

# The field of Machine Learning Operations (MLOps)

May 2024

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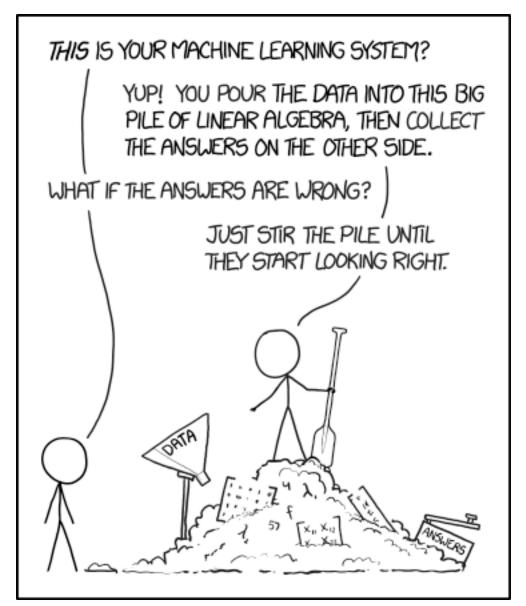
Prepared for the U.S. Department of Energy Under DOE Idaho Operations Office Contract DE-AC07-05ID14517 May 1, 2024 **Brandon Biggs** INL/CON-24-77463 The field of Machine Learning **Operations (MLOps) DevOps for Machine Learning** 

## Introduction

- Problem
- Goals
- MLOps background
- How this helps
- Application
- MLOps levels

### **Problem**

- Machine learning models get created
  - Funding dries up
  - Experts leave
  - Models are only ever run out of a Jupyter Notebook
  - Dependencies are too messy (or not available) for adoption later



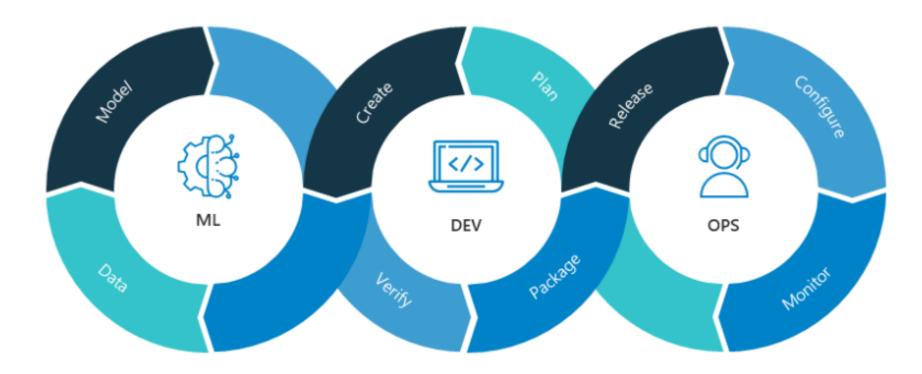
https://xkcd.com/1838/

#### Goals

- 1. Lower the barrier to entry for using AI/ML in everyday research work
- 2. Simplify the production process for AI/ML engineers and data scientists
- Enable collaboration and visibility via a central platform for uploading and sharing models
- Reduce repeated work by eliminating the need to recreate models from scratch to reproduce results
- 5. Increase impact, longevity, and reach of models created at INL through an accessible model hosting platform

## What is MLOps?

- "Like Dev Ops but need to be a data scientist and know about hardware"
  - John Lockman from Dell



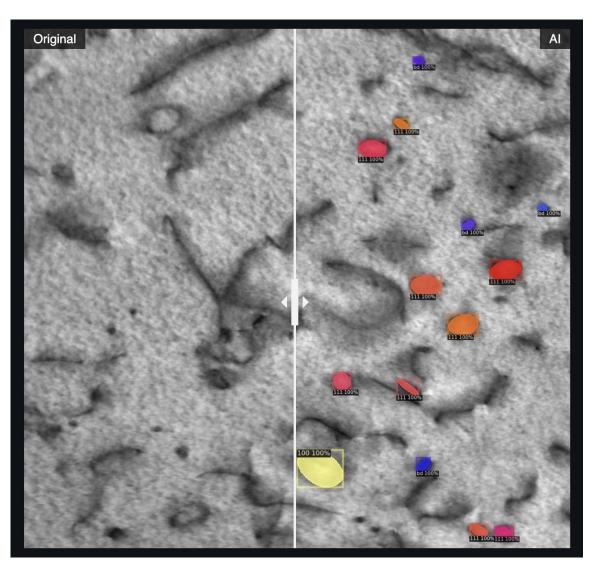
Source: Neal Analytics

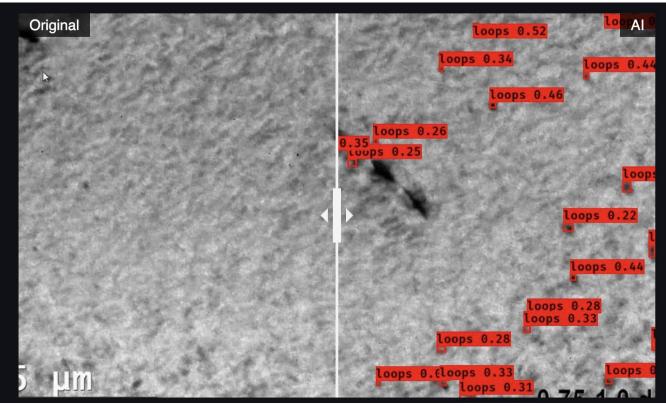
## How does this help?

#### Models get created

- Weights get saved to a central location and the model gets deployed
- Models get hosted in centralized location where they can be accessed from more than a Jupyter Notebook
- Dependencies can be containerized
- Performance of models can be tracked

## **Building Tools around the models**



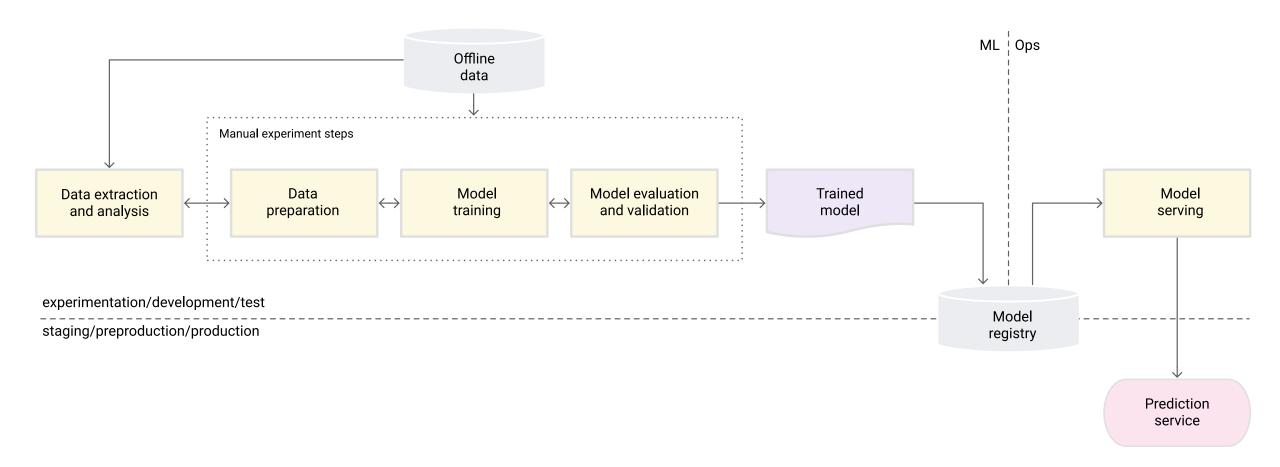


## **MLOps Levels**

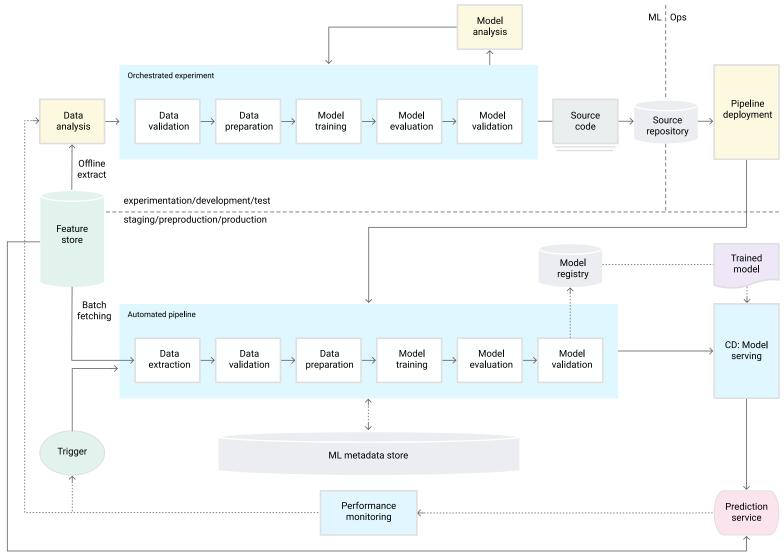
- Google defines 3 levels of MLOps
  - Level 0: Manual process
  - Level 1: ML pipeline automation
  - Level 2: CI/CD pipeline automation

https://cloud.google.com/architecture/mlops-continuous-delivery-and-automation-pipelines-in-machine-learning

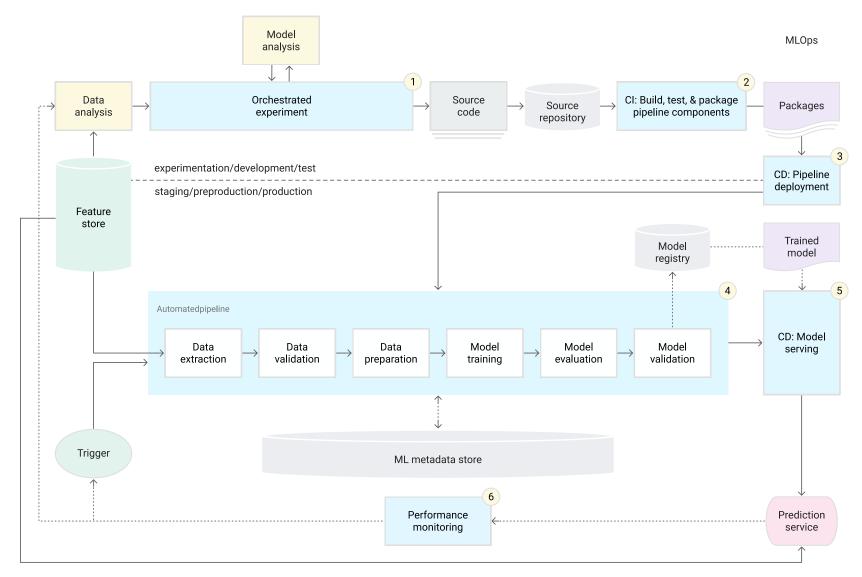
## **MLOps level 0: Manual process**



## **MLOps level 1: ML pipeline automation**



## MLOps level 2: CI/CD pipeline automation



## **Questions?**

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