

# What is MLOps?

MLOPS CONCEPTS



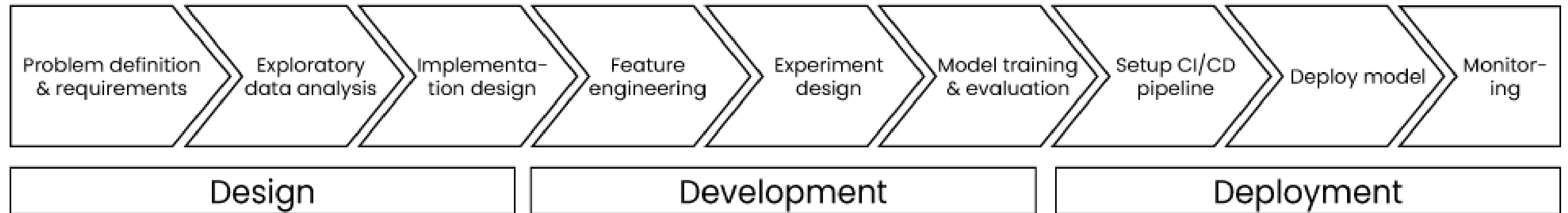
**Folkert Stijnman**

ML Engineer

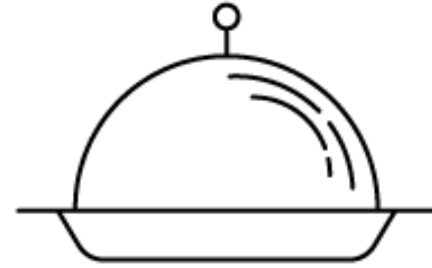
# Machine Learning Operations

*...is the set of practices to **design, deploy and maintain** machine learning in production **continuously, reliably, and efficiently**.*

- Focus on machine learning '*in production*'
- The full machine learning lifecycle



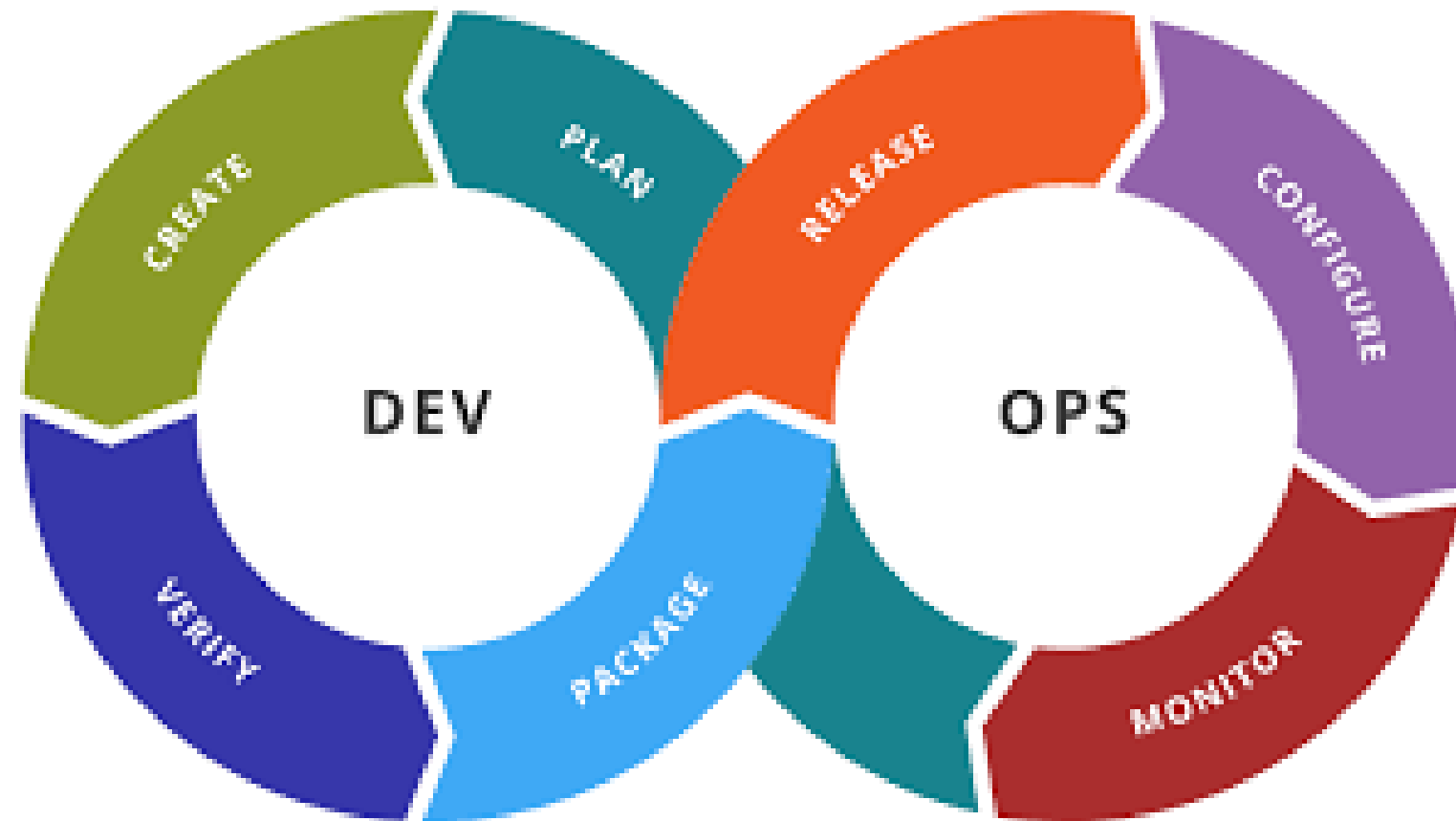
# Why MLOps



- Ingredients
- Guest taste
- People in service
- Kitchen equipment

- Business requirements and value
- Data
- Algorithm
- Computer

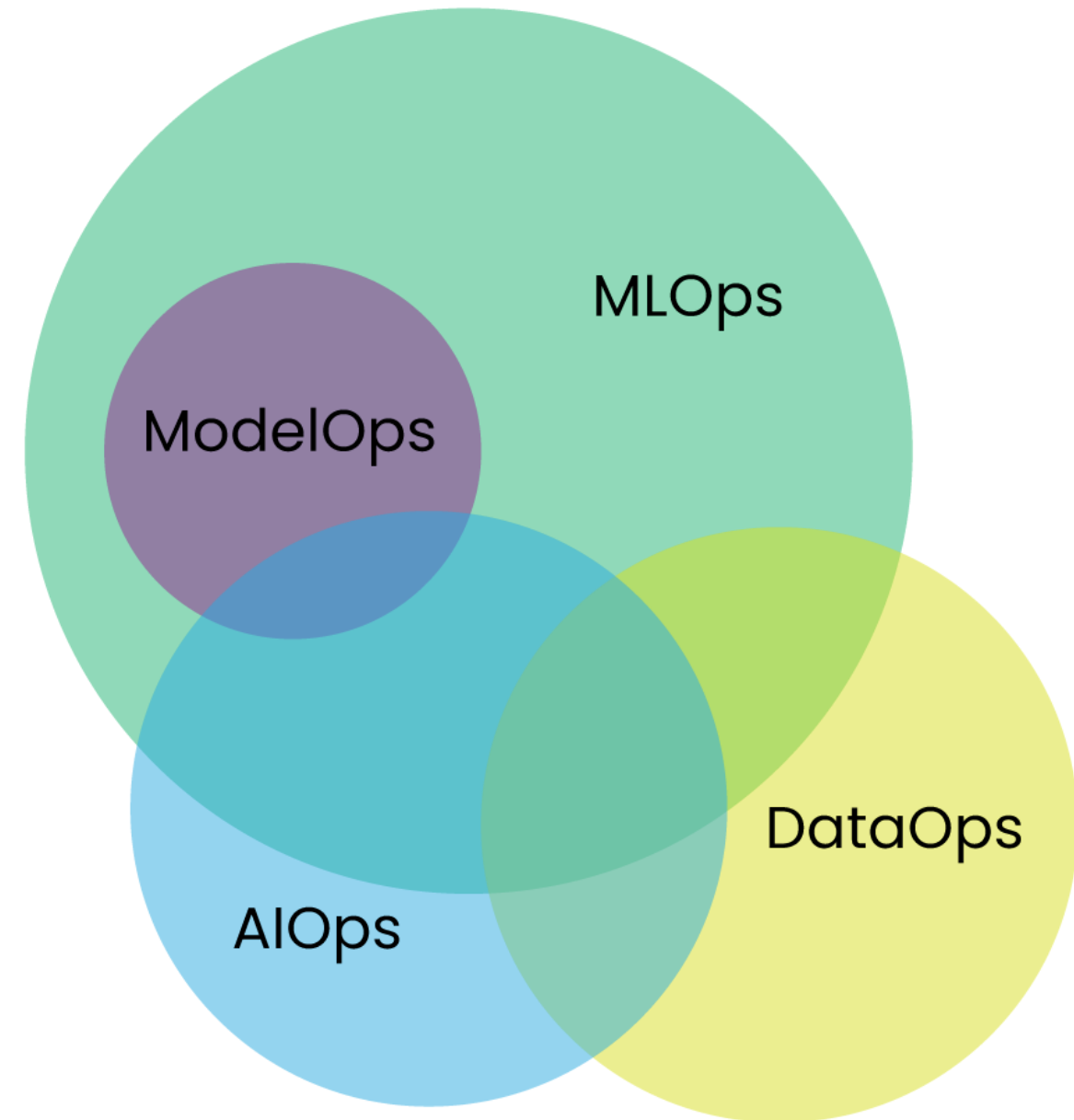
# The origin of MLOps



- Practices and tools to deliver software applications
- Development and Operations used to be separate
- Answer to low velocity in 'traditional' software development

# WhatOps?

- **ModelOps:** Model Operations
  - Is primarily focused on the machine learning model
- **DataOps:** Data Operations
  - Focuses on best practices in data quality and analytics
- **AIOps:** Artificial Intelligence for IT Operations
  - Focuses on using analytics, big data, and machine learning to solve IT issues without human assistance or intervention



# Benefits of MLOps

- Speed
- Reliability and security
- Improved collaboration

**Let's practice!**  
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# Different phases in MLOps

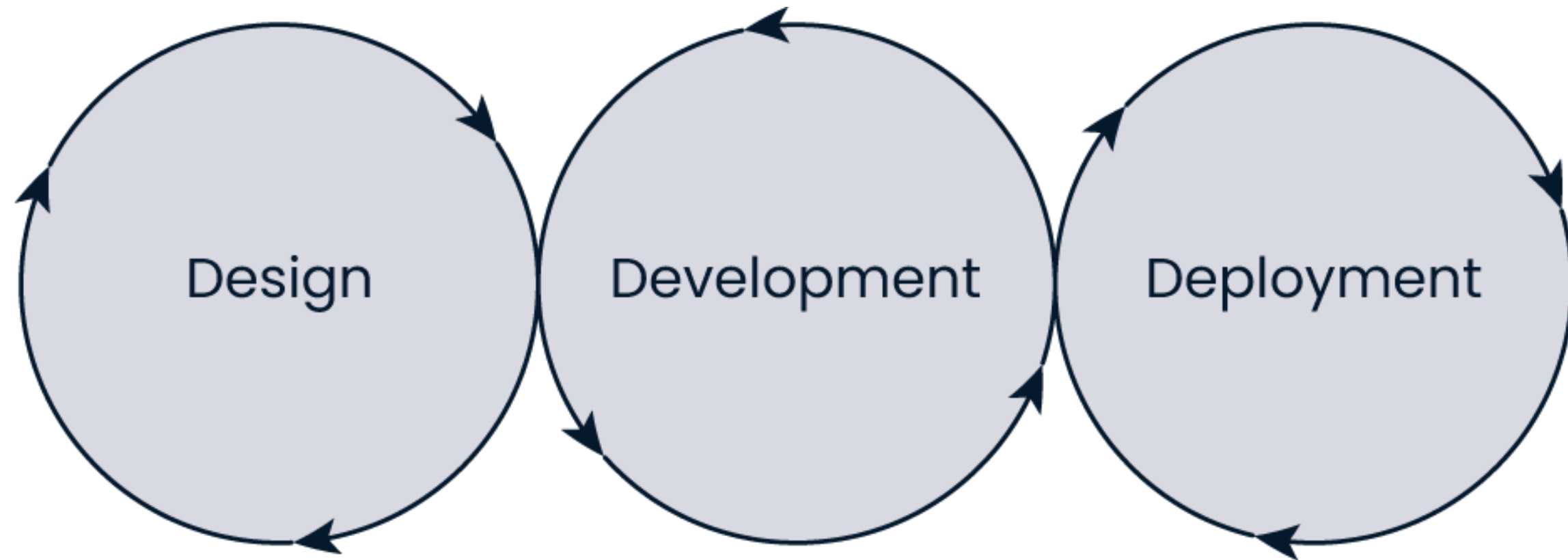
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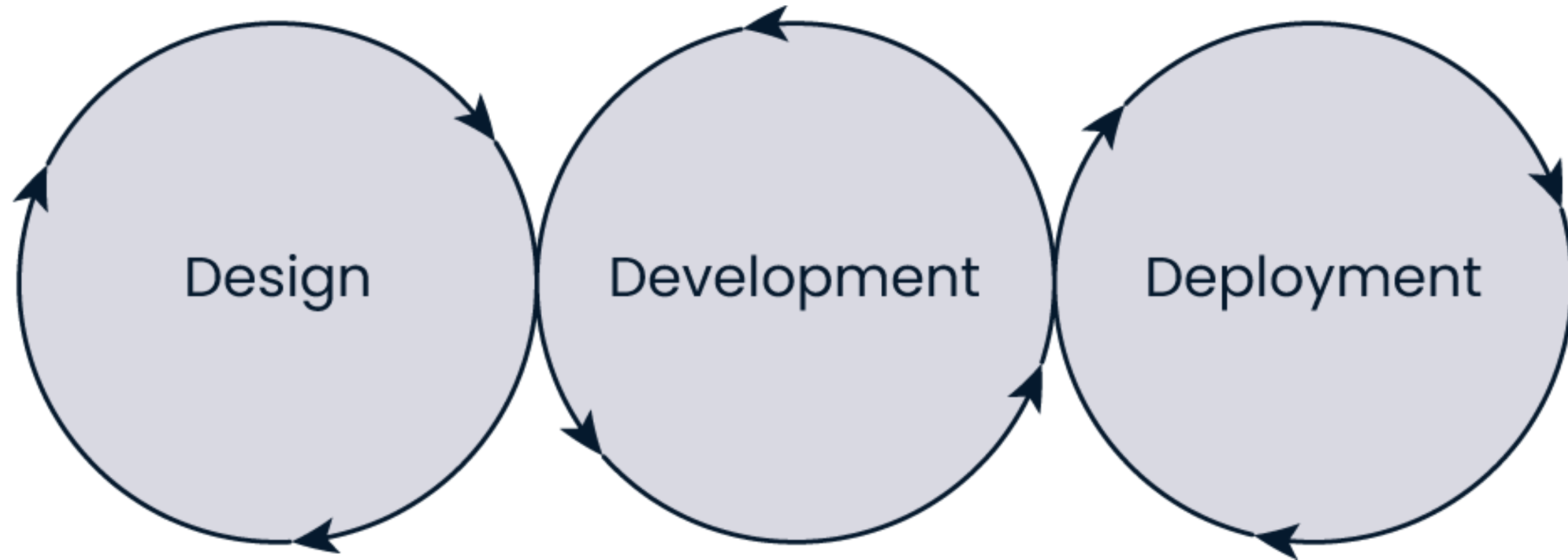


# MLOps lifecycle



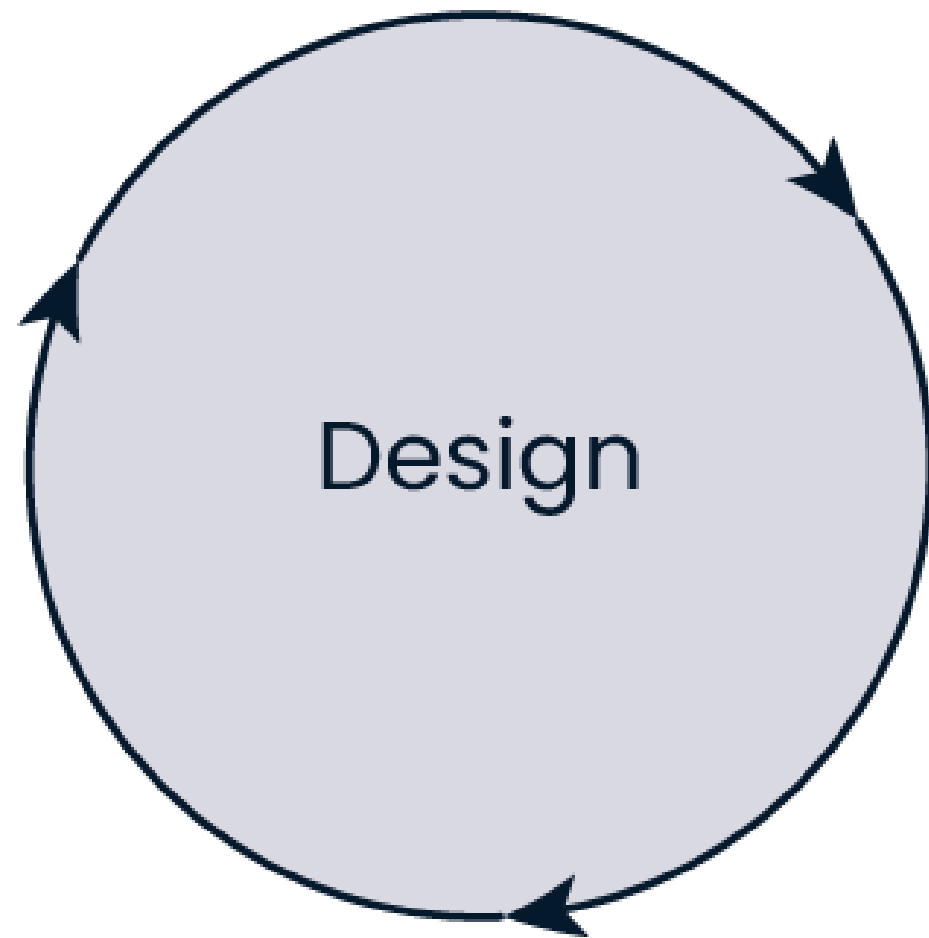
- It is not uncommon to go back and forth between phases
- It is important to constantly evaluate with stakeholders whether the machine learning project should be continued

# Why the machine learning lifecycle?



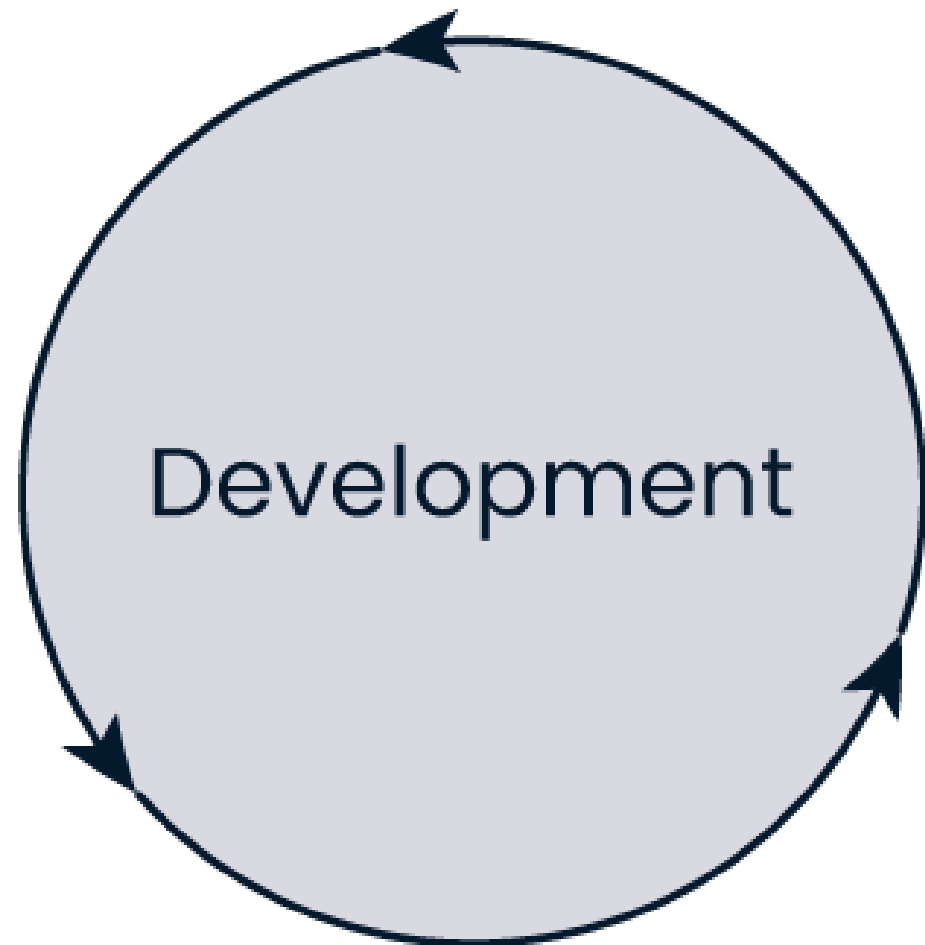
- Structure the process, don't 'just start'
- Defines who is required and when
- Applying practices and tools to specific phases

# Design phase



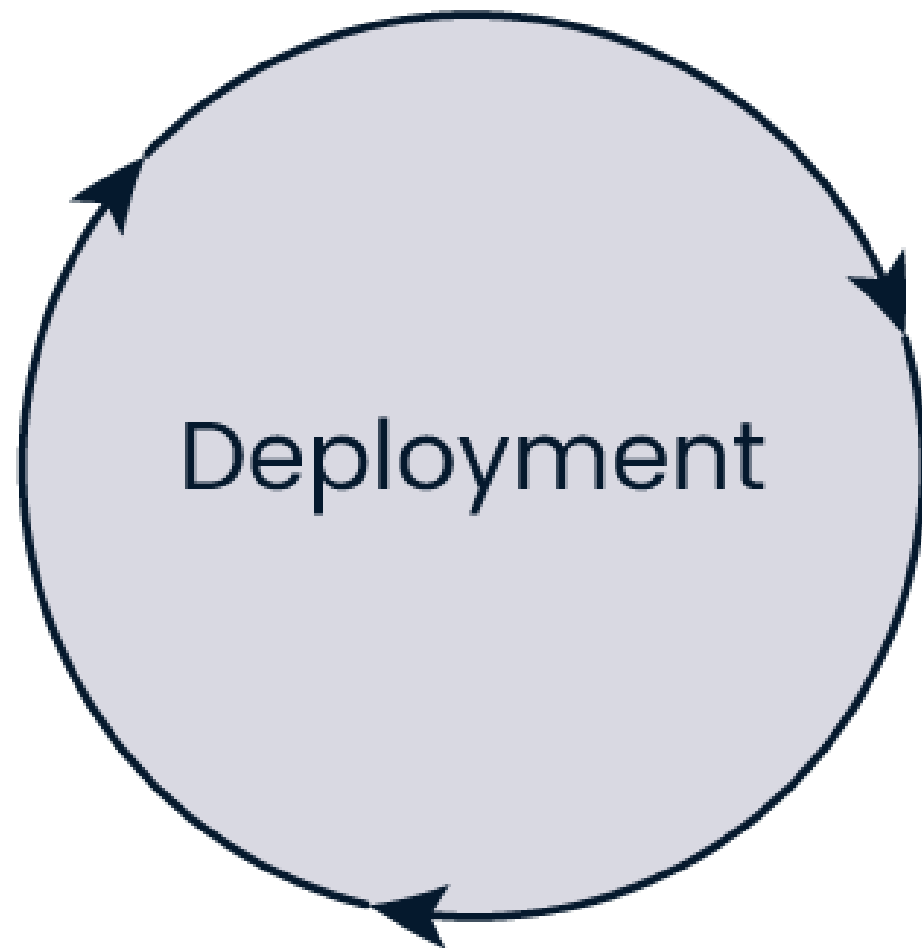
- Context of the problem
- Added value
- Business requirements
- Key metrics
- Data processing

# Development phase



- Develop machine learning model
- Combination of data, algorithms, and hyperparameters
- Model ready for deployment

# Deployment phase



- Integrate the machine learning model in business
- Deploying the model in production
- Monitoring the performance

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# Roles in MLOps

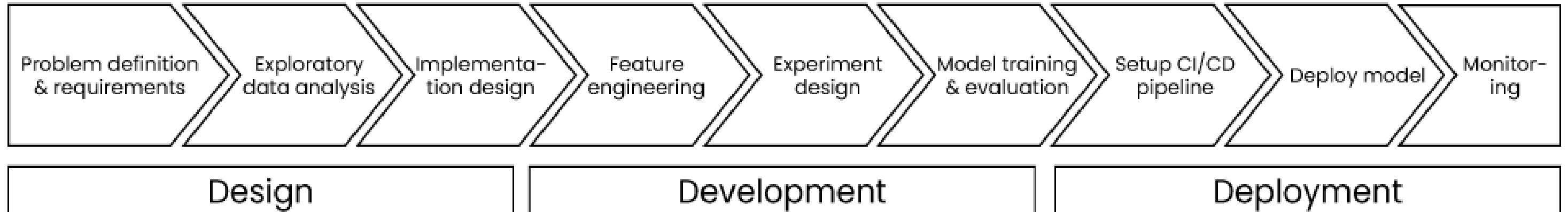
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# Machine learning lifecycle



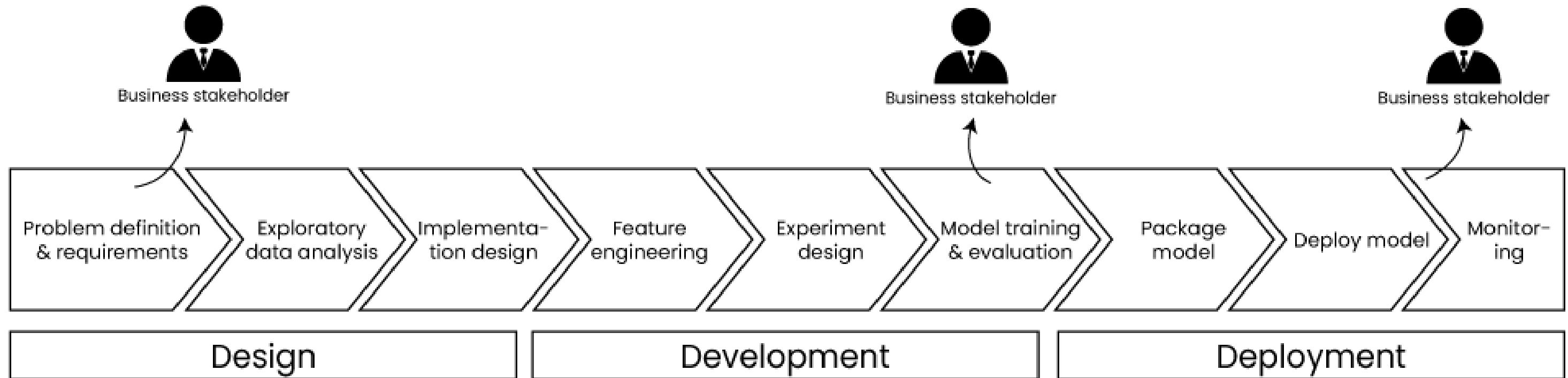
- Business roles
- Technical roles



# Business roles

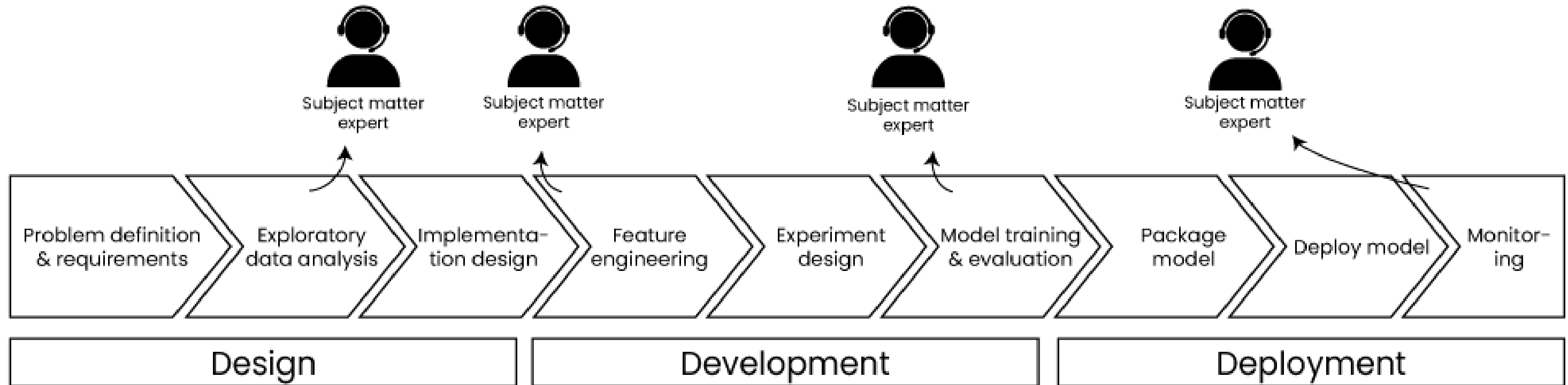
- Business stakeholder
- Subject matter expert

# Business roles: business stakeholder



- Budget decisions
- Vision of company
- Involved throughout the lifecycle

# Business roles: subject matter expert

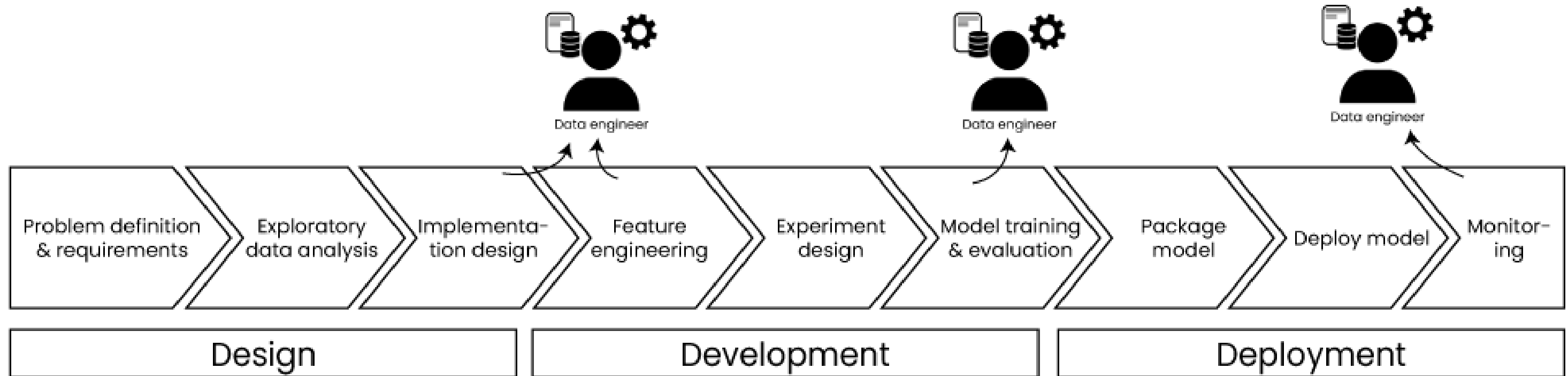


- Domain knowledge
- Involved throughout the lifecycle
- Interpret and validate data

# Technical roles

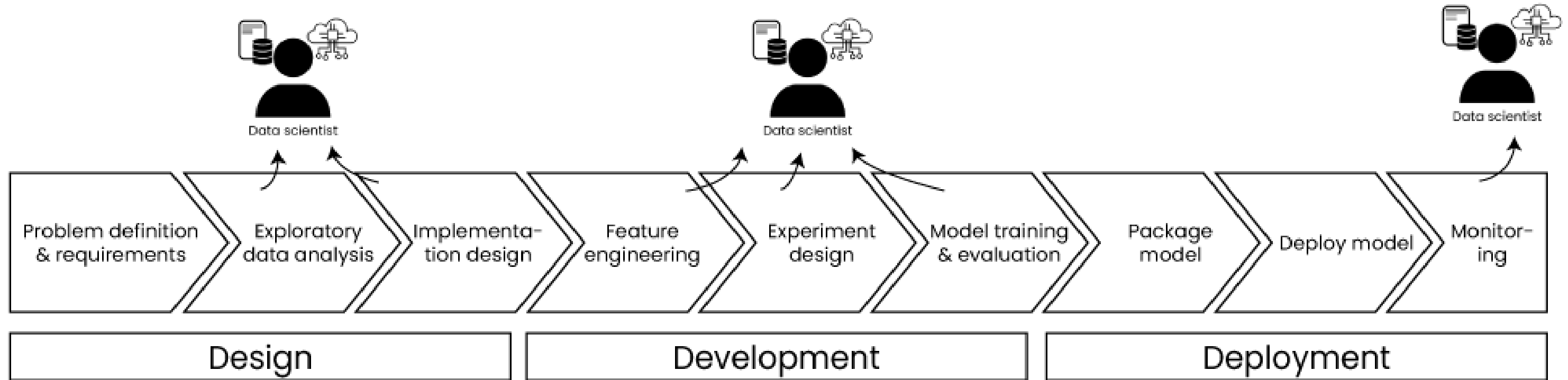
- Data engineer
- Data scientist
- Software engineer
- ML engineer
- Backend engineer

# Technical roles: data engineer



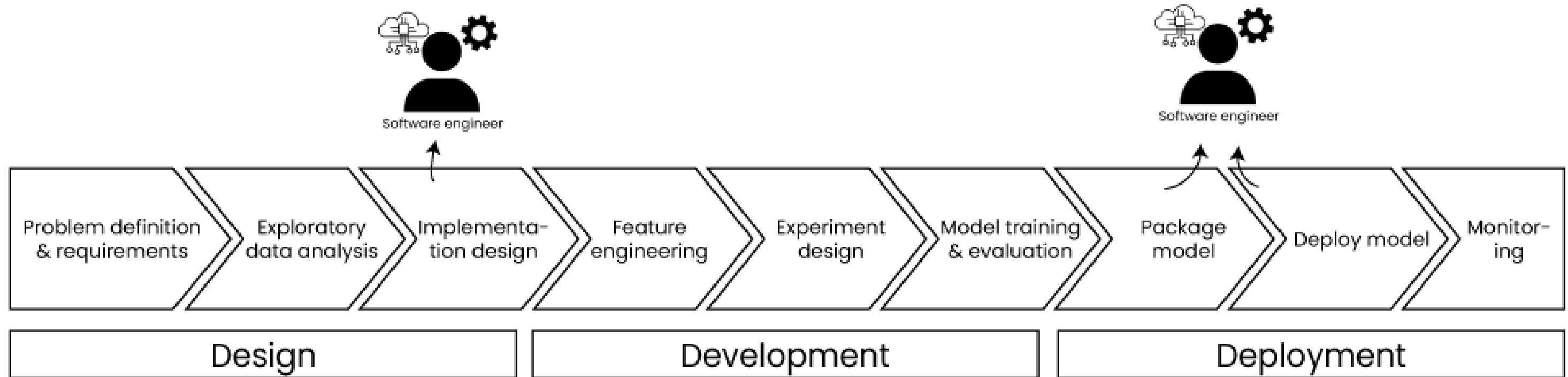
- Collecting, storing, and processing data
- Check and maintain data quality

# Technical roles: data scientist



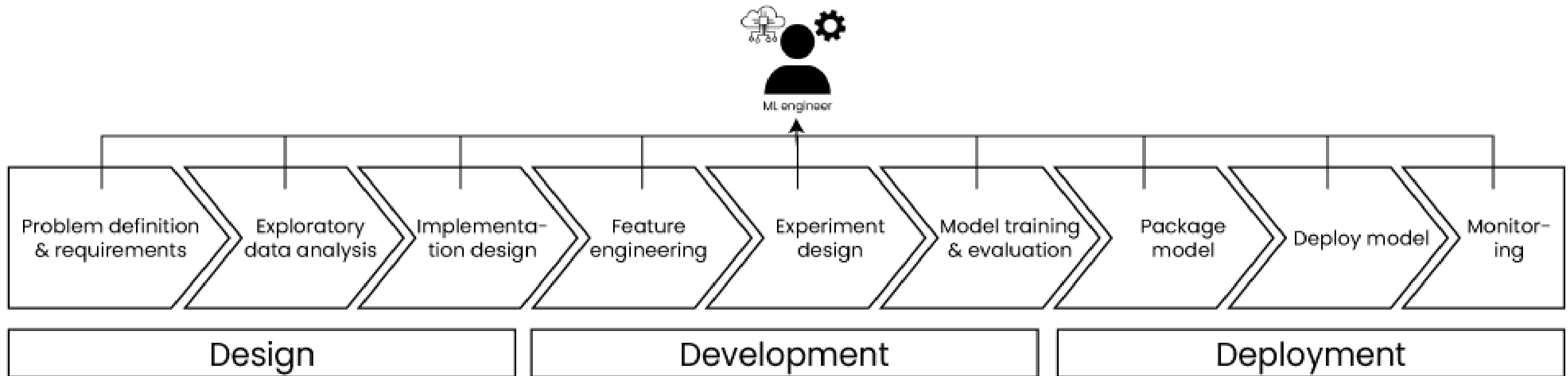
- Data analysis
- Model training and evaluation

# Technical roles: software engineer



- Write software for model deployment
- Make sure that code follows guidelines

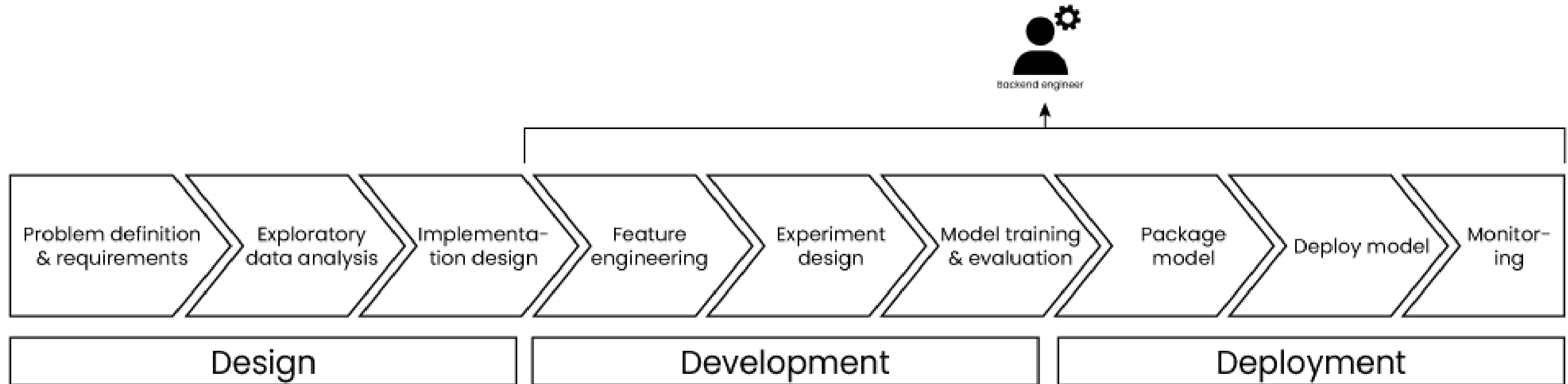
# Technical roles: ML engineer



- Versatile role
- Specifically designed for complete machine learning lifecycle



# Technical roles: backend engineer



- Enable development and deployment
- Cloud infrastructure

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