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% □
                                                                                                                                      Python 3 (ipykernel)
   #split data into training and test data
   data training=pd.DataFrame(df["Close"][0:int(len(df)*0.70)])
   data_testing=pd.DataFrame(df["Close"][int(len(df)*0.70):int(len(df))])
   print(data_training.shape)
   print(data_testing.shape)
   from sklearn.preprocessing import MinMaxScaler
   scaler=MinMaxScaler(feature range=(0,1))
   data_training_array=scaler.fit_transform(data_training)
   model=load model("Keras model.h5")
   past_100_days=data_training.tail(100)
   final_df=past_100_days.append(data_testing,ignore_index=True)
   input_data=scaler.fit_transform(final_df)
   x_test=[]
   y_test=[]
   for i in range(100,input data.shape[0]):
       x_test.append(input_data[i-100:i])
       y_test.append(input_data[i,0])
   x test,y test=np.array(x test),np.array(y test)
   y predicted=model.predict(x test)
   scaler.scale_
   scale factor=1/0.02099517
   y_predicted=y_predicted*scale_factor
   y_test=y_test*scale_factor
   st.subheader("Prediction VS Original")
   fig2=plt.figure(figsize=(12,6))
   plt.plot(y_test,"b",label="Original Price")
   plt.plot(y_predicted, "r", label="Predicted Price")
```

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from sklearn.preprocessing import MinMaxScaler
scaler=MinMaxScaler(feature range=(0,1))
data_training_array=scaler.fit_transform(data_training)
model=load_model("Keras_model.h5")
past 100 days=data training.tail(100)
final df=past 100 days.append(data testing,ignore index=True)
input_data=scaler.fit_transform(final_df)
x_test=[]
y_test=[]
for i in range(100,input_data.shape[0]):
    x test.append(input data[i-100:i])
   y_test.append(input_data[i,0])
x_test,y_test=np.array(x_test),np.array(y_test)
y_predicted=model.predict(x_test)
scaler.scale
scale_factor=1/0.02099517
y_predicted=y_predicted*scale_factor
y_test=y_test*scale_factor
st.subheader("Prediction VS Original")
fig2=plt.figure(figsize=(12,6))
plt.plot(y test,"b",label="Original Price")
plt.plot(y_predicted, "r", label="Predicted Price")
plt.xlabel("Time")
plt.ylabel("Price")
plt.legend()
st.pyplot(fig2
```

Actual vs Predicted Values 0.06 Actual Predicted 0.05 -0.04 Value 0.03 0.02 0.01 -

Figure 10: Line Graph

Sample

6

8

2

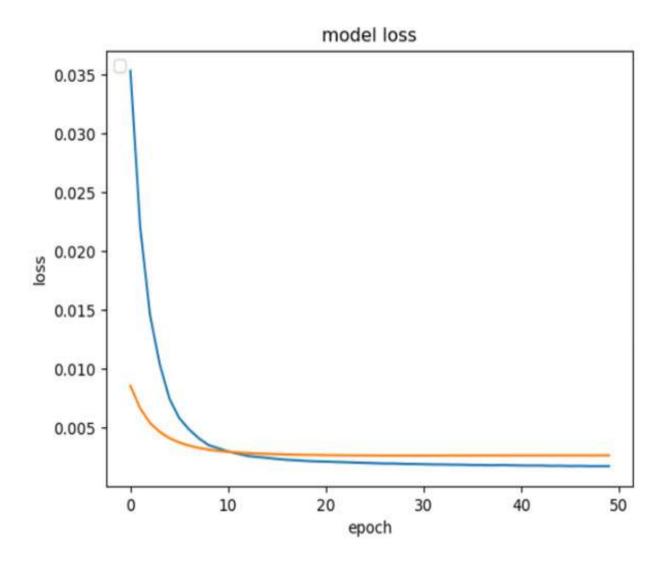


Figure 11: Line Graph of Model Accuracy

The line graph shows the loss which is the mean squared error (mse) on the y-axis and the number of epochs on the x-axis. It turns out that keeping the epoch at 50 instead of 10 and 20 improved the model's performance even more and helped it to become more

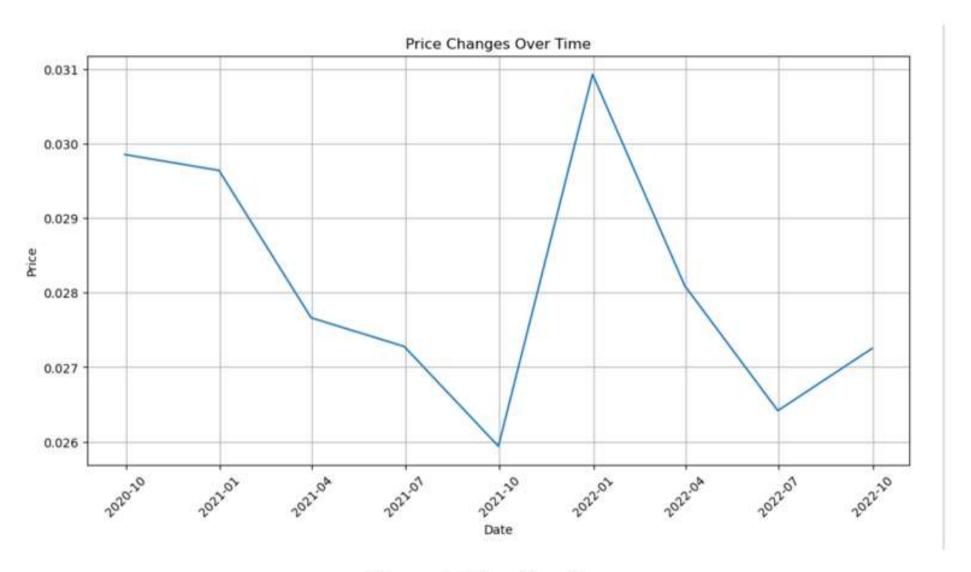


Figure 6: Line Graph

The above figure is a line graph shows the general changes in price that takes place throughout the 3 years of the products listed in the dataset.

