

RESEARCH @



Open Research Webinars



AI and Machine Learning Automation with Activeeon: from Models to MLOps

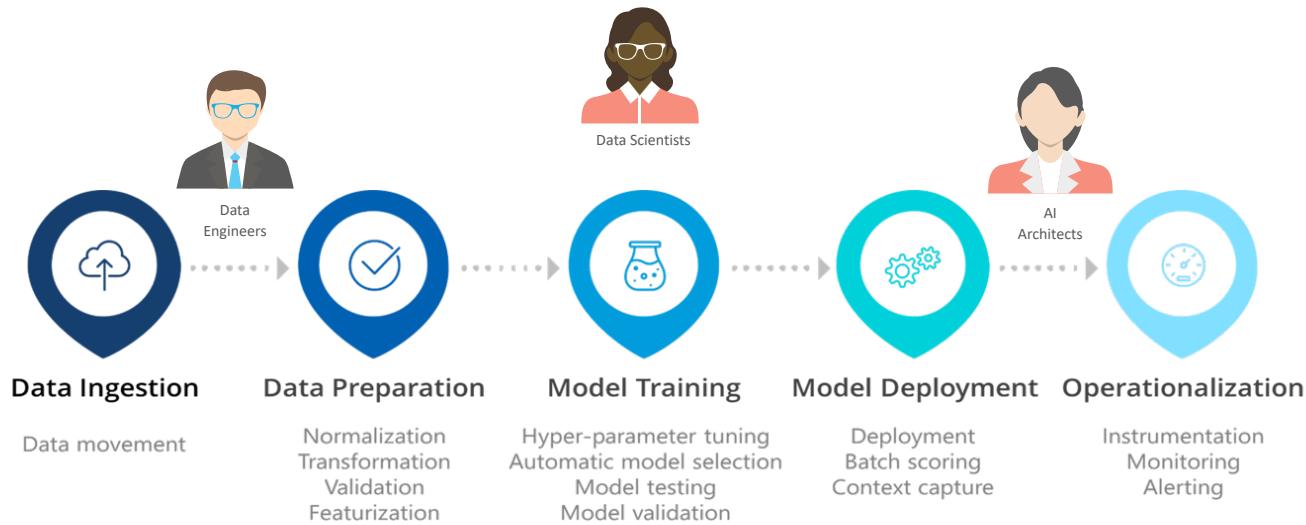
Andrews Cordolino Sobral, AI Solution Architect



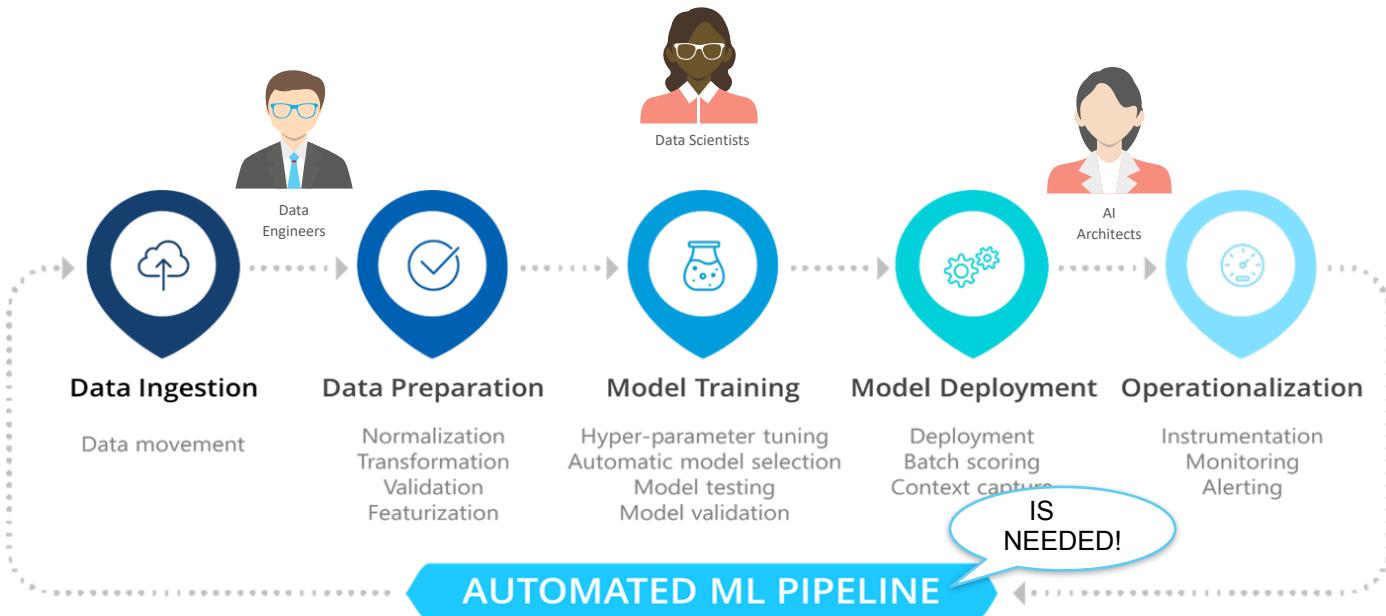
Agenda

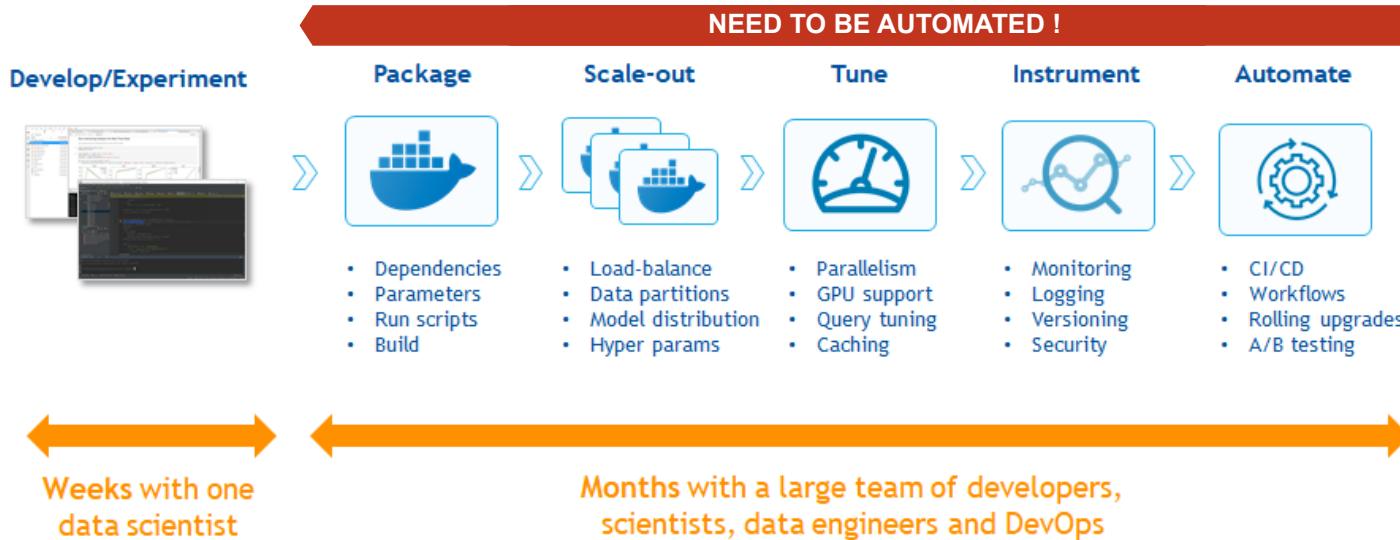
- Typical data science project (life cycle, roadmap, challenges, etc)
- What an enterprise level data science project needs?
- The evolution of the MLOps solutions
- Proactive Machine Learning (PML) at the core of data science projects
- Bringing up your data science project maturity level with PML
- Data science project pipelines with PML
- MLOps stack with PML

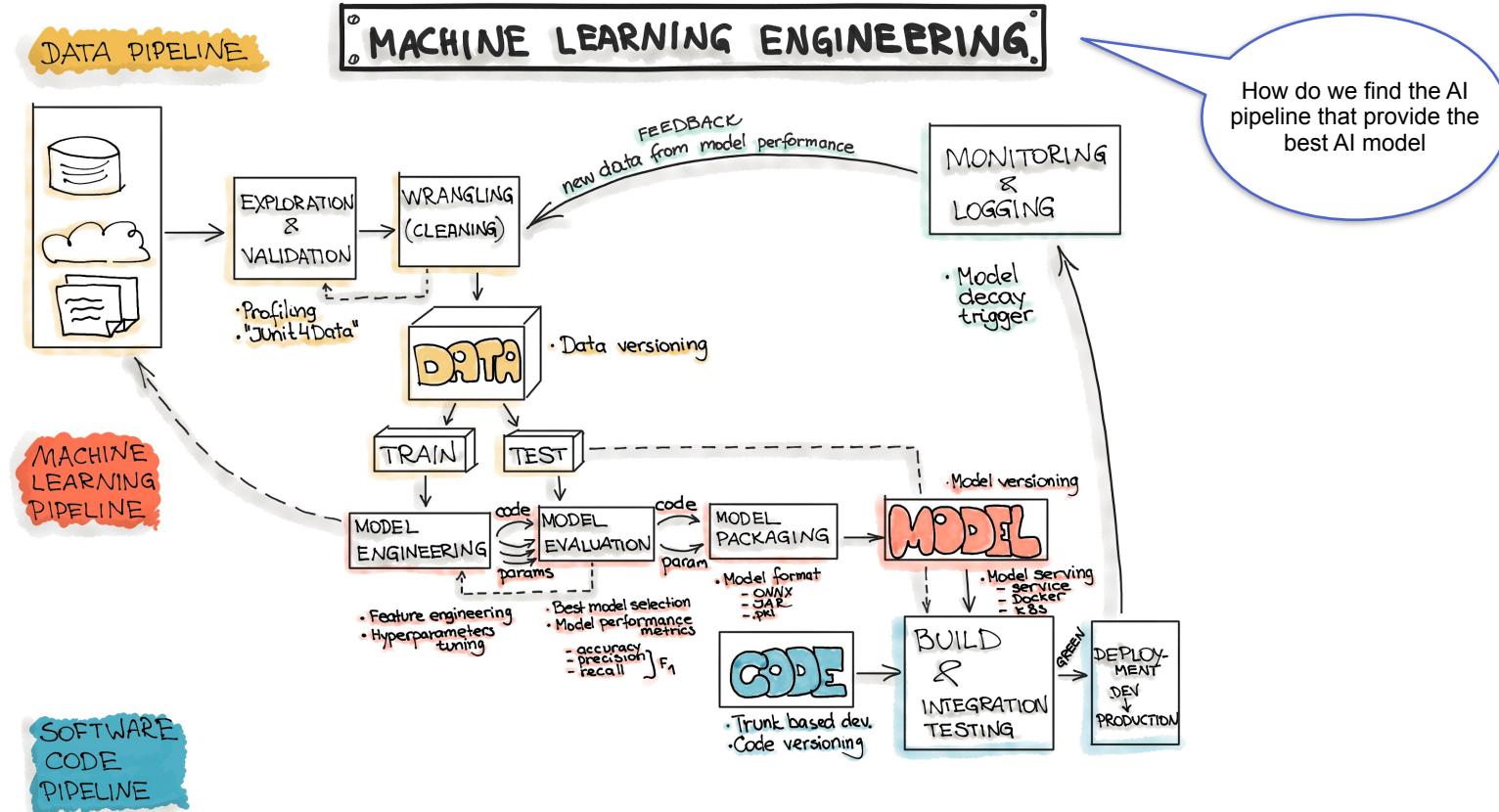
Data science project life cycle



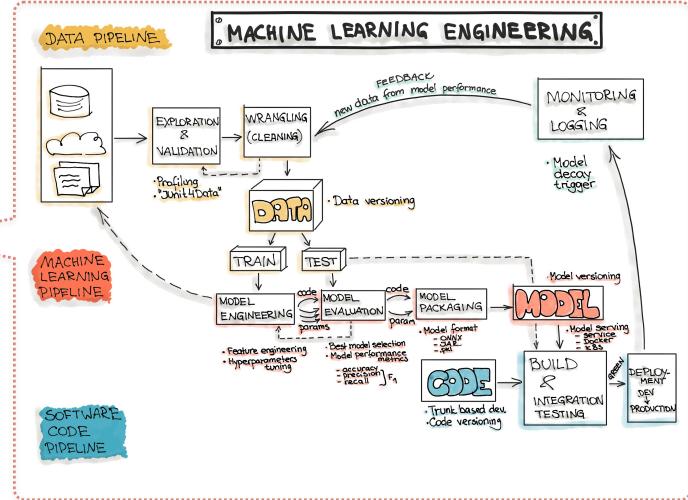
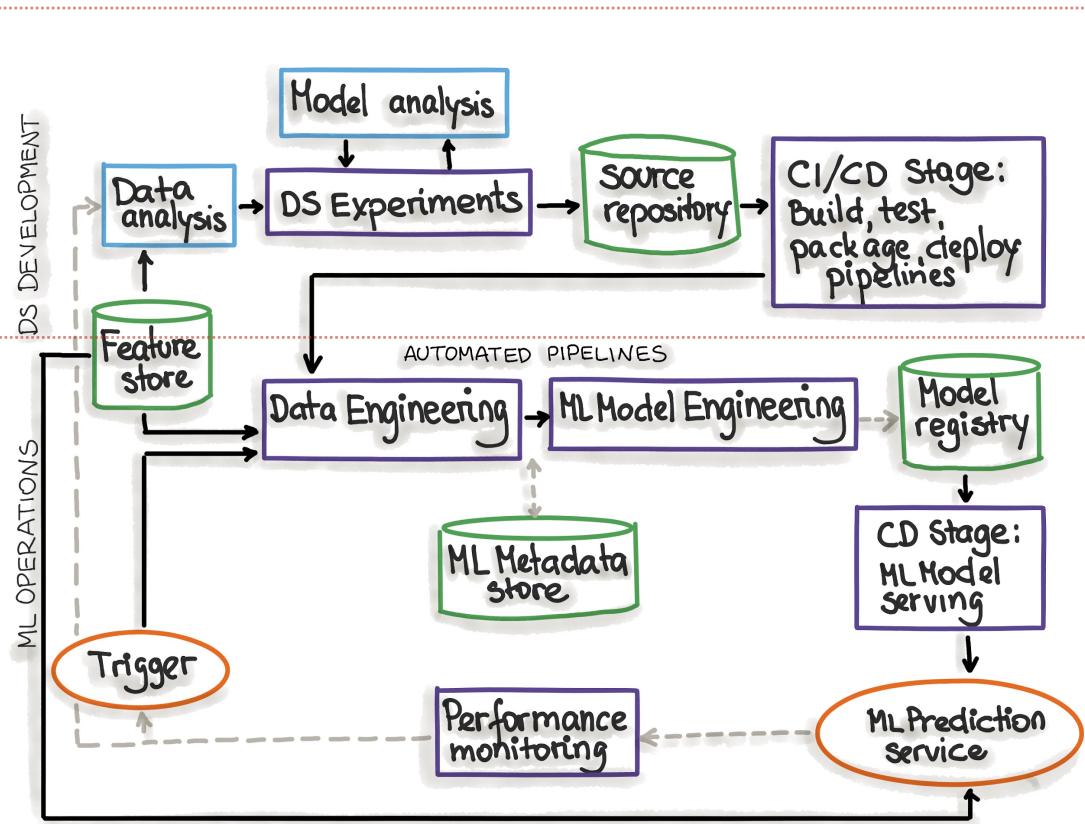
Data science project life cycle







Data science is...complex!



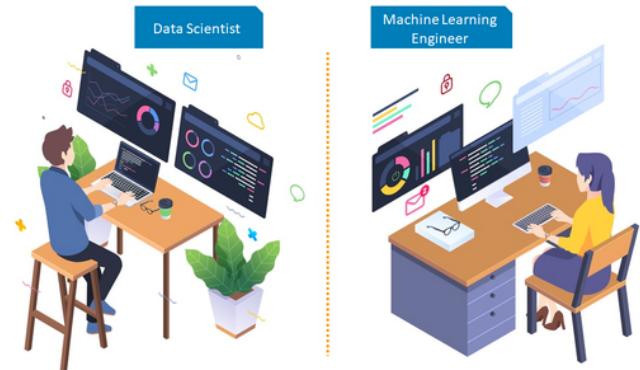
Data science is...complex!



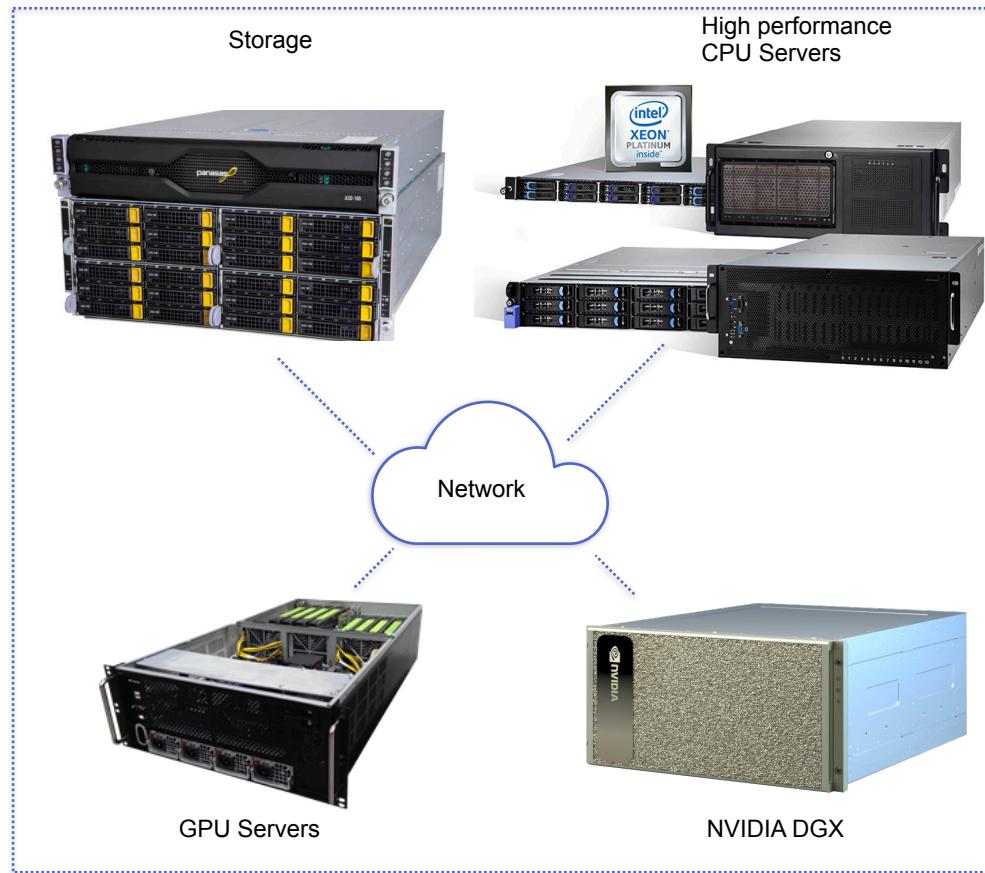
How do we bridge gaps between teams?



Data science is...complex!

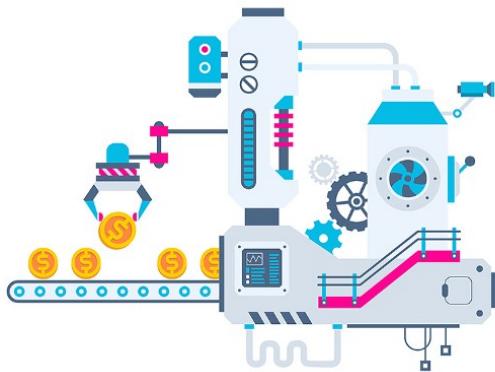


Hardware
Abstraction
Layer
is needed!

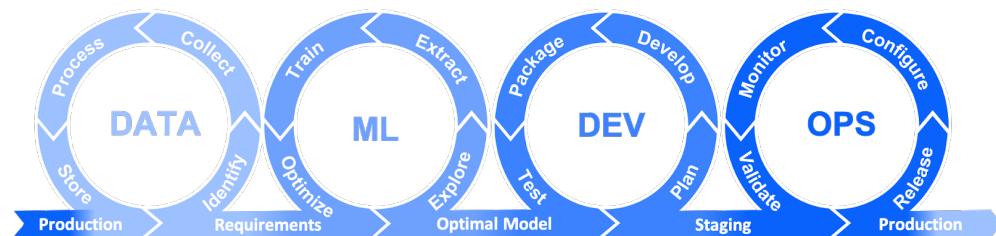


Data science needs...

A Modern Automation Platform

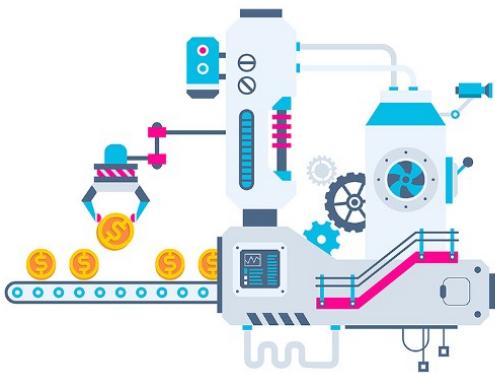


A Modern Methodology

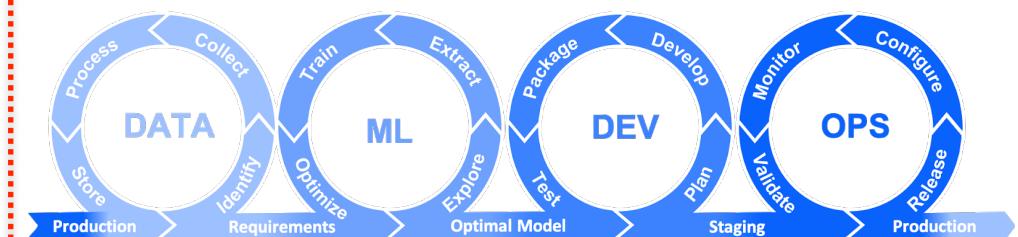


Data science needs...

A Modern Automation Platform

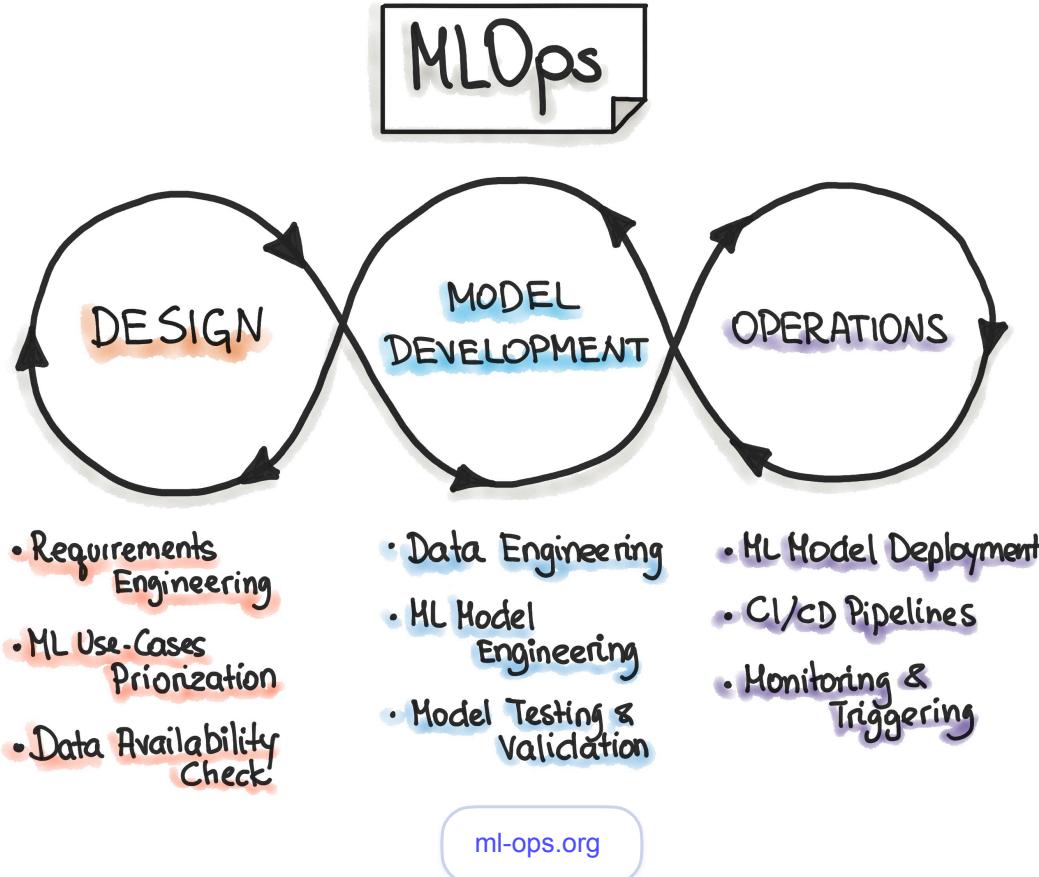


A Modern Methodology

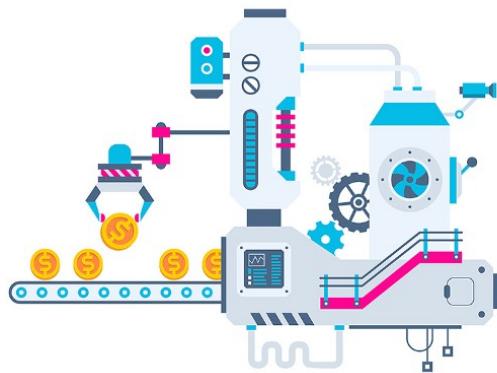


MLOps is the DevOps for data science projects

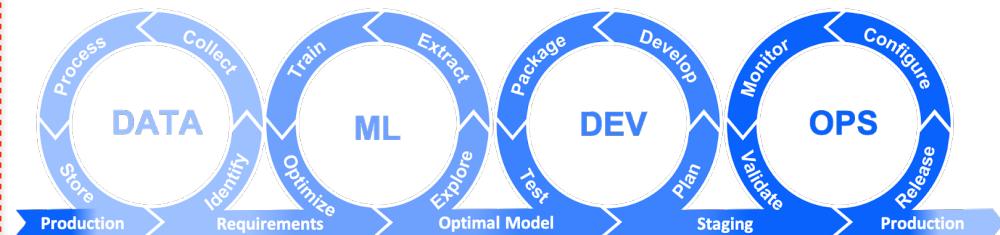
- MLOps aims to unify the release cycle for machine learning and software application release.
- MLOps enables automated testing of machine learning artifacts (e.g. data validation, ML model testing, and ML model integration testing)
- MLOps enables the application of agile principles to machine learning projects.
- MLOps enables supporting machine learning models and datasets to build these models as first-class citizens within CI/CD systems.
- MLOps reduces technical debt across machine learning models.
- MLOps must be a language-, framework-, platform-, and infrastructure-agnostic practice.

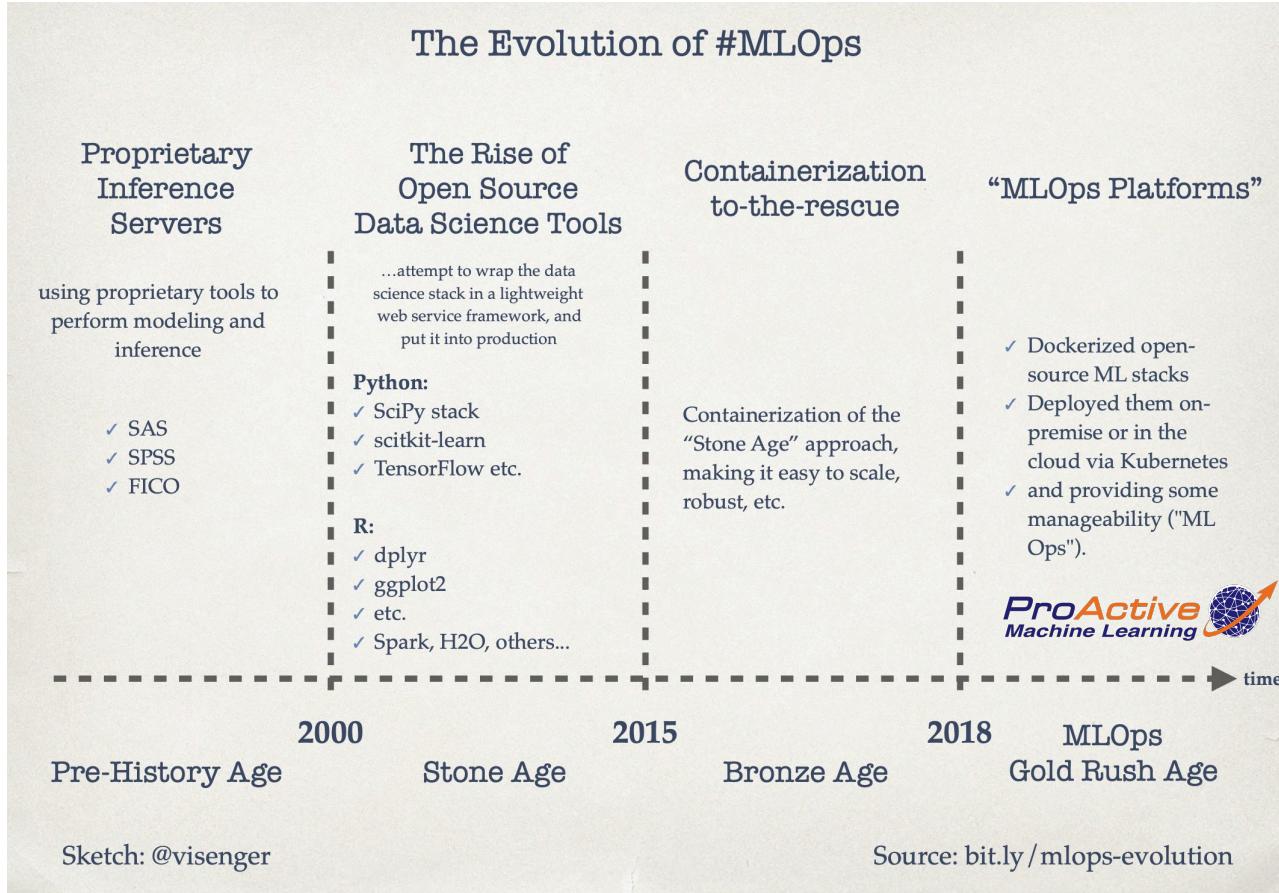


A Modern Automation Platform

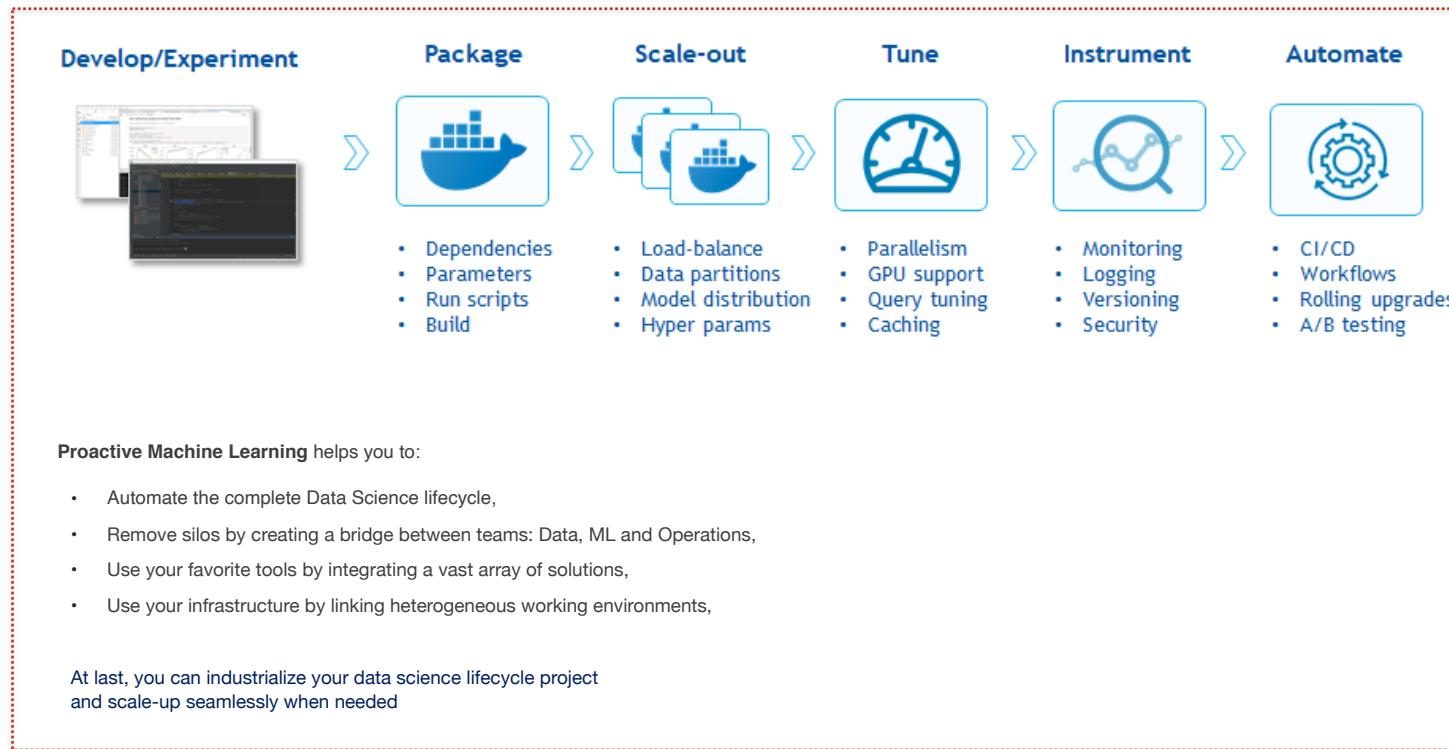


A Modern Methodology





ProActive Machine Learning





Stage 1 Manual process. This is a typical data science process, which is performed at the beginning of implementing ML. This level has an experimental and iterative nature. Every step in each pipeline, such as data preparation and validation, model training and testing, are executed manually. The common way to process is to use Rapid Application Development (RAD) tools, such as Jupyter Notebooks.

More than **80%** of the data science projects stays here!



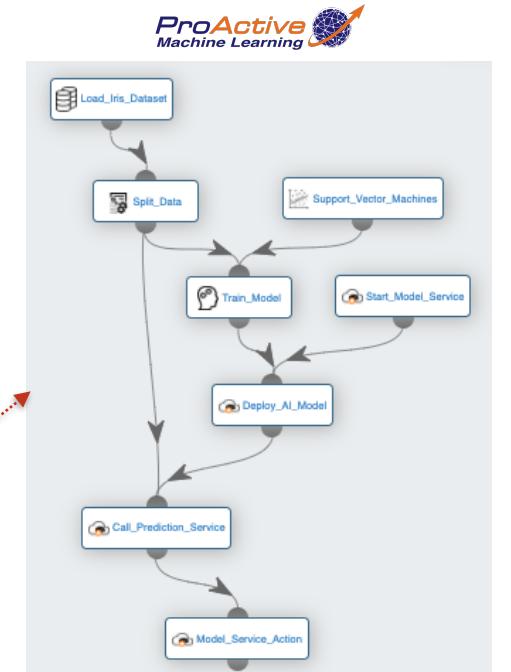
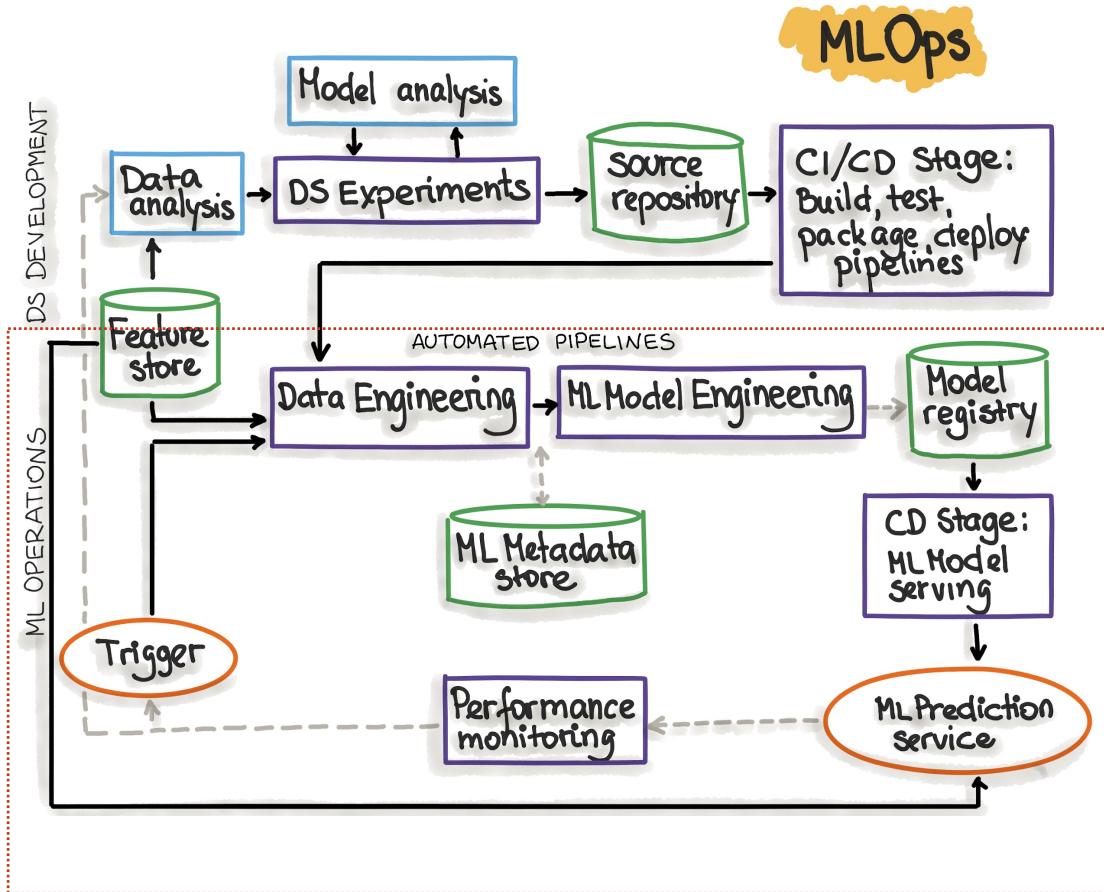
Stage 2 ML pipeline automation. The next level includes the execution of model training automatically. We introduce here the continuous training of the model. Whenever new data is available, the process of model retraining is triggered. This level of automation also includes data and model validation steps.

Less than **20%** of the data science projects comes here!

Stage 3 - CI/CD pipeline automation. In the final stage, we introduce a CI/CD system to perform fast and reliable ML model deployments in production. The core difference from the previous step is that we now automatically build, test, and deploy the Data, ML Model, and the ML training pipeline components.

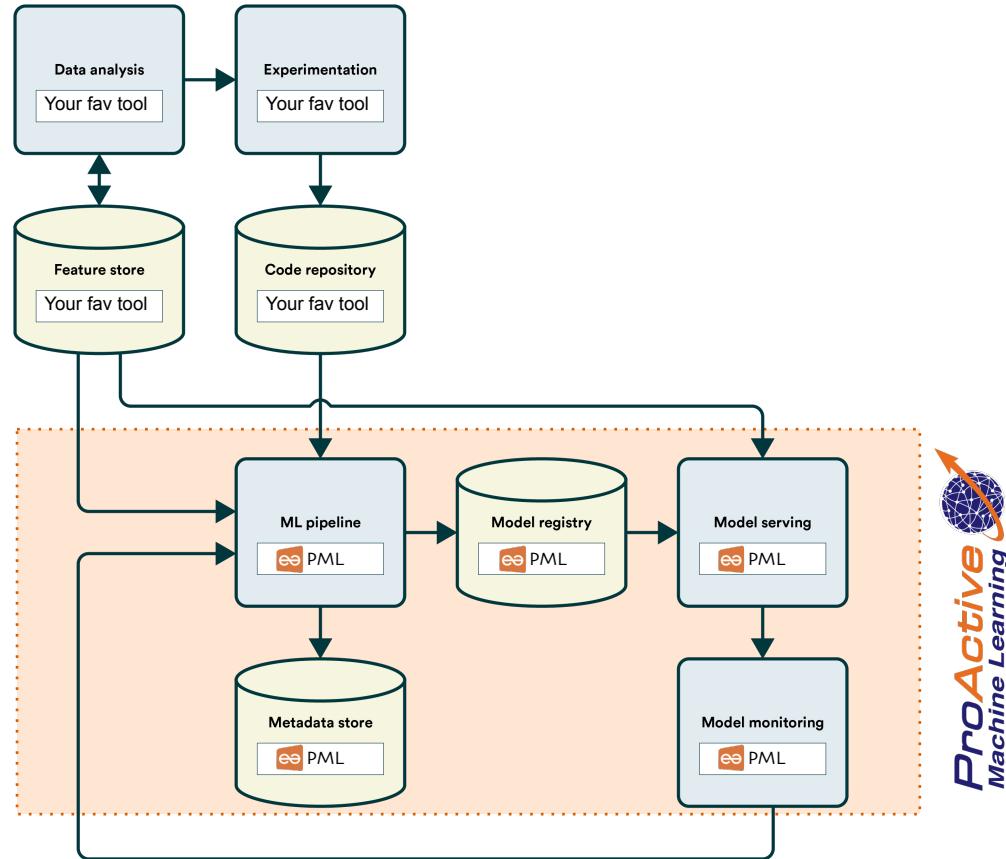
Less than **5%** of the data science projects comes here!

Data science project pipeline



Automated pipeline for model deployment

The MLOps stack with PML



Thank you for watching

ProActive Machine Learning (PML)

Accelerate the development and deployment of AI models with scalability

www.activeeon.com/products/proactive-machine-learning

RESEARCH @

ECLIPSE
FOUNDATION

Open Research Webinars

OW2