Importing libraries

```
import yfinance as yf
import pandas as pd
import numpy as np
import tensorflow as tf
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, LSTM, Dropout
from sklearn.preprocessing import MinMaxScaler
import matplotlib.pyplot as plt
Data importing and documentation
def download_data(ticker, start, end):
    data = yf.download(ticker, start=start, end=end)
    if isinstance(data.columns, pd.MultiIndex):
        data.columns = data.columns.get_level_values(0)
    return data
Relative Strength Index (RSI) calculation formula:
def compute RSI(series, period):
    delta = series.diff()
    up = delta.clip(lower=0)
    down = -delta.clip(upper=0)
    avg_gain = up.rolling(window=period).mean()
    avg_loss = down.rolling(window=period).mean()
    rs = avg_gain / avg_loss
    rsi = 100 - (100 / (1 + rs))
    return rsi
Moving Average Convergence Divergence (MACD) Line calculation:
def compute_MACD(data, short_period=12, long_period=26, signal_period=9):
    ema_short = data['Close'].ewm(span=short_period, adjust=False).mean()
    ema_long = data['Close'].ewm(span=long_period, adjust=False).mean()
    macd_line = ema_short - ema_long
    signal_line = macd_line.ewm(span=signal_period, adjust=False).mean()
    return macd line, signal line
Upper Bollinger Band Formula
def compute_BollingerBands(data, window=20, num_std=2):
    sma = data['Close'].rolling(window=window).mean()
    std = data['Close'].rolling(window=window).std()
    upper band = sma + num std * std
    lower_band = sma - num_std * std
    return sma, upper_band, lower_band
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Geodesic indicators calculation:
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def calculate indicators(data):
    data['SMA_short'] = data['Close'].rolling(window=20).mean()
    data['SMA_long'] = data['Close'].rolling(window=50).mean()
    macd line, signal line = compute MACD(data)
    data['MACD'] = macd_line
    data['MACD_signal'] = signal_line
    data['RSI'] = compute RSI(data['Close'], 14)
    bb_sma, upper_band, lower_band = compute_BollingerBands(data)
    data['BB_MA'] = bb_sma
    data['BB_upper'] = upper_band
    data['BB_lower'] = lower_band
    data = data.dropna()
    return data
Reshaping for LSTM Input Format
def prepare_lstm_data(data, seq_len):
    scaler = MinMaxScaler(feature range=(0, 1))
    scaled = scaler.fit_transform(data['Close'].values.reshape(-1, 1))
   X, y = [], []
    for i in range(seq len, len(scaled)):
       X.append(scaled[i - seq_len:i, 0])
       y.append(scaled[i, 0])
   X, y = np.array(X), np.array(y)
    X = np.reshape(X, (X.shape[0], X.shape[1], 1))
    return X, y, scaler
Model Compilation
def build_lstm_model(input_shape):
    model = Sequential()
    model.add(LSTM(50, return_sequences=True, input_shape=input_shape))
    model.add(LSTM(50))
   model.add(Dense(25))
    model.add(Dense(1))
    model.compile(optimizer='adam', loss='mean_squared_error')
    return model
Signal Generation Using numPy first-order diff:
def generate_lstm_signals(predictions):
    signals = np.where(np.diff(predictions.flatten(), prepend=predictions[0]) > 0, 1, -1)
    return signals
def backtest_strategy(data, signals):
    initial_capital = 10000.0
    portfolio = pd.DataFrame(index=data.index)
    portfolio['signal'] = signals
    portfolio['price'] = data['Close']
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portfolio['holdings'] = 0.0
    portfolio['cash'] = initial capital
    portfolio['total'] = initial_capital
    in_position = False
    shares = 0.0
    for i in range(1, len(portfolio)):
        if signals.iloc[i] == 1 and not in position:
            shares = portfolio['cash'].iloc[i - 1] / portfolio['price'].iloc[i]
           in_position = True
           portfolio.iloc[i, portfolio.columns.get loc('holdings')] = shares * portfolio['price'].iloc[i]
           portfolio.iloc[i, portfolio.columns.get_loc('cash')] = 0.0
        elif signals.iloc[i] == -1 and in_position:
           in position = False
           portfolio.iloc[i, portfolio.columns.get_loc('cash')] = shares * portfolio['price'].iloc[i]
           portfolio.iloc[i, portfolio.columns.get loc('holdings')] = 0.0
        else:
           if in position:
               portfolio.iloc[i, portfolio.columns.get loc('holdings')] = shares * portfolio['price'].iloc[i]
               portfolio.iloc[i, portfolio.columns.get loc('cash')] = 0.0
           else:
               portfolio.iloc[i, portfolio.columns.get_loc('cash')] = portfolio['cash'].iloc[i - 1]
        portfolio.iloc[i, portfolio.columns.get_loc('total')] = portfolio['cash'].iloc[i] + portfolio['holdings'].iloc[i]
    return portfolio
Drawdown Calculation
def compute max drawdown(series):
    cum max = series.cummax()
    dd = (series - cum_max) / cum_max
    return dd.min()
def evaluate_performance(portfolio):
    total return = portfolio['total'].iloc[-1] / portfolio['total'].iloc[0] - 1
    returns = portfolio['total'].pct_change().dropna()
    sharpe = (returns.mean() / returns.std()) * np.sqrt(252) if returns.std() != 0 else 0
    max dd = compute max drawdown(portfolio['total'])
    return total_return, sharpe, max_dd
Generating Scatter Plot Visualization
import matplotlib.pyplot as plt
def plot_price(data, signals):
    buys = data.index[signals == 1]
    sells = data.index[signals == -1]
    plt.figure(figsize=(14, 7))
    plt.plot(data.index, data['Close'], label='Close Price', color='blue')
    plt.plot(data.index, data['SMA_short'], label='SMA Short', color='green')
    plt.plot(data.index, data['SMA long'], label='SMA Long', color='red')
    # Scatter plot for Buy and Sell signals
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plt.scatter(buys, data.loc[buys, 'Close'], marker='^', color='green', label='Buy Signal', s=100)
    plt.scatter(sells, data.loc[sells, 'Close'], marker='v', color='red', label='Sell Signal', s=100)
    plt.title("AAPL Price with SMAs and Trading Signals")
    plt.xlabel("Date")
    plt.ylabel("Price")
    plt.legend()
    plt.grid(True)
    plt.show()
Portfolio Visualization
def plot_portfolio(portfolio):
    plt.figure(figsize=(14,7))
    plt.plot(portfolio.index, portfolio['total'], label='Total Portfolio Value', color='orange')
    plt.title("Portfolio Value Over Time")
    plt.xlabel("Date")
    plt.ylabel("Total Value ($)")
    plt.legend()
    plt.show()
LSTM final price prediction
def plot_predictions(test_data, y_test_inv, predictions_inv):
    plt.figure(figsize=(14,7))
    plt.plot(test data.index, y test inv.flatten(), label='Actual Price', color='blue')
    plt.plot(test_data.index, predictions_inv.flatten(), label='Predicted Price', color='red')
    plt.title("LSTM Price Prediction")
    plt.xlabel("Date")
    plt.ylabel("Price")
    plt.legend()
    plt.show()
Trading Signal Visualization
def main():
    ticker = 'AAPL'
    start = '2010-01-10'
    end = '2025-01-01'
    data = download_data(ticker, start, end)
    data = calculate_indicators(data)
    seq len = 60
    X, y, scaler = prepare_lstm_data(data, seq_len)
    split = int(0.8 * len(X))
   X_train, y_train = X[:split], y[:split]
   X_test, y_test = X[split:], y[split:]
    model = build_lstm_model((X_train.shape[1], 1))
    model.fit(X train, y train, epochs=10, batch size=32, verbose=0)
    predictions = model.predict(X_test)
    predictions_inv = scaler.inverse_transform(predictions)
    y test inv = scaler.inverse transform(y test.reshape(-1, 1))
    test_data = data[-len(y_test_inv):]
    lstm_signals = generate_lstm_signals(predictions_inv)
    lstm signals series = pd.Series(lstm signals, index=test data.index)
```

```
portfolio_lstm = backtest_strategy(test_data, lstm_signals_series)
ret_lstm, sharpe_lstm, dd_lstm = evaluate_performance(portfolio_lstm)
print("LSTM Strategy Return:", ret_lstm)
print("LSTM Strategy Sharpe:", sharpe_lstm)
print("LSTM Strategy Max Drawdown:", dd_lstm)
plot_predictions(test_data, y_test_inv, predictions_inv)
plot_portfolio(portfolio_lstm)
plot_price(test_data, lstm_signals_series)

if __name__ == '__main__':
    main()
```

🏂 /tmp/ipython-input-3-4269589612.py:2: FutureWarning: YF.download() has changed argument auto_adjust default to True data = yf.download(ticker, start=start, end=end)

/usr/local/lib/python3.11/dist-packages/keras/src/layers/rnn/rnn.py:200: UserWarning: Do not pass an `input_shape`/`input_dim` argument to a layer. When using Sequential models, prefe super().__init__(**kwargs)

• **0s** 11ms/step LSTM Strategy Return: 0.520233499397089 LSTM Strategy Sharpe: 0.8979670150205906 LSTM Strategy Max Drawdown: -0.16916856432516608

LSTM Price Prediction





