

Building an Agentic Financial Analyst with LangGraph and OpenAI

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In the world of stock trading, investors rely on various tools and methods to make informed decisions. One such approach is **fundamental analysis**, which evaluates a company's financial health and stock performance to provide actionable insights. With the advancement of AI and machine learning, stock analysis can now be automated to a great extent. In this post, we will explore how to create a **stock performance analysis agent** using LangChain, LangGraph, and Yahoo Finance, leveraging real-time stock data and key technical indicators.

Whether you're a finance enthusiast, developer, or data scientist, this step-by-step tutorial will empower you to create your own intelligent agent. Let's dive in!

What this Agentic Financial Analyst will do?

- Fetches stock price data using Yahoo Finance.
- Calculates technical indicators like RSI, MACD, VWAP, and more.
- Evaluate financial metrics such as P/E ratio, Debt-to-Equity, and Profit Margins.
- Provides a structured, AI-generated analysis using OpenAI's powerful language models.

Tools We'll Use

1. **LangGraph**: A library for orchestrating tools and building conversational agents.
2. **OpenAI GPT-4**: For generating intelligent and structured financial insights.
3. **yfinance**: To retrieve stock prices and financial ratios.
4. **ta (Technical Analysis Library)**: For calculating key technical indicators.
5. **Python libraries**: `pandas`, `dotenv`, and `datetime` for data manipulation and environment setup.

Step 1: Setting Up the Environment

Start by installing the required libraries:

```
pip install -U langgraph langchain langchain_openai pandas ta python-dotenv yfin
```

Set up a `.env` file to securely store your OpenAI API key:

```
OPENAI_API_KEY=your_openai_api_key_here
```

Step 2: Tools for Analyst

Fetching Stock Prices: This tool fetches the stock's historical data and computes several technical indicators.

```
from typing import Union, Dict, Set, List, TypedDict, Annotated
import pandas as pd
from langchain_core.tools import tool
```

```

import yfinance as yf
from ta.momentum import RSIIndicator, StochasticOscillator
from ta.trend import SMAIndicator, EMAIndicator, MACD
from ta.volume import volume_weighted_average_price

@tool
def get_stock_prices(ticker: str) -> Union[Dict, str]:
    """Fetches historical stock price data and technical indicator for a given t
    try:
        data = yf.download(
            ticker,
            start=dt.datetime.now() - dt.timedelta(weeks=24*3),
            end=dt.datetime.now(),
            interval='1wk'
        )
        df= data.copy()
        data.reset_index(inplace=True)
        data.Date = data.Date.astype(str)

        indicators = {}

        rsi_series = RSIIndicator(df['Close'], window=14).rsi().iloc[-12:]
        indicators["RSI"] = {date.strftime('%Y-%m-%d'): int(value)
                             for date, value in rsi_series.dropna().to_dict().items()}

        sto_series = StochasticOscillator(
            df['High'], df['Low'], df['Close'], window=14).stoch().iloc[-12:]
        indicators["Stochastic_Oscillator"] = {
            date.strftime('%Y-%m-%d'): int(value)
            for date, value in sto_series.dropna().to_dict().items()}

        macd = MACD(df['Close'])
        macd_series = macd.macd().iloc[-12:]
        indicators["MACD"] = {date.strftime('%Y-%m-%d'): int(value)
                              for date, value in macd_series.to_dict().items()}

        macd_signal_series = macd.macd_signal().iloc[-12:]
        indicators["MACD_Signal"] = {date.strftime('%Y-%m-%d'): int(value)
                                      for date, value in macd_signal_series.to_dict().items()}

        vwap_series = volume_weighted_average_price(
            high=df['High'], low=df['Low'], close=df['Close'],
            volume=df['Volume'],
        ).iloc[-12:]
        indicators["vwap"] = {date.strftime('%Y-%m-%d'): int(value)
                              for date, value in vwap_series.to_dict().items()}

        return {'stock_price': data.to_dict(orient='records'),
                'indicators': indicators}

    except Exception as e:
        return f"Error fetching price data: {str(e)}"

```

Financial Ratios: This tool retrieves key financial health ratios.

```
@tool
def get_financial_metrics(ticker: str) -> Union[Dict, str]:
    """Fetches key financial ratios for a given ticker."""
    try:
        stock = yf.Ticker(ticker)
        info = stock.info
        return {
            'pe_ratio': info.get('forwardPE'),
            'price_to_book': info.get('priceToBook'),
            'debt_to_equity': info.get('debtToEquity'),
            'profit_margins': info.get('profitMargins')
        }
    except Exception as e:
        return f"Error fetching ratios: {str(e)}"
```

Step 3: Building the LangGraph

LangGraph allows us to orchestrate tools and manage conversational logic efficiently.

1. Defining the Graph

We start by defining a `StateGraph` to manage the flow:

```
from langgraph.graph import StateGraph, START, END

class State(TypedDict):
    messages: Annotated[list, add_messages]
    stock: str

graph_builder = StateGraph(State)
```

2. Defining OpenAI and Binding Tools

We integrate the tools into LangGraph and create a feedback loop for analysis

```
import dotenv
dotenv.load_dotenv()
```

```
from langchain_openai import ChatOpenAI

llm = ChatOpenAI(model='gpt-4o-mini')

tools = [get_stock_prices, get_financial_metrics]
llm_with_tool = llm.bind_tools(tools)
```

3. Analyst Node

The prompt ensures the AI understands its role and delivers structured output.

```
FUNDAMENTAL_ANALYST_PROMPT = """
You are a fundamental analyst specializing in evaluating company (whose symbol is {company}).

You have access to the following tools:
1. **get_stock_prices**: Retrieves the latest stock price, historical price data
2. **get_financial_metrics**: Retrieves key financial metrics, such as revenue,

### Your Task:
1. **Input Stock Symbol**: Use the provided stock symbol to query the tools and
2. **Analyze Data**: Evaluate the results from the tools and identify potential
3. **Provide Summary**: Write a concise, well-structured summary that highlights
    - Recent stock price movements, trends and potential resistance.
    - Key insights from technical indicators (e.g., whether the stock is overbought)
    - Financial health and performance based on financial metrics.

### Constraints:
- Use only the data provided by the tools.
- Avoid speculative language; focus on observable data and trends.
- If any tool fails to provide data, clearly state that in your summary.

### Output Format:
Respond in the following format:
"stock": "<Stock Symbol>",
"price_analysis": "<Detailed analysis of stock price trends>",
"technical_analysis": "<Detailed time series Analysis from ALL technical indicators>",
"financial_analysis": "<Detailed analysis from financial metrics>",
"final Summary": "<Full Conclusion based on the above analyses>"
"Asked Question Answer": "<Answer based on the details and analysis above>"

Ensure that your response is objective, concise, and actionable."""

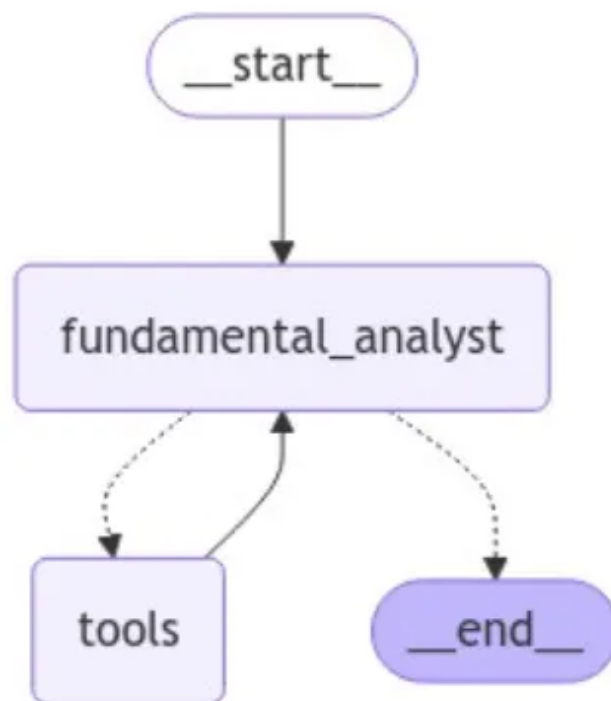
def fundamental_analyst(state: State):
    messages = [
        SystemMessage(content=FUNDAMENTAL_ANALYST_PROMPT.format(company=state['symbol'])),
        HumanMessage(content=state['messages'])
    ]
    return {
        'messages': llm_with_tool.invoke(messages)
    }
```

```
graph_builder.add_node('fundamental_analyst', fundamental_analyst)
graph_builder.add_edge(START, 'fundamental_analyst')
```

4. Adding tool to graph and compile

```
graph_builder.add_node(ToolNode(tools))
graph_builder.add_conditional_edges('fundamental_analyst', tools_condition)
graph_builder.add_edge('tools', 'fundamental_analyst')

graph = graph_builder.compile()
```



graph

5. Executing the Graph

```
events = graph.stream({'messages': [('user', 'Should I buy this stock?')],
                      'stock': 'TSLA'}, stream_mode='values')
for event in events:
    if 'messages' in event:
        event['messages'][-1].pretty_print()
```

Sample Output


```
{  
  "stock": "TSLA",  
  "price_analysis": "The recent stock price for TSLA has shown volatility, with  
  "technical_analysis": "The technical indicators present a mixed outlook. The R  
  "financial_analysis": "TSLA's financial metrics indicate a high valuation rela  
  "final Summary": "In summary, TSLA shows strong recent price recovery with pot  
  "Asked Question Answer": "Given the current overbought indicators and high val  
}
```

Future Improvements

Incorporating the idea of a **full portfolio manager agent** is a fantastic improvement to the project! With multiple specialized teams working under one umbrella, the agent can be significantly enhanced to cover a wide range of areas and provide a comprehensive portfolio management tool.

Conclusion

Building an Agentic Financial Analyst is not only a great way to learn about AI and financial analysis but also a stepping stone to creating powerful, real-world applications. Try it yourself and see the power of automation in action!

Check out the complete project on [GitHub](#). Fork it, experiment, and let me know your thoughts!

What features would you add to this project? Drop a comment below, and let's discuss!

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Written by Abhinav Kumar

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AI enthusiast, designing and deploying AI agents that extract meaningful insights and automate complex tasks is my idea of a good time.