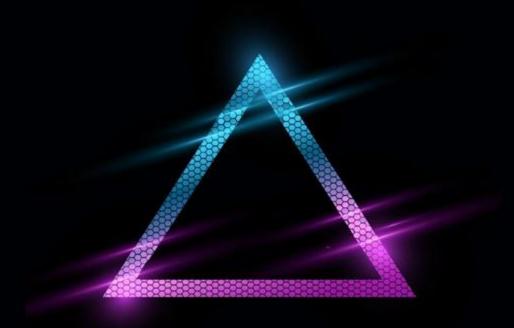


Workshop

Mastering Real-World Multi-Agent Systems



Speaker

Alessandro Romano Senior Data Scientist



Agenda

Objectives and Outcomes

Intro to CrewAl

Assignment 1 – Your First Agent

Assignment 2 - Content Creation with Guardrails

Assignment 3 - Fraud Detection Workflow

Assignment 4 – Al-Powered Onboarding System

Assignment 5 – Al-Powered Interview System

Wrap-Up





About Me

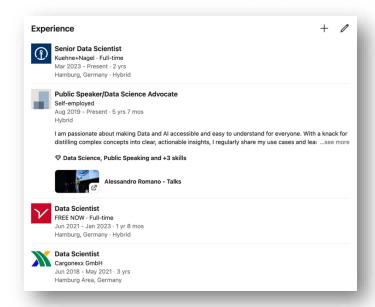


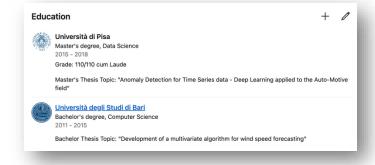
Musician

Hamburg, Hamburg



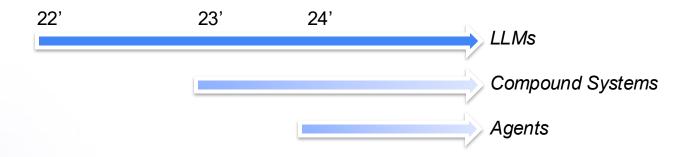








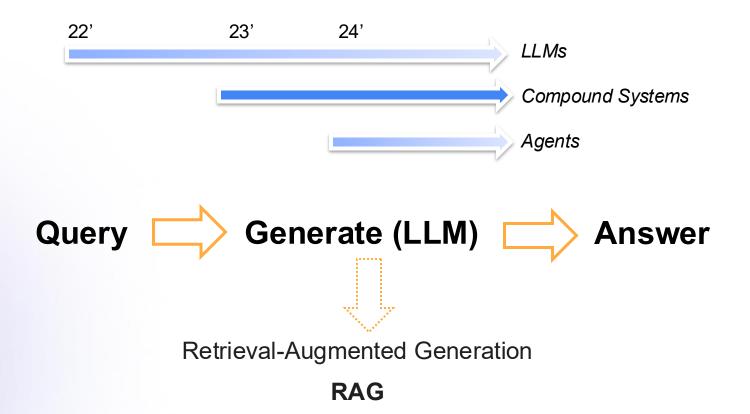
LLMs





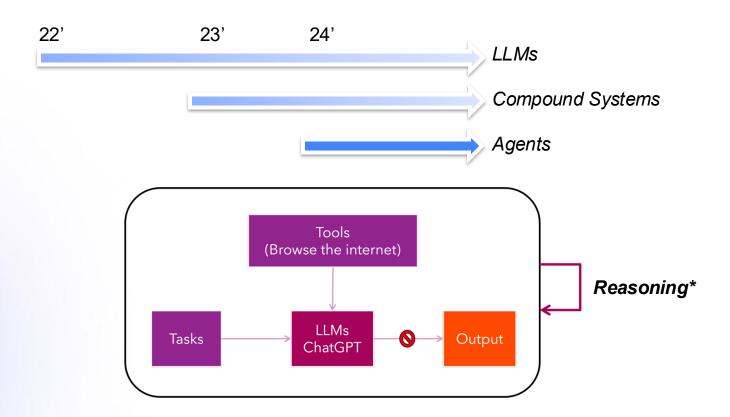


Compound Systems





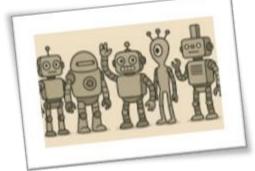
Al Agents

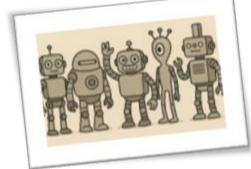






The "Agents" Family







Rule-based agents (e.g., **expert systems** in the 1980s)



Reinforcement learning agents in Al (e.g., Deep Q-learning agents, AlphaGo)



Robotic agents that act in the physical world



Multi-agent systems in simulations and games



Browser automation bots, crawlers, and personal assistants (like Clippy or Siri v1)



Al Agents vs. Agentic Al





Agentic AI refers to systems designed to act autonomously toward goals, while **AI agents** are the individual entities that perceive, reason, and take actions within those systems.



What does »Agentic AI» Solve?



Multi-step reasoning

Example: Market research → Competitor comparison → Strategy suggestion



Dynamic workflows

Example: Customer onboarding with conditional steps and retries



Planning and goal decomposition

Example: Travel planner that books, rebooks, and adapts



Tool/API interaction

Example: SEO optimizer that audits, edits, and deploys via CMS



Multi-role collaboration

Example: Product strategy crew (researcher → analyst → copywriter)



CrewAl

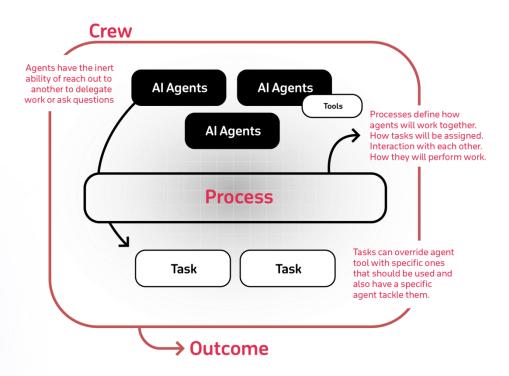
CrewAI is a framework for building AI agents that reason, collaborate, and act autonomously—so you can focus on outcomes, not infrastructure.

- Implements reasoning and orchestration for you
- Agentic Al under the hood—no need to build it from scratch
- **LLM-agnostic**: use any model you prefer
- Focus on solving your problem, not on building agent infrastructure



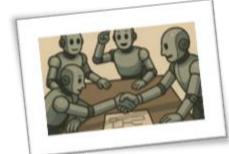


CrewAl Architecture



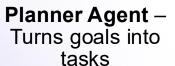


Build Your Team!











Content Creator Agent – Writes content



API Caller Agent Connects to **APIs**



Evaluator Agent - Checks results

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,
)
```



Tasks

In the crewAl 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',
)
```



Crews

A **crew** in crewAl 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,
)
```



Tools

CrewAl **tools** empower agents with capabilities like web searching, data analysis, collaboration, and task delegation, allowing users to build custom tools or leverage CrewAl 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()
```

More Advanced Features

CrewAl **Flow** is used for creating conditional logics. Also, you have full control on what type of **memory** you want to use.

```
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()
```



Hand-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!

Ski Trip Use Case





The Problem

When I started skiing, I struggled with fragmented information and lacked the time and knowledge to plan effectively.

My main challenges were:

- Finding a ski spot accessible to at least three friends
- Renting equipment
- Choosing the best travel option
- Ensuring it was affordable for everyone
- Matching our different ski levels

Optimizing for all these factors made choosing the right destination a complex task.





Why Orchestrating Al Agents?

Extensive information to process

Optimization challenges

Nondeterministic optimization problem Timeconsuming internet research

Reproducible for future trips

Scalable solution

I want to use LLMs for something practical



Ski Destination Researcher Agent

```
destination_finder = Agent(
    role='Ski Destination Researcher',
    goal='Identify the best ski destination reachable by the provided users based on travel costs and accessibility.',
    backstory="""You specialize in finding the most suitable locations for ski trips based on user needs,
    including travel costs, accessibility, and preferences.""",
    verbose=True,
    allow_delegation=False,
    tools=[search_tool]
)
```



Accommodation Specialist Agent

```
hotel_advisor = Agent(
    role='Accommodation Specialist',
    goal='Recommend suitable hotels near the ski destination.',
    backstory="""You are an experienced hotel advisor who knows how to find great deals on comfortable stays, close to ski resorts.""",
    verbose=True,
    allow_delegation=False,
    tools=[search_tool]
)
```



Ski Rental Advisor Agent

```
rental_advisor = Agent(
    role='Ski Rental Advisor',
    goal='Provide information on where to rent ski gear near the selected destination.',
    backstory="""You are an expert in identifying reliable and affordable ski rental shops, ensuring users get the right gear for their skill
level.""",
    verbose=True,
    allow_delegation=False,
    tools=[search_tool]
)
```

(Assistant) To The Regional Manager

A manager agent ensures coordination and adjusts decisions, like switching destinations if hotels are too costly, optimizing the trip within budget.

Why **I don't want to implement a manager** from scratch:

- Manual coordination and decision-making required by the manager.
- Error-prone logic using simple keyword checks for conditions.
- No built-in support for handling dependencies between agents.
- Difficult to scale and reuse for complex workflows.

```
manager_agent = Agent(
    role='Trip Planning Manager',
    goal='Oversee the entire planning process and adjust the destination based on cost considerations.',
    backstory="""You are a strategic planner who ensures that the trip plan is feasible within budget constraints.
    You can make decisions to change the destination based on feedback from the other agents.""",
    verbose=True,
    allow_delegation=True,
)
```



Tasks

Tell the **destination_finder** agent what to do, focusing on what we want to optimize for.

```
task1 = Task(
    description=f"""Research a ski destination that is accessible by {', '.join(names)} starting from their locations: {',
'.join(start_locations)}.
    Consider the cost of flights and their preferred skiing levels.""",
    expected_output="Destination recommendation with details on how it fits the requirements.",
    agent=destination_finder
)
```

```
manager_task = Task(
    description="""0versee the destination selection and hotel evaluation processes.
    If hotels are too expensive at a given destination, adjust the destination choice to ensure an affordable overall trip plan."",
    expected_output="Final destination, hotel, and rental plan.",
    agent=manager_agent,
    subtasks=[task1, task2, task3]
)
```

Here we implement our trip **constraints**. The Manager will switch to a different location if necessary.

Assembling The Crew

The Crew puts together the agents and the (manager) tasks. Finally, we also decide which kind of **process** to go for.

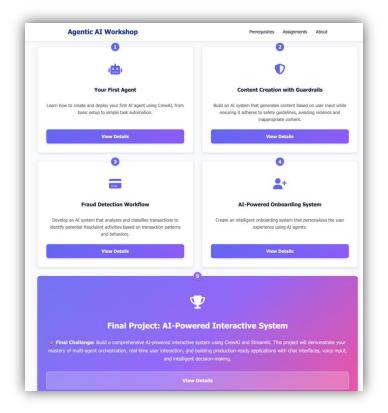
```
crew = Crew(
   agents=[manager_agent, destination_finder, hotel_advisor, rental_advisor],
   tasks=[manager_task],
   verbose=True,
   process=Process.hierarchical # Using hierarchical process for manager-subordinate relationships)
```



Final Output

```
• • •
Manager starts planning the ski trip...
Destination Finder suggests: Chamonix
Hotel Advisor suggests: expensive hotel near Chamonix
Hotel at Chamonix is too expensive. Re-evaluating destination...
Destination Finder suggests: Chamonix
Hotel Advisor suggests: affordable hotel near Chamonix
Rental Advisor suggests: budget rental option for ski gear
All details have been saved in 'ski_trip_details.txt'.
Trip planning is complete
```





https://pigna90.github.io/crewai-workshops

Thank You

