

**AGENTIC AI**

# **BUILD A MULTI-AGENT APPLICATION WITH CREWAI**

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# AGENDA

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Objectives and Outcomes (2 min)

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Intro to CrewAI (15 min)

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Assignment 1 - Your First Agent (15 min)

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Assignment 2 - Content Creation with Guardrails (20 min)

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Assignment 3 - Fraud Detection Workflow (20 min)

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Assignment 4 - AI-Powered Onboarding System (15 min)


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Wrap-Up (3 min)

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# ABOUT ME



**Alessandro Romano**  
Data Scientist | Speaker |  
Musician  
Hamburg, Hamburg  
 Kuehne+Nagel

## Experience



### Senior Data Scientist

Kuehne+Nagel · Full-time

Mar 2023 - Present · 2 yrs

Hamburg, Germany · Hybrid



### Public Speaker/Data Science Advocate

Self-employed

Aug 2019 - Present · 5 yrs 7 mos

Hybrid

I am passionate about making Data and AI accessible and easy to understand for everyone. With a knack for distilling complex concepts into clear, actionable insights, I regularly share my use cases and learnings. ...see more

 Data Science, Public Speaking and +3 skills



Alessandro Romano - Talks



### Data Scientist

FREE NOW · Full-time

Jun 2021 - Jan 2023 · 1 yr 8 mos

Hamburg, Germany · Hybrid



### Data Scientist

Cargonexx GmbH

Jun 2018 - May 2021 · 3 yrs

Hamburg Area, Germany

## Education



### Università di Pisa

Master's degree, Data Science

2015 - 2018

Grade: 110/110 cum Laude

Master's Thesis Topic: "Anomaly Detection for Time Series data - Deep Learning applied to the Auto-Motive field"



### Università degli Studi di Bari

Bachelor's degree, Computer Science

2011 - 2015

Bachelor Thesis Topic: "Development of a multivariate algorithm for wind speed forecasting"

# OBJECTIVES AND OUTCOMES

## **Objectives**

- Understand CrewAI's architecture and agentic AI concepts
- Learn to define agents, tasks, crews, and tools
- Explore structured outputs, flow control, and async execution
- Apply concepts through hands-on projects

## **Expected Outcomes**

- Ability to build and run agent workflows with CrewAI
- Experience designing multi-agent flows with guardrails and logic
- Practical understanding of hierarchical and conditional orchestration
- Ready-to-use code for real-world use cases (content, fraud, onboarding)

# AGENTIC AI



Agents take actions



Use tools



Implement reasoning through LLM



Access an additional knowledge



Interact with other agents to accomplish final goal



Able to take decisions in a dynamic environment

# IMPLEMENT REASONING

- How do we reason?
- Can we call it reasoning?
- Agents are not new!
- Reasoning is implemented through an LLM



# CAN WE CALL IT AN AGENT?



A deterministic agent or service that implements a deterministic logic with no llm, could also be classified as an agent, with the difference that is probably confined in a specific environment, where the laws in it don't change.

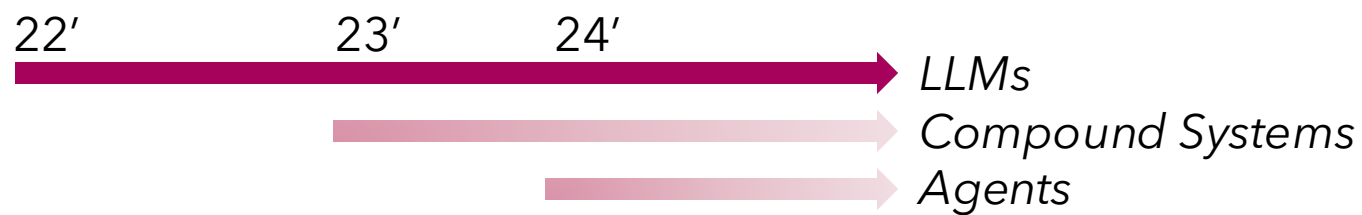


Going from point A to point B is a trivial task, but if something unexpected happens in the middle, an Agent can read the environment and reason through an LLM in order to take the most appropriate next action.



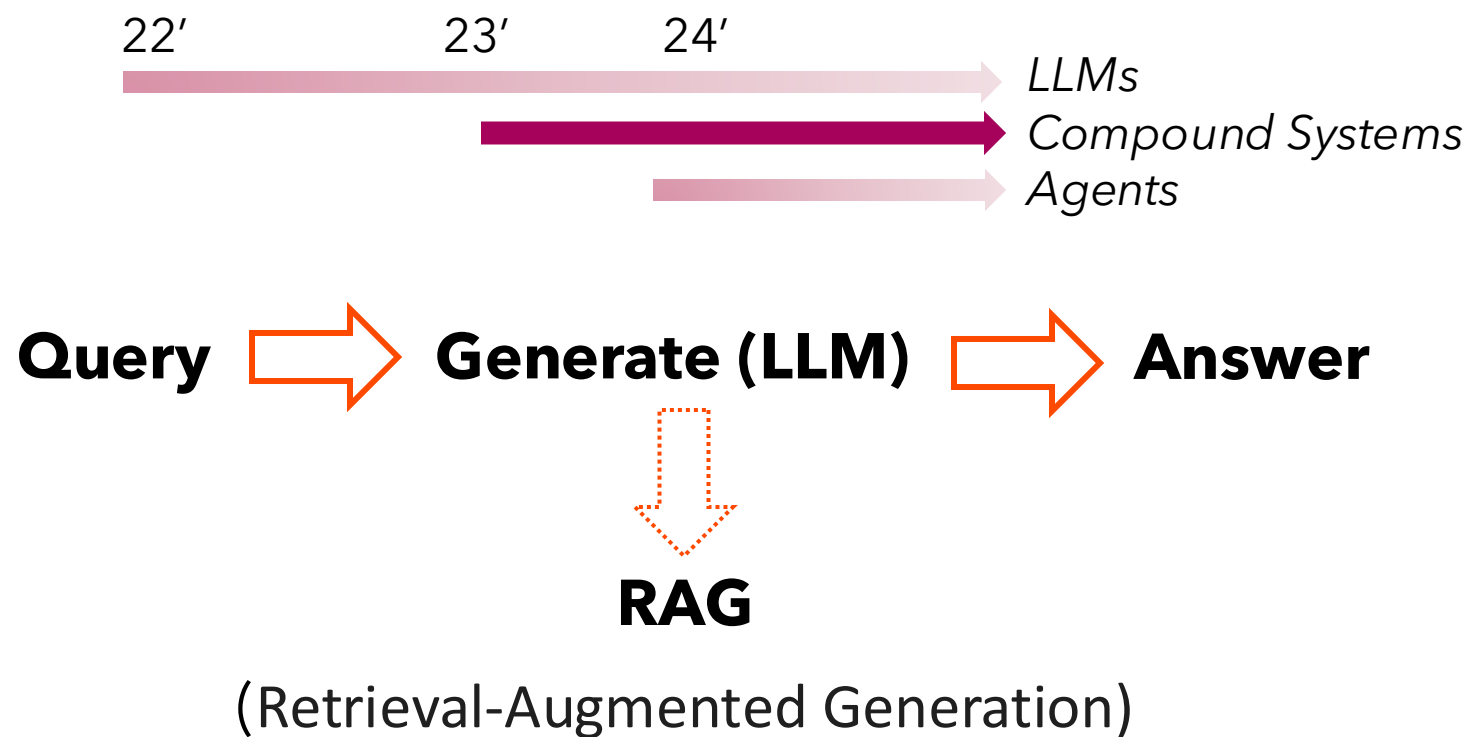
Therefore, Agents are not always needed.

# THE DAWN OF LLMS

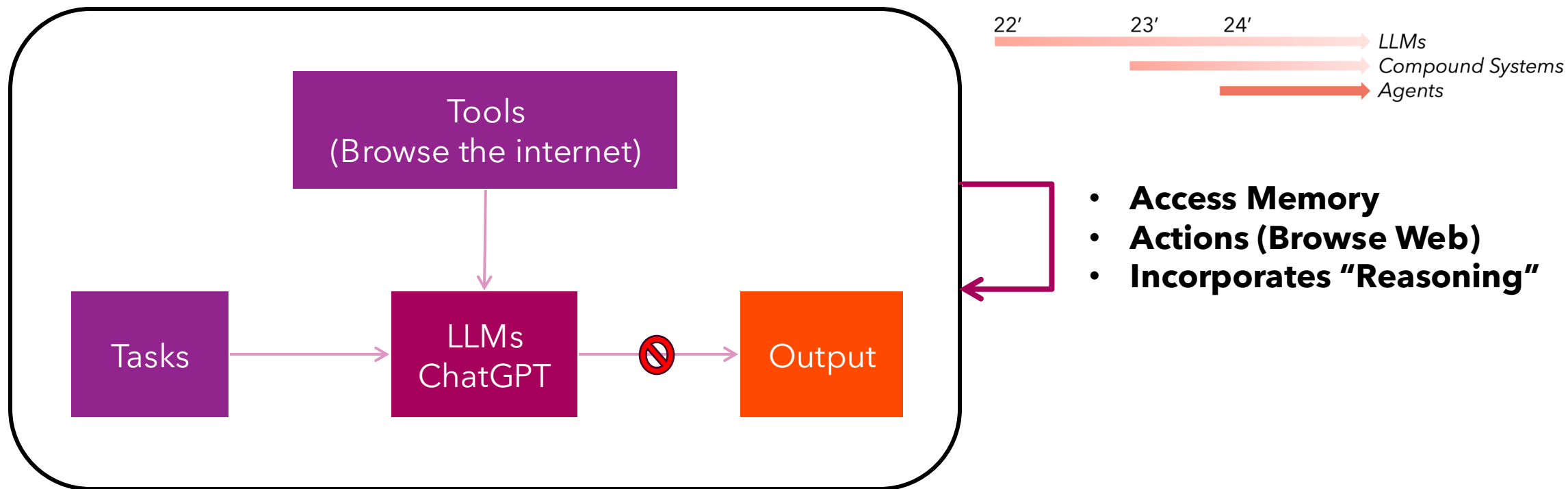




# COMPOUND SYSTEMS



# LLM AGENTS



# LLM AGENTS FRAMEWORKS

Some interesting mentions:

1. **LangChain**: Versatile for building applications with LLMs, including chains, agents, and memory.
2. **Hugging Face Transformers**: A leading library with extensive model options and community support.
3. **LlamaIndex**: Known for integrating LLMs with external data sources like databases and documents.
4. **Haystack**: Strong in question-answering and retrieval-based applications.
5. **DeepSpeed**: Optimizes large models for efficiency and scalability in production environments.



**Hugging Face**

# CREWAI

*Cutting-edge framework for orchestrating role-playing, autonomous AI agents. By fostering collaborative intelligence, CrewAI empowers agents to work together seamlessly, tackling complex tasks:*

- Provide **simple** tools for dealing with LLMs
- Implement "**orchestration**" strategies
- Core Concepts like **Agents, Tasks, and Crews**



# AGENTS

An **agent** is an **autonomous unit** programmed to:

- Perform tasks
- Make decisions
- Communicate with other agents

Think of an agent as a member of a team, with specific skills and a particular job to do. Agents can have different roles like *Researcher*, *Writer*, or *Customer Support*, each contributing to the overall goal of the crew.

```
from crewai import Agent

agent2 = Agent(
    role="agent role",
    goal="summarize the short bio for {input} and if needed do more research",
    backstory="agent backstory",
    verbose=True,
)
```

<https://docs.crewai.com/core-concepts/Agents/>

# TASKS

In the crewAI framework, **tasks** are assignments given to agents, including details like descriptions, responsible agents, and required tools. Tasks can be collaborative, involving multiple agents working together, with the Crew's process managing coordination for better teamwork and efficiency.

```
from crewai import Task

task = Task(
    description='Find and summarize the latest and most relevant news on AI',
    agent=sales_agent,
    expected_output='A bullet list summary of the top 5 most important AI news',
)
```

<https://docs.crewai.com/core-concepts/Tasks/>

# CREWS

A **crew** in crewAI represents a collaborative group of agents working together to achieve a set of tasks. Each crew defines the strategy for task execution, agent collaboration, and the overall workflow.

```
from crewai import Crew

# Assemble the crew with a sequential process
my_crew = Crew(
    agents=[researcher, writer],
    tasks=[research_task, write_article_task],
    process=Process.sequential,
    full_output=True,
    verbose=True,
)
```

<https://docs.crewai.com/core-concepts/Crews/>

# PROCESSES

In CrewAI, **processes** orchestrate the execution of tasks by agents, akin to project management in human teams. These processes ensure tasks are distributed and executed efficiently, in alignment with a predefined strategy:

- **Sequential:** Executes tasks one after another in a specific order.
- **Hierarchical:** Organizes tasks using a chain of command, where a manager model or agent oversees and delegates tasks.
- **Consensual** (Planned): Intended for future development, this process will enable agents to collaborate and make decisions democratically on task execution.

```
from crewai import Crew

# Assemble the crew with a sequential process
my_crew = Crew(
    agents=[researcher, writer],
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    full_output=True,
    verbose=True,
)
```



Process.sequential



# TOOLS

CrewAI **tools** empower agents with capabilities like web searching, data analysis, collaboration, and task delegation, allowing users to build custom tools or leverage CrewAI and LangChain tools for seamless teamwork and complex interactions.

```
from crewai import Agent, Task, Crew
from crewai_tools import SerperDevTool

# Set up the agent with a web search tool
researcher = Agent(
    role='Market Research Analyst',
    tools=[SerperDevTool()],
    verbose=True
)

# Define a single task
research = Task(description='Research AI industry trends.', agent=researcher)

# Assemble and run the crew
Crew(agents=[researcher], tasks=[research], planning=True).kickoff()
```

# GUARDRAILS

With **Flow**, we can create conditional logics that help us to control the application's internal flow.

For instance, we can handle those cases where the **LLM hallucinates** and trigger the agent N number of times until we get a "safe" output.

```
from crewai.flow.flow import Flow, and_, listen, start

class AndExampleFlow(Flow):

    @start()
    def start_method(self):
        self.state["greeting"] = "Hello from the start method"

    @listen(start_method)
    def second_method(self):
        self.state["joke"] = "What do computers eat? Microchips."

    @listen(and_(start_method, second_method))
    def logger(self):
        print("---- Logger ----")
        print(self.state)

flow = AndExampleFlow()
flow.kickoff()
```

<https://docs.crewai.com/concepts/flows>

# CONDITIONAL TASKS

**Conditional Tasks** in CrewAI let workflows adapt based on previous task results, enabling selective execution for more flexible and efficient processes.

```
class EventOutput(BaseModel):
    events: List[str]

task1 = Task(
    description="Fetch data about events in San Francisco using Serper tool",
    expected_output="List of 10 things to do in SF this week",
    agent=data_fetcher_agent,
    output_pydantic=EventOutput,
)

conditional_task = ConditionalTask(
    description="""
        Check if data is missing. If we have less than 10 events,
        fetch more events using Serper tool so that
        we have a total of 10 events in SF this week..
    """,
    expected_output="List of 10 Things to do in SF this week",
    condition=is_data_missing,
    agent=data_processor_agent,
)
```

# TASKS INPUT/OUTPUT

```
get_recipe = Task(
    description="Give me recipe given my food allergies {food_allergies}.",
    agent=chef,
)

...

crew.kickoff(inputs={'food_allergies': 'Peanuts'})
```

Input:

- Use *inputs* field in *kickoff()*
- Wrap your tasks into a function

The expected output is a safe yet powerfull way to make sure the next agent is able access the input data correctly.

```
task = Task(
    ...
    expected_output="image_url: URL of the generated image",
    ...
)
```

# ACCESS TASKS OUTPUT

```
from crewai import Agent, Crew, Task

# Define an agent
weather_agent = Agent(
    role="Weather Analyst",
    goal="Fetch and summarize today's weather forecast"
)

# Example task
task = Task(
    description='Summarize today's weather forecast for New York City',
    expected_output='A short weather summary including temperature and conditions',
    agent=weather_agent
)


# Execute the crew
crew = Crew(
    agents=[weather_agent],
    tasks=[task],
    verbose=True
)

result = crew.kickoff()

# Accessing the task output
output = task.output

print(f"Task Description: {output.description}")
print(f"Task Summary: {output.summary}")
print(f"Raw Output: {output.raw}")
if output.json_dict:
    print(f"JSON Output: {output.json_dict}")
if output.pydantic:
    print(f"Pydantic Output: {output.pydantic}")
```

# ASYNCHRONOUS CREW KICKOFF



```
# Async function to kickoff multiple crews asynchronously and wait for all to finish
async def async_multiple_crews():
    result_1 = crew_1.kickoff_async(inputs={"ages": [25, 30, 35, 40, 45]})
    result_2 = crew_2.kickoff_async(inputs={"ages": [20, 22, 24, 28, 30]})

    # Wait for both crews to finish
    results = await asyncio.gather(result_1, result_2)

    for i, result in enumerate(results, 1):
        print(f"Crew {i} Result:", result)

# Run the async function
asyncio.run(async_multiple_crews())
```

# HANDS-ON GUIDELINES



Explore the  
slides and the  
website



Feel free to ask  
me anything



Share your ideas  
with fellow  
attendees



Lend a hand to  
your neighbors



Get creative –  
and most  
importantly, have  
fun!



<https://pigna90.github.io/crewai-workshop-pyconde-2025/>



The background features a gradient of pink and orange hues. A large, semi-transparent semi-circle is positioned on the left side, overlapping the gradient. The text "THANK YOU" is centered in the lower half of the image.

**THANK YOU**