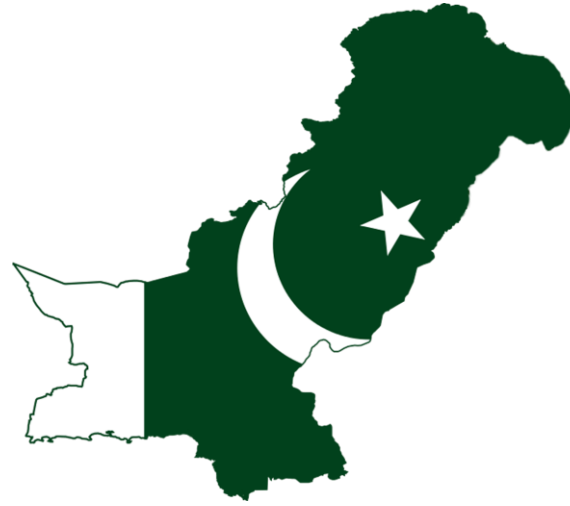




PRIME  
MINISTER'S **Kamyab**  
**Jawan**

کامیاب جوان NATIONAL YOUTH  
DEVELOPMENT  
PROGRAMME

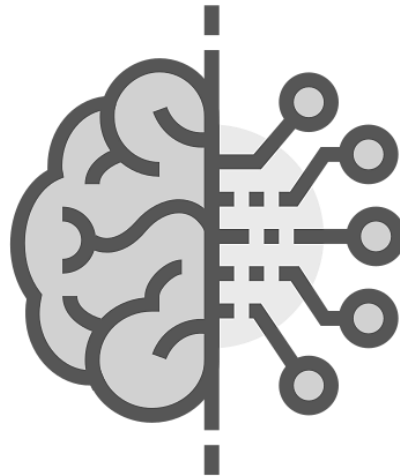
کامیاب جوان کامیاب پاکستان



**Skills for all - Batch III 2022**

# Artificial Intelligence (Robotics)

## Lecture 1: Introduction

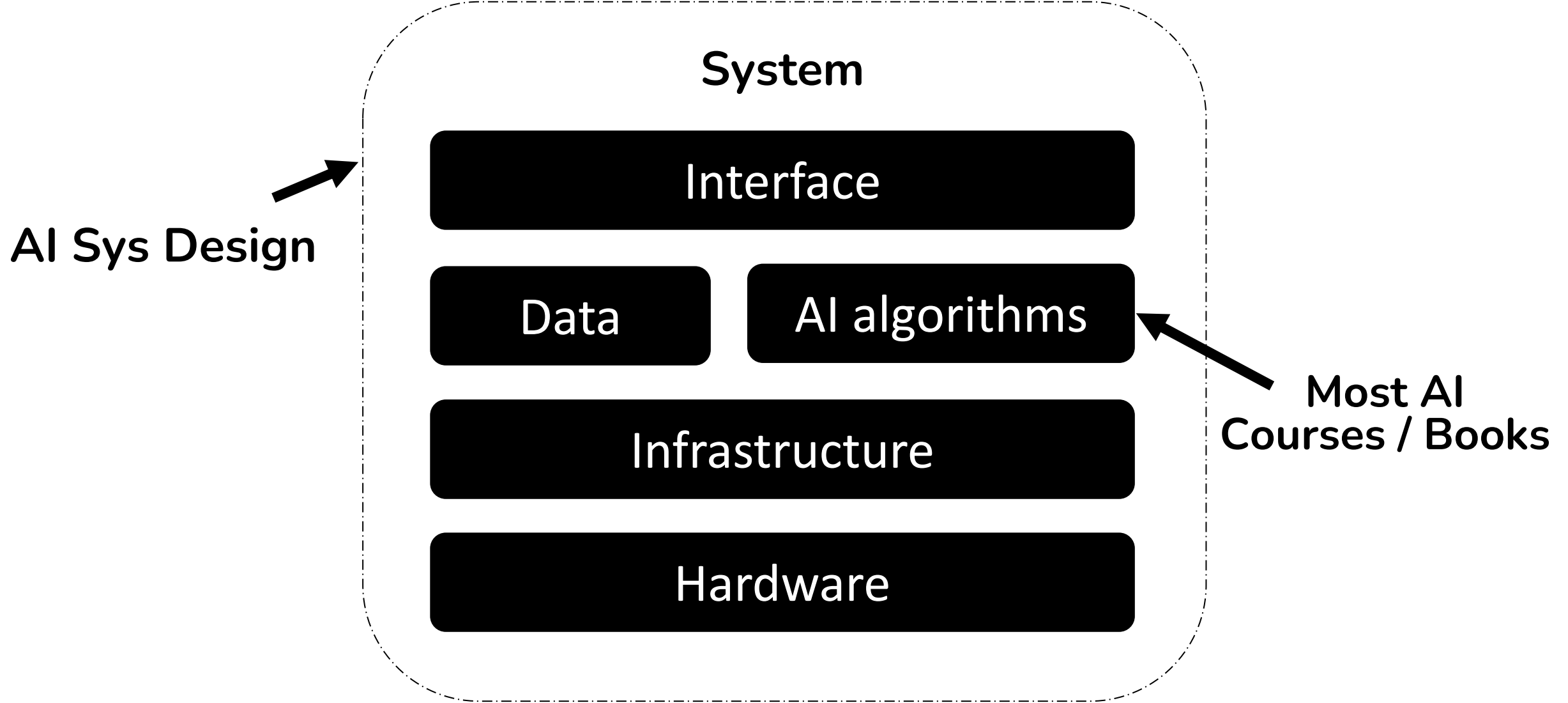


Muhammad Rizwan – Corvit Systems

# Artificial Intelligence: Expectations



This course won't teach  
you how to do this 🤖



Data-Centric AI

# Data-Centric AI


- Major Movement of AI
- Most recent developments shared by multiple experts in AI field
- Latest cutting edge techniques
- Build successful Machine Learning Systems (ML Sys)

***Data-centric AI is  
the discipline of systematically engineering  
the data used to build an AI system.***

# Model-centric approach

$$\text{AI System} = \text{Code} + \text{Data}$$

(algorithm/model)




**Work on this**



# Data-centric approach

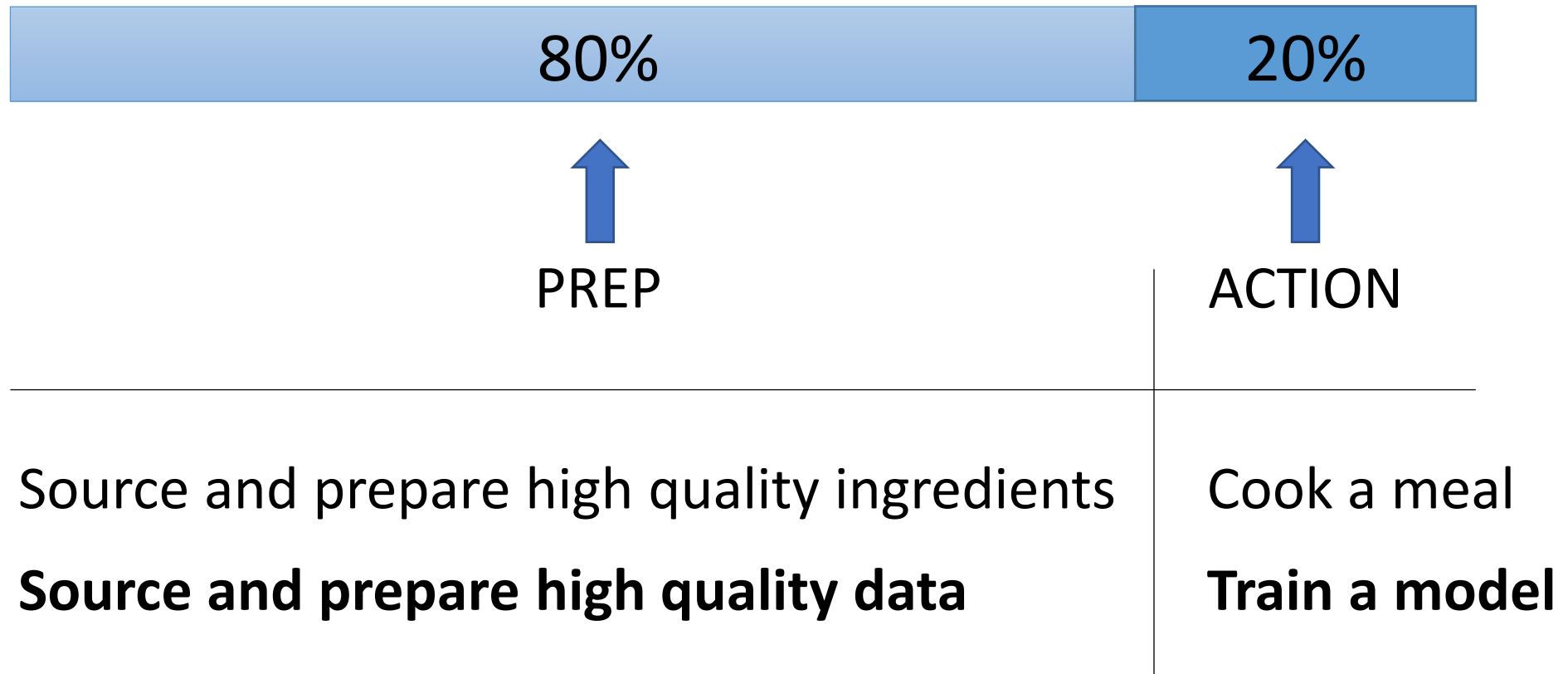
$$\text{AI System} = \text{Code} + \text{Data}$$

(algorithm/model)

The word 'Data' in the equation is circled in orange. An orange arrow points from the text 'Work on this' below to the circled 'Data'.

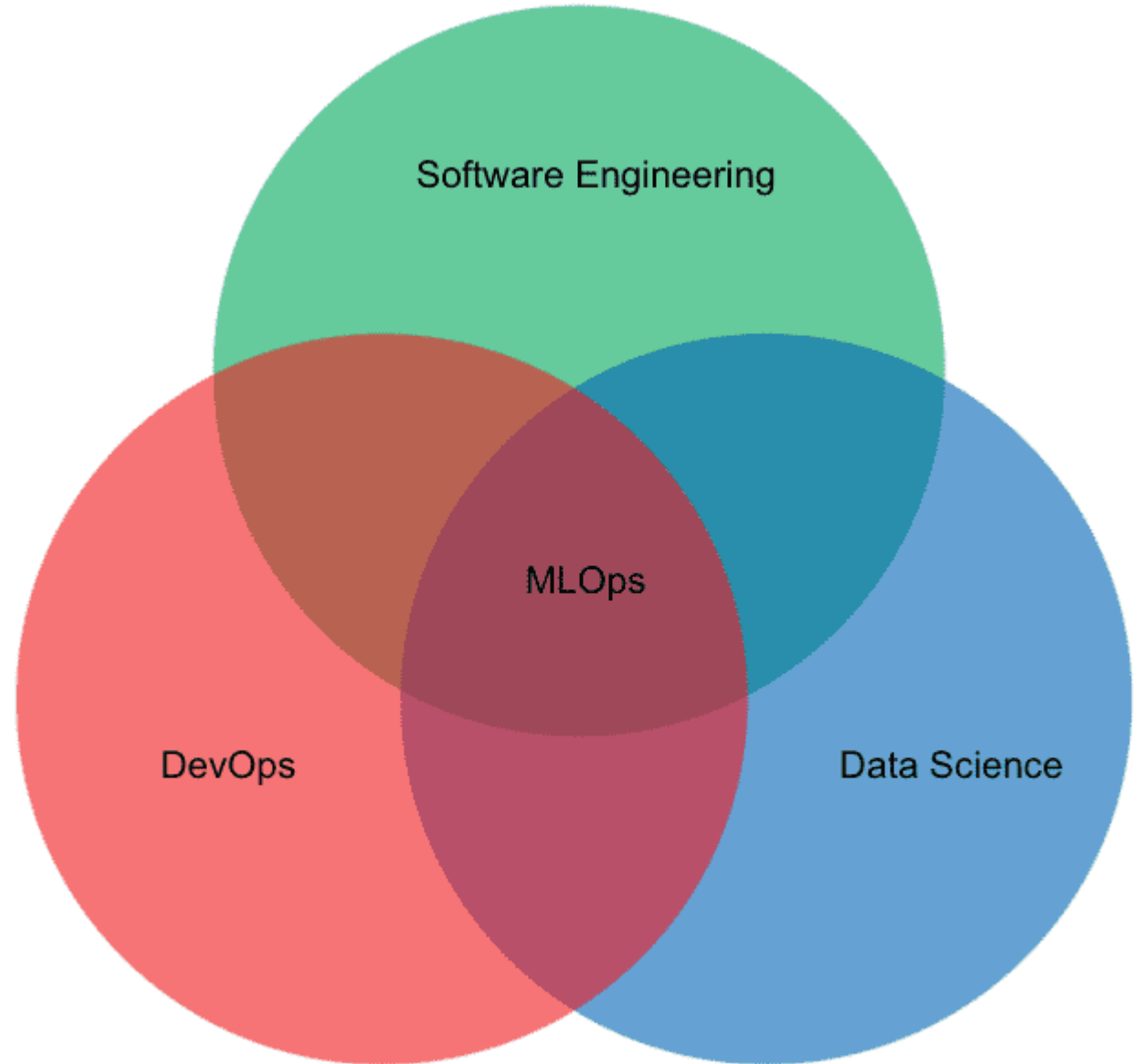
**Work on this**

# Data is Food of AI



ML + DevOps = 

# MLOps



# Making it systematic: The rise of **MLOps**

AI Systems = Code + Data

Creation

Software  
Engineers

ML Engineers

Quality /  
Infrastructure

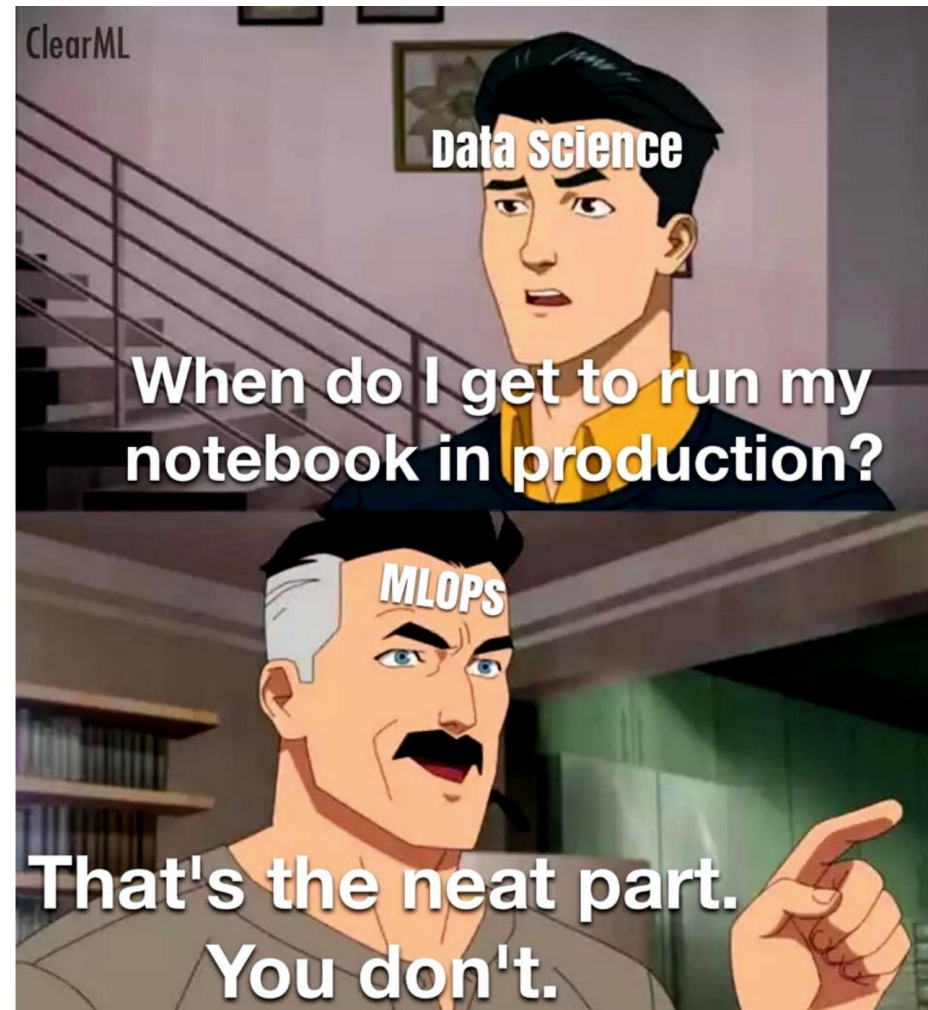
DevOps

MLOps

# From Big Data to Good Data

*MLOps most important task: **Ensure consistently high-quality data in all phases of the AI / ML project lifecycle.***

# Machine Learning in Production



# ML in production: Expectation

1. Collect data
2. Train model
3. Deploy model





# ML in production: Reality

1. Choose a metric to optimize
2. Collect data
3. Train model
4. Realize many labels are wrong -> relabel data
5. Train model
6. Model performs poorly on one class > collect more data for that class
7. Train model
8. Model performs poorly on most recent data -> collect more recent data
9. Train model

# ML in production: Reality

10. Deploy model
11. Dream about \$\$\$
12. Wake up at 2am to complaints that model biases against one group  
-> revert to older version
13. Get more data, train more, do more testing
14. Deploy model
- 15. Pray**
16. Model performs well but revenue decreasing
- 17. Cry**
18. Choose a different metric
19. Start over

# Unexpected problems in production





Amazon SageMaker



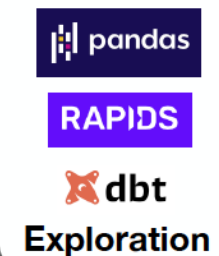
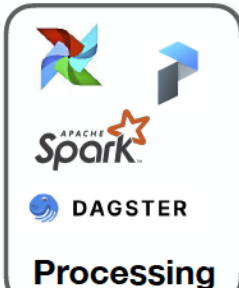
gradient<sup>o</sup>  
by Paperspace

FLOYD

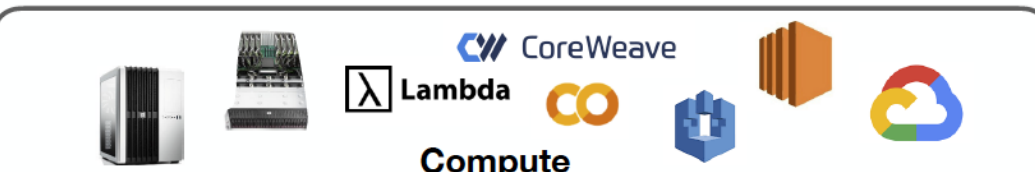
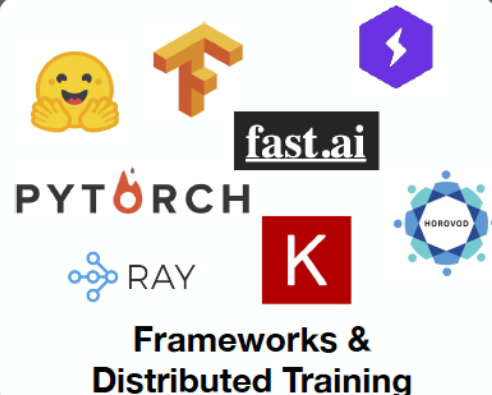
DOMINO  
DATA LAB

# ML Ecosystem

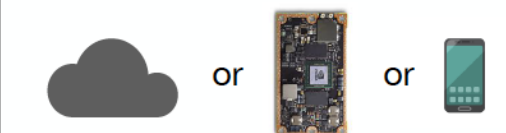
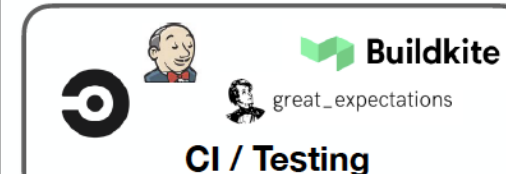
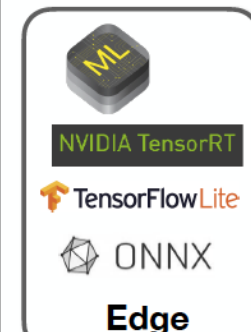
“All-in-one”



Data

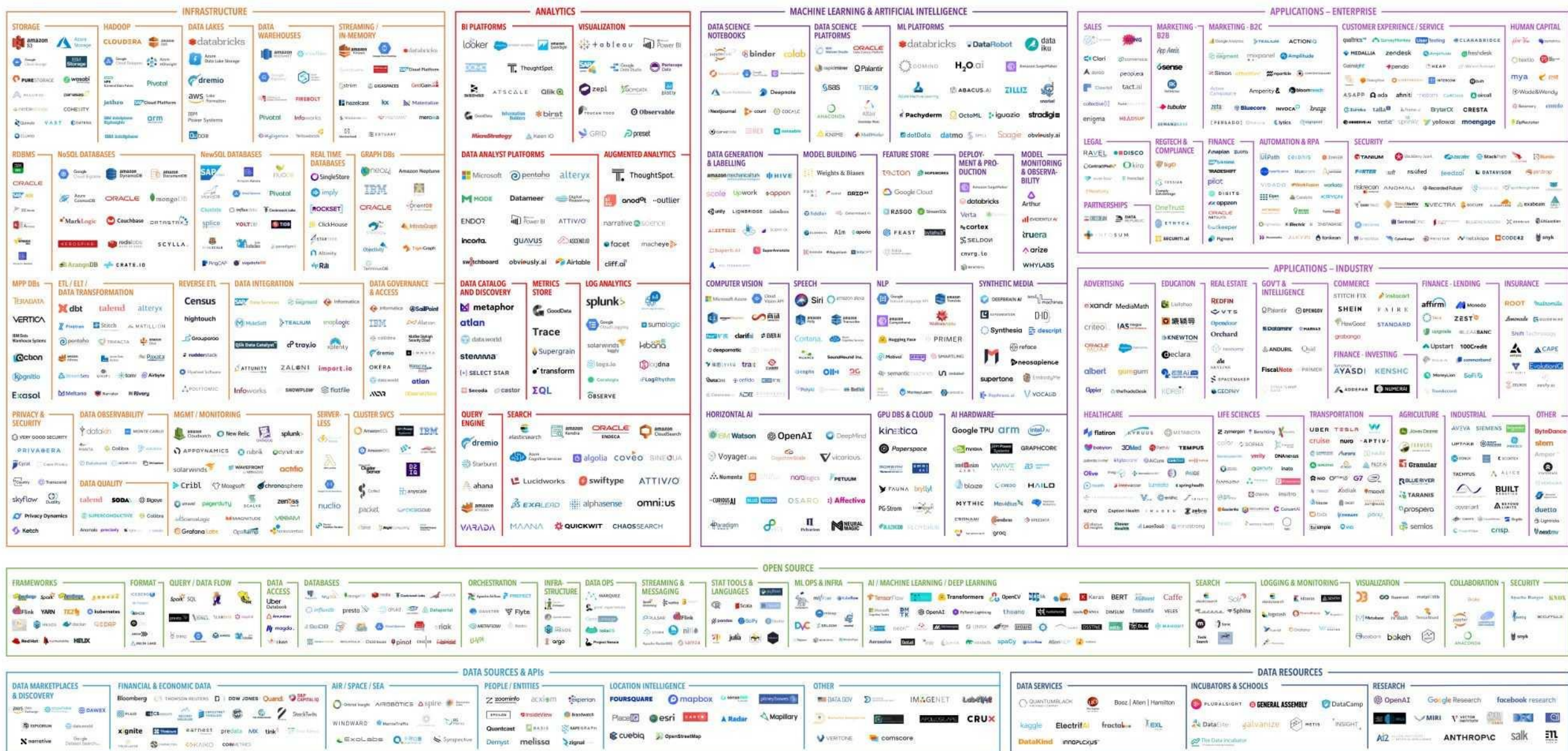


Training/Evaluation

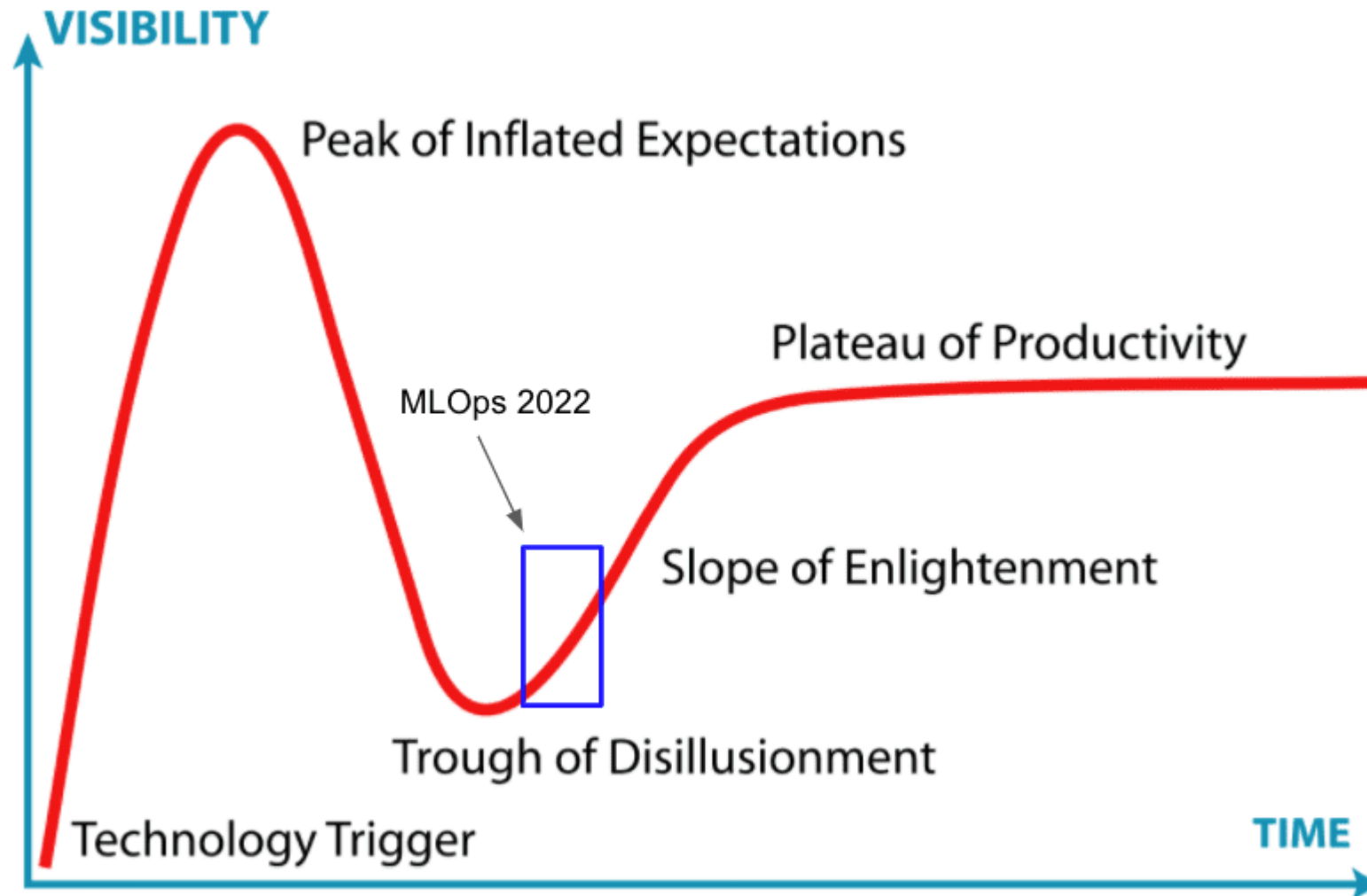




## MACHINE LEARNING, ARTIFICIAL INTELLIGENCE, AND DATA (MAD) LANDSCAPE 2021

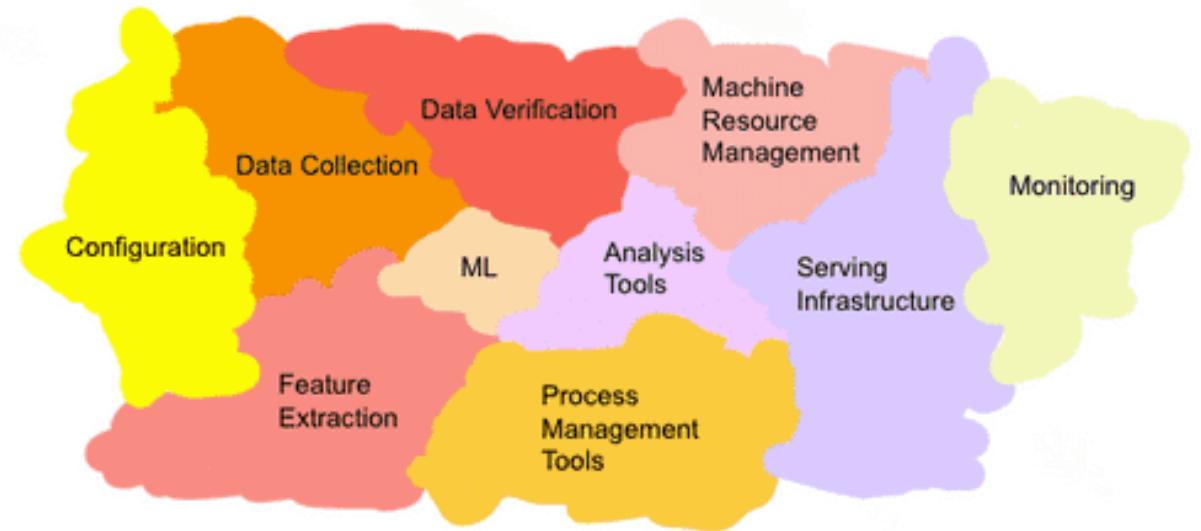
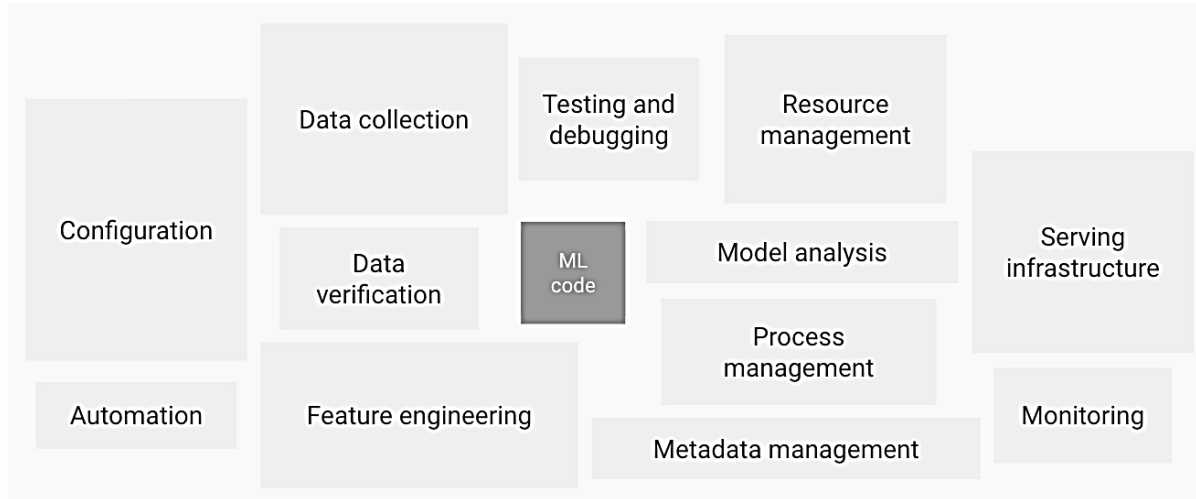


# MLOps: Gartner hype cycle paradigm

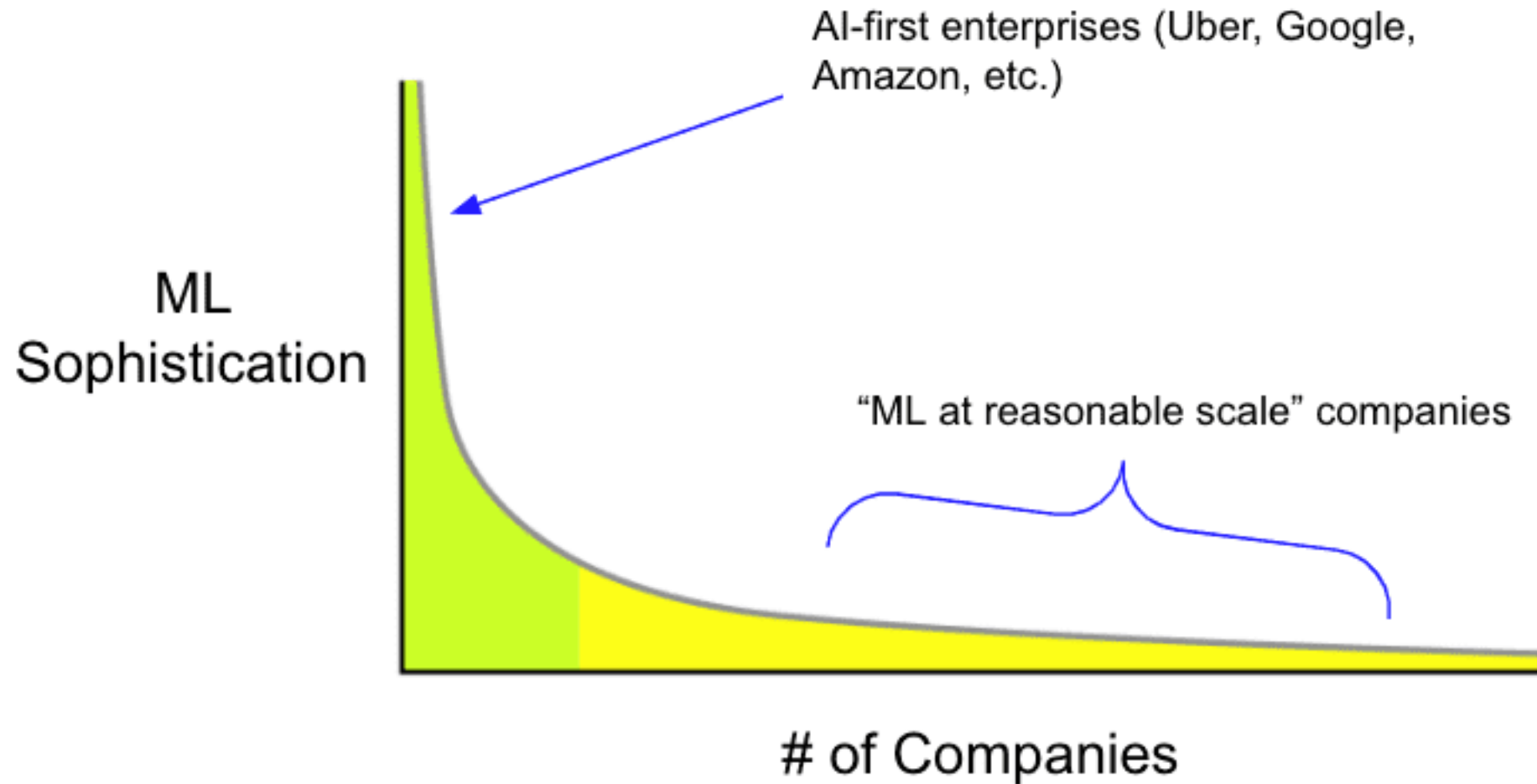


# MLOps pipeline: MLOps Amoeba

## Expectation vs. Reality



# You can start with ML at Reasonable Scale

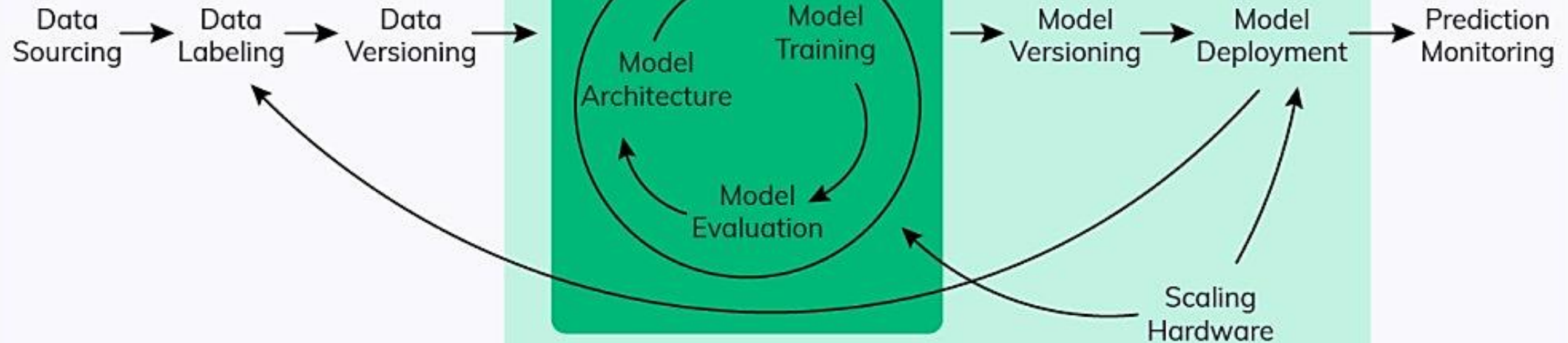




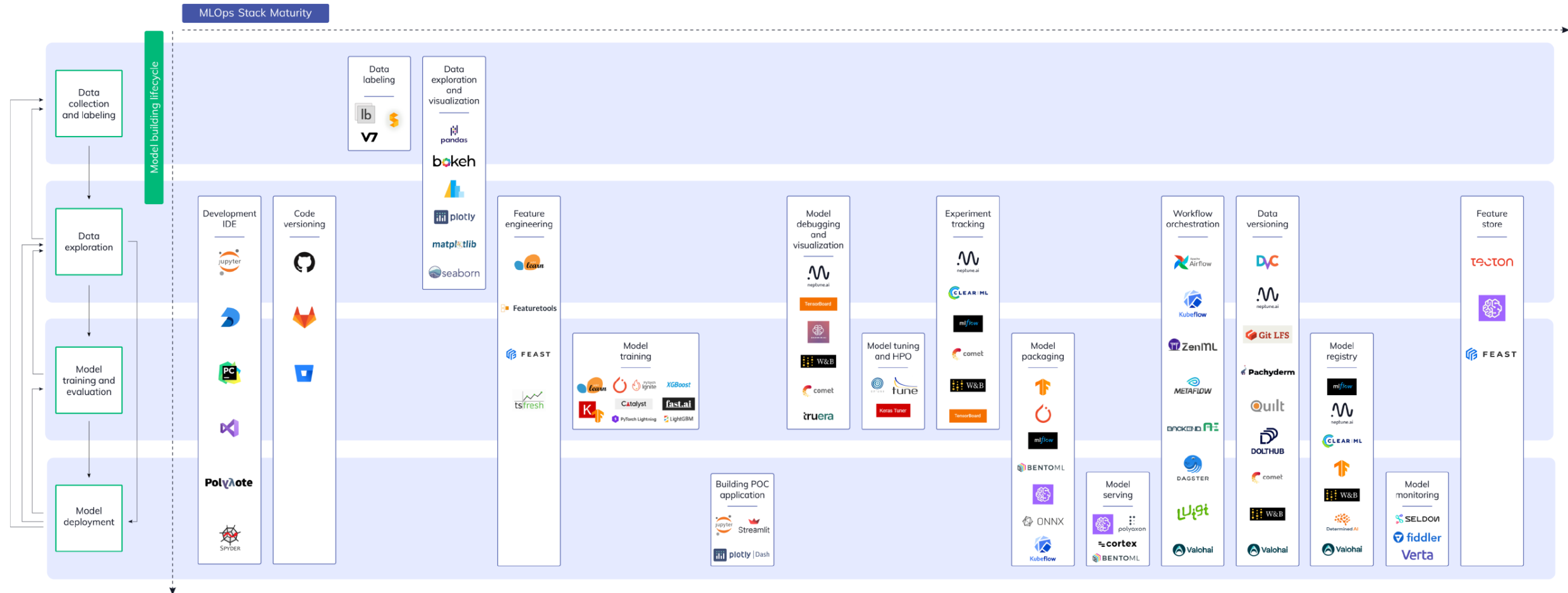
MLOPS

MODEL MANAGEMENT

EXPERIMENT TRACKING

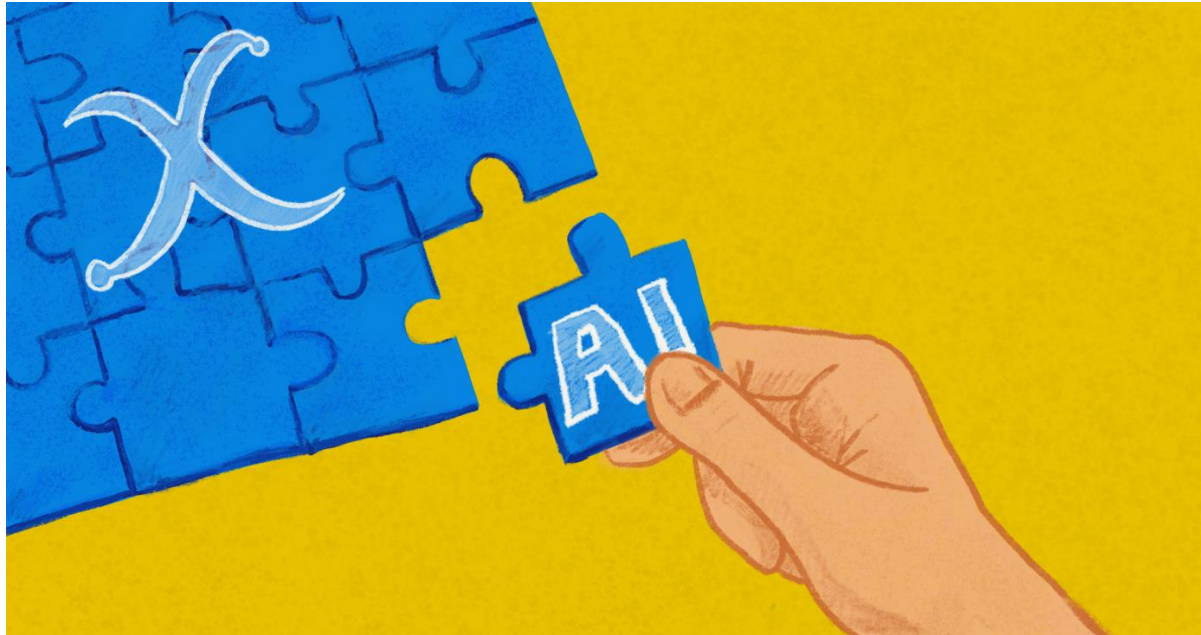


# MLOps tool stack



# AI + X: Don't switch careers, Add AI

Companies need professionals who have domain expertise, augmented with AI skills.



# AI + X: Don't switch careers, Add AI

*Many professionals and students want to jump into the world of AI.*

*But rather than abandoning your current career track to become a **Data Scientist** or a **Machine Learning Engineer**, consider developing AI skills to complement your existing subject matter expertise.*