoch 19/25
360/360          1s 1ms/step -
loss: 0.0023
Epoch 20/25
360/360          1s 1ms/step -
loss: 0.0023
Epoch 21/25
360/360          1s 1ms/step -
loss: 0.0018
Epoch 22/25
360/360          1s 1ms/step -
loss: 0.0023
Epoch 23/25
360/360          1s 1ms/step -
loss: 0.0019
Epoch 24/25
360/360          1s 2ms/step -
loss: 0.0018
Epoch 25/25
360/360          1s 2ms/step -
loss: 0.0020
7/7              0s 12ms/step
Gelecekteki Fiyat Tahminleri:
[[22398.5  ]
 [22296.781]
 [22259.416]
 [22454.293]
 [23343.756]
 [23243.453]
 [23577.275]
 [23502.238]
 [23387.04 ]

```

[23531.291]  
[23329.557]  
[23227.174]  
[23403.05 ]  
[23218.203]  
[23480.533]  
[23641.809]  
[23674.71 ]  
[23639.84 ]  
[23615.111]  
[23540.223]  
[23376.188]  
[23270.998]  
[23281.873]  
[23284.605]  
[23369.55 ]  
[23319.342]  
[23245.57 ]  
[23249.979]  
[23251.07 ]  
[23289.248]  
[23479.434]  
[23536.045]  
[23566.945]  
[23577.33 ]  
[23587.682]  
[23560.318]  
[23636.326]  
[23658.148]  
[23733.852]  
[23810.648]  
[23860.564]  
[23834.748]  
[23869.453]  
[23784.695]  
[23706.098]  
[23603.559]  
[23645.254]  
[23681.838]  
[23700.053]  
[23649.645]  
[23656.654]  
[23811.914]  
[23739.418]  
[23723.646]  
[23722.172]  
[24595.209]  
[23856.977]

[24619.748]  
[23966.97 ]  
[24068.727]  
[23638.729]  
[24003.91 ]  
[24126.848]  
[24169.371]  
[24352.326]  
[24377.621]  
[24177.627]  
[24138.02 ]  
[24000.844]  
[23970.684]  
[24984.43 ]  
[24115.607]  
[24187.77 ]  
[24213.21 ]  
[24561.465]  
[25061.975]  
[25123.455]  
[25235.727]  
[25213.75 ]  
[25225.018]  
[25272.158]  
[25265.979]  
[25421.371]  
[25375.94 ]  
[25487.58 ]  
[25491.195]  
[25444.1 ]  
[25364.791]  
[25386.824]  
[25535.37 ]  
[25435.213]  
[25385.527]  
[25372.328]  
[25297.115]  
[25197.287]  
[25286.295]  
[25410.457]  
[25281.01 ]  
[25287.955]  
[25323.363]  
[25254.576]  
[25309.654]  
[25342.129]  
[25127.793]  
[24902.135]

[24504.094]  
[25857.479]  
[24988.135]  
[24923.488]  
[24860.79 ]  
[24853.533]  
[24970.82 ]  
[25127.691]  
[25181.5 ]  
[25338.672]  
[25343.715]  
[25367.117]  
[25404.404]  
[25449.229]  
[25631.543]  
[25710.16 ]  
[25825.346]  
[25787.97 ]  
[25771.336]  
[25642.053]  
[25468.703]  
[25456.299]  
[25533.506]  
[25563.82 ]  
[25678.611]  
[25733.629]  
[25775.719]  
[25728.486]  
[25728.182]  
[25695.889]  
[25699.512]  
[25611.975]  
[25693.064]  
[25744.053]  
[25738.83 ]  
[25686.578]  
[25794.088]  
[25851.297]  
[25816.326]  
[25816.658]  
[25655.652]  
[25609.29 ]  
[25422.734]  
[25241.873]  
[25076.246]  
[24817.256]  
[26087.295]  
[24726.951]

[24453.3 ]  
[24523.06 ]  
[24842.635]  
[25068.916]  
[25179.574]  
[25183.36 ]  
[25274.545]  
[24582.74 ]  
[23964.86 ]  
[23837.309]  
[23774.064]  
[24122.44 ]  
[23883.639]  
[23456.344]  
[23821.264]  
[23666.46 ]  
[24673.162]  
[23904.941]  
[24047.078]  
[25003.928]  
[23940.768]  
[24750.49 ]  
[24173.76 ]  
[24478.002]  
[24057.836]  
[24039.863]  
[24103.256]  
[24080.95 ]  
[23922.79 ]  
[23676.326]  
[23916.38 ]  
[23948.422]  
[24167.537]  
[24082.541]  
[23891.357]  
[23931.99 ]  
[23917.955]  
[24014.52 ]  
[23932.953]  
[23684.727]  
[23600.287]  
[23753.803]  
[23609.873]  
[23704.23 ]  
[23648.152]  
[23655.424]  
[23800.592]  
[23872.613]]

72/72                      0s 1ms/step  
RMSE: 0.05041986269688317  
MAE: 0.023680437711474354

ail\_frx kısmı bu kodlar aracılığı ile yapıldı.

```
[5]: import pandas as pd
import numpy as np
import tensorflow as tf
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import MinMaxScaler
from sklearn.metrics import mean_squared_error, mean_absolute_error

veri = pd.read_csv(r"C:/Users/ASUS/Desktop/AI CLUB DATATHON/CSV/ail_frx.csv")

veri['open'] = veri['open'].str.replace('"', '').astype(float)
veri['high'] = veri['high'].str.replace('"', '').astype(float)
veri['low'] = veri['low'].str.replace('"', '').astype(float)
veri['close'] = veri['close'].str.replace('"', '').astype(float)

veri['open'].fillna(veri['open'].mean(), inplace=True)
veri['high'].fillna(veri['high'].mean(), inplace=True)
veri['low'].fillna(veri['low'].mean(), inplace=True)
veri['close'].fillna(veri['close'].mean(), inplace=True)

veri = veri[['open', 'low', 'high', 'close']]

X = veri[['close', 'open', 'high']].values
y = veri['open'].values

scaler = MinMaxScaler()
X_scaled = scaler.fit_transform(X)
y_scaled = scaler.fit_transform(y.reshape(-1,1)).reshape(-1)

X_train, X_test, y_train, y_test = train_test_split(X_scaled, y_scaled,
↳test_size=0.2, random_state=42)

model = tf.keras.models.Sequential([
    tf.keras.layers.Dense(64, activation='relu', input_shape=(X_train.
↳shape[1],)),
    tf.keras.layers.Dense(32, activation='relu'),
    tf.keras.layers.Dense(1)
])

model.compile(optimizer='adam', loss='mean_squared_error')

model.fit(X_scaled, y_scaled, epochs=25, batch_size=32, verbose=1)
```



```

gelecek_veri = X_scaled[-201:]
tahminler = model.predict(gelecek_veri)

tahminler = scaler.inverse_transform(tahminler)

print("Gelecekteki Fiyat Tahminleri:")
print(tahminler)

y_pred = model.predict(X_test)
mse = mean_squared_error(y_test, y_pred)
rmse = np.sqrt(mse)
mae = mean_absolute_error(y_test, y_pred)
print("RMSE:", rmse)
print("MAE:", mae)

```

Epoch 1/25

C:\Users\ASUS\anaconda3\Lib\site-packages\keras\src\layers\core\dense.py:87:  
 UserWarning: Do not pass an `input\_shape`/`input\_dim` argument to a layer. When  
 using Sequential models, prefer using an `Input(shape)` object as the first  
 layer in the model instead.

```
super().__init__(activity_regularizer=activity_regularizer, **kwargs)
```

354/354 3s 2ms/step -

loss: 0.0314

Epoch 2/25

354/354 1s 2ms/step -

loss: 2.5997e-06

Epoch 3/25

354/354 1s 2ms/step -

loss: 2.7128e-06

Epoch 4/25

354/354 1s 2ms/step -

loss: 1.9581e-06

Epoch 5/25

354/354 1s 2ms/step -

loss: 2.0711e-06

Epoch 6/25

354/354 1s 2ms/step -

loss: 2.1813e-06

Epoch 7/25

354/354 1s 2ms/step -

loss: 2.1230e-06

Epoch 8/25

354/354 1s 2ms/step -

loss: 1.9547e-06

Epoch 9/25

```

354/354          1s 2ms/step -
loss: 2.2674e-06
Epoch 10/25
354/354          1s 2ms/step -
loss: 3.5260e-06
Epoch 11/25
354/354          1s 2ms/step -
loss: 2.4798e-06
Epoch 12/25
354/354          1s 2ms/step -
loss: 1.6496e-06
Epoch 13/25
354/354          1s 2ms/step -
loss: 2.5276e-06
Epoch 14/25
354/354          1s 2ms/step -
loss: 2.1326e-06
Epoch 15/25
354/354          1s 2ms/step -
loss: 2.3528e-06
Epoch 16/25
354/354          1s 2ms/step -
loss: 2.2273e-06
Epoch 17/25
354/354          1s 2ms/step -
loss: 1.7572e-06
Epoch 18/25
354/354          1s 2ms/step -
loss: 2.0166e-06
Epoch 19/25
354/354          1s 2ms/step -
loss: 2.8404e-06
Epoch 20/25
354/354          1s 2ms/step -
loss: 3.9821e-06
Epoch 21/25
354/354          1s 2ms/step -
loss: 2.4832e-06
Epoch 22/25
354/354          1s 1ms/step -
loss: 1.6092e-06
Epoch 23/25
354/354          1s 2ms/step -
loss: 1.9422e-06
Epoch 24/25
354/354          1s 2ms/step -
loss: 1.7988e-06
Epoch 25/25

```

354/354 1s 2ms/step -

loss: 5.2469e-06

7/7 0s 11ms/step

Gelecekteki Fiyat Tahminleri:

[14.265686 ]  
[14.264134 ]  
[14.263401 ]  
[14.283654 ]  
[14.2631035]  
[14.2626915]  
[14.261282 ]  
[14.262968 ]  
[14.309646 ]  
[14.266759 ]  
[14.267619 ]  
[14.269239 ]  
[14.272859 ]  
[14.273415 ]  
[14.278101 ]  
[14.280927 ]  
[14.282512 ]  
[14.281318 ]  
[14.282728 ]  
[14.286574 ]  
[14.287751 ]  
[14.280328 ]  
[14.296904 ]  
[14.300331 ]  
[14.299552 ]  
[14.299396 ]  
[14.305658 ]  
[14.311224 ]  
[14.314793 ]  
[14.3160515]  
[14.318283 ]  
[14.318148 ]  
[14.320589 ]  
[14.320487 ]  
[14.31974 ]  
[14.318864 ]  
[14.317104 ]  
[14.319963 ]  
[14.31289 ]  
[14.311619 ]  
[14.316535 ]  
[14.312172 ]  
[14.318509 ]  
[14.339562 ]

[14.361766 ]  
[14.362633 ]  
[14.363131 ]  
[14.365618 ]  
[14.36849 ]  
[14.367969 ]  
[14.364935 ]  
[14.368728 ]  
[14.368736 ]  
[14.368006 ]  
[14.377705 ]  
[14.371225 ]  
[14.373147 ]  
[14.373659 ]  
[14.375245 ]  
[14.377011 ]  
[14.361865 ]  
[14.359163 ]  
[14.351724 ]  
[14.357038 ]  
[14.380824 ]  
[14.391841 ]  
[14.393525 ]  
[14.393555 ]  
[14.391149 ]  
[14.390387 ]  
[14.389868 ]  
[14.389391 ]  
[14.39129 ]  
[14.389207 ]  
[14.395382 ]  
[14.390063 ]  
[14.389803 ]  
[14.38999 ]  
[14.389649 ]  
[14.391044 ]  
[14.416995 ]  
[14.39312 ]  
[14.394411 ]  
[14.396838 ]  
[14.392902 ]  
[14.393635 ]  
[14.396023 ]  
[14.404543 ]  
[14.4111395]  
[14.41001 ]  
[14.408494 ]  
[14.4072485]

[14.40841 ]  
[14.409466 ]  
[14.409305 ]  
[14.4084015]  
[14.408863 ]  
[14.409138 ]  
[14.409238 ]  
[14.409391 ]  
[14.409852 ]  
[14.4096365]  
[14.411124 ]  
[14.413535 ]  
[14.414159 ]  
[14.418303 ]  
[14.415914 ]  
[14.483509 ]  
[14.418657 ]  
[14.409276 ]  
[14.416183 ]  
[14.445926 ]  
[14.435576 ]  
[14.458225 ]  
[14.495775 ]  
[14.45127 ]  
[14.496017 ]  
[14.446349 ]  
[14.429961 ]  
[14.431455 ]  
[14.430305 ]  
[14.431639 ]  
[14.431279 ]  
[14.431582 ]  
[14.431098 ]  
[14.432488 ]  
[14.433274 ]  
[14.43289 ]  
[14.432374 ]  
[14.432464 ]  
[14.429837 ]  
[14.42838 ]  
[14.427289 ]  
[14.42974 ]  
[14.438323 ]  
[14.453854 ]  
[14.445852 ]  
[14.446012 ]  
[14.4451275]  
[14.4458275]

[14.445273 ]  
[14.444143 ]  
[14.445065 ]  
[14.446023 ]  
[14.445467 ]  
[14.446286 ]  
[14.4453745]  
[14.444784 ]  
[14.445197 ]  
[14.442513 ]  
[14.444031 ]  
[14.446415 ]  
[14.446733 ]  
[14.45224 ]  
[14.461165 ]  
[14.472392 ]  
[14.480552 ]  
[14.479935 ]  
[14.54758 ]  
[14.57854 ]  
[14.521246 ]  
[14.490207 ]  
[14.536542 ]  
[14.484169 ]  
[14.48731 ]  
[14.484205 ]  
[14.484649 ]  
[14.48595 ]  
[14.48663 ]  
[14.48737 ]  
[14.487092 ]  
[14.485684 ]  
[14.485431 ]  
[14.483172 ]  
[14.472586 ]  
[14.466604 ]  
[14.456683 ]  
[14.477485 ]  
[14.506206 ]  
[14.527969 ]  
[14.526661 ]  
[14.527497 ]  
[14.52143 ]  
[14.542024 ]  
[14.523696 ]  
[14.523769 ]  
[14.523716 ]  
[14.523457 ]

```
[14.524108 ]
[14.524186 ]
[14.524357 ]
[14.525089 ]
[14.524102 ]
[14.523304 ]
[14.519765 ]
[14.524507 ]
[14.523908 ]
[14.535981 ]
[14.535927 ]
[14.540493 ]
[14.541479 ]]
71/71          0s 1ms/step
RMSE: 0.0011945772846795094
MAE: 0.0007662062744950194
```

```
[6]: import pandas as pd
import numpy as np
import tensorflow as tf
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import MinMaxScaler
from sklearn.metrics import mean_squared_error, mean_absolute_error

veri = pd.read_csv(r"C:/Users/ASUS/Desktop/AI CLUB DATATHON/CSV/ail_frx.csv")

veri['open'] = veri['open'].str.replace('"', '').astype(float)
veri['high'] = veri['high'].str.replace('"', '').astype(float)
veri['low'] = veri['low'].str.replace('"', '').astype(float)
veri['close'] = veri['close'].str.replace('"', '').astype(float)

veri['open'].fillna(veri['open'].mean(), inplace=True)
veri['high'].fillna(veri['high'].mean(), inplace=True)
veri['low'].fillna(veri['low'].mean(), inplace=True)
veri['close'].fillna(veri['close'].mean(), inplace=True)

veri = veri[['open', 'low', 'high', 'close']]

X = veri[['open', 'low', 'high']].values
y = veri['close'].values

scaler = MinMaxScaler()
X_scaled = scaler.fit_transform(X)
y_scaled = scaler.fit_transform(y.reshape(-1,1)).reshape(-1)
```

```

X_train, X_test, y_train, y_test = train_test_split(X_scaled, y_scaled,
↳test_size=0.2, random_state=42)

model = tf.keras.models.Sequential([
    tf.keras.layers.Dense(64, activation='relu', input_shape=(X_train.
↳shape[1],)),
    tf.keras.layers.Dense(32, activation='relu'),
    tf.keras.layers.Dense(1)
])

model.compile(optimizer='adam', loss='mean_squared_error')

model.fit(X_scaled, y_scaled, epochs=40, batch_size=32, verbose=1)

gelecek_veri = X_scaled[-201:]
tahminler = model.predict(gelecek_veri)

tahminler = scaler.inverse_transform(tahminler)

print("Gelecekteki Fiyat Tahminleri:")
print(tahminler)

y_pred = model.predict(X_test)
mse = mean_squared_error(y_test, y_pred)
rmse = np.sqrt(mse)
mae = mean_absolute_error(y_test, y_pred)
print("RMSE:", rmse)
print("MAE:", mae)

```

Epoch 1/40

C:\Users\ASUS\anaconda3\Lib\site-packages\keras\src\layers\core\dense.py:87:  
UserWarning: Do not pass an `input\_shape`/`input\_dim` argument to a layer. When  
using Sequential models, prefer using an `Input(shape)` object as the first  
layer in the model instead.

```
super().__init__(activity_regularizer=activity_regularizer, **kwargs)
```

354/354 3s 2ms/step -

loss: 0.0096

Epoch 2/40

354/354 1s 2ms/step -

loss: 1.3635e-06

Epoch 3/40

354/354 1s 2ms/step -

loss: 1.4124e-06



Epoch 4/40  
 354/354 1s 2ms/step -  
 loss: 1.0876e-06  
 Epoch 5/40  
 354/354 1s 2ms/step -  
 loss: 1.3421e-06  
 Epoch 6/40  
 354/354 1s 2ms/step -  
 loss: 1.2658e-06  
 Epoch 7/40  
 354/354 1s 2ms/step -  
 loss: 8.4336e-07  
 Epoch 8/40  
 354/354 1s 2ms/step -  
 loss: 8.8987e-07  
 Epoch 9/40  
 354/354 1s 2ms/step -  
 loss: 6.6911e-07  
 Epoch 10/40  
 354/354 1s 2ms/step -  
 loss: 5.3682e-07  
 Epoch 11/40  
 354/354 1s 2ms/step -  
 loss: 6.5779e-07  
 Epoch 12/40  
 354/354 1s 2ms/step -  
 loss: 7.0393e-07  
 Epoch 13/40  
 354/354 1s 2ms/step -  
 loss: 6.4482e-07  
 Epoch 14/40  
 354/354 1s 2ms/step -  
 loss: 4.2942e-07  
 Epoch 15/40  
 354/354 1s 2ms/step -  
 loss: 5.0371e-07  
 Epoch 16/40  
 354/354 1s 2ms/step -  
 loss: 9.2690e-07  
 Epoch 17/40  
 354/354 1s 2ms/step -  
 loss: 1.0376e-06  
 Epoch 18/40  
 354/354 1s 2ms/step -  
 loss: 1.8039e-06  
 Epoch 19/40  
 354/354 1s 2ms/step -  
 loss: 1.1519e-06

```

Epoch 20/40
354/354          1s 2ms/step -
loss: 3.5819e-06
Epoch 21/40
354/354          1s 2ms/step -
loss: 1.7386e-06
Epoch 22/40
354/354          1s 2ms/step -
loss: 1.0832e-06
Epoch 23/40
354/354          1s 2ms/step -
loss: 4.2494e-06
Epoch 24/40
354/354          1s 2ms/step -
loss: 5.6845e-07
Epoch 25/40
354/354          1s 2ms/step -
loss: 1.4131e-06
Epoch 26/40
354/354          1s 2ms/step -
loss: 1.4671e-06
Epoch 27/40
354/354          1s 2ms/step -
loss: 1.2906e-06
Epoch 28/40
354/354          1s 2ms/step -
loss: 9.4586e-07
Epoch 29/40
354/354          1s 2ms/step -
loss: 4.0270e-07
Epoch 30/40
354/354          1s 2ms/step -
loss: 2.2781e-06
Epoch 31/40
354/354          1s 2ms/step -
loss: 4.4542e-07
Epoch 32/40
354/354          1s 2ms/step -
loss: 2.1125e-06
Epoch 33/40
354/354          1s 2ms/step -
loss: 1.2365e-06
Epoch 34/40
354/354          1s 2ms/step -
loss: 2.2970e-07
Epoch 35/40
354/354          1s 2ms/step -
loss: 2.1297e-06

```

Epoch 36/40  
354/354 1s 2ms/step -  
loss: 2.3426e-06  
Epoch 37/40  
354/354 1s 2ms/step -  
loss: 3.7339e-07  
Epoch 38/40  
354/354 1s 2ms/step -  
loss: 1.4670e-06  
Epoch 39/40  
354/354 1s 2ms/step -  
loss: 3.9501e-07  
Epoch 40/40  
354/354 1s 1ms/step -  
loss: 1.3417e-06  
7/7 0s 10ms/step

Gelecekteki Fiyat Tahminleri:

[[14.37092 ]  
[14.36142 ]  
[14.363637 ]  
[14.398416 ]  
[14.356109 ]  
[14.352097 ]  
[14.351755 ]  
[14.352729 ]  
[14.477439 ]  
[14.356011 ]  
[14.356479 ]  
[14.362084 ]  
[14.362558 ]  
[14.365112 ]  
[14.368308 ]  
[14.370106 ]  
[14.370973 ]  
[14.370783 ]  
[14.371469 ]  
[14.377347 ]  
[14.379084 ]  
[14.387599 ]  
[14.388406 ]  
[14.388657 ]  
[14.389968 ]  
[14.391331 ]  
[14.400975 ]  
[14.40249 ]  
[14.408512 ]  
[14.40699 ]  
[14.407174 ]

[14.408674 ]  
[14.409769 ]  
[14.41008 ]  
[14.409033 ]  
[14.408136 ]  
[14.410127 ]  
[14.409014 ]  
[14.403575 ]  
[14.400427 ]  
[14.414823 ]  
[14.404262 ]  
[14.419423 ]  
[14.452425 ]  
[14.455838 ]  
[14.452818 ]  
[14.453421 ]  
[14.455773 ]  
[14.457028 ]  
[14.457166 ]  
[14.45724 ]  
[14.458043 ]  
[14.458874 ]  
[14.457919 ]  
[14.484898 ]  
[14.4625225]  
[14.464486 ]  
[14.463327 ]  
[14.465419 ]  
[14.466299 ]  
[14.451173 ]  
[14.44933 ]  
[14.450191 ]  
[14.452917 ]  
[14.484275 ]  
[14.482959 ]  
[14.485825 ]  
[14.486707 ]  
[14.48096 ]  
[14.479844 ]  
[14.479516 ]  
[14.479736 ]  
[14.484439 ]  
[14.479234 ]  
[14.493387 ]  
[14.479494 ]  
[14.479502 ]  
[14.479248 ]  
[14.479599 ]

[14.483168 ]  
[14.546634 ]  
[14.482888 ]  
[14.485623 ]  
[14.486433 ]  
[14.4826145]  
[14.482985 ]  
[14.489262 ]  
[14.504268 ]  
[14.502048 ]  
[14.498484 ]  
[14.498242 ]  
[14.497439 ]  
[14.497676 ]  
[14.498901 ]  
[14.499404 ]  
[14.497835 ]  
[14.498845 ]  
[14.499626 ]  
[14.499127 ]  
[14.500405 ]  
[14.499769 ]  
[14.499552 ]  
[14.502291 ]  
[14.502725 ]  
[14.507642 ]  
[14.507665 ]  
[14.506074 ]  
[14.678766 ]  
[14.508283 ]  
[14.506937 ]  
[14.511333 ]  
[14.602653 ]  
[14.534112 ]  
[14.57933 ]  
[14.587334 ]  
[14.575702 ]  
[14.653853 ]  
[14.567198 ]  
[14.520914 ]  
[14.522655 ]  
[14.520396 ]  
[14.52156 ]  
[14.521769 ]  
[14.521181 ]  
[14.521222 ]  
[14.524241 ]  
[14.523802 ]

[14.522474 ]  
[14.522286 ]  
[14.521803 ]  
[14.521521 ]  
[14.51876 ]  
[14.518377 ]  
[14.530593 ]  
[14.53433 ]  
[14.5607395]  
[14.535752 ]  
[14.536734 ]  
[14.535107 ]  
[14.535716 ]  
[14.53538 ]  
[14.534724 ]  
[14.536947 ]  
[14.5357065]  
[14.535623 ]  
[14.53747 ]  
[14.534415 ]  
[14.535145 ]  
[14.535058 ]  
[14.533175 ]  
[14.533653 ]  
[14.538173 ]  
[14.537494 ]  
[14.551138 ]  
[14.580138 ]  
[14.57268 ]  
[14.572558 ]  
[14.5722065]  
[14.748278 ]  
[14.776507 ]  
[14.678849 ]  
[14.584242 ]  
[14.707205 ]  
[14.576754 ]  
[14.583414 ]  
[14.575291 ]  
[14.575097 ]  
[14.577669 ]  
[14.577321 ]  
[14.579139 ]  
[14.57815 ]  
[14.576984 ]  
[14.575767 ]  
[14.573488 ]  
[14.565437 ]

```

[14.561133 ]
[14.587017 ]
[14.594309 ]
[14.620886 ]
[14.61996  ]
[14.620692 ]
[14.618488 ]
[14.617715 ]
[14.666649 ]
[14.61486  ]
[14.614808 ]
[14.615795 ]
[14.614849 ]
[14.6158905]
[14.61534  ]
[14.615852 ]
[14.616315 ]
[14.614916 ]
[14.616552 ]
[14.615568 ]
[14.615507 ]
[14.615392 ]
[14.645705 ]
[14.654919 ]
[14.637547 ]
[14.633227 ]]
71/71          0s 815us/step
RMSE: 0.007185994284786264
MAE: 0.006429869016247972

```

```

[7]: import pandas as pd
import numpy as np
import tensorflow as tf
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import MinMaxScaler
from sklearn.metrics import mean_squared_error, mean_absolute_error

veri = pd.read_csv(r"C:/Users/ASUS/Desktop/AI CLUB DATATHON/CSV/ail_frx.csv")

veri['open'] = veri['open'].str.replace('"', '').astype(float)
veri['high'] = veri['high'].str.replace('"', '').astype(float)
veri['low'] = veri['low'].str.replace('"', '').astype(float)
veri['close'] = veri['close'].str.replace('"', '').astype(float)

veri['open'].fillna(veri['open'].mean(), inplace=True)
veri['high'].fillna(veri['high'].mean(), inplace=True)
veri['low'].fillna(veri['low'].mean(), inplace=True)

```

```

veri['close'].fillna(veri['close'].mean(), inplace=True)

veri = veri[['open', 'low', 'high', 'close']]

X = veri[['close', 'open', 'high']].values
y = veri['low'].values

scaler = MinMaxScaler()
X_scaled = scaler.fit_transform(X)
y_scaled = scaler.fit_transform(y.reshape(-1,1)).reshape(-1)

X_train, X_test, y_train, y_test = train_test_split(X_scaled, y_scaled,
    ↪test_size=0.2, random_state=42)

model = tf.keras.models.Sequential([
    tf.keras.layers.Dense(64, activation='relu', input_shape=(X_train.
    ↪shape[1],)),
    tf.keras.layers.Dense(32, activation='relu'),
    tf.keras.layers.Dense(1)
])

model.compile(optimizer='adam', loss='mean_squared_error')

model.fit(X_scaled, y_scaled, epochs=25, batch_size=32, verbose=1)

gelecek_veri = X_scaled[-201:]
tahminler = model.predict(gelecek_veri)

tahminler = scaler.inverse_transform(tahminler)

print("Gelecekteki Fiyat Tahminleri:")
print(tahminler)

y_pred = model.predict(X_test)
mse = mean_squared_error(y_test, y_pred)
rmse = np.sqrt(mse)
mae = mean_absolute_error(y_test, y_pred)
print("RMSE:", rmse)
print("MAE:", mae)

```

Epoch 1/25

C:\Users\ASUS\anaconda3\Lib\site-packages\keras\src\layers\core\dense.py:87:  
 UserWarning: Do not pass an `input\_shape`/`input\_dim` argument to a layer. When  
 using Sequential models, prefer using an `Input(shape)` object as the first



layer in the model instead.

```
super().__init__(activity_regularizer=activity_regularizer, **kwargs)

354/354          3s 2ms/step -
loss: 0.0234
Epoch 2/25
354/354          1s 2ms/step -
loss: 9.4186e-06
Epoch 3/25
354/354          1s 2ms/step -
loss: 7.2204e-06
Epoch 4/25
354/354          1s 2ms/step -
loss: 6.4341e-06
Epoch 5/25
354/354          1s 2ms/step -
loss: 6.9352e-06
Epoch 6/25
354/354          1s 2ms/step -
loss: 8.1289e-06
Epoch 7/25
354/354          1s 2ms/step -
loss: 6.6905e-06
Epoch 8/25
354/354          1s 2ms/step -
loss: 7.2243e-06
Epoch 9/25
354/354          1s 2ms/step -
loss: 6.0264e-06
Epoch 10/25
354/354          1s 2ms/step -
loss: 6.4109e-06
Epoch 11/25
354/354          1s 2ms/step -
loss: 7.2272e-06
Epoch 12/25
354/354          1s 2ms/step -
loss: 5.9459e-06
Epoch 13/25
354/354          1s 2ms/step -
loss: 6.7903e-06
Epoch 14/25
354/354          1s 2ms/step -
loss: 7.1161e-06
Epoch 15/25
354/354          1s 2ms/step -
loss: 7.8001e-06
Epoch 16/25
```

```

354/354          1s 2ms/step -
loss: 1.0129e-05
Epoch 17/25
354/354          1s 2ms/step -
loss: 8.5468e-06
Epoch 18/25
354/354          1s 2ms/step -
loss: 9.7683e-06
Epoch 19/25
354/354          1s 2ms/step -
loss: 6.2865e-06
Epoch 20/25
354/354          1s 2ms/step -
loss: 7.9681e-06
Epoch 21/25
354/354          1s 2ms/step -
loss: 6.0829e-06
Epoch 22/25
354/354          1s 2ms/step -
loss: 9.2659e-06
Epoch 23/25
354/354          1s 2ms/step -
loss: 8.6202e-06
Epoch 24/25
354/354          1s 2ms/step -
loss: 7.0496e-06
Epoch 25/25
354/354          1s 2ms/step -
loss: 7.1247e-06
7/7              0s 13ms/step
Gelecekteki Fiyat Tahminleri:
[[14.245492 ]
 [14.244752 ]
 [14.244617 ]
 [14.260704 ]
 [14.244955 ]
 [14.244984 ]
 [14.243921 ]
 [14.245583 ]
 [14.277243 ]
 [14.249397 ]
 [14.250165 ]
 [14.251717 ]
 [14.255325 ]
 [14.256014 ]
 [14.260753 ]
 [14.263605 ]
 [14.264639 ]

```

[14.263676 ]  
[14.265212 ]  
[14.221781 ]  
[14.267796 ]  
[14.2625 ]  
[14.279585 ]  
[14.282902 ]  
[14.281944 ]  
[14.282043 ]  
[14.288142 ]  
[14.293682 ]  
[14.296921 ]  
[14.298638 ]  
[14.300764 ]  
[14.30082 ]  
[14.303091 ]  
[14.302999 ]  
[14.302261 ]  
[14.301406 ]  
[14.2520895]  
[14.301395 ]  
[14.29517 ]  
[14.294153 ]  
[14.297182 ]  
[14.293831 ]  
[14.300293 ]  
[14.321107 ]  
[14.343628 ]  
[14.29673 ]  
[14.297234 ]  
[14.348319 ]  
[14.302365 ]  
[14.350209 ]  
[14.347521 ]  
[14.351366 ]  
[14.35123 ]  
[14.350514 ]  
[14.357326 ]  
[14.35381 ]  
[14.355614 ]  
[14.307557 ]  
[14.357907 ]  
[14.357887 ]  
[14.343611 ]  
[14.339692 ]  
[14.33396 ]  
[14.33783 ]  
[14.362902 ]

[14.374081 ]  
[14.375165 ]  
[14.374935 ]  
[14.373369 ]  
[14.372967 ]  
[14.372306 ]  
[14.371986 ]  
[14.373256 ]  
[14.371862 ]  
[14.376369 ]  
[14.372649 ]  
[14.3723755]  
[14.372553 ]  
[14.372229 ]  
[14.373706 ]  
[14.391855 ]  
[14.37568 ]  
[14.377093 ]  
[14.378947 ]  
[14.3754635]  
[14.326288 ]  
[14.378501 ]  
[14.385924 ]  
[14.393429 ]  
[14.3408375]  
[14.390923 ]  
[14.38987 ]  
[14.391066 ]  
[14.340555 ]  
[14.391903 ]  
[14.391075 ]  
[14.391309 ]  
[14.340282 ]  
[14.391819 ]  
[14.391965 ]  
[14.392368 ]  
[14.392275 ]  
[14.393826 ]  
[14.396051 ]  
[14.396786 ]  
[14.400792 ]  
[14.398482 ]  
[14.443202 ]  
[14.400525 ]  
[14.391491 ]  
[14.397818 ]  
[14.418255 ]  
[14.417663 ]

[14.441353 ]  
[14.471614 ]  
[14.4303 ]  
[14.462458 ]  
[14.423309 ]  
[14.412408 ]  
[14.413685 ]  
[14.412933 ]  
[14.414119 ]  
[14.413848 ]  
[14.414157 ]  
[14.413817 ]  
[14.415073 ]  
[14.415735 ]  
[14.415449 ]  
[14.414965 ]  
[14.414712 ]  
[14.411559 ]  
[14.410741 ]  
[14.40756 ]  
[14.359483 ]  
[14.4208355]  
[14.433473 ]  
[14.428318 ]  
[14.428353 ]  
[14.427469 ]  
[14.428238 ]  
[14.427742 ]  
[14.426618 ]  
[14.370765 ]  
[14.428617 ]  
[14.428111 ]  
[14.428742 ]  
[14.370719 ]  
[14.427348 ]  
[14.427864 ]  
[14.425084 ]  
[14.426619 ]  
[14.372066 ]  
[14.42909 ]  
[14.433265 ]  
[14.442803 ]  
[14.454555 ]  
[14.46312 ]  
[14.401499 ]  
[14.531583 ]  
[14.47449 ]  
[14.49105 ]

```
[14.472423 ]
[14.502382 ]
[14.466605 ]
[14.468961 ]
[14.46705 ]
[14.46751 ]
[14.468765 ]
[14.469529 ]
[14.470069 ]
[14.46971 ]
[14.468418 ]
[14.466985 ]
[14.4640045]
[14.454479 ]
[14.444294 ]
[14.433194 ]
[14.459487 ]
[14.4884815]
[14.510889 ]
[14.509279 ]
[14.510408 ]
[14.440607 ]
[14.518515 ]
[14.506907 ]
[14.506943 ]
[14.506952 ]
[14.506682 ]
[14.507342 ]
[14.507402 ]
[14.507675 ]
[14.508214 ]
[14.507296 ]
[14.5062275]
[14.439 ]
[14.507622 ]
[14.507055 ]
[14.514324 ]
[14.4498625]
[14.523104 ]
[14.524891 ]]
```

71/71 0s 2ms/step

RMSE: 0.00261864428211394

MAE: 0.0016339600934450968

```
[8]: import pandas as pd
import numpy as np
import tensorflow as tf
```

```

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

veri = pd.read_csv(r"C:/Users/ASUS/Desktop/AI CLUB DATATHON/CSV/ail_fr.x.csv")

veri['open'] = veri['open'].str.replace('\"', '').astype(float)
veri['high'] = veri['high'].str.replace('\"', '').astype(float)
veri['low'] = veri['low'].str.replace('\"', '').astype(float)
veri['close'] = veri['close'].str.replace('\"', '').astype(float)

veri['open'].fillna(veri['open'].mean(), inplace=True)
veri['high'].fillna(veri['high'].mean(), inplace=True)
veri['low'].fillna(veri['low'].mean(), inplace=True)
veri['close'].fillna(veri['close'].mean(), inplace=True)

veri = veri[['open', 'low', 'high', 'close']]

X = veri[['close', 'open', 'high']].values
y = veri['close'].values

scaler = MinMaxScaler()
X_scaled = scaler.fit_transform(X)
y_scaled = scaler.fit_transform(y.reshape(-1,1)).reshape(-1)

X_train, X_test, y_train, y_test = train_test_split(X_scaled, y_scaled,
↳test_size=0.2, random_state=42)

model = tf.keras.models.Sequential([
    tf.keras.layers.Dense(64, activation='relu', input_shape=(X_train.
↳shape[1],)),
    tf.keras.layers.Dense(32, activation='relu'),
    tf.keras.layers.Dense(1)
])

model.compile(optimizer='adam', loss='mean_squared_error')

```

```

model.fit(X_scaled, y_scaled, epochs=25, batch_size=32, verbose=1)

gelecek_veri = X_scaled[-201:]
tahminler = model.predict(gelecek_veri)

tahminler = scaler.inverse_transform(tahminler)

print("Gelecekteki Fiyat Tahminleri:")
print(tahminler)

y_pred = model.predict(X_test)
mse = mean_squared_error(y_test, y_pred)
rmse = np.sqrt(mse)
mae = mean_absolute_error(y_test, y_pred)
print("RMSE:", rmse)
print("MAE:", mae)

```

Epoch 1/25

C:\Users\ASUS\anaconda3\Lib\site-packages\keras\src\layers\core\dense.py:87:  
 UserWarning: Do not pass an `input\_shape`/`input\_dim` argument to a layer. When  
 using Sequential models, prefer using an `Input(shape)` object as the first  
 layer in the model instead.

```
super().__init__(activity_regularizer=activity_regularizer, **kwargs)
```

354/354                    3s 2ms/step -

loss: 0.0459

Epoch 2/25

354/354                    1s 2ms/step -

loss: 2.0051e-06

Epoch 3/25

354/354                    1s 2ms/step -

loss: 3.4334e-07

Epoch 4/25

354/354                    1s 2ms/step -

loss: 2.3760e-07

Epoch 5/25

354/354                    1s 2ms/step -

loss: 1.7649e-07

Epoch 6/25

354/354                    1s 2ms/step -

loss: 1.5046e-07

Epoch 7/25

354/354                    1s 2ms/step -

loss: 1.8930e-07



```

Epoch 8/25
354/354          1s 2ms/step -
loss: 2.0901e-07
Epoch 9/25
354/354          1s 2ms/step -
loss: 1.6161e-07
Epoch 10/25
354/354          1s 2ms/step -
loss: 1.2695e-06
Epoch 11/25
354/354          1s 2ms/step -
loss: 2.6158e-07
Epoch 12/25
354/354          1s 2ms/step -
loss: 2.6054e-07
Epoch 13/25
354/354          1s 2ms/step -
loss: 3.0494e-06
Epoch 14/25
354/354          1s 2ms/step -
loss: 3.0851e-07
Epoch 15/25
354/354          1s 2ms/step -
loss: 3.0842e-07
Epoch 16/25
354/354          1s 2ms/step -
loss: 2.1421e-06
Epoch 17/25
354/354          1s 2ms/step -
loss: 5.3798e-08
Epoch 18/25
354/354          1s 2ms/step -
loss: 2.0522e-06
Epoch 19/25
354/354          1s 2ms/step -
loss: 8.8629e-07
Epoch 20/25
354/354          1s 2ms/step -
loss: 4.2907e-06
Epoch 21/25
354/354          1s 2ms/step -
loss: 1.7286e-06
Epoch 22/25
354/354          1s 2ms/step -
loss: 1.3526e-06
Epoch 23/25
354/354          1s 2ms/step -
loss: 1.3628e-05

```

```
Epoch 24/25
354/354          1s 2ms/step -
loss: 1.2759e-07
Epoch 25/25
354/354          1s 2ms/step -
loss: 1.4876e-07
7/7              0s 12ms/step
Gelecekteki Fiyat Tahminleri:
[[14.245327 ]
 [14.242742 ]
 [14.25019  ]
 [14.248877 ]
 [14.248027 ]
 [14.247018 ]
 [14.249766 ]
 [14.251161 ]
 [14.24859  ]
 [14.254551 ]
 [14.254421 ]
 [14.25936  ]
 [14.258871 ]
 [14.263444 ]
 [14.266914 ]
 [14.269058 ]
 [14.264736 ]
 [14.2668915]
 [14.268669 ]
 [ 9.937164 ]
 [14.254604 ]
 [14.282385 ]
 [14.287262 ]
 [14.286259 ]
 [14.2862215]
 [14.289896 ]
 [14.298087 ]
 [14.299222 ]
 [14.302453 ]
 [14.304588 ]
 [14.303806 ]
 [14.306985 ]
 [14.306754 ]
 [14.306972 ]
 [14.30582  ]
 [14.30519  ]
 [ 9.937077 ]
 [14.296471 ]
 [14.298399 ]
 [14.297739 ]
```

[14.29599 ]  
[14.2936735]  
[14.3098545]  
[14.341825 ]  
[14.346879 ]  
[ 9.937212 ]  
[ 9.937205 ]  
[14.353946 ]  
[ 9.93728 ]  
[14.351195 ]  
[14.354583 ]  
[14.355508 ]  
[14.354511 ]  
[14.353707 ]  
[14.3562 ]  
[14.359046 ]  
[14.358816 ]  
[ 9.937238 ]  
[14.363243 ]  
[14.347876 ]  
[14.340597 ]  
[14.328246 ]  
[14.344483 ]  
[14.334461 ]  
[14.376925 ]  
[14.376521 ]  
[14.374313 ]  
[14.373195 ]  
[14.374669 ]  
[14.376617 ]  
[14.375126 ]  
[14.376198 ]  
[14.376116 ]  
[14.376818 ]  
[14.376522 ]  
[14.376175 ]  
[14.376336 ]  
[14.375905 ]  
[14.376289 ]  
[14.380788 ]  
[14.3778305]  
[14.379567 ]  
[14.383424 ]  
[14.378994 ]  
[14.379214 ]  
[ 9.937256 ]  
[14.385197 ]  
[14.390794 ]

[14.395933 ]  
[ 9.937296 ]  
[14.393457 ]  
[14.394428 ]  
[14.394912 ]  
[ 9.937254 ]  
[14.395618 ]  
[14.395222 ]  
[14.394082 ]  
[ 9.937246 ]  
[14.39561 ]  
[14.396006 ]  
[14.395873 ]  
[14.39651 ]  
[14.399791 ]  
[14.398732 ]  
[14.404558 ]  
[14.403137 ]  
[14.402344 ]  
[14.394735 ]  
[14.398231 ]  
[14.400667 ]  
[14.3994665]  
[14.411894 ]  
[14.425948 ]  
[14.480757 ]  
[14.417612 ]  
[14.44209 ]  
[14.41267 ]  
[14.4156065]  
[14.41619 ]  
[14.4159155]  
[14.417256 ]  
[14.417037 ]  
[14.417921 ]  
[14.417455 ]  
[14.4186535]  
[14.420136 ]  
[14.418953 ]  
[14.418623 ]  
[14.418788 ]  
[14.415086 ]  
[14.410498 ]  
[14.4131155]  
[14.394306 ]  
[ 9.936829 ]  
[14.430076 ]  
[14.431245 ]

[14.430923 ]  
[14.430892 ]  
[14.429256 ]  
[14.430428 ]  
[14.430557 ]  
[14.429913 ]  
[ 9.937212 ]  
[14.432052 ]  
[14.432179 ]  
[14.430978 ]  
[ 9.937283 ]  
[14.43122 ]  
[14.432046 ]  
[14.42941 ]  
[14.4299 ]  
[ 9.937174 ]  
[14.431658 ]  
[14.433568 ]  
[14.467821 ]  
[14.465318 ]  
[14.466438 ]  
[ 9.937229 ]  
[14.682763 ]  
[ 9.948185 ]  
[14.46753 ]  
[14.474325 ]  
[14.464332 ]  
[14.468856 ]  
[14.469473 ]  
[14.470972 ]  
[14.470945 ]  
[14.4731045]  
[14.472972 ]  
[14.473156 ]  
[14.471646 ]  
[14.471724 ]  
[14.45963 ]  
[14.450654 ]  
[14.453568 ]  
[14.411774 ]  
[14.431269 ]  
[14.485884 ]  
[14.514363 ]  
[14.512608 ]  
[14.5113125]  
[14.511154 ]  
[ 9.937832 ]  
[14.508514 ]

```

[14.510259 ]
[14.510249 ]
[14.51153  ]
[14.510822 ]
[14.510973 ]
[14.510752 ]
[14.5117445]
[14.510882 ]
[14.510395 ]
[14.510301 ]
[ 9.937806 ]
[14.510342 ]
[14.510727 ]
[14.500871 ]
[ 9.936631 ]
[14.527299 ]
[14.529329 ]]
71/71          0s 1ms/step
RMSE: 0.0001340586521505379
MAE: 0.00010264055692906284

```

crp\_ail kısmı bu kodlar aracılığı ile yapıldı.

```

[9]: import pandas as pd
import numpy as np
import tensorflow as tf
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import MinMaxScaler
from sklearn.metrics import mean_squared_error, mean_absolute_error

veri = pd.read_csv(r"C:/Users/ASUS/Desktop/AI CLUB DATATHON/CSV/crp_ail.csv")

veri['open'] = veri['open'].str.replace('"', '').astype(float)
veri['high'] = veri['high'].str.replace('"', '').astype(float)
veri['low'] = veri['low'].str.replace('"', '').astype(float)
veri['close'] = veri['close'].str.replace('"', '').astype(float)

veri['open'].fillna(veri['open'].mean(), inplace=True)
veri['high'].fillna(veri['high'].mean(), inplace=True)
veri['low'].fillna(veri['low'].mean(), inplace=True)
veri['close'].fillna(veri['close'].mean(), inplace=True)

veri = veri[['open', 'low', 'high', 'close']]

X = veri[['close', 'low', 'high']].values

```

```

y = veri['open'].values

scaler = MinMaxScaler()
X_scaled = scaler.fit_transform(X)
y_scaled = scaler.fit_transform(y.reshape(-1,1)).reshape(-1)

X_train, X_test, y_train, y_test = train_test_split(X_scaled, y_scaled,
    ↪test_size=0.2, random_state=42)

model = tf.keras.models.Sequential([
    tf.keras.layers.Dense(64, activation='relu', input_shape=(X_train.
    ↪shape[1],)),
    tf.keras.layers.Dense(32, activation='relu'),
    tf.keras.layers.Dense(1)
])

model.compile(optimizer='adam', loss='mean_squared_error')

model.fit(X_scaled, y_scaled, epochs=20, batch_size=32, verbose=1)

gelecek_veri = X_scaled[-201:]
tahminler = model.predict(gelecek_veri)

tahminler = scaler.inverse_transform(tahminler)

print("Gelecekteki Fiyat Tahminleri:")
print(tahminler)

y_pred = model.predict(X_test)
mse = mean_squared_error(y_test, y_pred)
rmse = np.sqrt(mse)
mae = mean_absolute_error(y_test, y_pred)
print("RMSE:", rmse)
print("MAE:", mae)

```

Epoch 1/20

C:\Users\ASUS\anaconda3\Lib\site-packages\keras\src\layers\core\dense.py:87:  
UserWarning: Do not pass an `input\_shape`/`input\_dim` argument to a layer. When  
using Sequential models, prefer using an `Input(shape)` object as the first  
layer in the model instead.

```
super().__init__(activity_regularizer=activity_regularizer, **kwargs)
```

346/346                      3s 2ms/step -

loss: 0.0134

Epoch 2/20

346/346                      1s 2ms/step -

```

loss: 0.0032
Epoch 3/20
346/346          1s 2ms/step -
loss: 0.0035
Epoch 4/20
346/346          1s 2ms/step -
loss: 0.0031
Epoch 5/20
346/346          1s 2ms/step -
loss: 0.0032
Epoch 6/20
346/346          1s 2ms/step -
loss: 0.0031
Epoch 7/20
346/346          1s 2ms/step -
loss: 0.0034
Epoch 8/20
346/346          1s 2ms/step -
loss: 0.0037
Epoch 9/20
346/346          1s 2ms/step -
loss: 0.0037
Epoch 10/20
346/346          1s 2ms/step -
loss: 0.0034
Epoch 11/20
346/346          1s 2ms/step -
loss: 0.0034
Epoch 12/20
346/346          1s 2ms/step -
loss: 0.0033
Epoch 13/20
346/346          1s 2ms/step -
loss: 0.0033
Epoch 14/20
346/346          1s 2ms/step -
loss: 0.0034
Epoch 15/20
346/346          1s 2ms/step -
loss: 0.0031
Epoch 16/20
346/346          1s 2ms/step -
loss: 0.0031
Epoch 17/20
346/346          1s 2ms/step -
loss: 0.0032
Epoch 18/20
346/346          1s 2ms/step -

```



loss: 0.0031  
Epoch 19/20  
346/346 1s 2ms/step -  
loss: 0.0038  
Epoch 20/20  
346/346 1s 2ms/step -  
loss: 0.0034  
7/7 0s 13ms/step

Gelecekteki Fiyat Tahminleri:

[[302.71118]  
[302.5584 ]  
[302.4678 ]  
[302.44077]  
[302.36328]  
[302.29083]  
[302.13602]  
[302.15274]  
[302.2982 ]  
[302.2537 ]  
[302.3346 ]  
[302.31302]  
[303.16376]  
[303.55057]  
[304.47092]  
[304.0437 ]  
[304.81683]  
[304.90115]  
[305.18546]  
[302.97073]  
[305.13684]  
[305.2045 ]  
[304.964 ]  
[305.27948]  
[305.54535]  
[306.32504]  
[306.28555]  
[304.0476 ]  
[306.35812]  
[306.2134 ]  
[304.45078]  
[306.1448 ]  
[306.26617]  
[306.89972]  
[305.97653]  
[305.97095]  
[306.34097]  
[306.4167 ]  
[306.54343]

[306.61942]  
[306.6148 ]  
[306.69037]  
[306.83087]  
[306.9505 ]  
[306.69928]  
[306.29266]  
[306.2607 ]  
[306.3829 ]  
[306.65143]  
[306.79385]  
[307.41068]  
[307.79355]  
[307.8712 ]  
[307.78925]  
[307.74637]  
[308.3831 ]  
[308.02277]  
[308.4479 ]  
[308.96317]  
[310.4981 ]  
[309.34644]  
[308.50623]  
[308.97342]  
[309.15387]  
[309.2964 ]  
[307.22144]  
[309.99265]  
[309.77966]  
[309.20184]  
[309.2088 ]  
[309.44058]  
[309.29105]  
[309.20477]  
[309.43918]  
[309.37573]  
[309.23865]  
[309.22748]  
[309.116 ]  
[309.193 ]  
[309.42392]  
[309.35312]  
[309.77744]  
[309.61346]  
[309.5499 ]  
[309.26675]  
[309.6984 ]  
[309.6406 ]

[309.71375]  
[309.75308]  
[309.70328]  
[309.5902 ]  
[309.48566]  
[309.10495]  
[309.14084]  
[308.9551 ]  
[308.77463]  
[308.8638 ]  
[309.01093]  
[309.07498]  
[308.91916]  
[308.54132]  
[308.381 ]  
[307.1792 ]  
[306.9235 ]  
[307.27612]  
[306.8153 ]  
[306.03375]  
[305.95984]  
[306.40826]  
[306.49313]  
[306.5877 ]  
[306.3989 ]  
[306.54538]  
[306.52167]  
[306.67468]  
[306.61118]  
[306.5244 ]  
[306.67554]  
[306.71628]  
[306.68274]  
[304.6995 ]  
[306.57925]  
[306.86758]  
[307.1107 ]  
[307.29825]  
[307.63928]  
[308.25906]  
[308.36102]  
[308.9467 ]  
[308.7693 ]  
[308.5441 ]  
[308.62326]  
[308.66455]  
[308.7685 ]  
[308.75977]

[308.7375 ]  
[308.58823]  
[308.28275]  
[308.09772]  
[307.89673]  
[307.81332]  
[307.93347]  
[308.05444]  
[307.99612]  
[307.92725]  
[307.46918]  
[307.13922]  
[306.85098]  
[306.43335]  
[306.53635]  
[306.95746]  
[306.89584]  
[307.00403]  
[307.09268]  
[306.97614]  
[306.9731 ]  
[306.9168 ]  
[306.97253]  
[307.2831 ]  
[307.0247 ]  
[305.1517 ]  
[307.337 ]  
[307.59897]  
[307.8759 ]  
[307.83136]  
[307.71957]  
[308.08875]  
[308.14465]  
[307.7586 ]  
[307.09882]  
[306.7474 ]  
[306.908 ]  
[306.83752]  
[306.46112]  
[306.5236 ]  
[306.45364]  
[306.31418]  
[305.9411 ]  
[306.12387]  
[306.23428]  
[303.9462 ]  
[305.00873]  
[305.02844]

```

[305.03036]
[305.12247]
[305.38367]
[305.713  ]
[306.0129 ]
[306.27936]
[306.70395]
[306.86816]
[304.54895]
[306.6244 ]
[306.44397]
[306.6805 ]
[306.70712]
[306.74442]
[306.70386]
[306.75064]
[306.89584]
[306.89642]]
70/70          0s 1ms/step
RMSE: 0.06217185445865172
MAE: 0.027066357016339693

```

```

[10]: import pandas as pd
import numpy as np
import tensorflow as tf
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import MinMaxScaler
from sklearn.metrics import mean_squared_error, mean_absolute_error

veri = pd.read_csv(r"C:/Users/ASUS/Desktop/AI CLUB DATATHON/CSV/crp_ail.csv")

veri['open'] = veri['open'].str.replace('\"', '').astype(float)
veri['high'] = veri['high'].str.replace('\"', '').astype(float)
veri['low'] = veri['low'].str.replace('\"', '').astype(float)
veri['close'] = veri['close'].str.replace('\"', '').astype(float)

veri['open'].fillna(veri['open'].mean(), inplace=True)
veri['high'].fillna(veri['high'].mean(), inplace=True)
veri['low'].fillna(veri['low'].mean(), inplace=True)
veri['close'].fillna(veri['close'].mean(), inplace=True)

veri = veri[['open', 'low', 'high', 'close']]

```

```

X = veri[['close', 'low', 'high']].values
y = veri['high'].values

scaler = MinMaxScaler()
X_scaled = scaler.fit_transform(X)
y_scaled = scaler.fit_transform(y.reshape(-1,1)).reshape(-1)

X_train, X_test, y_train, y_test = train_test_split(X_scaled, y_scaled,
    ↪test_size=0.2, random_state=42)

model = tf.keras.models.Sequential([
    tf.keras.layers.Dense(64, activation='relu', input_shape=(X_train.
    ↪shape[1],)),
    tf.keras.layers.Dense(32, activation='relu'),
    tf.keras.layers.Dense(1)
])

model.compile(optimizer='adam', loss='mean_squared_error')

model.fit(X_scaled, y_scaled, epochs=20, batch_size=32, verbose=1)

gelecek_veri = X_scaled[-201:]
tahminler = model.predict(gelecek_veri)

tahminler = scaler.inverse_transform(tahminler)

print("Gelecekteki Fiyat Tahminleri:")
print(tahminler)

y_pred = model.predict(X_test)
mse = mean_squared_error(y_test, y_pred)
rmse = np.sqrt(mse)
mae = mean_absolute_error(y_test, y_pred)
print("RMSE:", rmse)
print("MAE:", mae)

```

Epoch 1/20

C:\Users\ASUS\anaconda3\Lib\site-packages\keras\src\layers\core\dense.py:87:  
 UserWarning: Do not pass an `input\_shape`/`input\_dim` argument to a layer. When  
 using Sequential models, prefer using an `Input(shape)` object as the first  
 layer in the model instead.

```
super().__init__(activity_regularizer=activity_regularizer, **kwargs)
```

346/346                      3s 2ms/step -

loss: 0.0247

Epoch 2/20

346/346	1s 2ms/step -
loss: 1.3462e-05	
Epoch 3/20	
346/346	1s 2ms/step -
loss: 8.8164e-06	
Epoch 4/20	
346/346	1s 2ms/step -
loss: 7.0121e-06	
Epoch 5/20	
346/346	1s 2ms/step -
loss: 5.7155e-06	
Epoch 6/20	
346/346	1s 2ms/step -
loss: 4.9056e-06	
Epoch 7/20	
346/346	1s 2ms/step -
loss: 4.4036e-06	
Epoch 8/20	
346/346	1s 2ms/step -
loss: 3.8157e-06	
Epoch 9/20	
346/346	1s 2ms/step -
loss: 2.6242e-06	
Epoch 10/20	
346/346	1s 2ms/step -
loss: 1.9687e-06	
Epoch 11/20	
346/346	1s 1ms/step -
loss: 1.8634e-06	
Epoch 12/20	
346/346	1s 2ms/step -
loss: 9.4455e-07	
Epoch 13/20	
346/346	1s 2ms/step -
loss: 8.2270e-07	
Epoch 14/20	
346/346	1s 2ms/step -
loss: 6.2134e-07	
Epoch 15/20	
346/346	1s 2ms/step -
loss: 6.2863e-07	
Epoch 16/20	
346/346	1s 2ms/step -
loss: 3.6657e-07	
Epoch 17/20	
346/346	1s 2ms/step -
loss: 3.4710e-07	
Epoch 18/20	

```
346/346          1s 2ms/step -
loss: 4.0628e-07
Epoch 19/20
346/346          1s 2ms/step -
loss: 4.6238e-07
Epoch 20/20
346/346          1s 2ms/step -
loss: 6.8817e-07
7/7              0s 10ms/step
```

Gelecekteki Fiyat Tahminleri:

```
[[304.3568 ]
 [304.2312 ]
 [303.95346]
 [304.00488]
 [303.95926]
 [303.9359 ]
 [303.76617]
 [303.7061 ]
 [303.93616]
 [303.8753 ]
 [304.1085 ]
 [304.0291 ]
 [304.94028]
 [305.5903 ]
 [306.9397 ]
 [306.43857]
 [307.04343]
 [307.06985]
 [307.3823 ]
 [306.73523]
 [306.8234 ]
 [307.03   ]
 [306.926  ]
 [307.36792]
 [307.43005]
 [308.7068 ]
 [308.32184]
 [307.9275 ]
 [308.4707 ]
 [308.1869 ]
 [308.28555]
 [308.17746]
 [308.32953]
 [309.176  ]
 [308.44156]
 [308.28006]
 [308.33362]
 [308.3984 ]
```



[308.621 ]  
[308.57547]  
[308.52707]  
[308.51328]  
[308.8729 ]  
[309.011 ]  
[308.75626]  
[308.27936]  
[308.31964]  
[308.2786 ]  
[308.79462]  
[308.775 ]  
[309.73816]  
[310.07355]  
[310.0033 ]  
[309.88306]  
[309.8304 ]  
[312.07095]  
[310.58536]  
[310.95822]  
[311.45844]  
[313.49496]  
[312.04874]  
[311.03067]  
[311.10013]  
[311.19986]  
[311.56427]  
[311.43585]  
[312.6367 ]  
[312.29776]  
[311.41464]  
[311.4053 ]  
[311.58856]  
[311.58893]  
[311.36682]  
[311.73373]  
[311.77542]  
[311.57468]  
[311.46387]  
[311.2902 ]  
[311.45157]  
[312.07242]  
[311.81332]  
[312.16263]  
[312.0145 ]  
[311.93353]  
[311.49432]  
[311.94315]

[311.80804]  
[311.7964 ]  
[312.0403 ]  
[312.004 ]  
[311.8173 ]  
[311.76154]  
[311.22906]  
[311.2589 ]  
[311.23596]  
[310.9508 ]  
[310.99133]  
[311.1368 ]  
[311.22656]  
[311.02505]  
[310.8622 ]  
[311.59 ]  
[309.86197]  
[309.1935 ]  
[309.42297]  
[309.31348]  
[308.00922]  
[308.11694]  
[308.35953]  
[308.27847]  
[308.70535]  
[308.38672]  
[308.57278]  
[308.4464 ]  
[308.72784]  
[308.60416]  
[308.46307]  
[308.59274]  
[308.70453]  
[308.6192 ]  
[308.57367]  
[308.53537]  
[309.02322]  
[309.1038 ]  
[309.40442]  
[310.1368 ]  
[310.57056]  
[310.59778]  
[311.33078]  
[310.86783]  
[310.72577]  
[310.60236]  
[310.7178 ]  
[310.82755]

[310.87854]  
[310.9301 ]  
[310.7136 ]  
[310.2956 ]  
[310.21292]  
[309.96146]  
[309.79153]  
[309.9463 ]  
[310.21274]  
[310.05893]  
[310.05637]  
[309.9374 ]  
[309.50888]  
[309.04187]  
[308.8031 ]  
[308.54346]  
[309.1988 ]  
[308.8449 ]  
[308.99786]  
[309.0576 ]  
[308.80096]  
[308.8546 ]  
[308.81122]  
[308.9498 ]  
[309.35278]  
[308.93237]  
[309.0574 ]  
[309.28033]  
[309.81824]  
[309.94968]  
[309.9232 ]  
[309.84882]  
[310.3161 ]  
[310.30566]  
[309.83087]  
[309.53638]  
[308.9746 ]  
[309.10458]  
[308.99442]  
[308.58344]  
[308.54028]  
[308.42593]  
[308.26086]  
[307.94797]  
[308.065 ]  
[308.1504 ]  
[307.8904 ]  
[306.8049 ]

```

[306.8346 ]
[306.83466]
[306.947  ]
[307.38278]
[307.69827]
[307.91608]
[308.28583]
[308.6862 ]
[309.00934]
[308.50143]
[308.6846 ]
[308.60498]
[308.67014]
[308.6021 ]
[308.64764]
[308.56342]
[308.55823]
[308.86697]
[308.81738]]
70/70          0s 1ms/step
RMSE: 0.0007563394079618845
MAE: 0.0007179458762785808

```

```

[11]: import pandas as pd
import numpy as np
import tensorflow as tf
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import MinMaxScaler
from sklearn.metrics import mean_squared_error, mean_absolute_error

veri = pd.read_csv(r"C:/Users/ASUS/Desktop/AI CLUB DATATHON/CSV/crp_ail.csv")

veri['open'] = veri['open'].str.replace('"', '').astype(float)
veri['high'] = veri['high'].str.replace('"', '').astype(float)
veri['low'] = veri['low'].str.replace('"', '').astype(float)
veri['close'] = veri['close'].str.replace('"', '').astype(float)

veri['open'].fillna(veri['open'].mean(), inplace=True)
veri['high'].fillna(veri['high'].mean(), inplace=True)
veri['low'].fillna(veri['low'].mean(), inplace=True)
veri['close'].fillna(veri['close'].mean(), inplace=True)

veri = veri[['open', 'low', 'high', 'close']]

X = veri[['close', 'low', 'high']].values
y = veri['low'].values

```

```

scaler = MinMaxScaler()
X_scaled = scaler.fit_transform(X)
y_scaled = scaler.fit_transform(y.reshape(-1,1)).reshape(-1)

X_train, X_test, y_train, y_test = train_test_split(X_scaled, y_scaled,
↳test_size=0.2, random_state=42)

model = tf.keras.models.Sequential([
    tf.keras.layers.Dense(64, activation='relu', input_shape=(X_train.
↳shape[1],)),
    tf.keras.layers.Dense(32, activation='relu'),
    tf.keras.layers.Dense(1)
])

model.compile(optimizer='adam', loss='mean_squared_error')

model.fit(X_scaled, y_scaled, epochs=20, batch_size=32, verbose=1)

gelecek_veri = X_scaled[-201:]
tahminler = model.predict(gelecek_veri)

tahminler = scaler.inverse_transform(tahminler)

print("Gelecekteki Fiyat Tahminleri:")
print(tahminler)

y_pred = model.predict(X_test)
mse = mean_squared_error(y_test, y_pred)
rmse = np.sqrt(mse)
mae = mean_absolute_error(y_test, y_pred)
print("RMSE:", rmse)
print("MAE:", mae)

```

Epoch 1/20

C:\Users\ASUS\anaconda3\Lib\site-packages\keras\src\layers\core\dense.py:87:  
UserWarning: Do not pass an `input\_shape`/`input\_dim` argument to a layer. When  
using Sequential models, prefer using an `Input(shape)` object as the first  
layer in the model instead.

```
super().__init__(activity_regularizer=activity_regularizer, **kwargs)
```

346/346                      3s 2ms/step -

loss: 0.0365

Epoch 2/20

346/346                      1s 2ms/step -

loss: 2.6020e-05

Epoch 3/20

346/346	1s 2ms/step -
loss: 1.5432e-05	
Epoch 4/20	
346/346	1s 2ms/step -
loss: 9.7915e-06	
Epoch 5/20	
346/346	1s 2ms/step -
loss: 7.0072e-06	
Epoch 6/20	
346/346	1s 2ms/step -
loss: 3.8594e-06	
Epoch 7/20	
346/346	1s 2ms/step -
loss: 2.3884e-06	
Epoch 8/20	
346/346	1s 2ms/step -
loss: 1.7009e-06	
Epoch 9/20	
346/346	1s 2ms/step -
loss: 8.6841e-07	
Epoch 10/20	
346/346	1s 2ms/step -
loss: 5.1009e-07	
Epoch 11/20	
346/346	1s 2ms/step -
loss: 2.6750e-07	
Epoch 12/20	
346/346	1s 2ms/step -
loss: 2.7685e-07	
Epoch 13/20	
346/346	1s 1ms/step -
loss: 1.6991e-07	
Epoch 14/20	
346/346	1s 2ms/step -
loss: 4.7751e-07	
Epoch 15/20	
346/346	1s 2ms/step -
loss: 2.9043e-07	
Epoch 16/20	
346/346	1s 2ms/step -
loss: 2.0052e-06	
Epoch 17/20	
346/346	1s 2ms/step -
loss: 1.5541e-07	
Epoch 18/20	
346/346	1s 2ms/step -
loss: 8.0056e-07	
Epoch 19/20	

346/346 1s 2ms/step -

loss: 9.1724e-07

Epoch 20/20

346/346 1s 2ms/step -

loss: 3.8163e-06

7/7 0s 13ms/step

Gelecekteki Fiyat Tahminleri:

[[304.11304]

[303.90225]

[304.025 ]

[303.89487]

[303.76694]

[303.625 ]

[303.45346]

[303.56934]

[303.64346]

[303.60175]

[303.51993]

[303.5633 ]

[304.476 ]

[304.5941 ]

[305.14713]

[304.74075]

[305.84985]

[306.03748]

[306.33472]

[306.2368 ]

[306.9035 ]

[306.81 ]

[306.36255]

[306.56705]

[307.13483]

[307.43054]

[307.82645]

[307.4503 ]

[307.81616]

[307.804 ]

[308.00848]

[307.6551 ]

[307.75253]

[308.24673]

[306.92526]

[307.10263]

[307.93127]

[308.04108]

[308.06244]

[308.3104 ]

[308.35248]

[308.55524]  
[308.44434]  
[308.57025]  
[308.28482]  
[307.89273]  
[307.75534]  
[308.10394]  
[308.1062 ]  
[308.4777 ]  
[308.77548]  
[309.2785 ]  
[309.5574 ]  
[309.50687]  
[309.4768 ]  
[308.22684]  
[309.1971 ]  
[309.74707]  
[310.36966]  
[311.5539 ]  
[310.59753]  
[309.80298]  
[310.85306]  
[311.1982 ]  
[311.0679 ]  
[310.73196]  
[311.3882 ]  
[311.32428]  
[311.0214 ]  
[311.04105]  
[311.39685]  
[311.02124]  
[311.08612]  
[311.1852 ]  
[310.99997]  
[310.90045]  
[311.02173]  
[310.95938]  
[310.95502]  
[310.72324]  
[310.8769 ]  
[311.47116]  
[311.26965]  
[311.21948]  
[311.0672 ]  
[311.5627 ]  
[311.61816]  
[311.83496]  
[311.5762 ]



[311.50366]  
[311.46655]  
[311.27777]  
[311.0143 ]  
[311.06976]  
[310.64566]  
[310.56128]  
[310.72177]  
[310.90012]  
[310.94623]  
[310.8171 ]  
[310.11542]  
[308.81384]  
[308.0774 ]  
[308.26938]  
[308.83408]  
[307.8906 ]  
[307.6029 ]  
[307.273 ]  
[308.0634 ]  
[308.3706 ]  
[308.07162]  
[308.0126 ]  
[308.1302 ]  
[308.2314 ]  
[308.24902]  
[308.2524 ]  
[308.21448]  
[308.41965]  
[308.38364]  
[308.4057 ]  
[308.27124]  
[308.26187]  
[308.34186]  
[308.8332 ]  
[308.92056]  
[308.83658]  
[309.78842]  
[309.99564]  
[310.51456]  
[310.6529 ]  
[310.28882]  
[310.64233]  
[310.58368]  
[310.69937]  
[310.61032]  
[310.49878]  
[310.41074]

[310.18637]  
[309.84976]  
[309.67056]  
[309.67783]  
[309.77536]  
[309.73904]  
[309.78467]  
[309.62872]  
[308.68207]  
[308.40796]  
[308.2839 ]  
[307.58026]  
[308.147 ]  
[308.35754]  
[308.63992]  
[308.70618]  
[308.85422]  
[308.8878 ]  
[308.81586]  
[308.73184]  
[308.69168]  
[308.95145]  
[308.84192]  
[308.75793]  
[309.16498]  
[309.1191 ]  
[309.63358]  
[309.56543]  
[309.3761 ]  
[309.68723]  
[309.8476 ]  
[309.5023 ]  
[308.2777 ]  
[308.1218 ]  
[308.354 ]  
[308.32346]  
[307.9219 ]  
[308.1172 ]  
[308.09802]  
[307.96634]  
[307.44632]  
[307.74072]  
[307.90448]  
[307.254 ]  
[306.61978]  
[306.62848]  
[306.63416]  
[306.7158 ]

```

[306.79684]
[307.2048 ]
[307.6617 ]
[307.84015]
[308.36923]
[308.37888]
[307.999 ]
[308.18784]
[307.8396 ]
[308.33655]
[308.48785]
[308.51935]
[308.52438]
[308.6449 ]
[308.61313]
[308.6723 ]]
70/70          0s 2ms/step
RMSE: 0.00037814870053846196
MAE: 0.0003494904348502573

```

```

[12]: import pandas as pd
import numpy as np
import tensorflow as tf
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import MinMaxScaler
from sklearn.metrics import mean_squared_error, mean_absolute_error

veri = pd.read_csv(r"C:/Users/ASUS/Desktop/AI CLUB DATATHON/CSV/crp_ail.csv")

veri['open'] = veri['open'].str.replace('"', '').astype(float)
veri['high'] = veri['high'].str.replace('"', '').astype(float)
veri['low'] = veri['low'].str.replace('"', '').astype(float)
veri['close'] = veri['close'].str.replace('"', '').astype(float)

veri['open'].fillna(veri['open'].mean(), inplace=True)
veri['high'].fillna(veri['high'].mean(), inplace=True)
veri['low'].fillna(veri['low'].mean(), inplace=True)
veri['close'].fillna(veri['close'].mean(), inplace=True)

veri = veri[['open', 'low', 'high', 'close']]

X = veri[['close', 'low', 'high']].values
y = veri['close'].values

scaler = MinMaxScaler()
X_scaled = scaler.fit_transform(X)

```

```

y_scaled = scaler.fit_transform(y.reshape(-1,1)).reshape(-1)

X_train, X_test, y_train, y_test = train_test_split(X_scaled, y_scaled,
↳test_size=0.2, random_state=42)

model = tf.keras.models.Sequential([
    tf.keras.layers.Dense(64, activation='relu', input_shape=(X_train.
↳shape[1],)),
    tf.keras.layers.Dense(32, activation='relu'),
    tf.keras.layers.Dense(1)
])

model.compile(optimizer='adam', loss='mean_squared_error')

model.fit(X_scaled, y_scaled, epochs=20, batch_size=32, verbose=1)

gelecek_veri = X_scaled[-201:]
tahminler = model.predict(gelecek_veri)

tahminler = scaler.inverse_transform(tahminler)

print("Gelecekteki Fiyat Tahminleri:")
print(tahminler)

y_pred = model.predict(X_test)
mse = mean_squared_error(y_test, y_pred)
rmse = np.sqrt(mse)
mae = mean_absolute_error(y_test, y_pred)
print("RMSE:", rmse)
print("MAE:", mae)

```

Epoch 1/20

C:\Users\ASUS\anaconda3\Lib\site-packages\keras\src\layers\core\dense.py:87:  
UserWarning: Do not pass an `input\_shape`/`input\_dim` argument to a layer. When  
using Sequential models, prefer using an `Input(shape)` object as the first  
layer in the model instead.

```
super().__init__(activity_regularizer=activity_regularizer, **kwargs)
```

346/346 3s 2ms/step -

loss: 0.0369

Epoch 2/20

346/346 1s 2ms/step -

loss: 2.9132e-06

Epoch 3/20

```

346/346          1s 2ms/step -
loss: 1.5659e-06
Epoch 4/20
346/346          1s 2ms/step -
loss: 1.0600e-06
Epoch 5/20
346/346          1s 2ms/step -
loss: 5.4161e-07
Epoch 6/20
346/346          1s 2ms/step -
loss: 5.9794e-07
Epoch 7/20
346/346          1s 2ms/step -
loss: 3.3842e-07
Epoch 8/20
346/346          1s 2ms/step -
loss: 2.8209e-07
Epoch 9/20
346/346          1s 2ms/step -
loss: 2.5849e-07
Epoch 10/20
346/346          1s 2ms/step -
loss: 8.1465e-07
Epoch 11/20
346/346          1s 2ms/step -
loss: 4.4596e-07
Epoch 12/20
346/346          1s 2ms/step -
loss: 2.6183e-07
Epoch 13/20
346/346          1s 2ms/step -
loss: 1.7326e-06
Epoch 14/20
346/346          1s 2ms/step -
loss: 4.6759e-07
Epoch 15/20
346/346          1s 2ms/step -
loss: 6.7310e-07
Epoch 16/20
346/346          1s 2ms/step -
loss: 2.7283e-06
Epoch 17/20
346/346          1s 2ms/step -
loss: 7.9965e-07
Epoch 18/20
346/346          1s 2ms/step -
loss: 4.5928e-06
Epoch 19/20

```

346/346 1s 2ms/step -

loss: 8.8650e-08

Epoch 20/20

346/346 1s 2ms/step -

loss: 9.8152e-08

7/7 0s 12ms/step

Gelecekteki Fiyat Tahminleri:

[[304.3181 ]

[304.09055]

[304.08466]

[304.09818]

[303.93784]

[303.7332 ]

[303.8583 ]

[303.83618]

[303.7076 ]

[303.9915 ]

[303.67664]

[303.75266]

[304.98486]

[305.72015]

[306.5244 ]

[305.97 ]

[306.9756 ]

[306.48846]

[306.83017]

[270.4711 ]

[306.97546]

[307.0773 ]

[306.70297]

[307.33926]

[307.48883]

[308.42188]

[307.8853 ]

[270.49118]

[307.99478]

[308.33643]

[270.5042 ]

[308.10315]

[308.46143]

[308.62732]

[307.83505]

[308.1345 ]

[308.42087]

[308.25003]

[308.67337]

[308.46777]

[308.68784]

[308.64746]  
[308.94574]  
[308.7848 ]  
[308.3588 ]  
[308.00705]  
[308.33017]  
[308.34906]  
[308.84158]  
[308.8605 ]  
[309.40045]  
[309.96286]  
[309.88306]  
[309.8574 ]  
[309.53918]  
[310.28006]  
[310.21667]  
[311.11166]  
[311.6343 ]  
[312.27316]  
[310.73654]  
[311.0019 ]  
[311.15805]  
[311.33646]  
[311.1309 ]  
[270.59695]  
[312.48834]  
[311.5642 ]  
[311.16238]  
[311.46216]  
[311.63043]  
[311.23618]  
[311.4582 ]  
[311.85873]  
[311.1429 ]  
[311.59235]  
[311.2065 ]  
[311.3807 ]  
[311.13202]  
[311.78015]  
[311.80856]  
[312.13736]  
[311.64163]  
[311.48035]  
[311.63614]  
[311.73022]  
[311.92493]  
[311.90607]  
[311.72437]

[311.58426]  
[311.92844]  
[311.40256]  
[311.16995]  
[311.27573]  
[310.97693]  
[310.8659 ]  
[311.10202]  
[311.22098]  
[311.1011 ]  
[311.03818]  
[310.38882]  
[310.10413]  
[308.71478]  
[309.08826]  
[309.50565]  
[307.94894]  
[307.74228]  
[308.2841 ]  
[308.42572]  
[308.43594]  
[308.48773]  
[308.2328 ]  
[308.55966]  
[308.46915]  
[308.72684]  
[308.54593]  
[308.57318]  
[308.68372]  
[308.58627]  
[308.64365]  
[270.50803]  
[308.47192]  
[309.15176]  
[309.23245]  
[309.18222]  
[309.52396]  
[310.34943]  
[310.62143]  
[310.74637]  
[310.79663]  
[310.46277]  
[310.75156]  
[310.77097]  
[310.89398]  
[311.00046]  
[310.77863]  
[310.4962 ]



[310.35623]  
[309.92392]  
[309.9151 ]  
[309.89694]  
[310.04117]  
[310.1017 ]  
[310.18365]  
[309.87756]  
[309.05 ]  
[309.0154 ]  
[308.95743]  
[308.26825]  
[308.47278]  
[308.69342]  
[308.86826]  
[309.15646]  
[308.9687 ]  
[308.95377]  
[308.8919 ]  
[308.90305]  
[309.05948]  
[309.04977]  
[309.01315]  
[270.51752]  
[309.30923]  
[309.97556]  
[309.93774]  
[309.67636]  
[309.961 ]  
[310.3595 ]  
[309.99216]  
[309.6663 ]  
[308.9334 ]  
[308.70566]  
[308.67224]  
[308.49683]  
[308.20337]  
[308.56253]  
[308.22025]  
[308.0702 ]  
[307.95 ]  
[308.18375]  
[308.18716]  
[270.4846 ]  
[306.76868]  
[306.85995]  
[306.8229 ]  
[306.96286]

```
[307.5458 ]
[307.73682]
[307.92972]
[308.41483]
[308.8231 ]
[308.5313 ]
[270.49933]
[308.45694]
[308.66806]
[308.6646 ]
[308.5927 ]
[308.69977]
[308.7057 ]
[308.71057]
[308.85242]
[308.9755 ]]
70/70          0s 2ms/step
RMSE: 0.0006083523823731347
MAE: 0.0005342891759797481
```

bk\_frx kısmı bu kodlar aracılığı ile yapıldı.

```
[13]: from sklearn.ensemble import RandomForestRegressor
from sklearn.metrics import mean_squared_error
from sklearn.preprocessing import StandardScaler
from sklearn.model_selection import train_test_split
import pandas as pd
import numpy as np

veri = pd.read_csv(r"C:/Users/ASUS/Desktop/AI CLUB DATATHON/CSV/bk_frx.csv")
veri['open'] = veri['open'].str.replace('"', '').astype(float)
veri['high'] = veri['high'].str.replace('"', '').astype(float)
veri['low'] = veri['low'].str.replace('"', '').astype(float)
veri['close'] = veri['close'].str.replace('"', '').astype(float)

veri['FuturePrice'] = veri['open'].shift(-1)

veri.dropna(inplace=True)

X = veri[['open', 'low', 'high', 'close']].tail(4886)
y = veri['FuturePrice'].tail(4886)

scaler = StandardScaler()
X_scaled = scaler.fit_transform(X)
y_scaled = scaler.fit_transform(y.values.reshape(-1, 1)).reshape(-1)
```

```

X_train, X_test, y_train, y_test = train_test_split(X_scaled, y_scaled,
↳test_size=0.2, random_state=42)

rf_model = RandomForestRegressor()
rf_model.fit(X, y)

gelecek_veri = X.tail(89)[::-1]
rf_tahminler = rf_model.predict(gelecek_veri)

rf_mse = mean_squared_error(y[-88:], rf_tahminler[::-1])
rf_rmse = np.sqrt(rf_mse)

print("Random Forest Regressor ile Gelecekteki Fiyat Tahminleri:")
np.set_printoptions(precision=7)
print(rf_tahminler)
print("RMSE Değeri:", rf_rmse)

```

```

Random Forest Regressor ile Gelecekteki Fiyat Tahminleri:
[20140.377913  20190.1059696 20200.236113  20240.2605609 20321.4754565
 20327.9324696 20231.0618348 20380.6681609 20455.706713  20476.9304478
 20463.8101826 20519.492213  20524.4977174 20549.4236174 20549.1772826
 20518.0690478 20509.1887826 20620.0689304 20578.7358435 20582.468887
 20658.5480826 20611.009387  20659.2504478 20752.3343478 20949.4086261
 20955.8248522 20887.0930696 20998.0409652 21087.659587  21122.7266913
 21061.8165565 21065.8180565 21064.0722826 21243.4083826 21366.5034174
 21331.9361261 21585.6351435 21128.3430783 20960.5547565 20924.9161739
 20908.5587609 20942.2326565 20825.2058913 20721.5878957 20710.1666261
 20464.5950435 20291.0529957 20223.084413  20163.3969652 20339.6742739
 20397.5843783 20438.5476174 20385.1766174 20395.8754739 20465.9829522
 20396.8534391 20404.4619348 20437.4544348 20549.5563478 20637.1194913
 20611.4661087 20635.119087  20572.3398957 20545.5882609 20539.0858522
 20933.4844304 20902.1920783 21123.9788739 21164.477787  21232.2929
 21244.4287478 21125.5951435 21189.3725783 21009.707187  21059.3778652
 20979.1215783 20943.9770478 21103.1464174 21243.7607174 21163.9759
 21168.0518304 21325.4549174 21384.3312348 21301.5182739 21546.664013
 21546.7953609 21658.8191565 21620.8082478 21585.8598174]
RMSE Değeri: 755.593732240723

```

```

[14]: from sklearn.ensemble import RandomForestRegressor
      from sklearn.metrics import mean_squared_error
      from sklearn.preprocessing import StandardScaler
      from sklearn.model_selection import train_test_split
      import pandas as pd
      import numpy as np

```

```

veri = pd.read_csv(r"C:/Users/ASUS/Desktop/AI CLUB DATATHON/CSV/bk_frx.csv")
veri['open'] = veri['open'].str.replace('"', '').astype(float)
veri['high'] = veri['high'].str.replace('"', '').astype(float)
veri['low'] = veri['low'].str.replace('"', '').astype(float)
veri['close'] = veri['close'].str.replace('"', '').astype(float)

veri['FuturePrice'] = veri['high'].shift(-1)

veri.dropna(inplace=True)

X = veri[['open', 'low', 'high', 'close']].tail(4886)
y = veri['FuturePrice'].tail(4886)

scaler = StandardScaler()
X_scaled = scaler.fit_transform(X)
y_scaled = scaler.fit_transform(y.values.reshape(-1, 1)).reshape(-1)

X_train, X_test, y_train, y_test = train_test_split(X_scaled, y_scaled,
    ↪test_size=0.2, random_state=42)

rf_model = RandomForestRegressor()
rf_model.fit(X, y)

gelecek_veri = X.tail(89)[::-1]
rf_tahminler = rf_model.predict(gelecek_veri)

rf_mse = mean_squared_error(y[-88:], rf_tahminler[::-1])
rf_rmse = np.sqrt(rf_mse)

print("Random Forest Regressor ile Gelecekteki Fiyat Tahminleri:")
np.set_printoptions(precision=7)
print(rf_tahminler)
print("RMSE Değeri:", rf_rmse)

```

```

Random Forest Regressor ile Gelecekteki Fiyat Tahminleri:
[20151.4565261 20225.8049261 20243.7560261 20295.4697391 20357.227413
 20385.3243261 20366.6554217 20403.5231435 20367.456413 20464.727313
 20475.1763783 20522.1703261 20519.073513 20529.4518348 20483.1738478
 20548.1472391 20603.2859043 20577.7106391 20564.5984391 20589.9935043
 20664.1672348 20590.4345565 20607.800813 20755.3235478 20698.3918696
 20693.6433 20772.6704478 20964.967813 20963.2992348 20898.903313
 21131.7666478 21139.7885435 21107.4140696 21087.9706739 21076.4906435

```

```

21373.1450087 21392.2588826 21602.0374913 21097.6911043 21158.5127522
21080.2674696 21061.2718087 20982.3027304 20948.5813652 20999.264113
20948.5373522 20864.6305174 20804.176413 20734.3585913 20511.3820609
20451.2531 20241.4042087 20177.9871739 20402.6901217 20415.1128609
20472.8635043 20434.1188913 20426.1988478 20494.9251 20492.2776348
20464.8943783 20567.6989391 20661.6928261 20669.6140217 20657.1712348
20632.3649304 20612.705387 20549.9049957 20805.6772348 20942.0169696
21091.9694087 20989.9384174 21155.9701348 21178.9933348 21255.9417304
21273.0365348 21177.1491652 21200.3836739 21155.6862391 21091.6214522
21089.4786391 21095.3839913 20958.4156304 21132.0198957 21207.7163391
21173.2870783 21426.0171174 21389.2023957 21327.4067 ]

```

RMSE Değeri: 653.8977609132194

```

[15]: from sklearn.ensemble import RandomForestRegressor
from sklearn.metrics import mean_squared_error
from sklearn.preprocessing import StandardScaler
from sklearn.model_selection import train_test_split
import pandas as pd
import numpy as np

veri = pd.read_csv(r"C:/Users/ASUS/Desktop/AI CLUB DATATHON/CSV/bk_frx.csv")
veri['open'] = veri['open'].str.replace('"', '').astype(float)
veri['high'] = veri['high'].str.replace('"', '').astype(float)
veri['low'] = veri['low'].str.replace('"', '').astype(float)
veri['close'] = veri['close'].str.replace('"', '').astype(float)

veri['FuturePrice'] = veri['low'].shift(-1)

veri.dropna(inplace=True)

X = veri[['open', 'low', 'high', 'close']].tail(4886)
y = veri['FuturePrice'].tail(4886)

scaler = StandardScaler()
X_scaled = scaler.fit_transform(X)
y_scaled = scaler.fit_transform(y.values.reshape(-1, 1)).reshape(-1)

X_train, X_test, y_train, y_test = train_test_split(X_scaled, y_scaled,
    ↪test_size=0.2, random_state=42)

rf_model = RandomForestRegressor()
rf_model.fit(X, y)

gelecek_veri = X.tail(89)[::-1]
rf_tahminler = rf_model.predict(gelecek_veri)

rf_mse = mean_squared_error(y[-88:], rf_tahminler[:-1])

```

```

rf_rmse = np.sqrt(rf_mse)

print("Random Forest Regressor ile Gelecekteki Fiyat Tahminleri:")
np.set_printoptions(precision=7)
print(rf_tahminler)
print("RMSE Değeri:", rf_rmse)

```

Random Forest Regressor ile Gelecekteki Fiyat Tahminleri:

```

[20079.6389913 20125.6234043 20164.3304565 20211.5209652 20232.4790087
 20292.4473565 20238.6922217 20325.6707522 20303.2667304 20429.2191565
 20409.0980522 20411.1703522 20420.9949391 20505.5774087 20429.9118435
 20484.321413 20435.8628217 20465.9126696 20518.6098783 20558.5596652
 20535.3564957 20602.1438696 20625.9201304 20630.1755696 20756.8573217
 20953.3949174 20878.3989957 20868.0948739 21008.7312957 21061.5976435
 21035.767287 21009.0541348 21235.1530565 21295.9373739 21043.9871261
 21089.0624739 20996.5940652 20935.2389609 20898.9125304 20825.0752348
 20881.8801739 20942.7444522 20836.4577913 20667.0661043 20270.0193696
 20173.8752957 20153.6928957 20332.5669739 20363.2424522 20349.7180043
 20325.5294957 20332.2517391 20328.782313 20393.463513 20505.176913
 20544.7665304 20563.388887 20580.542913 20540.3537043 20479.2331826
 20522.017087 20645.213513 20860.8930957 20872.6602478 20944.6817174
 20921.470713 21140.7215739 21143.2844696 21142.0668522 21125.3043348
 20983.143587 20973.1913391 20942.9055261 20930.4201087 21157.0363565
 21310.7824435 21317.0903652 21390.4098739 21493.6176957 21476.4933957
 21535.6934739 21592.0006739 21548.8319435 21532.7863087 21405.0682913
 21505.138513 21516.162513 21533.9241957 21418.6833 ]

```

RMSE Değeri: 802.0540051505249

```

[16]: from sklearn.ensemble import RandomForestRegressor
from sklearn.metrics import mean_squared_error
from sklearn.preprocessing import StandardScaler
from sklearn.model_selection import train_test_split
import pandas as pd
import numpy as np

veri = pd.read_csv(r"C:/Users/ASUS/Desktop/AI CLUB DATATHON/CSV/bk_frx.csv")
veri['open'] = veri['open'].str.replace('\"', '').astype(float)
veri['high'] = veri['high'].str.replace('\"', '').astype(float)
veri['low'] = veri['low'].str.replace('\"', '').astype(float)
veri['close'] = veri['close'].str.replace('\"', '').astype(float)

veri['FuturePrice'] = veri['close'].shift(-1)

veri.dropna(inplace=True)

X = veri[['open', 'low', 'high', 'close']].tail(4886)

```

```

y = veri['FuturePrice'].tail(4886)

scaler = StandardScaler()
X_scaled = scaler.fit_transform(X)
y_scaled = scaler.fit_transform(y.values.reshape(-1, 1)).reshape(-1)

X_train, X_test, y_train, y_test = train_test_split(X_scaled, y_scaled,
    ↪test_size=0.2, random_state=42)

rf_model = RandomForestRegressor()
rf_model.fit(X, y)

gelecek_veri = X.tail(89)[::-1]
rf_tahminler = rf_model.predict(gelecek_veri)

rf_mse = mean_squared_error(y[-88:], rf_tahminler[:-1])
rf_rmse = np.sqrt(rf_mse)

print("Random Forest Regressor ile Gelecekteki Fiyat Tahminleri:")
np.set_printoptions(precision=7)
print(rf_tahminler)
print("RMSE Değeri:", rf_rmse)

```

```

Random Forest Regressor ile Gelecekteki Fiyat Tahminleri:
[20095.2958826 20167.8003435 20197.4672087 20225.9653087 20256.1234609
 20329.301713 20330.298113 20273.0673304 20359.5748522 20361.8994783
 20459.1344304 20449.7552739 20481.4326826 20456.4867348 20471.9081217
 20470.2250261 20535.6602087 20566.3861522 20512.3369826 20523.103587
 20536.2121 20546.0002783 20576.7034087 20566.7244304 20652.2927
 20666.1579609 20635.4935435 20678.7569826 20828.561713 20961.3684217
 20892.6306478 20895.3182087 21011.4428217 21084.8330739 21096.2363696
 21056.7769478 21059.7585435 21304.3867261 21339.7204435 21362.8671565
 21423.552387 21081.2873739 21113.4975478 21040.7610391 21001.9961478
 20932.5456391 20915.7672217 20897.2068522 20949.7234 20860.2944261
 20797.341487 20712.1829304 20451.852 20398.9992478 20217.2509565
 20164.9818957 20271.0466478 20357.1907739 20394.8138609 20397.7060478
 20376.6461522 20416.4175826 20461.8574783 20470.2192652 20413.4707826
 20591.3032957 20575.6079217 20612.1366739 20611.3937522 20613.8617087
 20552.2662609 20542.2040957 20684.9347087 20886.8243826 20936.2581739
 21018.7976087 20933.9584348 21095.4077261 21140.217613 21176.8698783
 21220.3923261 21109.7949217 21143.0966652 21125.1918826 21059.5281087
 21069.3022304 20994.3666826 20946.4855609 20966.0842696]
RMSE Değeri: 549.0391426919456

```