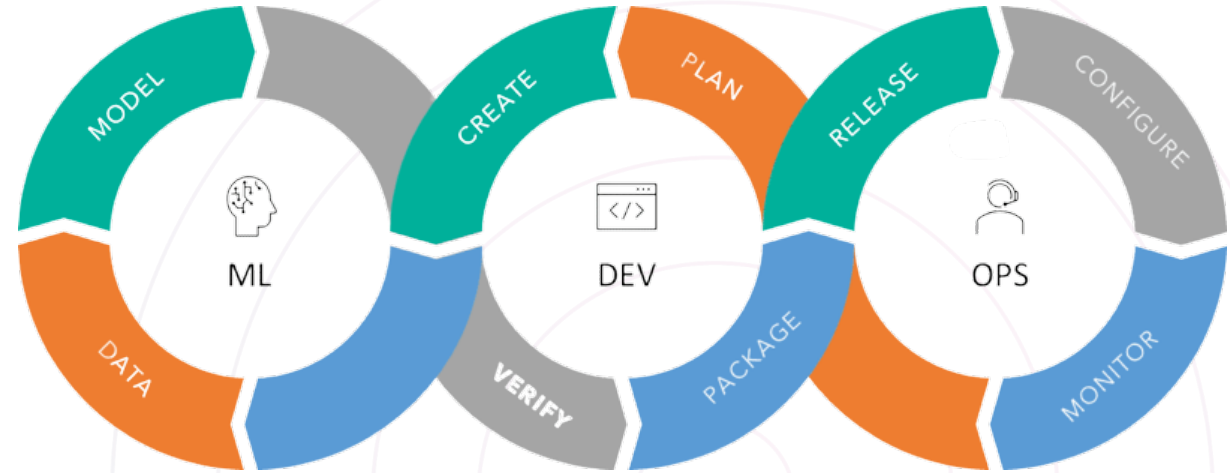




# 1 Introducing MLOps



# The road so far





# DevOps

## What is DevOps?

Tools and best practices to improve software development process and operations

## Advantages:

- Speed
- Rapid Delivery
- Reliability
- Improved collaboration
- Security





# Continuous improvement

CI/CD Deliver **faster** and **better**

## Continuous Integration

- Allows small incremental improvements
- Automates the build, test, and packaging of applications in a reliable and repeatable way.
- Streamlines code changes
- Set of practices performed as developers are writing code

## Continuous Delivery

- Automated delivery of code
- Allows for continuous deployment
- Set of practices performed after the code is completed



# Is DevOps enough?

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

Typical Software

## Typical software developer's flow

Save code changes → Refresh → Changed

## Typical data scientist's flow

Save code change → Spin a cluster → Deploy code  
→ Transfer Data → Model Training



# Why is DevOps not enough?

## Traditional software

Only Code

Easy to get feedback

Code is relatively static

## Machine learning systems

Code + Data

Code change → retrain model

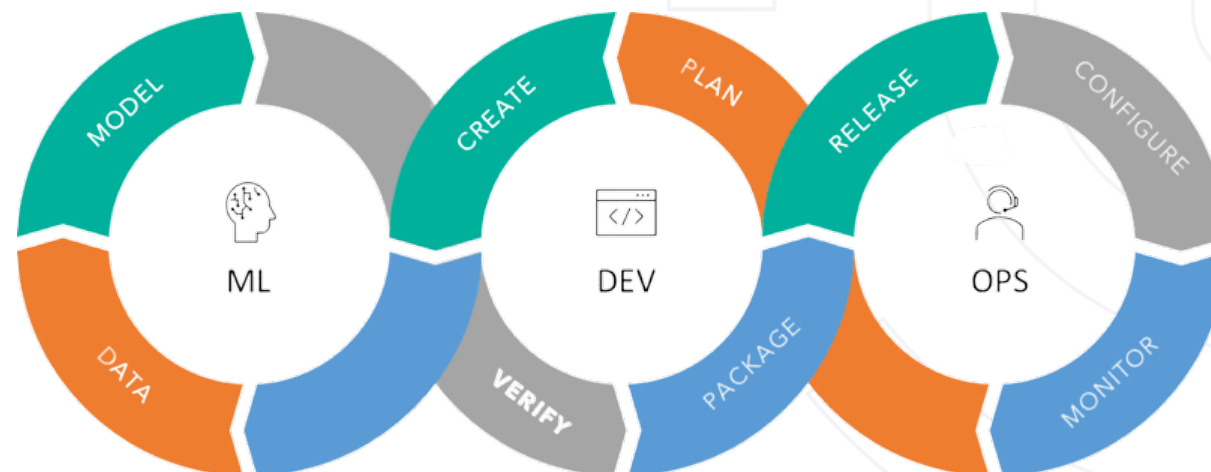
Models learn constantly



# Enters MLOps

## Machine Learning Operations (MLOps)

Practice that aims to make developing and maintenance of machine learning systems in production smooth and efficient, augmenting their long-term value while reducing the risk associated it with it.





# MLOps 4 pillars

**Collaborative**

**Reproducible**

**Scalable**

**Continuous**





# Collaborative



## Collaborative ML

- MLOps ensures that all steps in the ML system are transparent
- MLOPs eases collaboration through all the project life cycle

*Visible*



*Auditable*





# Reproducible



CartoonStock.com

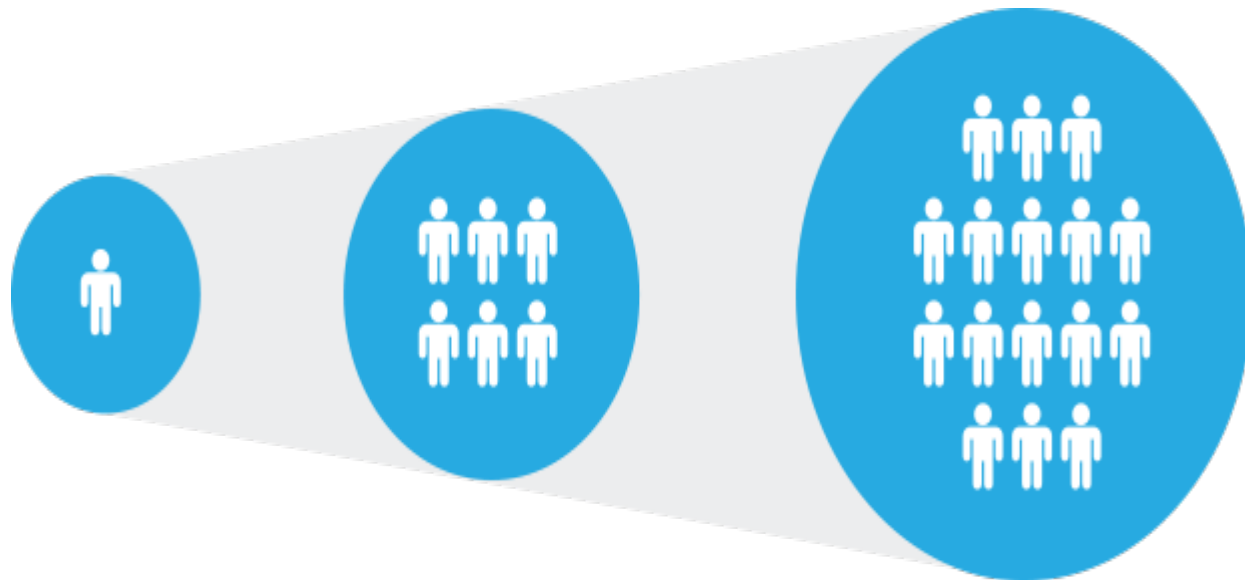
## Reproducible ML

- MLOps enforces storing of all artifacts
- Versioning more than code: data, models, meta data, etc.

***Process ≠ Experiment***



# Scalable



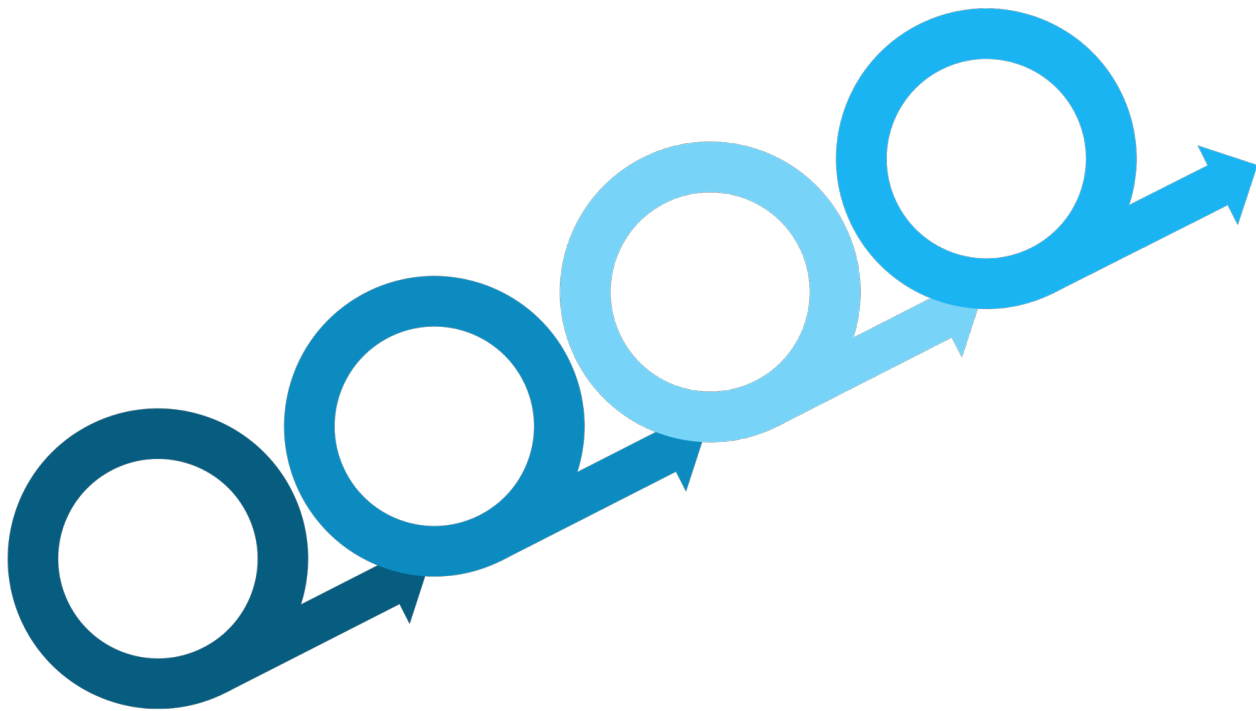
## Scalable ML

- MLOps eases expansion of infrastructure to scale projects
- Volumes of data can grow quickly, so the set up should grow naturally

*Natural growth*



# Continuous



## Continuous ML

- Ensure a CI/CD process is central to MLOps
- ML should be thought as a continuous process
- Retraining models should be effortless

*Ad-hoc < Automated*