



Fabrizio Lazzaretti & Marco Crisafulli

# DevOps für KI: Mehr als nur Modelle



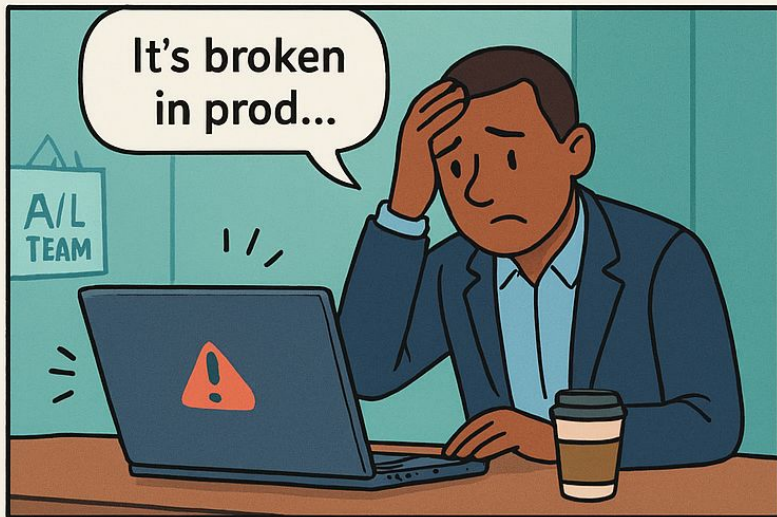
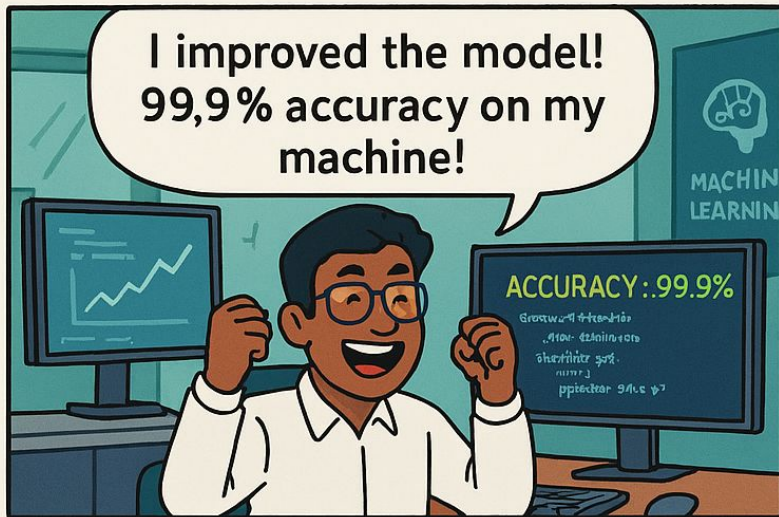
Fabrizio Lazzaretti  
Managing Consultant @ Wavestone |  
CNCF Ambassador | Speaker | Author

*Helping enterprises adopt cloud-native  
technologies while ensuring future-proof  
architectures and business-IT alignment*



Marco Crisafulli  
Co-Founder @ enki |  
ML Engineer | Father | Sometimes funny

*AI can solve real problems. Helping companies to understand and utilize this technology.*



**Gartner predicts ... 85% of AI projects will deliver erroneous outcomes due to bias in data, algorithms or the teams responsible for managing them.**

[Gartner](#)

**Zillow: [CEO] made reference to the company's lack of confidence in its home buying algorithm's ability to accurately predict fluctuations in home prices. ... leading to a write-down of over \$300 million**

[GeekWire](#)

**Google Flu Trends performing well for two to three years and then failing significantly and requiring substantial revision**

[Wired](#)

# Why is this needed?

*“The main challenges people face when developing ML capabilities are scale, version control, model reproducibility, and aligning stakeholders”*

2020 State of Enterprise Machine Learning Report

# From problem to solution

1. We will not solve all your AI problems in this talk
2. You will not be better than the big ones, but you can learn from it

Our goal for today:

- Understand your business, don't make wrong expectations and let the business understand the IT
- Reproducibility, Testing, Continuous Deployment with practical implementation in Kubeflow

Before we start...



## We need AI!

- What business problem do we want to solve?
- Which KPIs can we define to know if the project was successful?
- Do we have the necessary data, know-how and laws to be able to do this?
- Does the problem benefits from applying AI?

Who are we?

CEOs!



What do we want?

AI!



AI that does what?

We don't know!!!



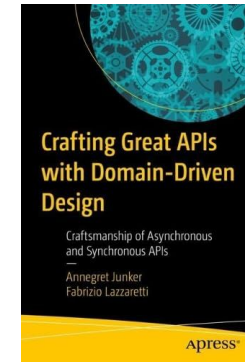
When do we want it?

NOW!!!







# How can we find out if we need this AI project?

- Qualify business ideas  
*Understand what we want to do, and what to prioritize*
  - Business Model Canvas
  - Capability Map
  - Wardley Map
- Gathering business requirements  
*Understand how we need to do it and get a common language and understanding*
  - Domain Storytelling
  - Visual Glossary
  - Event Storming
  - Context Map
- Tactical design of the components  
Define how we structure the solution
  - Bounded Context Canvas
  - Architecture Communication Canvas
  - API Product Canvas



# Getting started with the project

# The 4 phases

PoC 	MVP 	Product Strategy 	Scaled Approach 	
Minimal Concept	Concept	Concept	Concept	Governance
	Data Quality	Data Quality	Data Quality	Disaster Recovery
	Compliance	Compliance	Compliance	Cost Optimization
	Minimal SLAs	SLAs	SLAs	Training Plan
		Evolution	Evolution	

## Phase 1: PoC 🥰

We have uncertainty that we want to mitigate before doing an MVP:

- What do we want to prove?
- When did we prove it?
- How do we measure success?

=> Multiple PoCs can be needed, to prove different things

=> If everything is clear, no PoC is needed

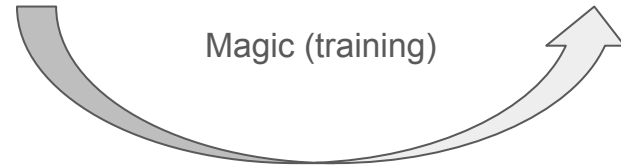
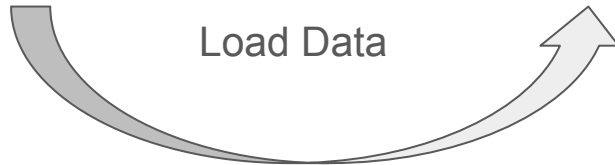
# Phase 2: MVP

PoC can be simple, but what do we need to know to go to prod?

- **Budget**
  - Initial costs
  - Operational costs
  - Iterative extension/update costs
- **When is the project successful?**
  - How will we define quality?
  - How do we measure KPIs and see if we achieve them?
- **Architecture**
  - Loosely coupled
  - Good APIs between components
  - Service cuts
- **SLA**
  - What happens if it does not work?
  - How fast do we need to recover after a problem?
- **Security & Compliance**
- **Security risk, business risk, risk acceptance, ...**
  - What happens if we do something wrong?
  - What happens if data get lost or stolen?

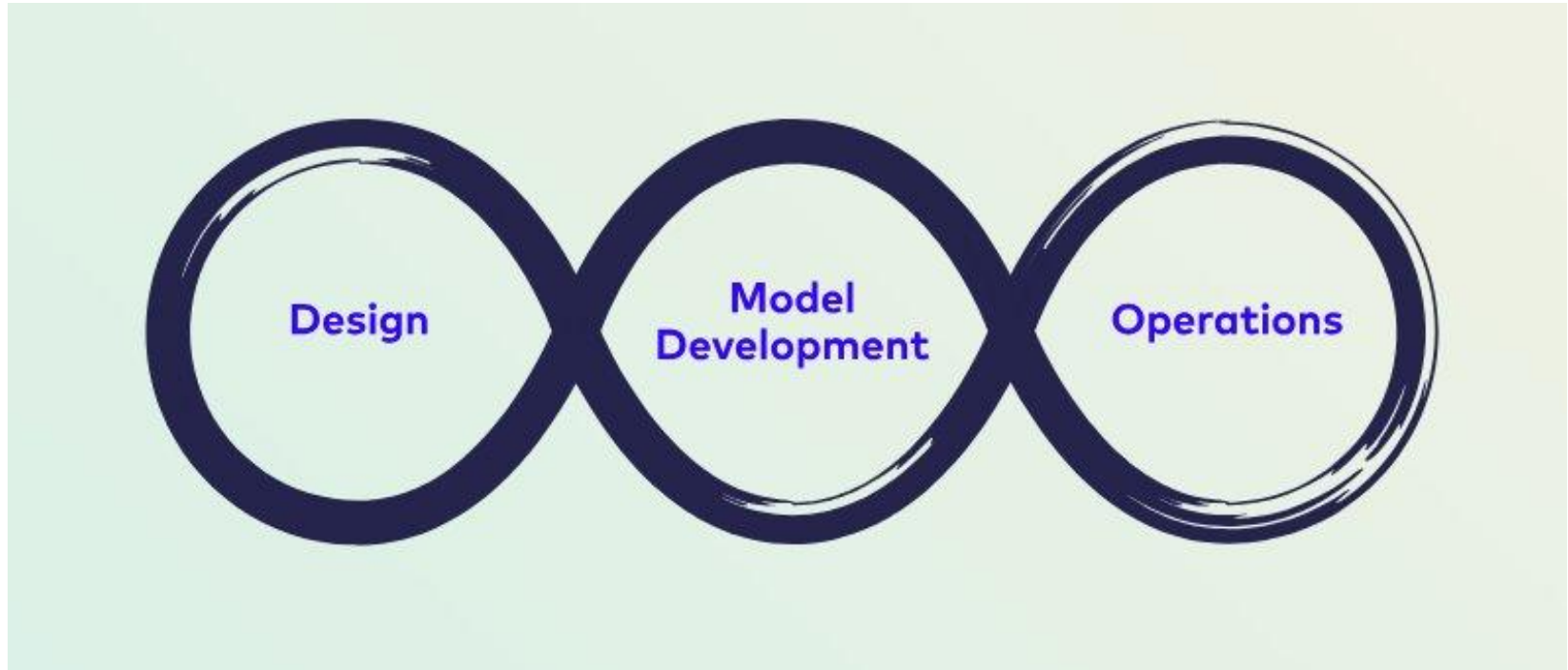
# From concept to implementation

# End-to-end ML, right?



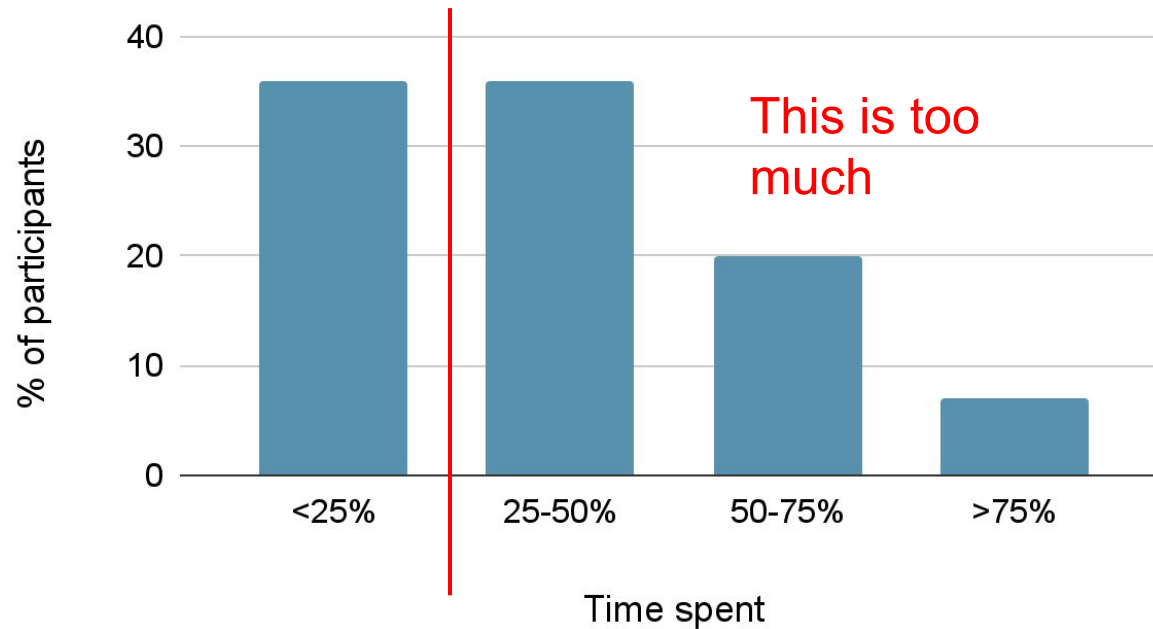


# MLOps - Machine Learning Operations

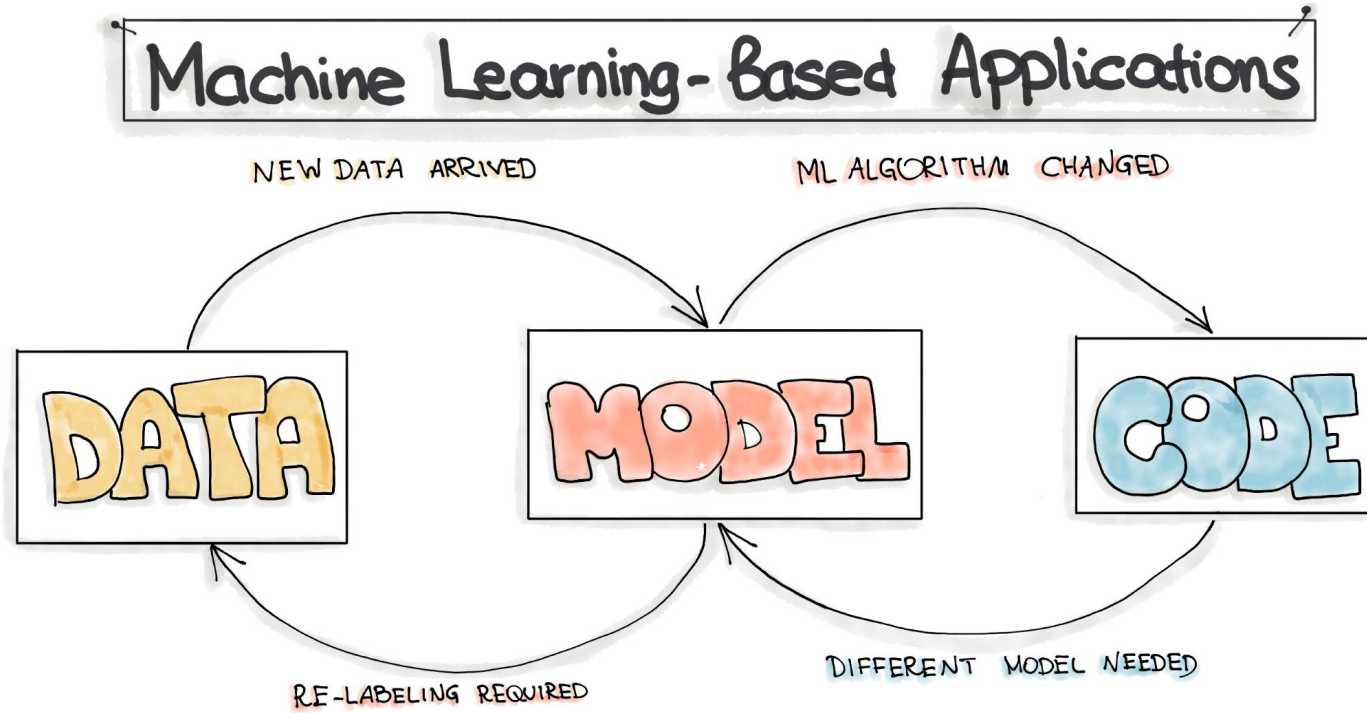


# Why is this needed?

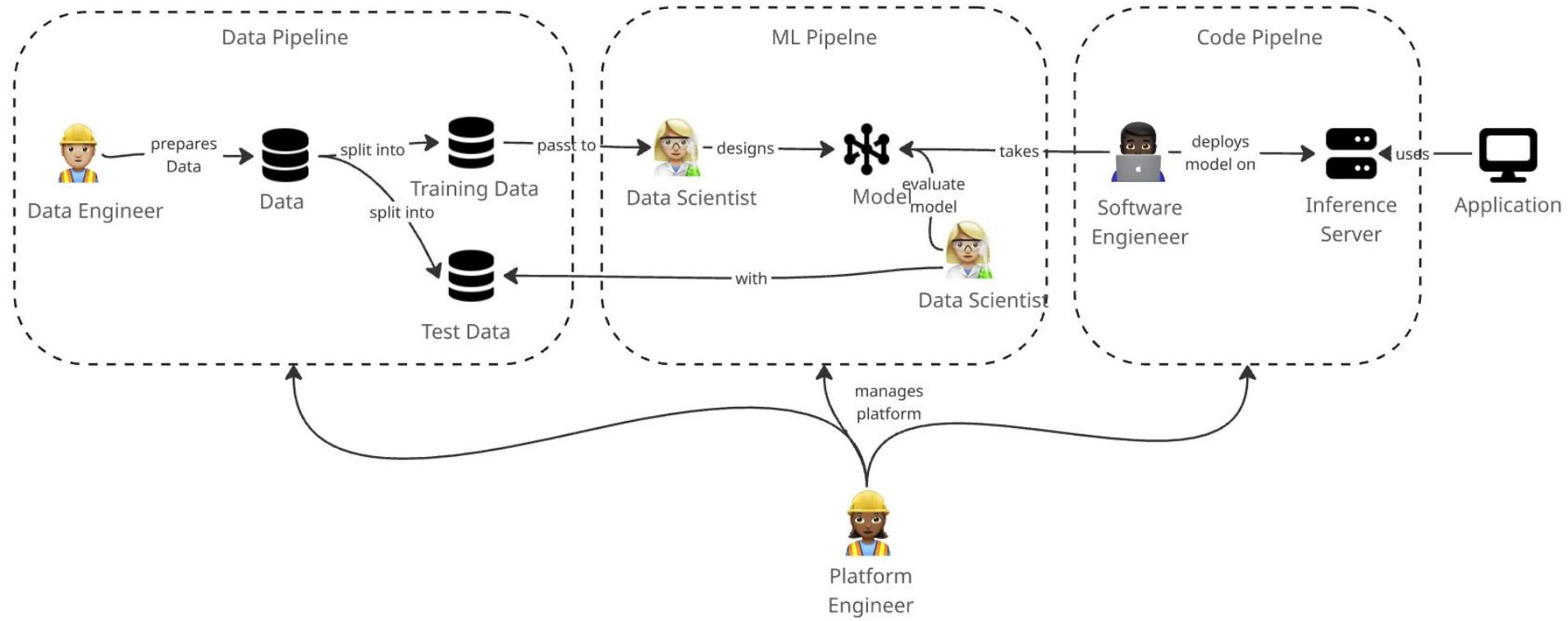
Percent time spent on deploying models



# MLOps is not DevOps



# MLOps in detail



# Applying MLOps with Kubeflow

# Kubeflow is a Cloud Native Computing Foundation project

- CNCF is a foundation is a Linux Foundation project
- Enable open-source projects in the cloud & adds governance
- Founded in 2015 with the announcement of Kubernetes 1.0 (donation from Google, before called Borg)
- The big projects: Kubernetes, Prometheus, Envoy, ArgoCD

Platinum (19)



Gold (27)

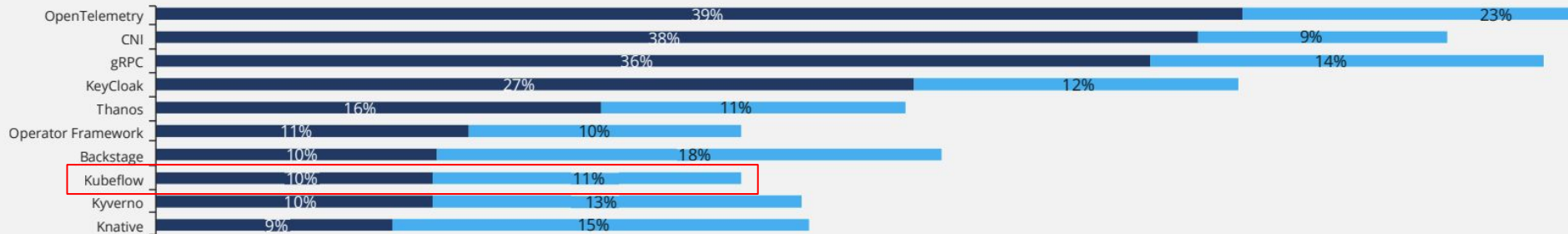


# Kubeflow an incubator project with traction

FIGURE 14

## CNCF INCUBATED PROJECTS IN USE OR IN EVALUATION

Which of these incubating CNCF projects is your organization using in production or evaluating?



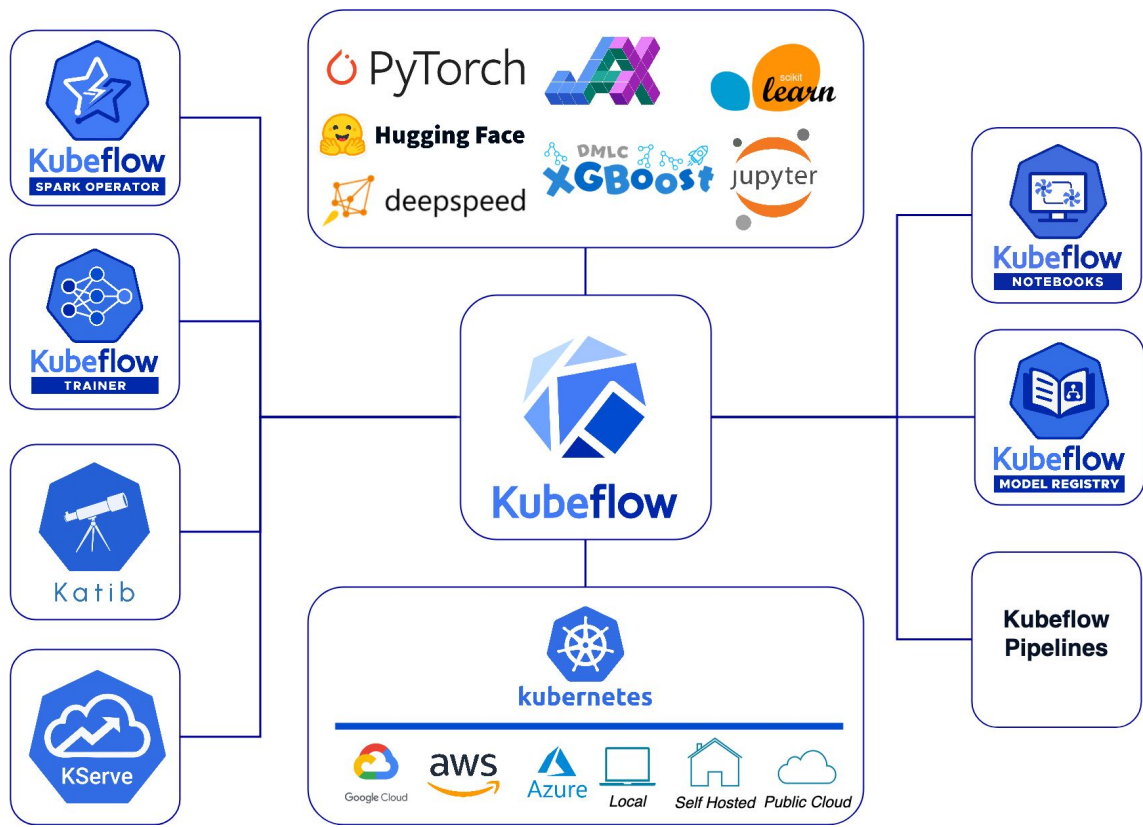
■ Using in Production ■ Evaluating

# Who uses Kubeflow?

The Kubeflow user survey [2023] drew responses from 90 members of the community mostly made up of members from the United States (43%), Europe (34%), and Asia-Pacific (10%).

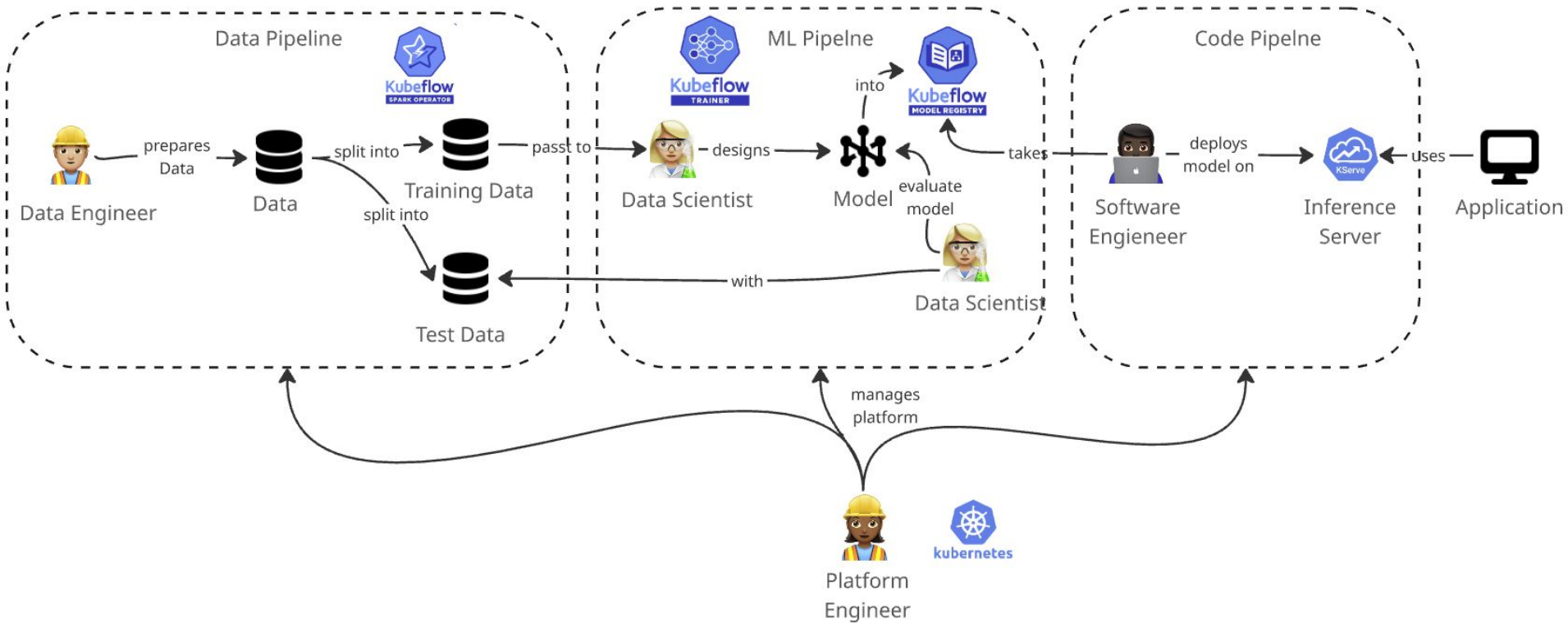
The majority of the respondents were from the Tech industry (49%), followed by Finance (13%) and Consulting (11%).



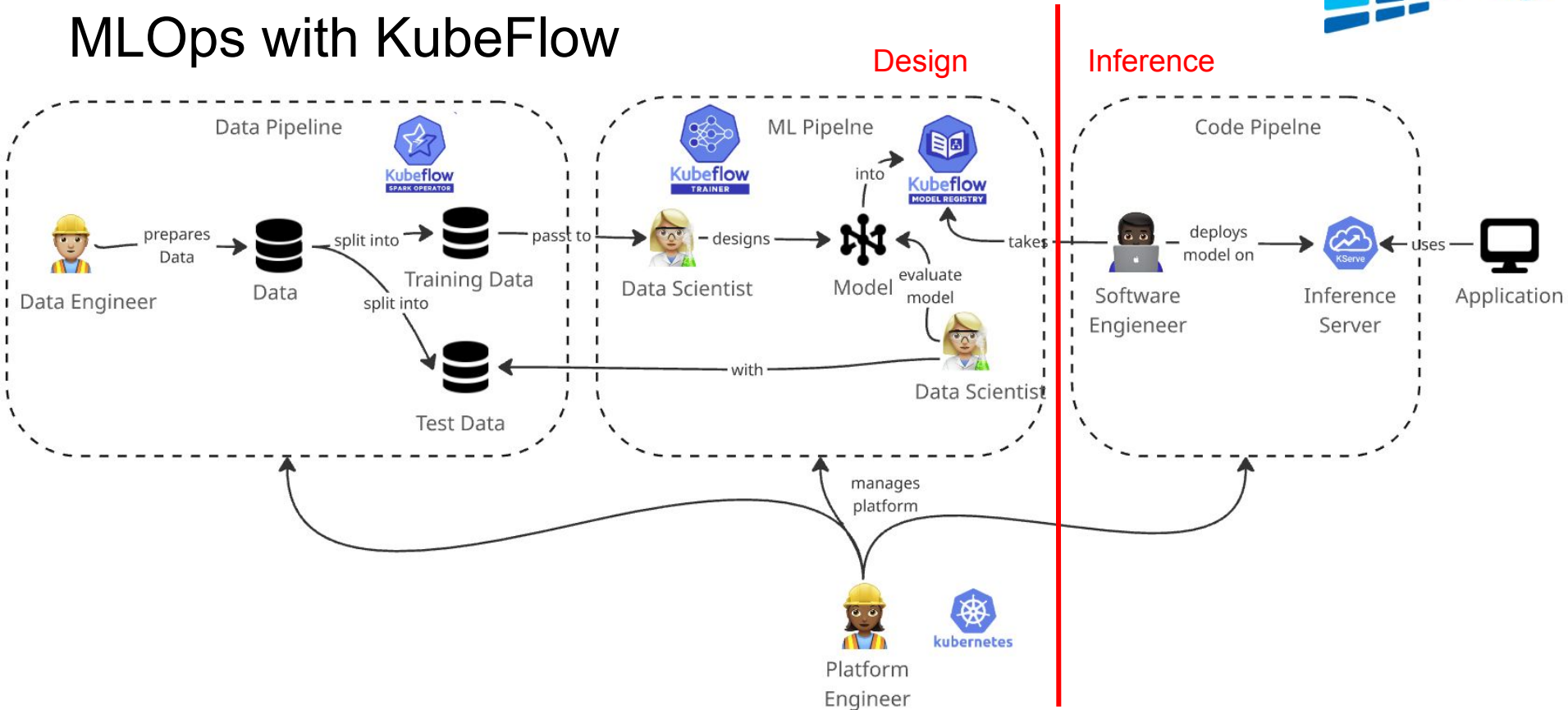


# Kubeflow and an own model

# MLOps with Kubeflow

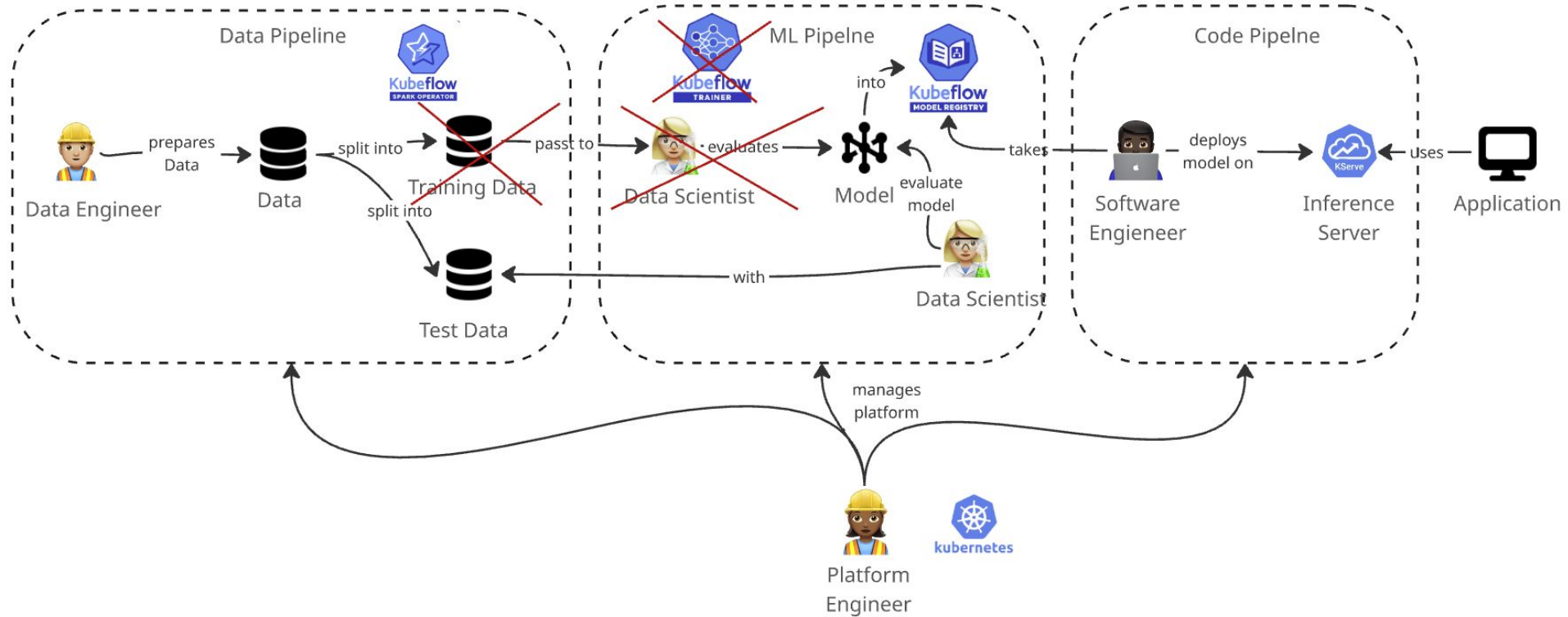


# MLOps with KubeFlow



I want an off the shelf model (ChatGPT, ..)

# MLOps with off the shelf models



# When to use Kubeflow?

# Pros and cons

Pros	Cons
Cloud-native, Kubernetes-integrated	High operational complexity
Highly scalable & production-ready	Steep learning curve
Modular, extensible components	Difficult upgrades/maintenance
Vendor-neutral, open source	Heavy infrastructure overhead
Works across clouds & on-prem	Slow for rapid prototyping
Strong community & ecosystem	Documentation needs improvement



# Lessons learned

- Start small
  - Kubeflow has a lot of components -> Start with one
  - Attention regarding over-engineering
- Own cluster
  - Managed if possible
- Manage Kubernetes Manifests -> DevOps
  - e.g kustomize
- It is not a magic platform
  - You still need to think about the process

# Conclusion

Don't throw out what you learned from classical IT projects

- Requirement engineering
- Phases
- KPIs

At the end people need to talk!

DevOps + MLOps + Any-Ops

Always evaluate your results, outcome can change dramatically after small changes!

## Follow up

Wavestone: Beyond traditional change management whitepaper:

<https://www.wavestone.com/en/insight/beyond-traditional-change-management/>

Kubeflow Introduction:

<https://www.kubeflow.org/docs/started/introduction/>

Kubeflow training from CNCF:

<https://trainingportal.linuxfoundation.org/courses/introduction-to-aiml-toolkits-with-kubeflow-lfs147>

# Thank you!

Kontakt Fabrizio

<https://github.com/Lazzaretti>

<https://www.linkedin.com/in/fabrizio-lazzaretti/>

Kontakt Marco

<https://github.com/mcrisafu>

<https://www.linkedin.com/in/mcrisafu/>



<https://lazzaretti.me/ittage25>