

# LangChain: Agents

## Outline:

- Using built in LangChain tools: DuckDuckGo search and Wikipedia
- Defining your own tools

```
In [18]: import os

from dotenv import load_dotenv, find_dotenv
_ = load_dotenv(find_dotenv()) # read local .env file

import warnings
warnings.filterwarnings("ignore")
```

## Built-in LangChain tools

```
In [19]: #!/pip install -U wikipedia
```

```
In [20]: from langchain.agents.agent_toolkits import create_python_agent
from langchain.agents import load_tools, initialize_agent
from langchain.agents import AgentType
from langchain.tools.python.tool import PythonREPLTool
from langchain.python import PythonREPL
from langchain.chat_models import ChatOpenAI
```

```
In [21]: llm = ChatOpenAI(temperature=0)
```

```
In [22]: tools = load_tools(["llm-math","wikipedia"], llm=llm)
```

```
In [23]: agent= initialize_agent(
    tools,
    llm,
    agent=AgentType.CHAT_ZERO_SHOT_REACT_DESCRIPTION,
    handle_parsing_errors=True,
    verbose = True)
```

```
In [24]: agent("What is the 25% of 300?")
```

> Entering new AgentExecutor chain...

Thought: We need to calculate 25% of 300, which means we need to multiply 300 by 0.25.

Action:

```
```  
{  
  "action": "Calculator",  
  "action_input": "300*0.25"  
}  
```
```

Observation: Answer: 75.0

Thought: The calculator tool returned the answer 75.0, which is 25% of 300.

Final Answer: 25% of 300 is 75.0.

> Finished chain.

```
{'input': 'What is the 25% of 300?', 'output': '25% of 300 is 75.0.'}
```

```
In [25]: question = "Tom M. Mitchell is an American computer scientist \
and the Founders University Professor at Carnegie Mellon University (CMU)\
what book did he write?"
result = agent(question)
```

> Entering new AgentExecutor chain...

Thought: I should use Wikipedia to find the answer to this question.

Action:

```

```
{
  "action": "Wikipedia",
  "action_input": "Tom M. Mitchell"
}
```

Observation: Page: Tom M. Mitchell

Summary: Tom Michael Mitchell (born August 9, 1951) is an American computer scientist and the Founders University Professor at Carnegie Mellon University (CMU). He is a founder and former Chair of the Machine Learning Department at CMU. Mitchell is known for his contributions to the advancement of machine learning, artificial intelligence, and cognitive neuroscience and is the author of the textbook Machine Learning. He is a member of the United States National Academy of Engineering since 2010. He is also a Fellow of the American Academy of Arts and Sciences, the American Association for the Advancement of Science and a Fellow and past President of the Association for the Advancement of Artificial Intelligence. In October 2018, Mitchell was appointed as the Interim Dean of the School of Computer Science at Carnegie Mellon.

Page: Tom Mitchell (Australian footballer)

Summary: Thomas Mitchell (born 31 May 1993) is a professional Australian rules footballer playing for the Collingwood Football Club in the Australian Football League (AFL). He previously played for the Sydney Swans from 2012 to 2016, and the Hawthorn Football Club between 2017 and 2022. Mitchell won the Brownlow Medal as the league's best and fairest player in 2018 and set the record for the most disposals in a VFL/AFL match, accruing 54 in a game against Collingwood during that season.

Thought:The book that Tom M. Mitchell wrote is "Machine Learning".

Action:

```

```
{
  "action": "Wikipedia",
  "action_input": "Machine Learning (book)"
}
```

Observation: Page: Machine learning

Summary: Machine learning (ML) is a field devoted to understanding and building methods that let machines "learn" – that is, methods that leverage data to improve computer performance on some set of tasks. Machine learning algorithms build a model based on sample data, known as training data, in order to make predictions or decisions without being explicitly programmed to do so. Machine learning algorithms are used in a wide variety of applications, such as in medicine, email filtering, speech recognition, agriculture, and computer vision, where it is difficult or unfeasible to develop conventional algorithms to perform the needed tasks. A subset of machine learning is closely related to computational statistics, which focuses on making predictions using computers,

but not all machine learning is statistical learning. The study of mathematical optimization delivers methods, theory and application domains to the field of machine learning. Data mining is a related field of study, focusing on exploratory data analysis through unsupervised learning. Some implementations of machine learning use data and artificial neural networks in a way that mimics the working of a biological brain. In its application across business problems, machine learning is also referred to as predictive analytics.

Page: Quantum machine learning

Summary: Quantum machine learning is the integration of quantum algorithms within machine learning programs. The most common use of the term refers to machine learning algorithms for the analysis of classical data executed on a quantum computer, i.e. quantum-enhanced machine learning. While machine learning algorithms are used to compute immense quantities of data, quantum machine learning utilizes qubits and quantum operations or specialized quantum systems to improve computational speed and data storage done by algorithms in a program. This includes hybrid methods that involve both classical and quantum processing, where computationally difficult subroutines are outsourced to a quantum device. These routines can be more complex in nature and executed faster on a quantum computer. Furthermore, quantum algorithms can be used to analyze quantum states instead of classical data. Beyond quantum computing, the term "quantum machine learning" is also associated with classical machine learning methods applied to data generated from quantum experiments (i.e. machine learning of quantum systems), such as learning the phase transitions of a quantum system or creating new quantum experiments. Quantum machine learning also extends to a branch of research that explores methodological and structural similarities between certain physical systems and learning systems, in particular neural networks. For example, some mathematical and numerical techniques from quantum physics are applicable to classical deep learning and vice versa. Furthermore, researchers investigate more abstract notions of learning theory with respect to quantum information, sometimes referred to as "quantum learning theory".

Page: Timeline of machine learning

Summary: This page is a timeline of machine learning. Major discoveries, achievements, milestones and other major events in machine learning are included. Thought: Tom M. Mitchell wrote the book "Machine Learning".

Final Answer: "Machine Learning".

> Finished chain.

## Python Agent

```
In [26]: agent = create_python_agent(  
        llm,  
        tool=PythonREPLTool(),  
        verbose=True  
    )
```

```
In [27]: customer_list = [
    ["Harrison", "Chase"],
    ["Lang", "Chain"],
    ["Dolly", "Too"],
    ["Elle", "Elem"],
    ["Geoff", "Fusion"],
    ["Trance", "Former"],
    ["Jen", "Ayai"]
]
```

```
In [28]: agent.run(f"""Sort these customers by \
last name and then first name \
and print the output: {customer_list}""")
```

> Entering new AgentExecutor chain...

I can use the sorted() function to sort the list of customers by last name and then first name. I will need to provide a key function to sorted() that returns a tuple of the last name and first name in that order.

Action: Python REPL

Action Input:

```

```
customers = [['Harrison', 'Chase'], ['Lang', 'Chain'], ['Dolly', 'Too'], ['Elle', 'Elem'], ['Geoff', 'Fusion'], ['Trance', 'Former'], ['Jen', 'Ayai']]
sorted_customers = sorted(customers, key=lambda x: (x[1], x[0]))
for customer in sorted_customers:
    print(customer)
```
```

Observation: ['Jen', 'Ayai']

['Lang', 'Chain']

['Harrison', 'Chase']

['Elle', 'Elem']

['Trance', 'Former']

['Geoff', 'Fusion']

['Dolly', 'Too']

Thought: The customers have been sorted by last name and then first name.

Final Answer: [['Jen', 'Ayai'], ['Lang', 'Chain'], ['Harrison', 'Chase'], ['Elle', 'Elem'], ['Trance', 'Former'], ['Geoff', 'Fusion'], ['Dolly', 'Too']]

> Finished chain.

"[['Jen', 'Ayai'], ['Lang', 'Chain'], ['Harrison', 'Chase'], ['Elle', 'Elem'], ['Trance', 'Former'], ['Geoff', 'Fusion'], ['Dolly', 'Too']]"

**View detailed outputs of the chains**

```
In [29]: import langchain
langchain.debug=True
agent.run(f""Sort these customers by \
last name and then first name \
and print the output: {customer_list}""")
langchain.debug=False
```

[chain/start] [1:chain:AgentExecutor] Entering Chain run with input:

```
{
  "input": "Sort these customers by last name and then first name and print the output: [['Harrison', 'Chase'], ['Lang', 'Chain'], ['Dolly', 'Too'], ['Elle', 'Elem'], ['Geoff', 'Fusion'], ['Trance', 'Former'], ['Jen', 'Aya i']]"
}
```

[chain/start] [1:chain:AgentExecutor > 2:chain:LLMChain] Entering Chain run with input:

```
{
  "input": "Sort these customers by last name and then first name and print the output: [['Harrison', 'Chase'], ['Lang', 'Chain'], ['Dolly', 'Too'], ['Elle', 'Elem'], ['Geoff', 'Fusion'], ['Trance', 'Former'], ['Jen', 'Aya i']]",
  "agent_scratchpad": "",
  "stop": [
    "\nObservation:",
    "\n\n\tObservation:"
  ]
},
```

## Define your own tool

```
In [30]: #!/pip install DateTime
```

```
In [31]: from langchain.agents import tool
from datetime import date
```

```
In [32]: @tool
def time(text: str) -> str:
    """Returns today's date, use this for any \
questions related to knowing today's date. \
The input should always be an empty string, \
and this function will always return today's \
date - any date mathematics should occur \
outside this function."""
    return str(date.today())
```

```
In [33]: agent = initialize_agent(
    tools + [time],
    llm,
    agent=AgentType.CHAT_ZERO_SHOT_REACT_DESCRIPTION,
    handle_parsing_errors=True,
    verbose = True)
```

**Note:**

The agent will sometimes come to the wrong conclusion (agents are a work in progress!).

If it does, please try running it again.

```
In [34]: try:
          result = agent("whats the date today?")
        except:
          print("exception on external access")
```

> Entering new AgentExecutor chain...

Thought: I need to use the `time` tool to get today's date.

Action:

```

```
{
  "action": "time",
  "action_input": ""
}
```

Observation: 2023-06-24

Thought: I have successfully retrieved today's date using the `time` tool.

Final Answer: Today's date is 2023-06-24.

> Finished chain.

In [ ]:

In [ ]:

In [ ]:

In [ ]:

In [ ]:

In [ ]:

In [ ]:

In [ ]:

In [ ]:



