



CLOUD NATIVE
COMPUTING FOUNDATION



From Problem to Production: DevOps for AI with Kubeflow

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*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.

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 outperform the big players, but you can learn from them

Our goal for today:

- Understand your business, avoid wrong expectations, and help the business understand IT
- Reproducibility, testing, and continuous deployment with a practical implementation in Kubeflow

Before we start...

We need AI!

- What **business problem** do we want to solve?
- Which **KPIs** can we define to determine if the project was successful?
- Do we have the **necessary data, know-how** and **legal framework** to do this?
- Does the problem **benefit** from applying AI?

Who are we?

CEOs!



What do we want?

AI!



AI that does what?

We don't know!!!



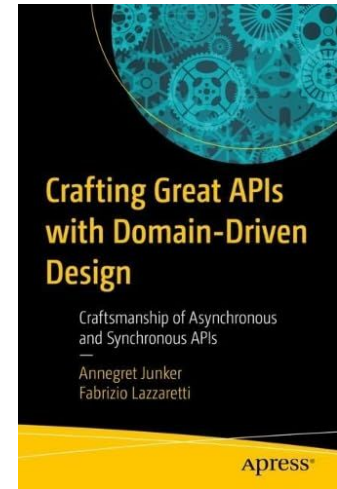
When do we want it?

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, API Product Canvas



Getting started with the project

The 4 phases


Awareness &
Active

 PoC
Minimal Concept


Active

 MVP
Concept
Data Quality
Compliance
Minimal SLAs
Basic Monitoring

Operational

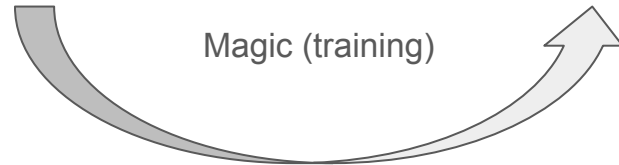
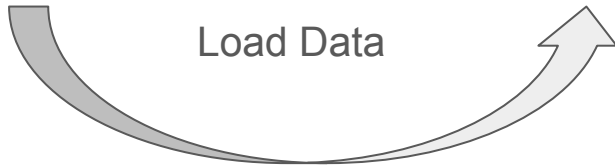
 Product Strategy
Concept
Data Quality
Compliance <small>(NIST AI RMF, ISO 42001)</small>
SLAs
Monitoring
Evolution <small>(A/B testing infrastructure)</small>

Systemic &
Transformational

 Scaled Approach	
Concept	Governance
Data Quality	Disaster Recovery & HA
Compliance <small>(NIST AI RMF, ISO 42001)</small>	Cost Optimization
SLAs <small>(99.9%+ availability)</small>	Training Plan
Monitoring	Observability
Evolution <small>(Canary deployment)</small>	

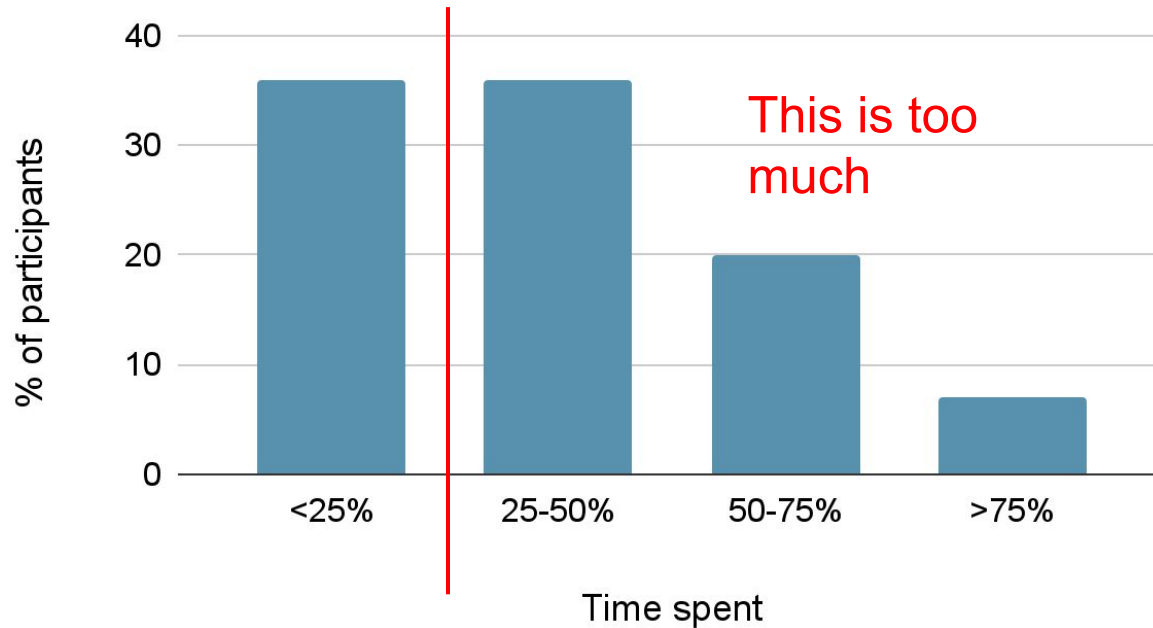
From concept to implementation

End-to-end ML, right?

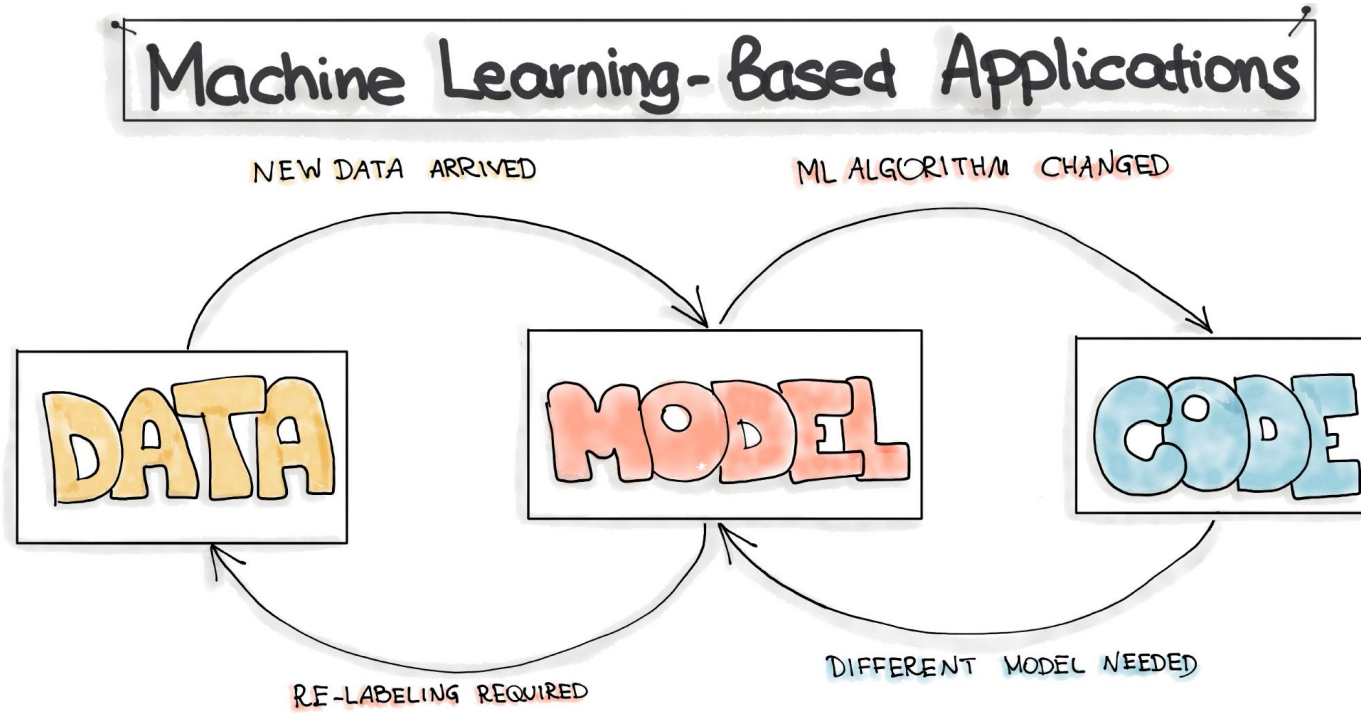


Why is this needed?

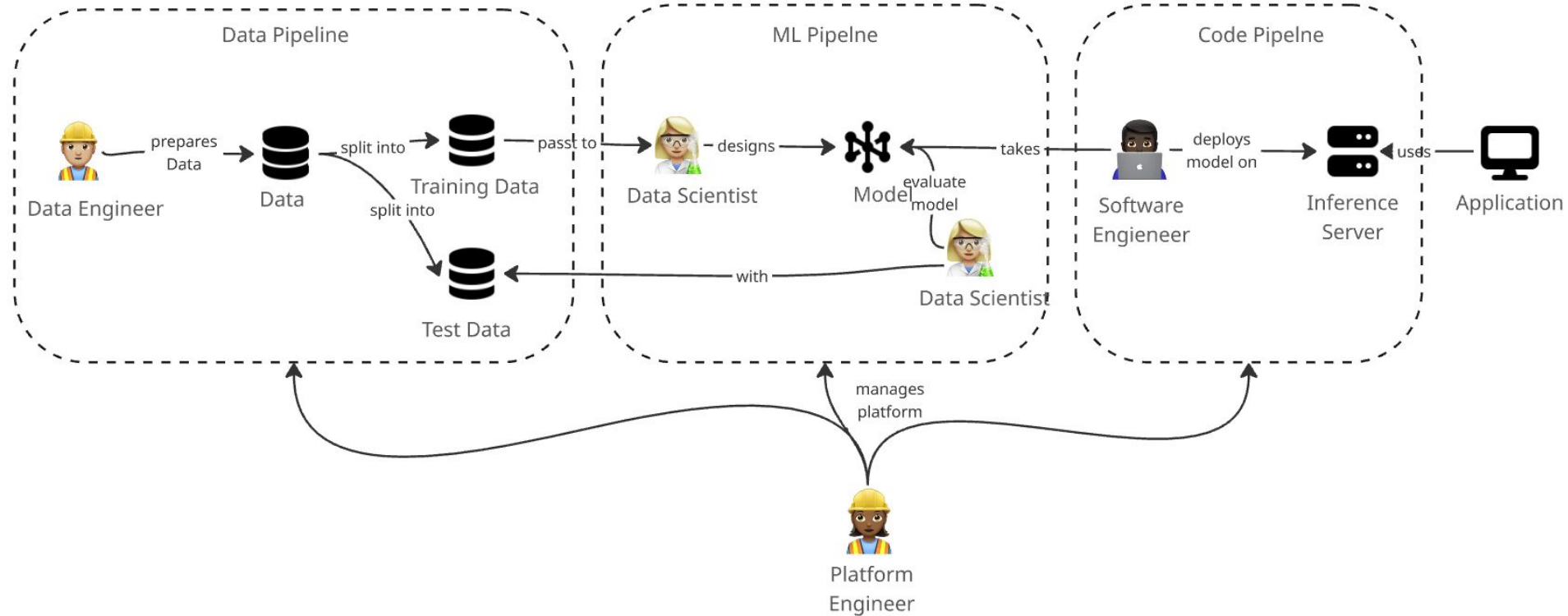
Percent time spent on deploying models



MLOps is not DevOps



MLOps in detail



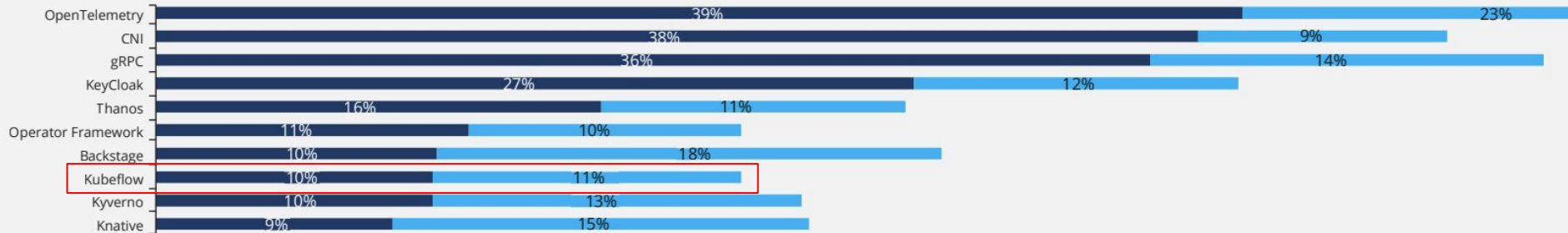
Applying MLOps with Kubeflow

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?

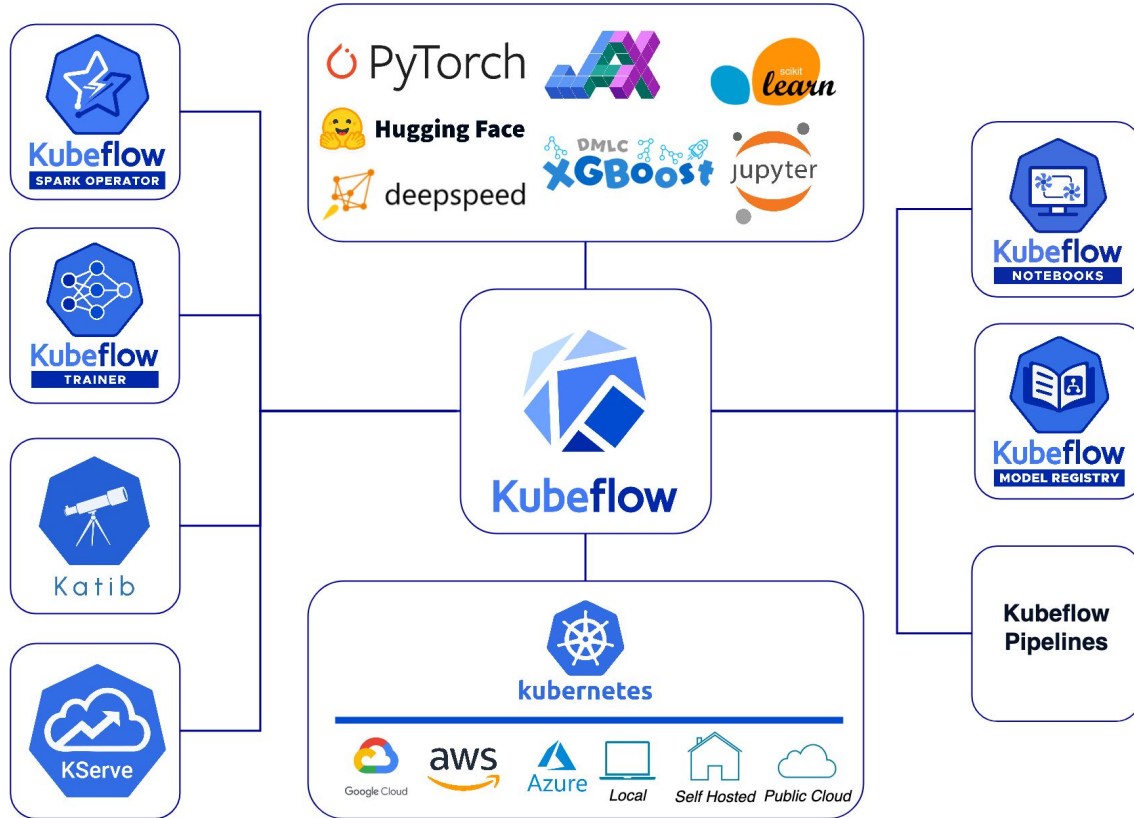


■ Using in Production ■ Evaluating

Who uses Kubeflow?

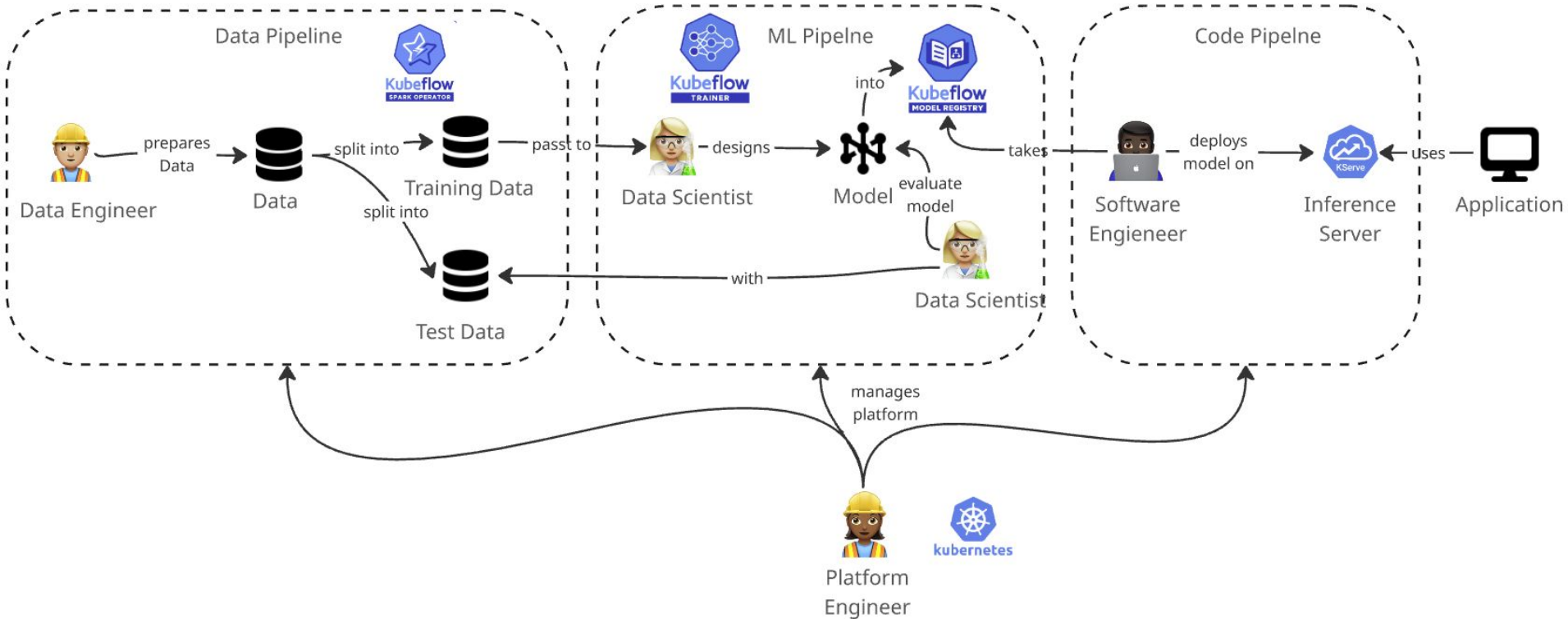
The Kubeflow user survey [2023] drew responses from 90 members of the community mostly made up of members from the United States (43%), Europe (34%), and Asia-Pacific (10%).

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

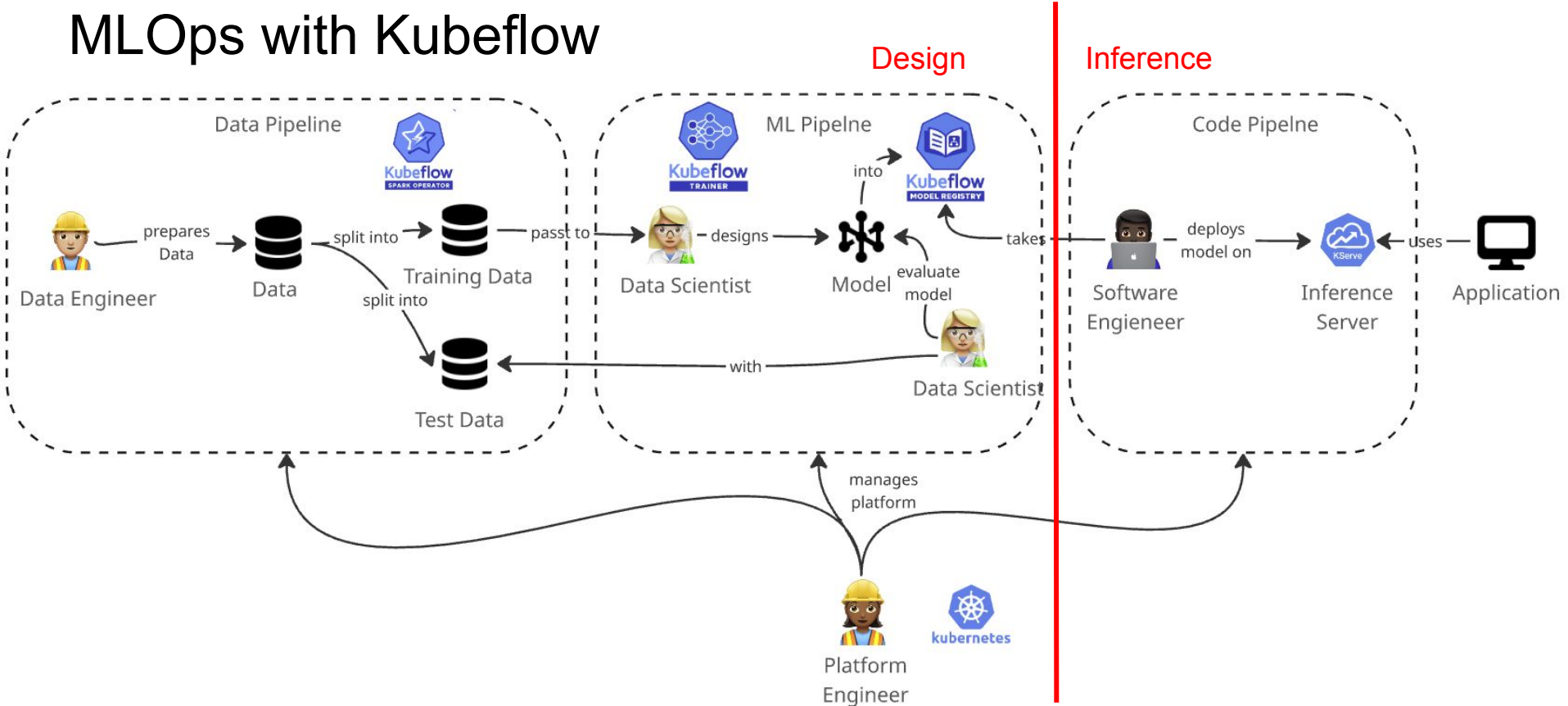


Kubeflow and an own model

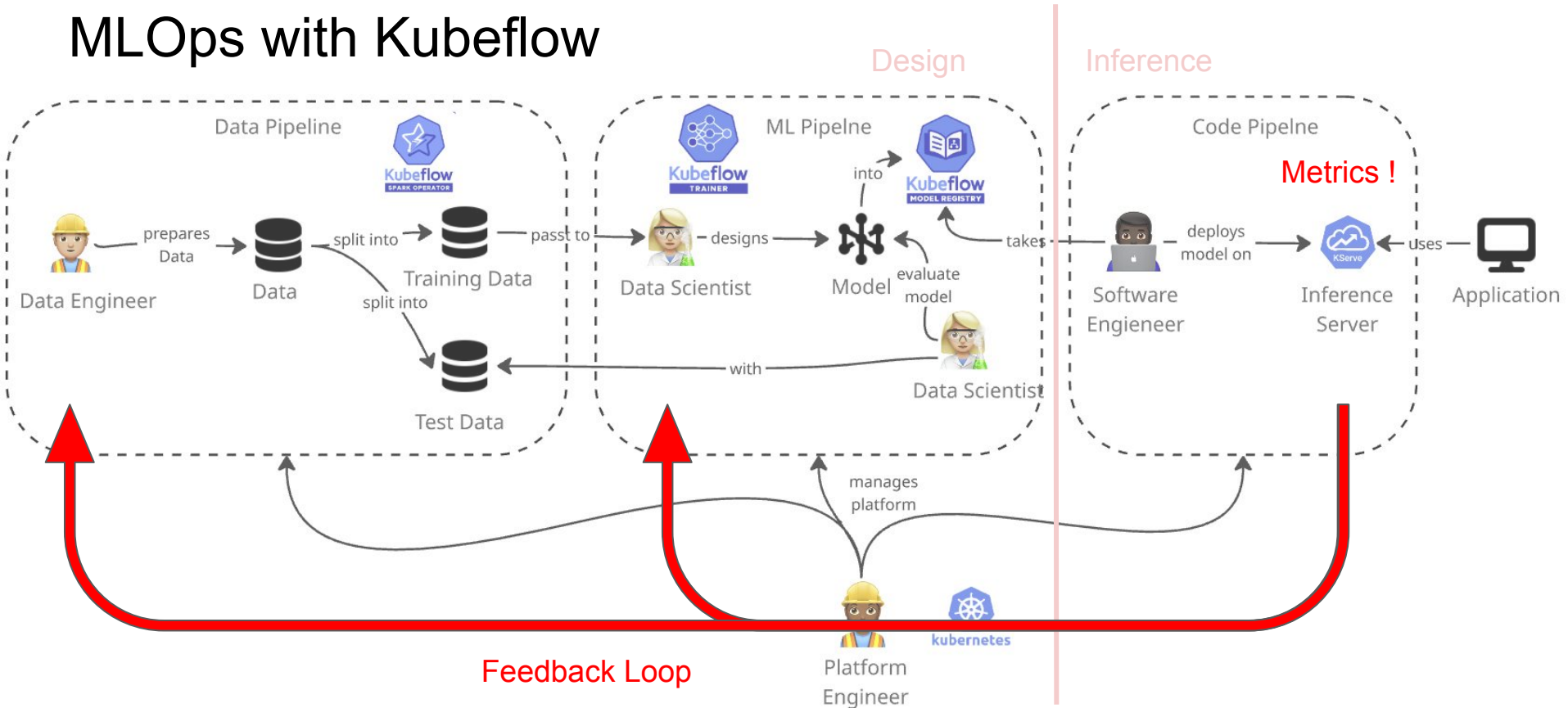
MLOps with Kubeflow



MLOps with Kubeflow

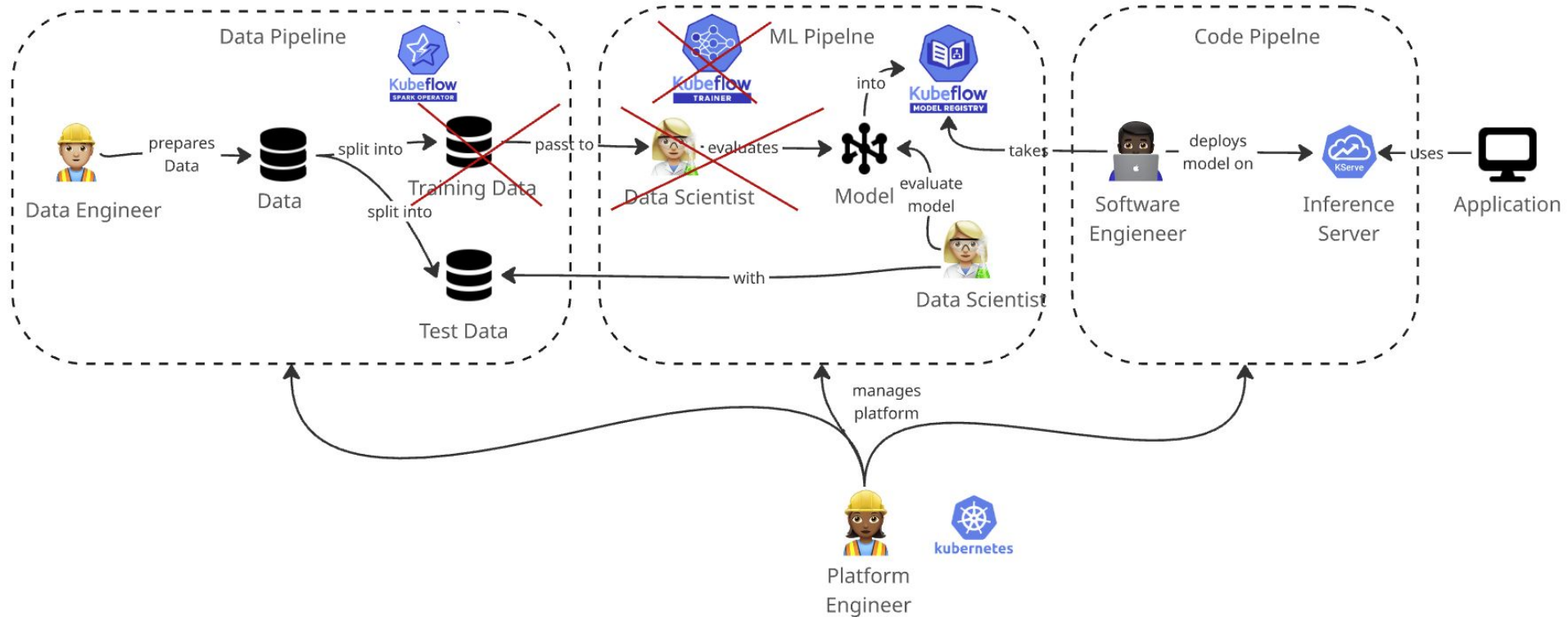


MLOps with Kubeflow



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 many components -> Start with one
 - Be careful about over-engineering
- Run a dedicated cluster
 - Use managed services if possible
- Manage Kubernetes manifests with DevOps practices
 - 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!



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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-ai/ml-toolkits-with-kubeflow-lfs147>