

# GENAI ENGINEERING

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Software Engineering in the AI Era: from ML to LLMs and Agentic AI  
Petra Heck – Associate Professor Fontys ICT - [p.heck@fontys.nl](mailto:p.heck@fontys.nl)

# Program

14.00 – 14.30

**What is GenAI Engineering?**

Petra Heck (Fontys)

14.30 – 15.00

**GenAI Engineering in Practice**

Bart Scheffer (Moonly)

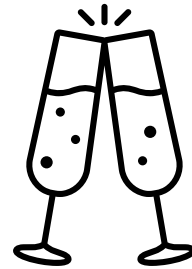
15.00 – 15.30

**How AI is Reshaping Software Engineering**

Pascal Widdershoven (Kabisa)

15.30 – 17.00

Drinks and time for discussion

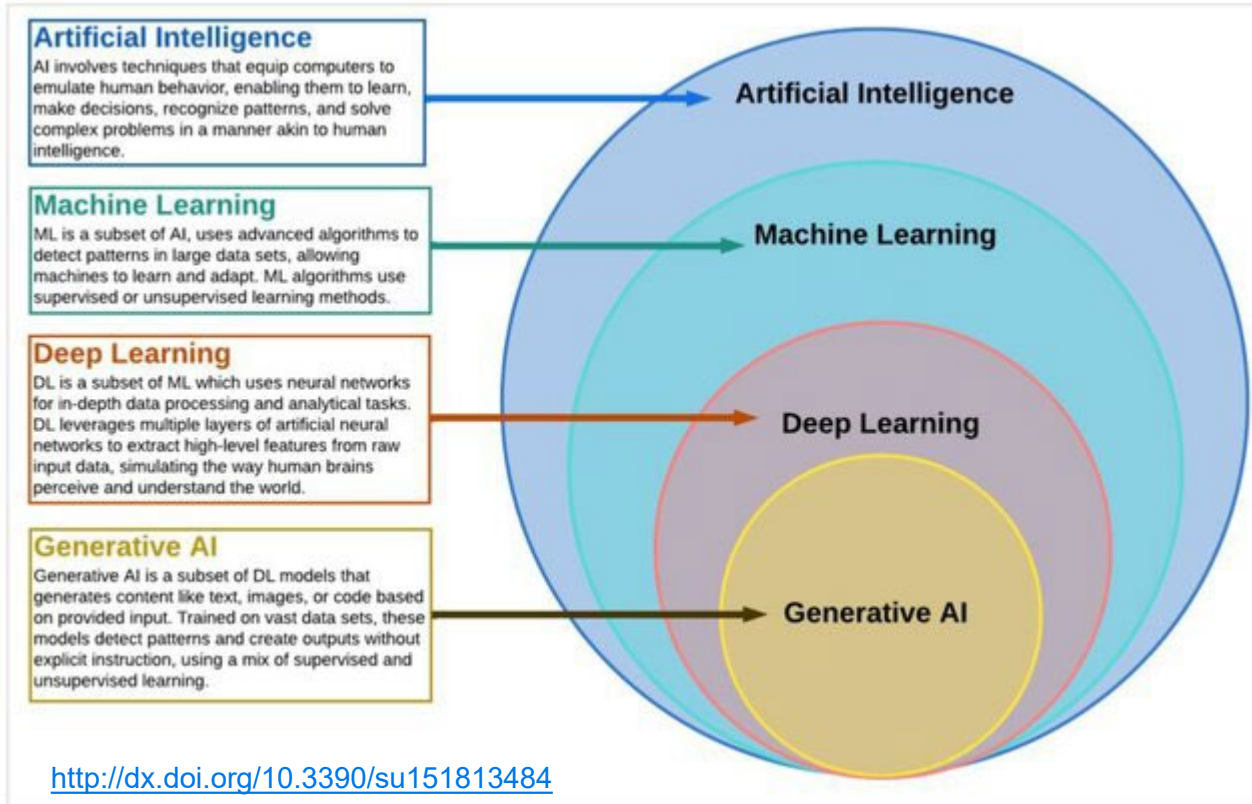


# About Me

- **Software engineer** (MSc Computer Science) since 2002
  - **Software quality** consultant since 2004
  - Lecturer **Fontys ICT** since 2012
    - 2016 PhD “Quality of JIT Requirements” (Computer Science, AI4RE)
    - 2019 Senior researcher AI Engineering (SE4AI)
    - **2023 DEMAND: Data Engineering & Data Management (SE4DATA)**
  - Associate Professor AI & Software Engineering (SE4AI & AI4SE) since 2025
- => Practice-oriented researcher with a Software Engineering view on AI & Data



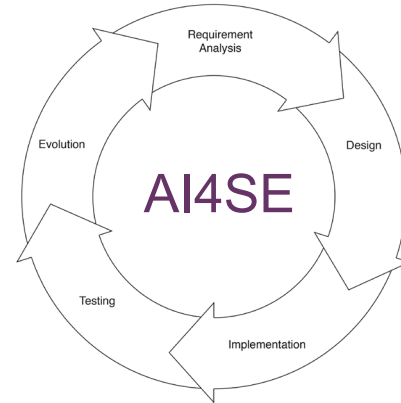
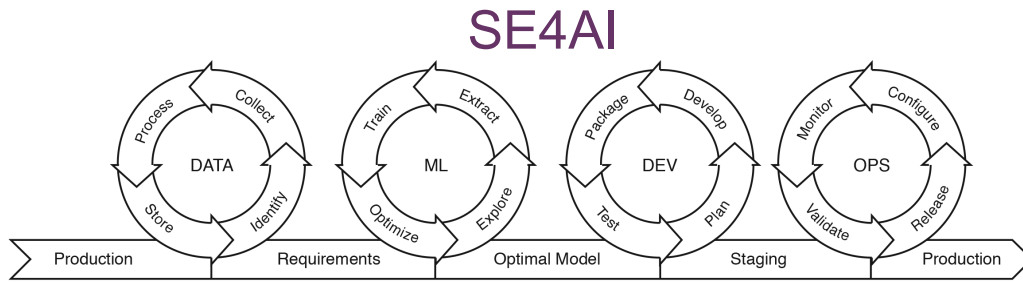
# 2019 till 2026: From AI to GenAI



- Emulate human behavior
- Pattern detection algorithms
- Neural networks
- Generate content

# From AI Engineering to AI-augmented Engineering

“How to build production-ready AI systems?” (Petra Heck, 2019)



“How to use AI to build production-ready systems?” (Petra Heck, 2024)

# GenAI Engineering = LLM Engineering =

**SE4LLM:** How to build production-ready LLM-enabled systems?

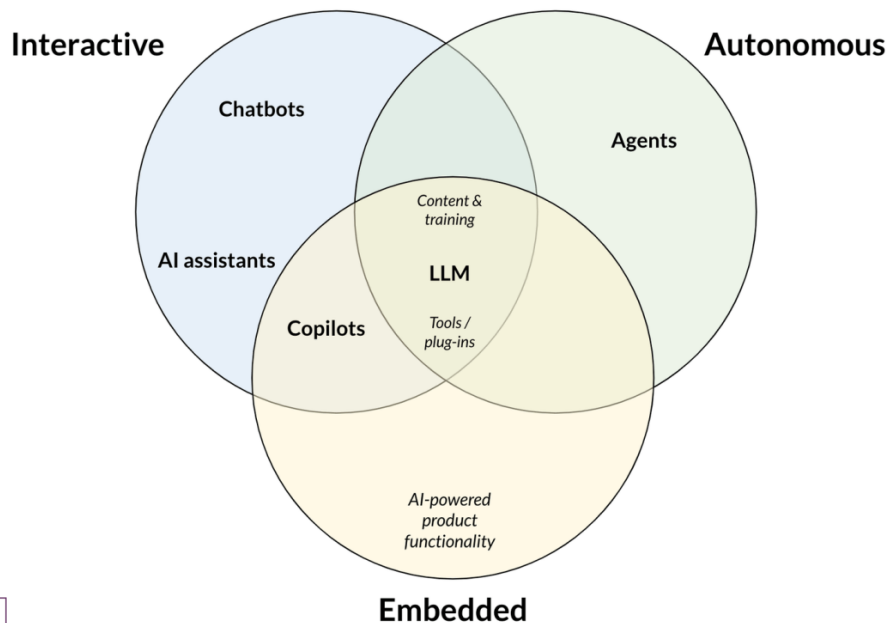
+

**LLM4SE:** How to use LLM-enabled tools to build production-ready systems?



# SE4LLM: BUILDING LLM-ENABLED SYSTEMS

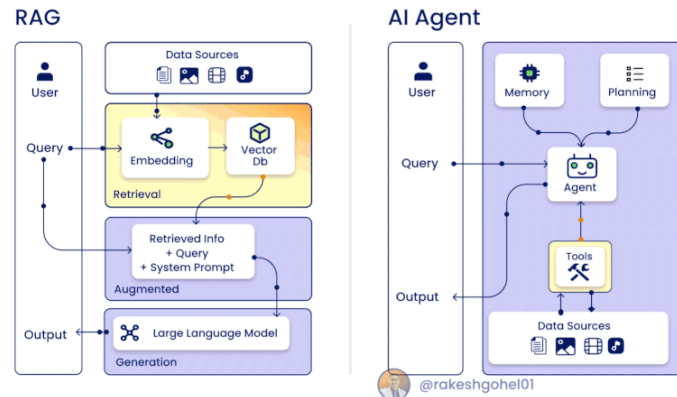
# GenAI Landscape



[A visual guide to today's GenAI landscape - MERL Tech](#)

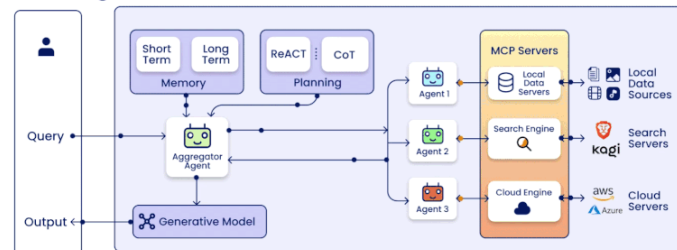


## RAG vs Agentic RAG



@rakeshgoel01

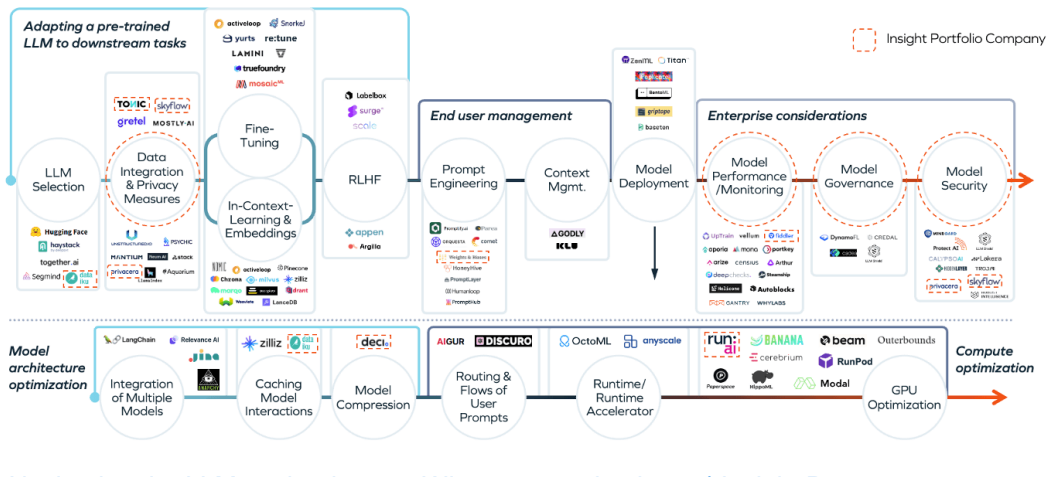
## Multi-Agent RAG



[\(9\) Post | LinkedIn](#)

# From DevOps to MLOps to LLMOps

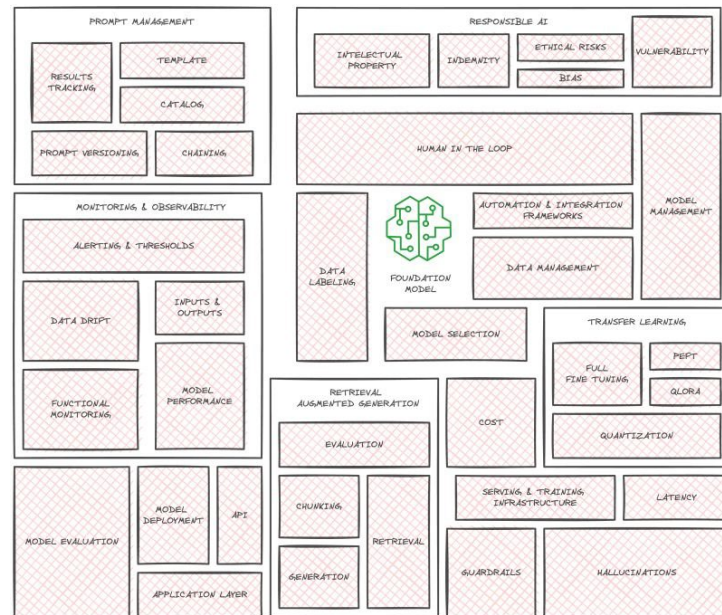
LLMops adapts the MLOps tech stack for generative AI use cases



INSIGHT PARTNERS

Insight Portfolio Company

## HIDDEN TECHNICAL DEBT IN GENERATIVE AI SYSTEMS



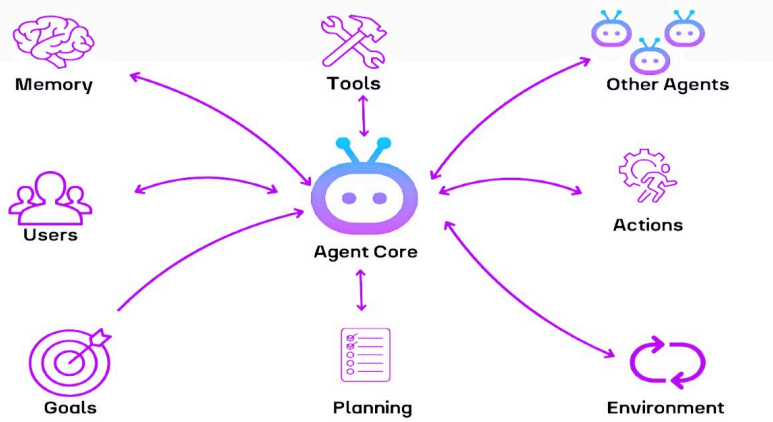
Navigating the LLMops landscape: What you need to know | Insight Partners

<https://fontysblogt.nl/llmops-engineering-trustworthy-llm-systems/>

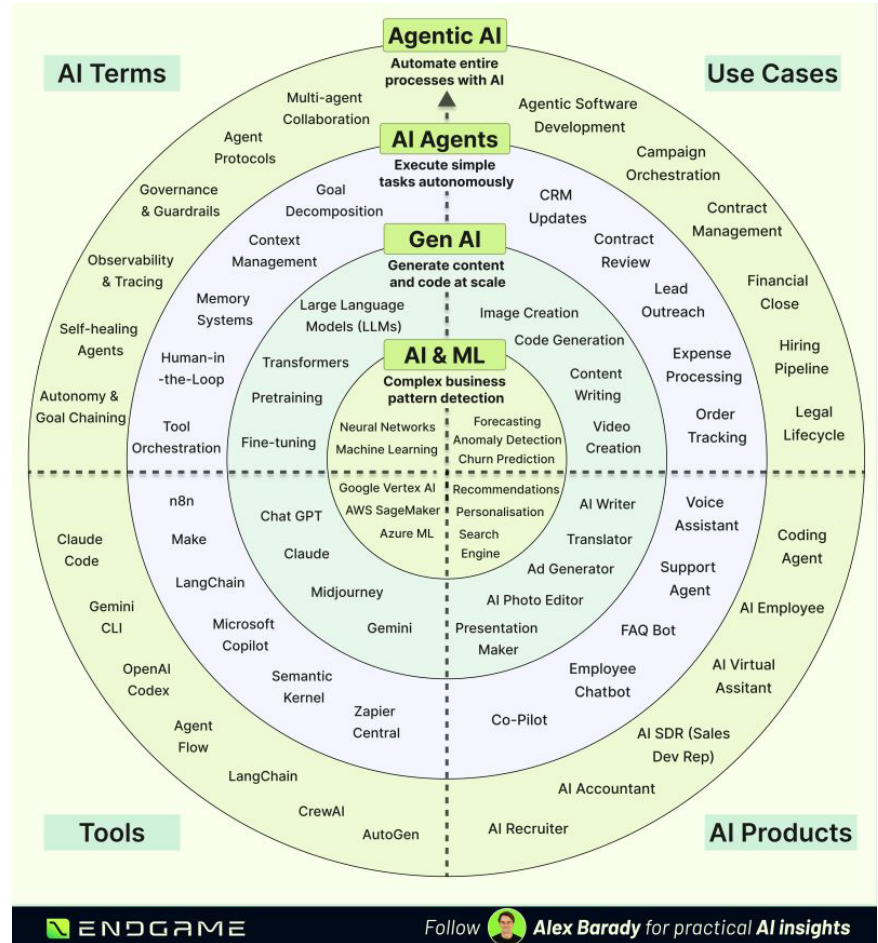


Visualisation by Eduardo Ordax (https://www.linkedin.com/in/eordax/) Original Idea by Eduardo Ordax (adapted from "Hidden Technical Debt in Machine Learning Systems")

# Agentic AI



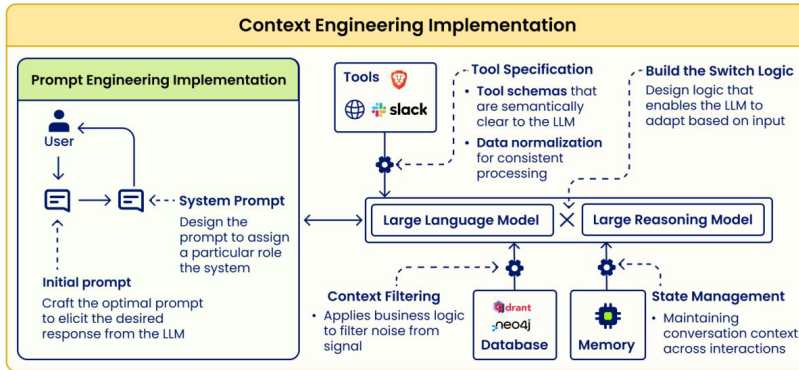
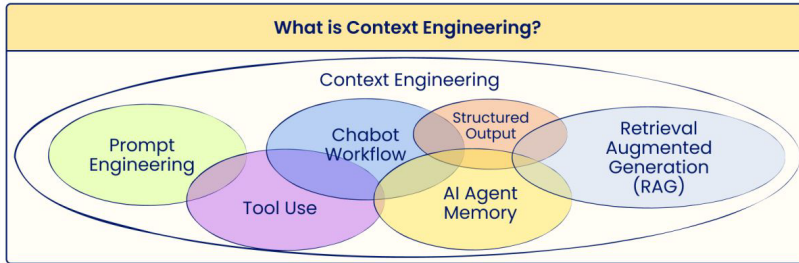
[What is Agentic AI?](#)



Follow Alex Barady for practical AI insights


[\(16\) Post | LinkedIn](#)


# Agentic AI – Terminology



(9) Post | LinkedIn

























**Rakesh Goel**   
@rakeshgoel01

## 20 – AI Agent Terms

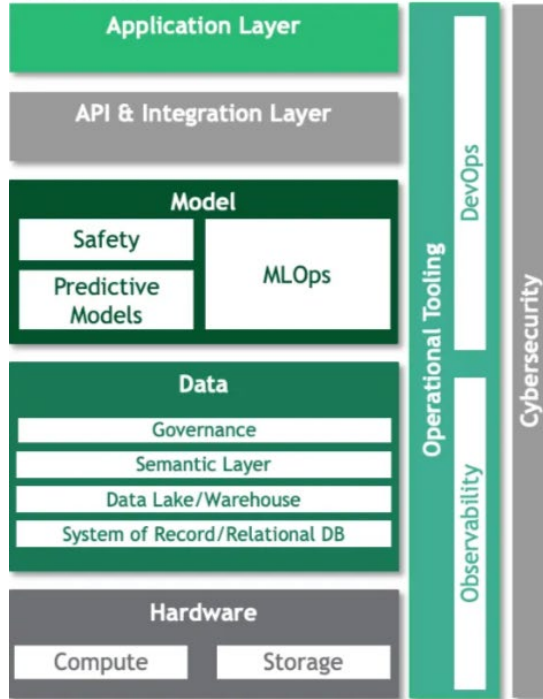
You should know

<p><b>1</b></p> <p><b>Agent</b></p>  <p>AI entity that uses prompts and environment to perceive and act on goals.</p>	<p><b>2</b></p> <p><b>Environment</b></p>  <p>Context or sandbox where an AI agent operates and interacts with other tools.</p>	<p><b>3</b></p> <p><b>Perception</b></p>  <p>The AI Agents' ability to understand and interpret environmental data.</p>	<p><b>4</b></p> <p><b>Action</b></p>  <p>The current process performed by an AI agent or group of agents.</p>
<p><b>5</b></p> <p><b>State</b></p>  <p>The current condition of an agent's environment or system.</p>	<p><b>6</b></p> <p><b>LLMs</b></p>  <p>Large language Models - the brain behind the Agents to perform.</p>	<p><b>7</b></p> <p><b>LRMs</b></p>  <p>Large Reasoning model - a reasoning type for more context based reasoning.</p>	<p><b>8</b></p> <p><b>Tools</b></p>  <p>Native or Third Party APIs used by Agents to perform their task.</p>
<p><b>9</b></p> <p><b>Memory</b></p>  <p>Storage for current context as well as past interactions.</p>	<p><b>10</b></p> <p><b>Knowledge Base</b></p>  <p>Database for Agents' knowledge used to fuel and generate outcomes.</p>	<p><b>11</b></p> <p><b>Orchestration</b></p>  <p>Process of developing agents' interaction, starting from input to output.</p>	<p><b>12</b></p> <p><b>Planning</b></p>  <p>The process of an AI agent determining a sequence of actions to achieve a goal.</p>
<p><b>13</b></p> <p><b>Evaluation</b></p>  <p>The assessment of an AI agent's performance, in achieving its goals.</p>	<p><b>14</b></p> <p><b>Architecture</b></p>  <p>The blueprint of an AI agent, defining how its components interact.</p>	<p><b>15</b></p> <p><b>CoT</b></p>  <p>A reasoning technique where an agent breaks down complex problems.</p>	<p><b>16</b></p> <p><b>ReAct</b></p>  <p>A reasoning Framework for combining reasoning and acting iteratively.</p>
<p><b>17</b></p> <p><b>Multi-Agent System</b></p>  <p>Multiple AI agents interacting in a shared space.</p>	<p><b>18</b></p> <p><b>Swarm</b></p>  <p>Collectively exhibit intelligent behavior through self-organized interactions.</p>	<p><b>19</b></p> <p><b>Handoffs</b></p>  <p>The transfer of tasks or responsibilities between Multiple AI agents.</p>	<p><b>20</b></p> <p><b>Agent Debate</b></p>  <p>Engage in structured argumentation to create better outcomes.</p>

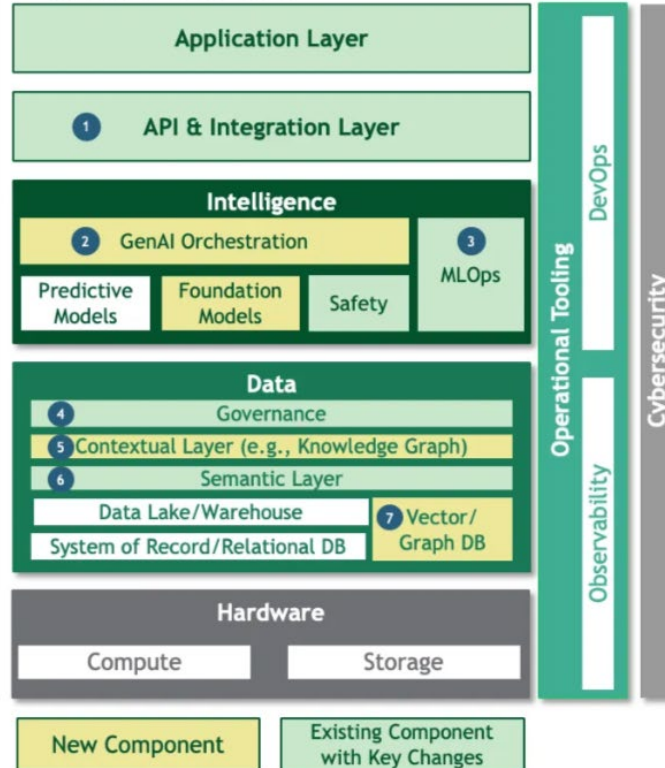
# GenAI Tech Stack



Cloud-native tech stack



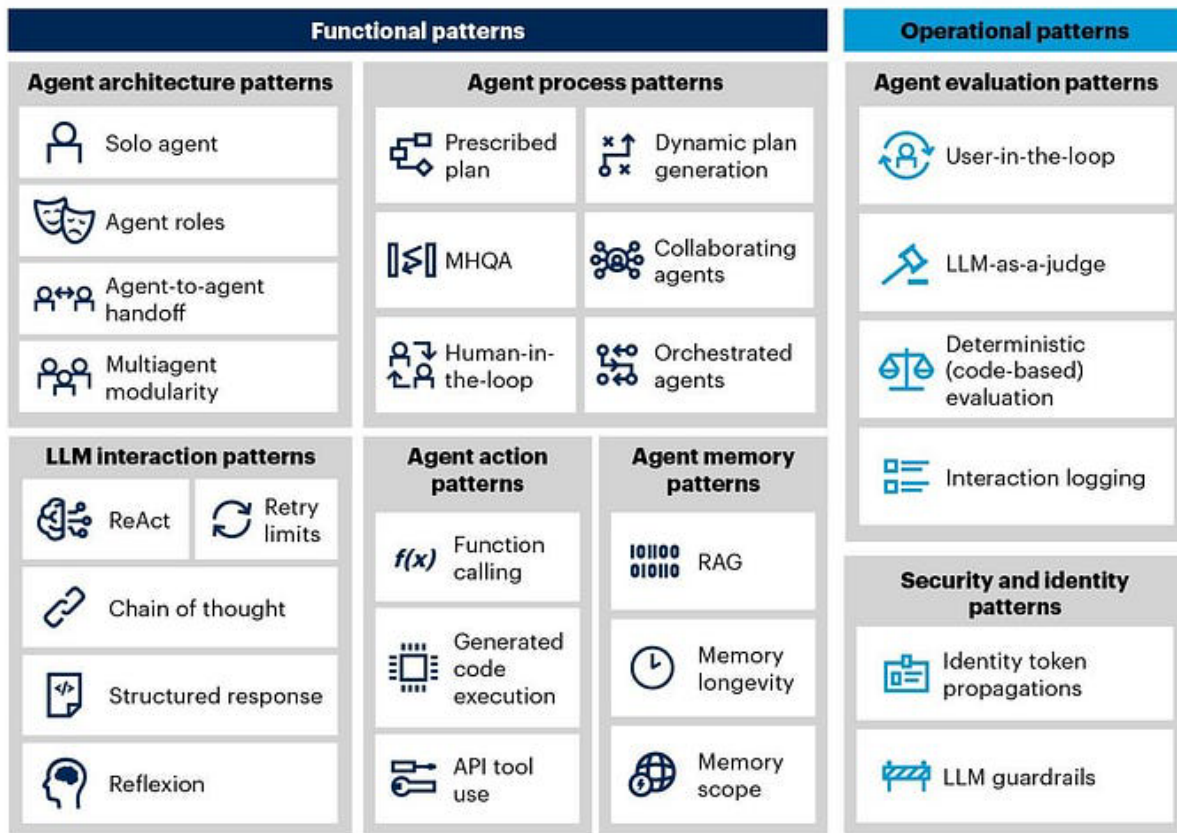
AI-native tech stack



1. Retrieval Augmented Generation 2. Natural Language  
Source: Market participant interviews

[From Apps to Agents: How the AI-Native Tech Stack Is Transforming Software | BCGonTech](#)

# Patterns for Building LLM-based Agents



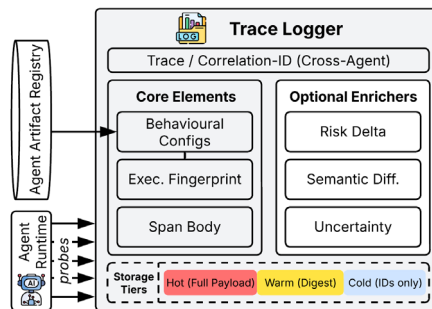
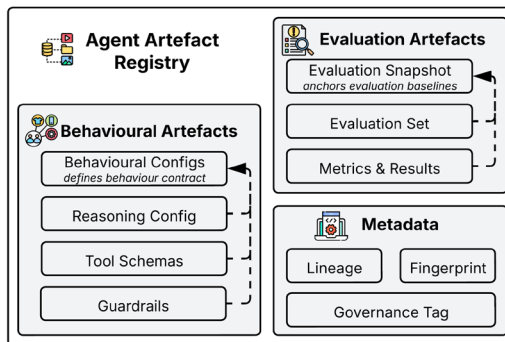
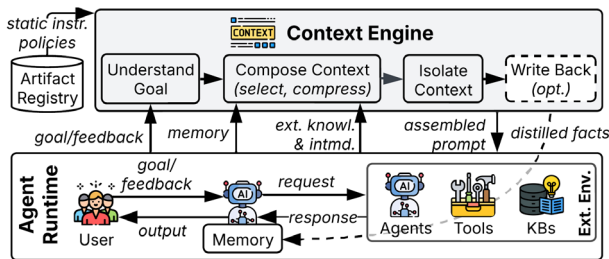
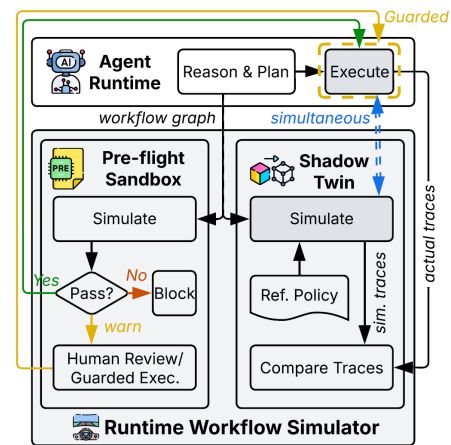
Emerging Patterns for Building LLM-Based AI Agents



# From LLMOps to AgentOps

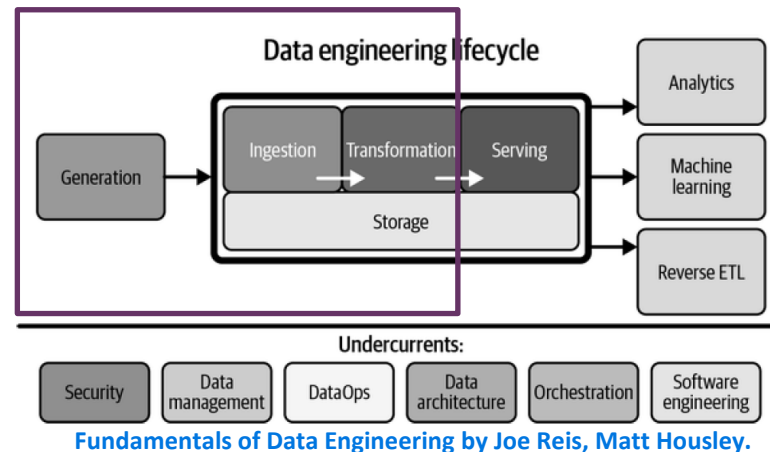
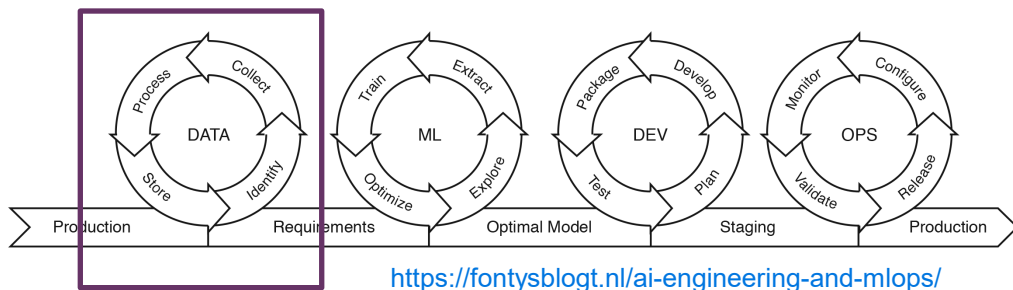
AgentOps Patterns by Category

Category	Pattern	Description
Artefact Management	Agent Artefact Registry	Versioned store for behavioural, eval, and metadata artefacts.
Execution	Context Engine	Assembles just-enough context within budget.
	Runtime Workflow Simulator	Shadow/sandbox execution to vet plans before/during exe.
Safety & Enforcement	Guardrail Enforcement Points	Policy checks at all critical agent action points.
Monitoring & Observability	Trace Logger	Structured spans for replay and audit.
	Cumulative Risk Ledger	Append-only per-step risk signals; shows accumulation.
Evaluation-Driven Learning	Dynamic Test Case Factory	Harvests operational insights into versioned tests.
	Marginal Risk Assessor	Compares candidate vs. baseline risk.
Interaction	Agent-as-a-Judge	Evaluator agent scoring runs against contracts.
	Interactive ChatOps	Conversational queries over ops data; safe controls.
Assurance & Compliance	Trust Calibration Controller	Calibrates user cues to measured reliability.
	Assurance Evidence Factory	Sealed bundles linking artefacts to bounded claims.



# LLM4SE: AI-AUGMENTED DATA ENGINEERING

# From MLOps to DataOps



“How to build production-ready AI systems?” (Petra Heck, 2019)

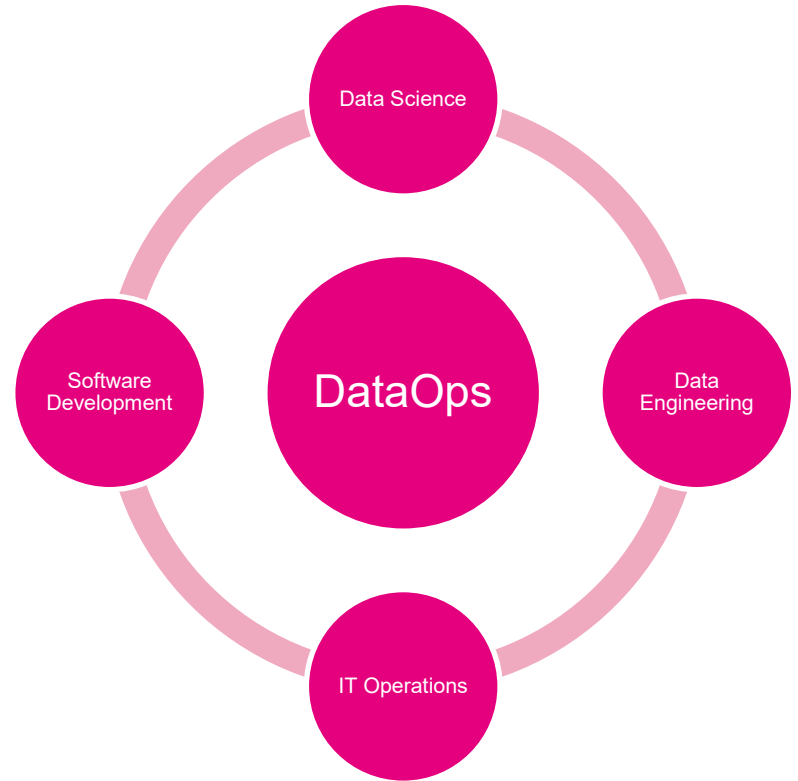


“How to build (and maintain!) production-ready data-driven (AI) systems?” (Petra Heck, 2024)

# DataOps

“an approach that accelerates the delivery of high-quality results by *automation and orchestration* of data life cycle stages.”

DataOps adopts the best practices, processes, tools and technologies from Agile software engineering and DevOps thereby promoting the culture of *collaboration and continuous improvement*.



Aiswarya Raj Munappy, David Issa Mattos, Jan Bosch, Helena Holmström Olsson, and Anas Dakkak. 2020.

**From ad-hoc data analytics to dataops.**

In Proceedings of the International Conference on Software and System Processes. 165–174.

# Framework for Unified DataOps Strategy

<http://dx.doi.org/10.1016/j.jimej.2025.100321>

## Comprehensive DataOps Strategy

### Structured Approaches

- End-to-end DataOps lifecycle integration.
- Workflow orchestration to streamline processes.
- Team collaboration.
- Automate routine tasks & reduce manual effort.
- Integrate data governance throughout.



### Policy and Governance

- Ensure data privacy & security with strong governance policies.
- Define clear data ownership and sovereignty responsibilities.
- Establish ethical data management guidelines.
- Incorporate continuous monitoring and testing.
- Use cross-organizational governance frameworks.



### Innovations and Tools

- Choose appropriate tools for each DataOps stage.
- Leverage scalable cloud-based solutions.
- Adopt real-time data processing tools (e.g., Kafka, Spark).
- Use machine learning for data quality management (e.g., DQSOps).
- Automate versioning & deployment using CI/CD pipelines.



**Jenkins**

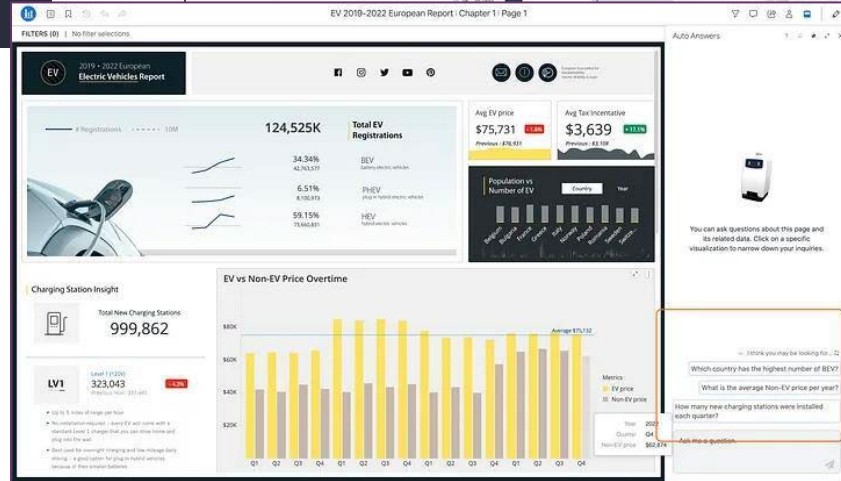
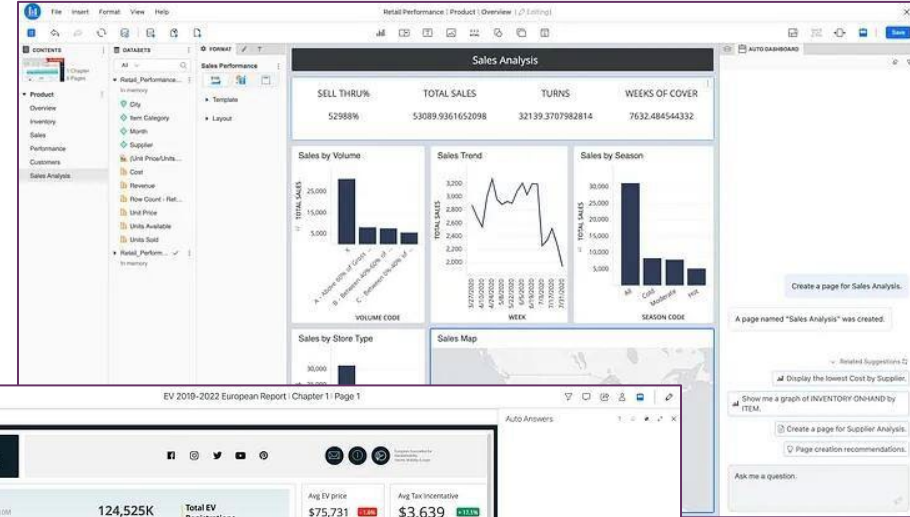
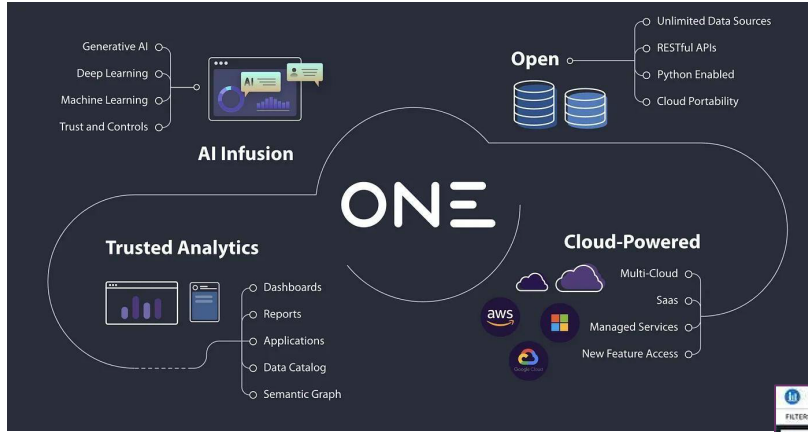


**kafka**

## Generative AI for DataOps

- Synthetic Data Generation
- Anomaly Detection
- Code Automation
- Data Quality Assurance
- Documentation Automation
- Intelligent Monitoring

# GenAI for Dashboards



[AI en Design Thinking in de Dashboardmeesterschap van MicroStrategy](#)



# GenAI for Self-Service Analytics

Which suppliers are most impacted by tariffs, ranked by total transaction costs?

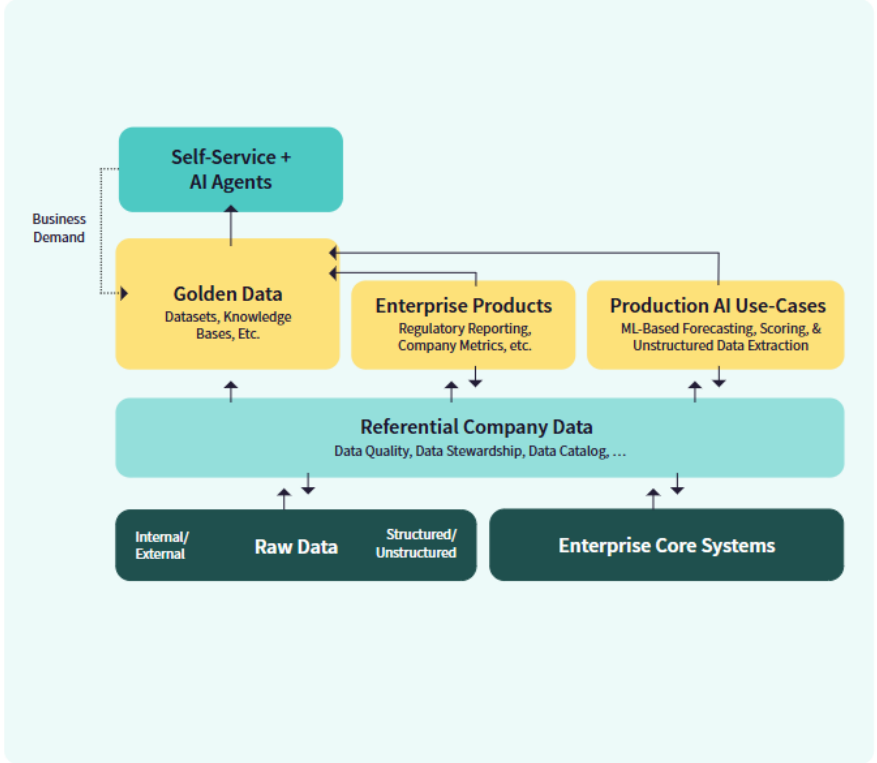
SPEEDLINE and GEARPRO are the most impacted suppliers by tariffs, with the highest total transaction costs.

Rank	Supplier	Total Transaction Cost (Impacted by Tariffs)
1	SPEEDLINE	7.00 billion
2	GEARPRO	4.94 billion
3	SHIFTSUPPLY	816 million
4	AUTOSUPPLY	558 million
5	PARTSYNC	470 million

Chat with Supply Chain Risk Agent

Ask Anything

Agents



Self-Service Analytics In The Age Of AI Agents

# Agentic AI for Data Lakehouse



Applied & Gen AI

Intelligent Automation

Cloud & Data

Digital Products & Platforms

Webinar | Towards Data Lakehouse Architecture

## 05 GenAI for Data Teams: AI-Assisted Engineering in Data Lakehouse

📅 January 27, 2026

“how **Agentic AI** shifts data teams from hand-built pipelines to **outcome-driven data products** guided by clear intent and intelligent assistants.

You’ll learn how **MCP integrations** enable AI assistants to tap your data catalogs, databases and transformation tools like dbt - dramatically accelerating development while preserving governance and trust.

We’ll show how **spec-driven development** turns business needs and technical requirements into unambiguous, testable data products - boosting delivery speed and confidence.

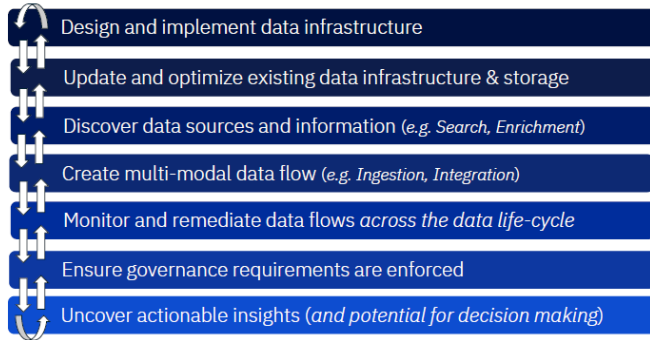
Finally, we’ll explore **conversational data interfaces** that allow both technical and business users to interact with data products using natural language.”

<https://events.xebia.com/5-towards-data-lakehouse-architecture-webinar-ai-assisted-engineering>

# Agentic DataOps

## Agentic DataOps

**Proposal:** Agentic data stack that can autonomously resolve data asset and insight requests from a stakeholder (including AI itself).  
Process creates a system that is comprised of all data stack stages:

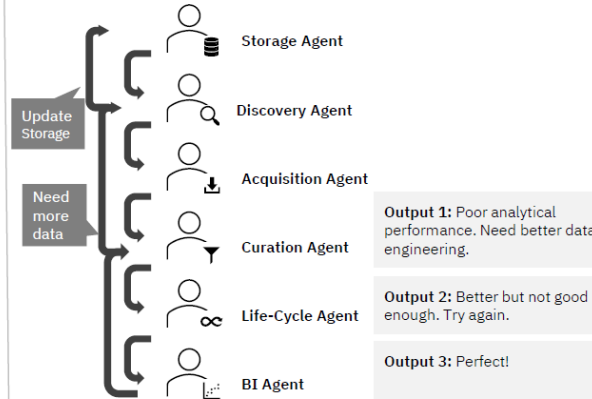


**Impact:** Democratize value creation from data at speed and scale. Bring time to value from weeks down to hours by decreasing task times and reducing human touchpoints / bottlenecks.

### Example: Fund Performance Forecasting

Create financial product that forecasts performance across Asian and European mutual funds.

*Additional requirements include low latency, GDPR compliance, price-performance, high data quality, restricted user access.*



[Keynote Lisa Amini-What's Next in AI for Data and Data Management--Pydata Global 2025 \[2512.07926\] Can AI autonomously build, operate, and use the entire data stack?](#)

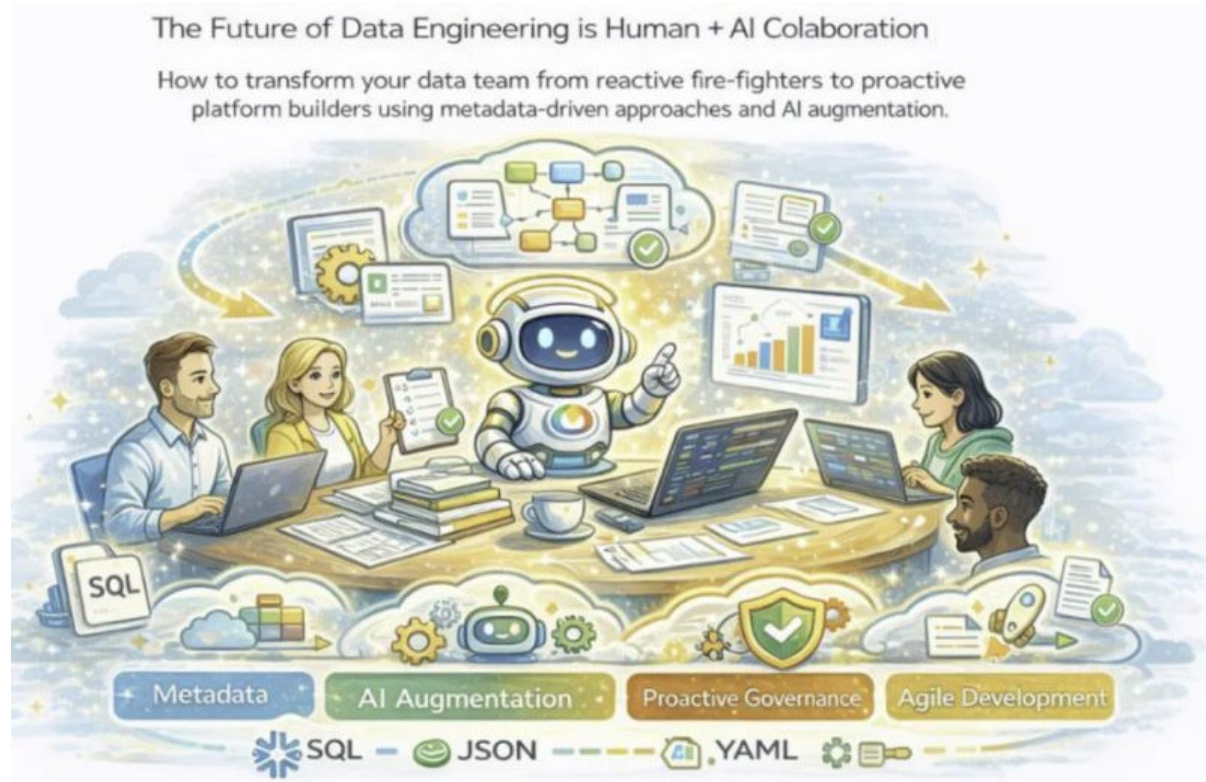
# Building AI-Enabled Metadata-Driven Data Platform Teams



Jaco van der Laan he/him  
905 followers

Exploring Business & Logical Data Modeling.  
Writing on Clarity, Structure & Creative  
Approaches to Data Architecture.

Follow



[\(13\) Post | LinkedIn](#)

[Where Should Data Documentation Live?](#)

[Building AI-Enabled, Metadata-Driven Data Platform Teams](#)

[The Knowledge Repository Pattern: One Folder Per Artifact](#)

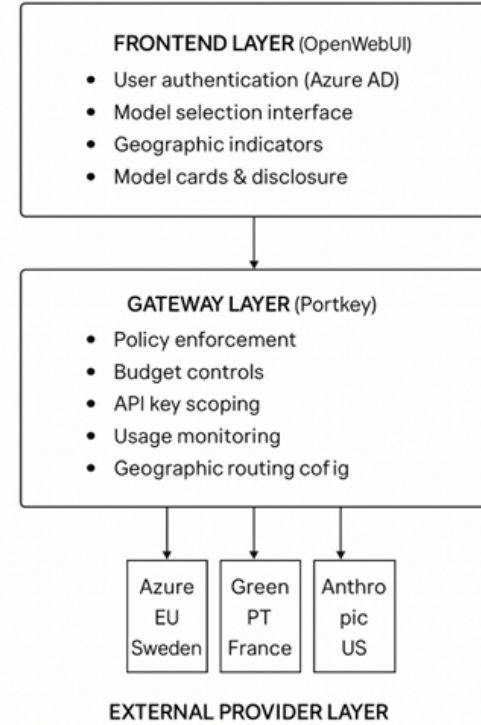
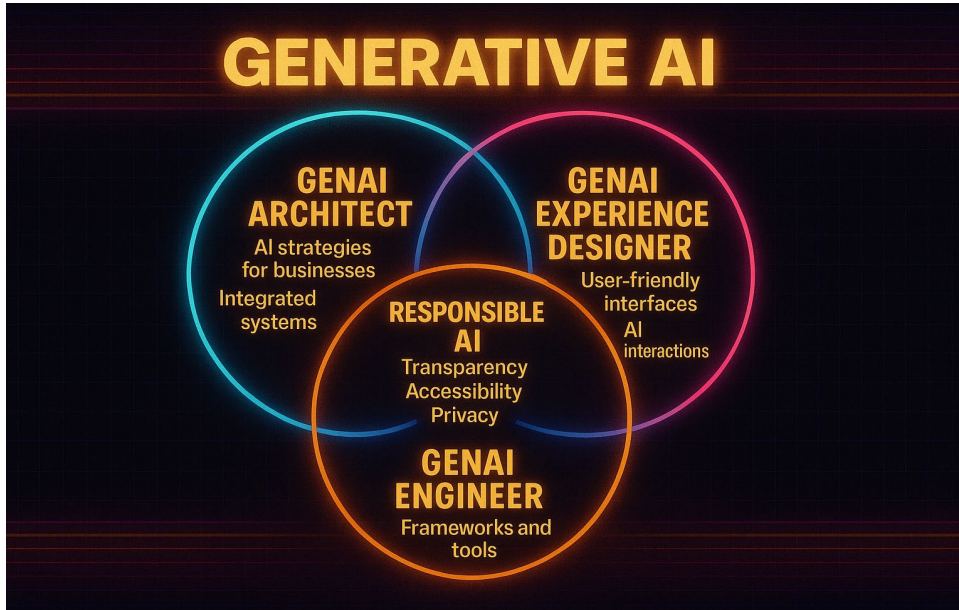


# Other Examples

- [AI Agents With Dataiku | Dataiku](#)
- [Introducing Genie Code | Databricks Blog](#)
- [Action introduceert Data Observability Agents | BI-Platform](#)
- [IBM Db2 Genius Hub voegt autonome mogelijkheden toe aan Db2 | BI-Platform](#)
- [LLMs4OM: Matching Ontologies with Large Language Models](#)

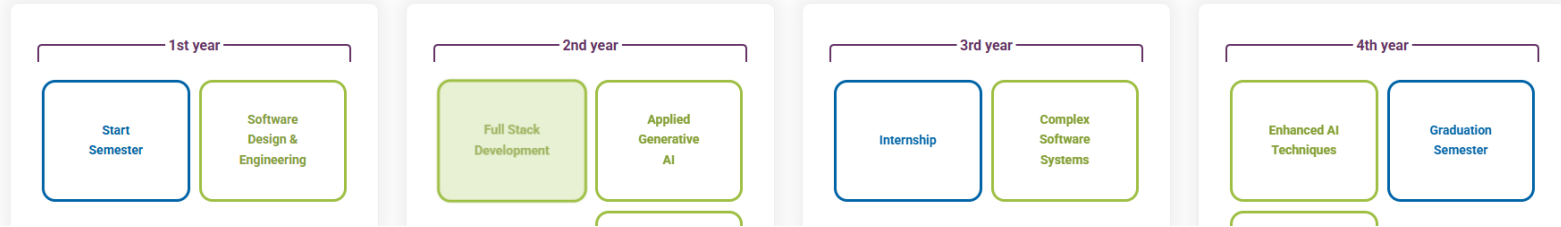
# GENAI ENGINEERING @ FONTYS ICT

# Fontys ICT – Applied GenAI

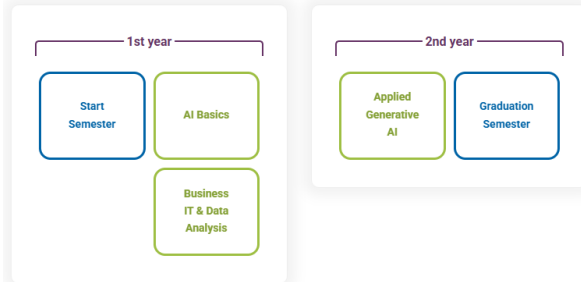


# Fontys ICT – Study Programmes

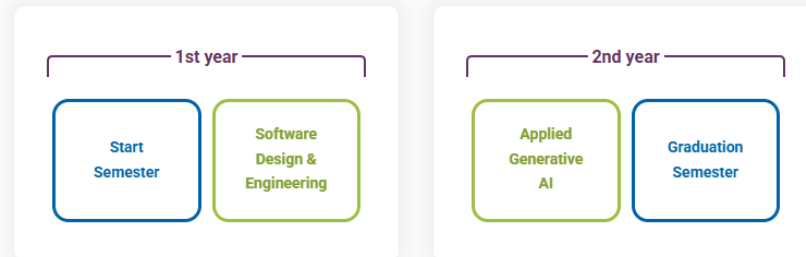
## Study programme Software Engineering



## Function Prompt Engineer



## Function Software Prompt (AI Integration) Engineer



# Applied Research Projects

- “A Validation Framework for LLM applications” – Professional Doctorate Leon Schrijvers
- “Architecture patterns for modifiable AI systems” – Phd Merel Veracx
- “Active Model Cards” – Koen Suilen & Ruud Huijts – under submission
- “What is a GenAI engineer?” – Petra Heck & Leon van Bokhorst
- DEMAND (Sprong) – Data engineering and data management in data chains – with HAN and Saxion
- ....
- **Several projects where we apply LLMs to innovate professional practice of other domains**

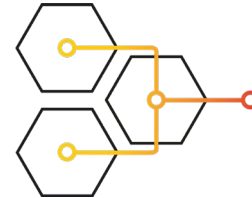
# CONCLUSION

# Lessons We Are Learning

- GenAI engineering is software engineering
  - New **architectures** for systems, new **tech stack**
  - From Devops to MLOps to **LLMOps** to **AgentOps**
- Don't forget about the **data** (and data engineering)
- From automation to augmentation, new processes?
- Today's impossible is tomorrow's possible
  - So start yesterday, learning by doing



Let's learn together! [p.heck@fontys.nl](mailto:p.heck@fontys.nl)



**DEMAND**



[DEMAND – Handling Data Safely, Responsibly, and Efficiently](#)

# Questions?



[Petra Heck, auteur op Fontys](#), [p.heck@fontys.nl](mailto:p.heck@fontys.nl), <https://www.linkedin.com/in/petraheck/>

The benefits of AI are not equally distributed!

*How do others use it? — What practices and patterns are developing? — How to avoid damaging the talent-pipeline? — How to use it to improve quality and productivity? — How to keep developers happy and healthy?*

We are starting a **qualitative research** effort about how software engineers use AI in their work (using rigorous research methods and frameworks such as DORA & SPACE)

 We would like to find **partners, contributors and participants** for going on this journey with us.

 You can help us shape this research, contribute expertise, and get answers

 Interested? Send a mail to [marc.vangrootel@fontys.nl](mailto:marc.vangrootel@fontys.nl) or scan this code

