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

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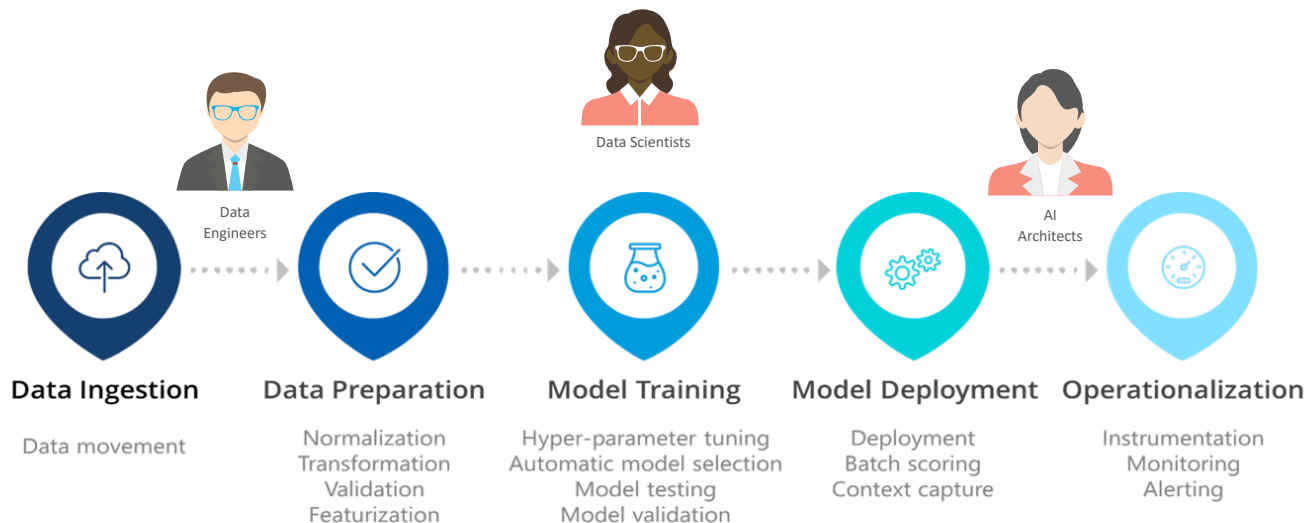


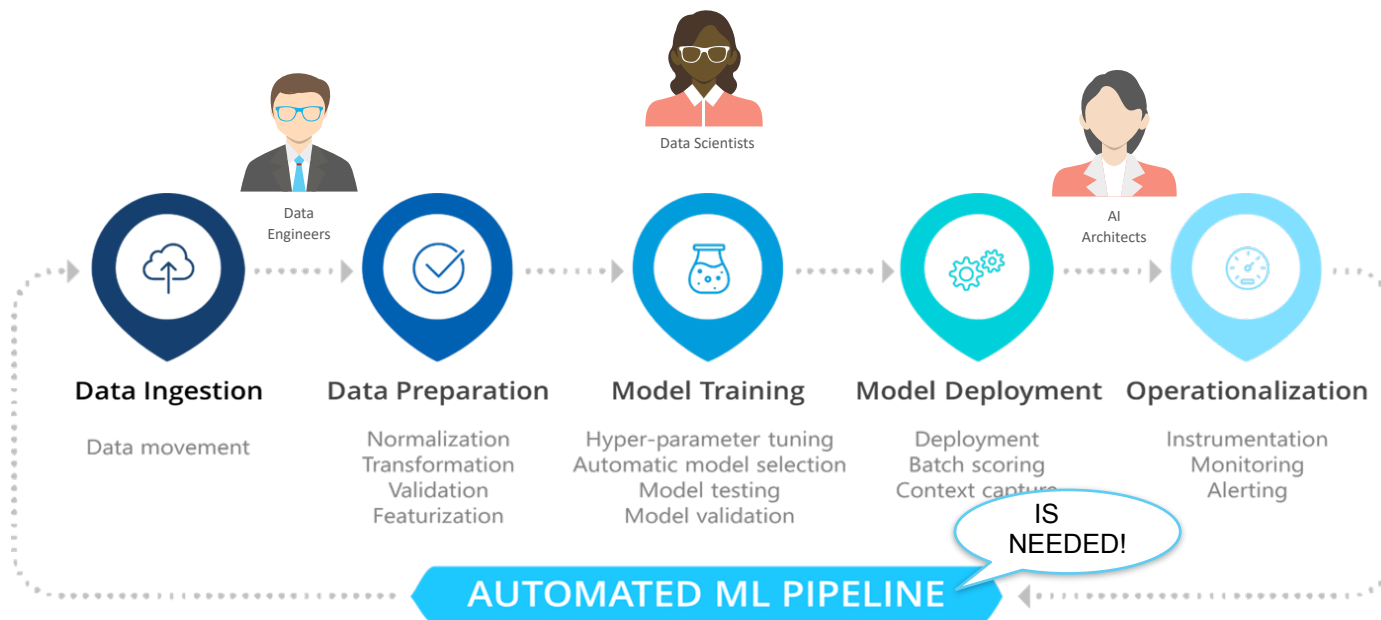
Activeeon
SCALE BEYOND LIMITS

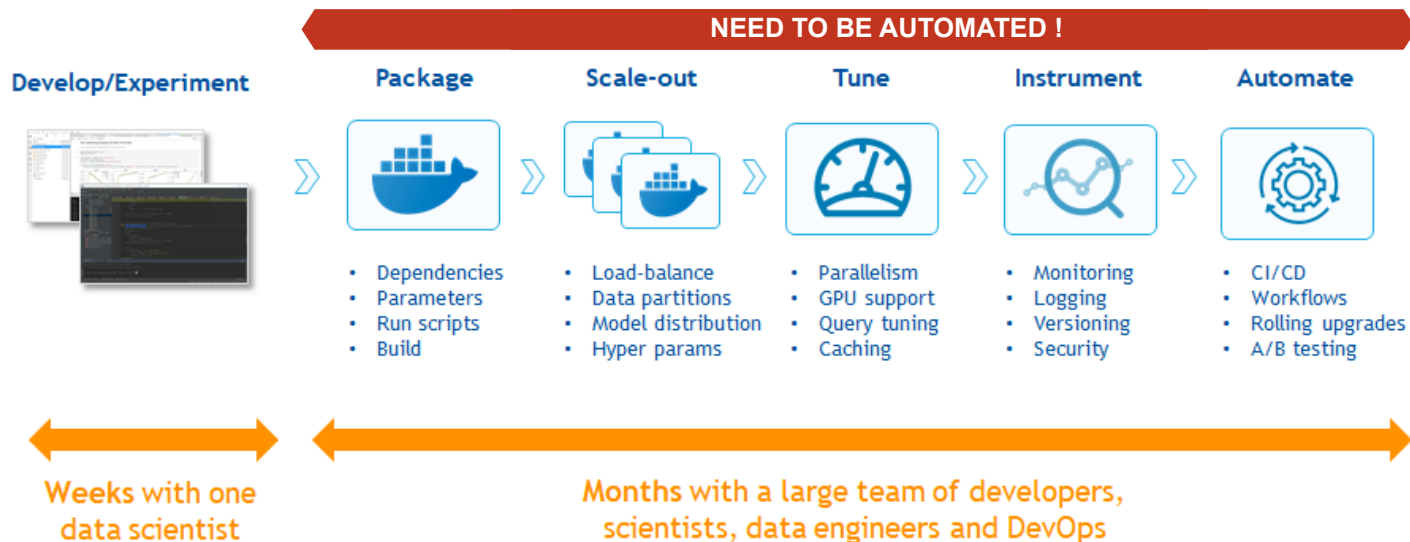
Agenda

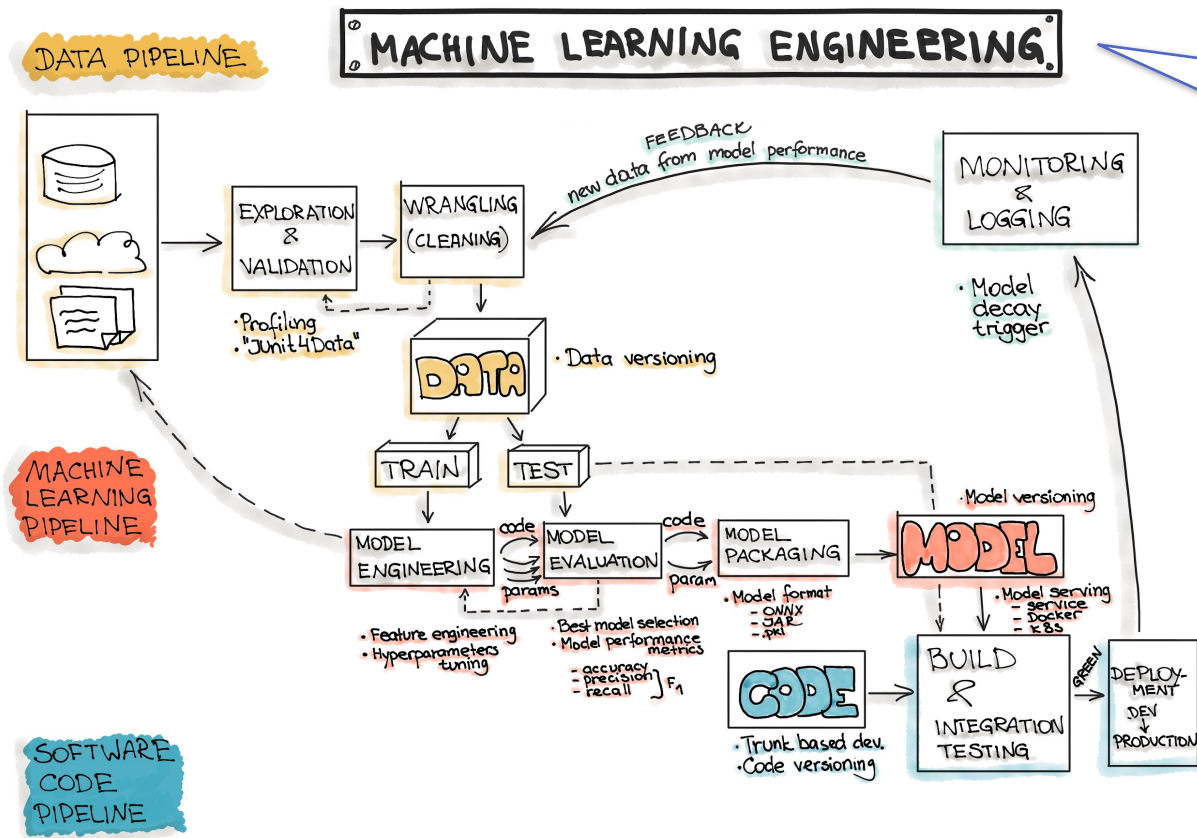
- Typical data science project (life cycle, roadmap, challenges, etc)
- What an enterprise level data science project needs?
- The evolution of the MLOps solutions
- Proactive Machine Learning (PML) at the core of data science projects
- Bringing up your data science project maturity level with PML
-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

Data science project life cycle

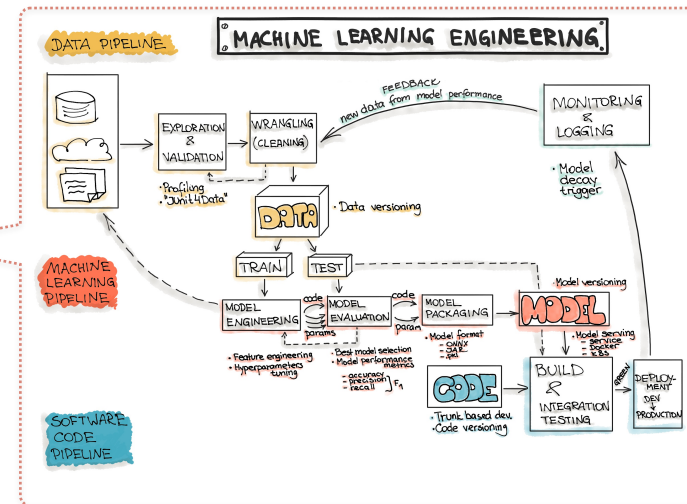
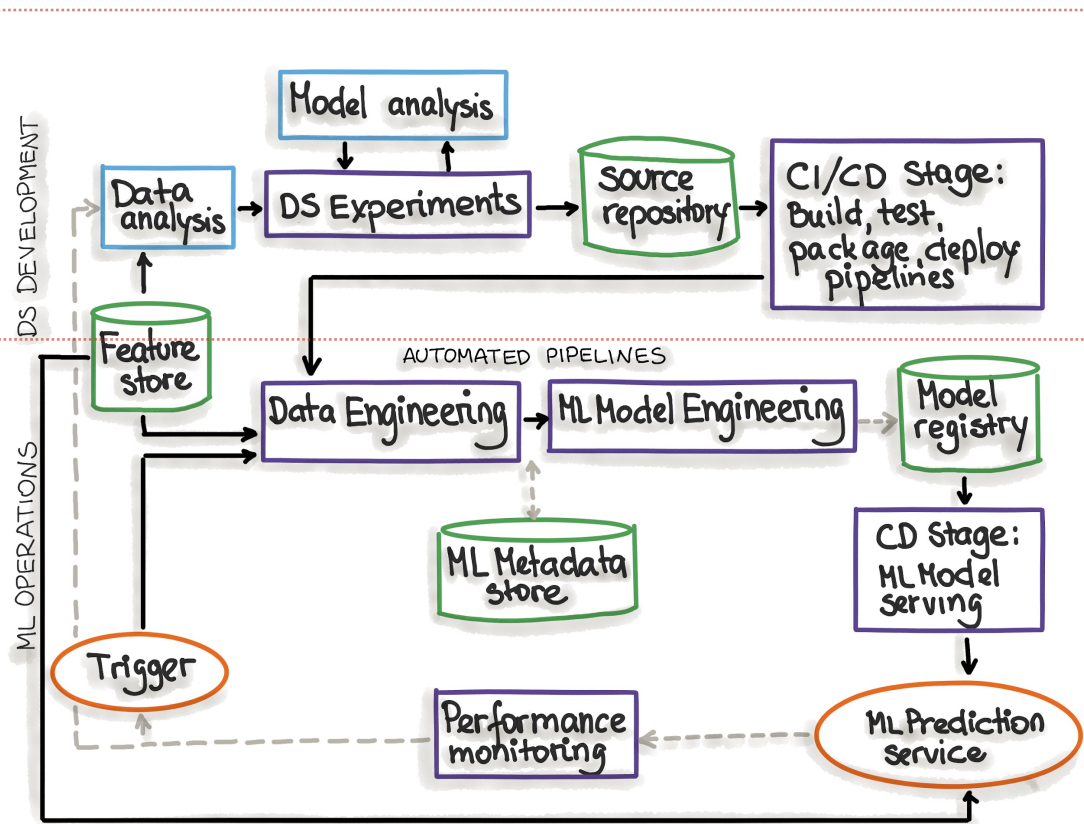




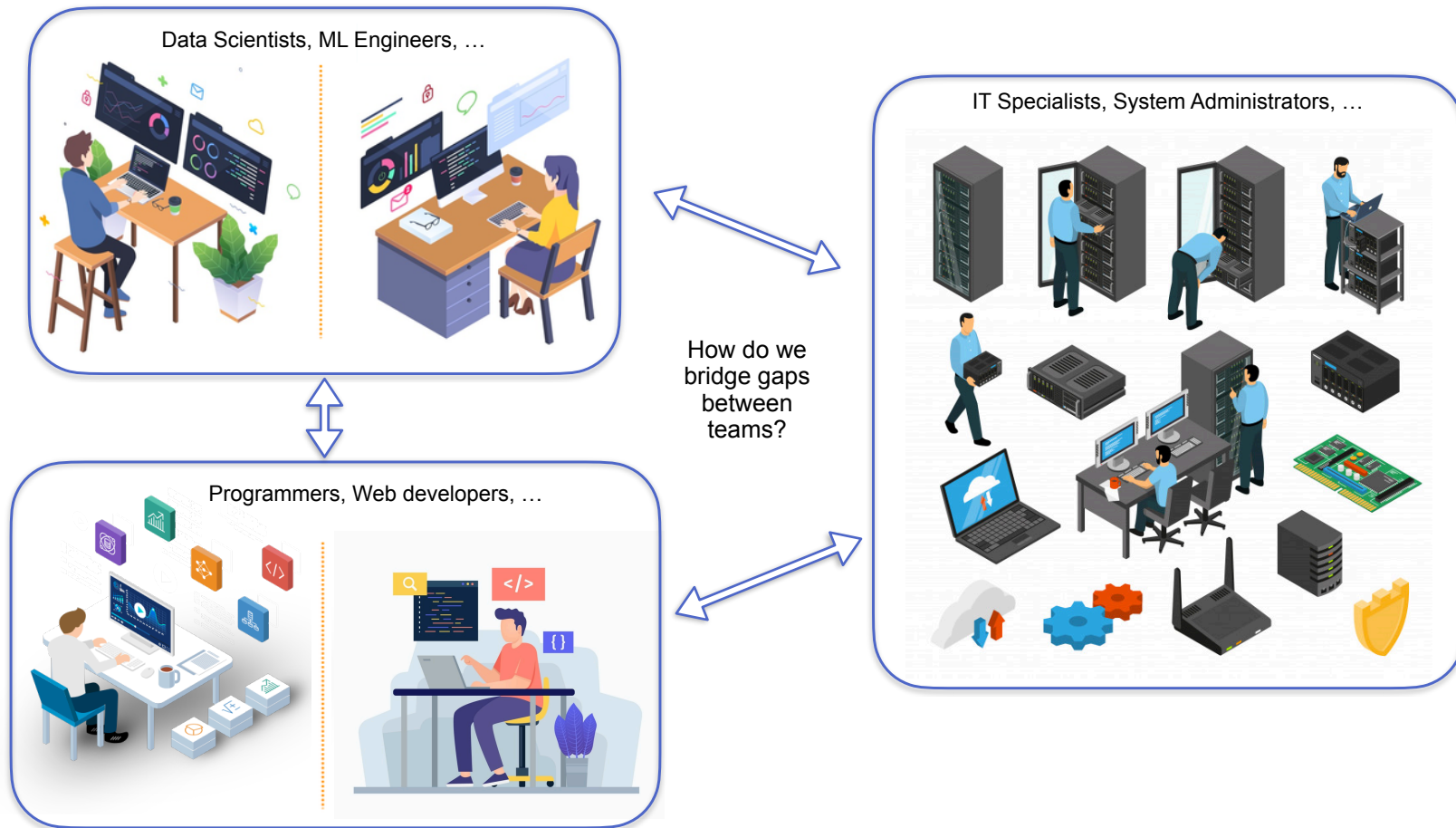




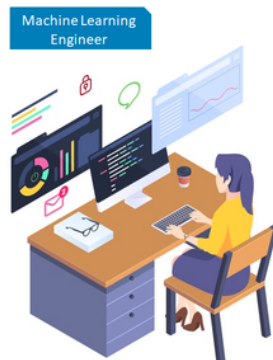
Data science is...complex!



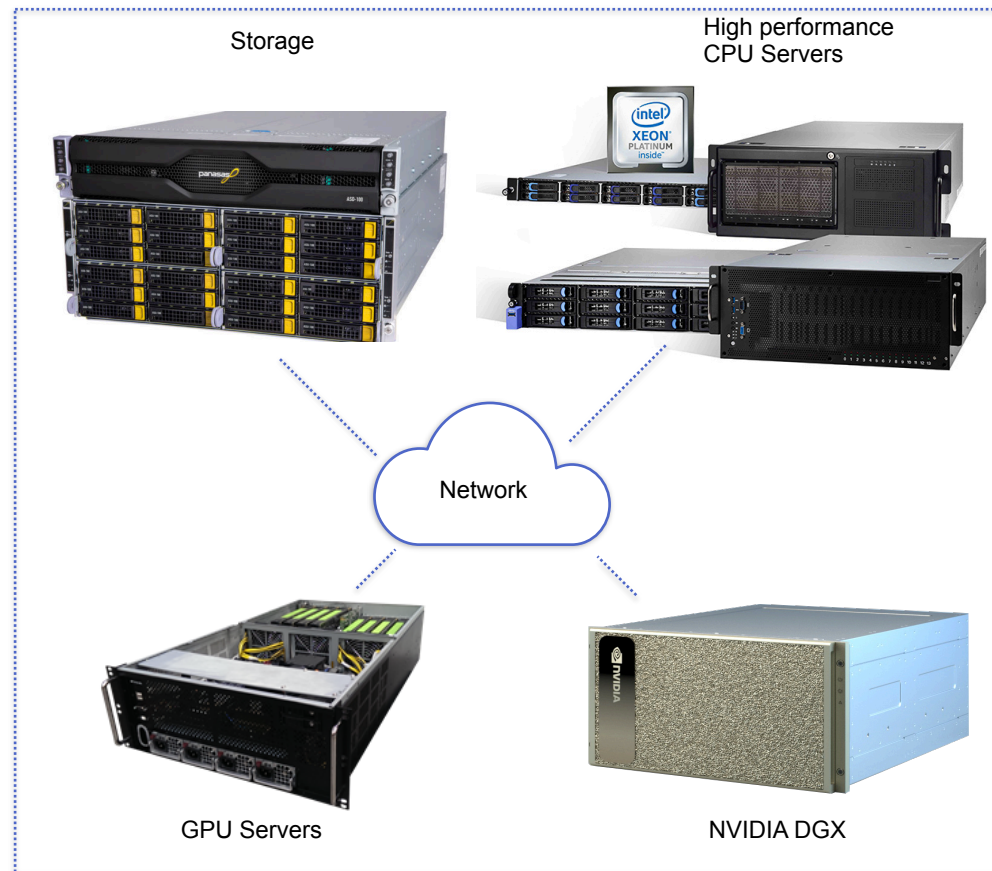
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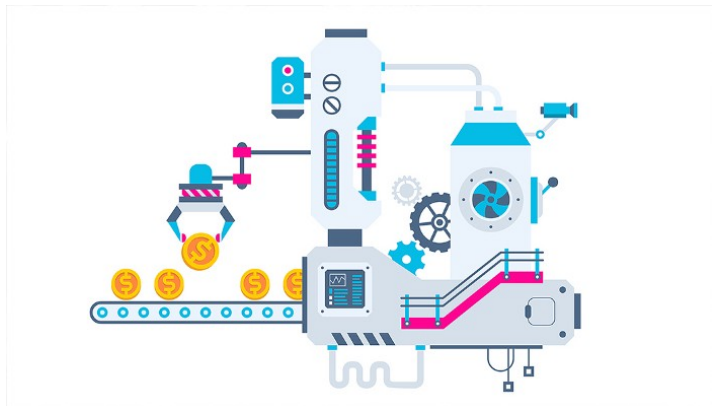
Data science is...complex!



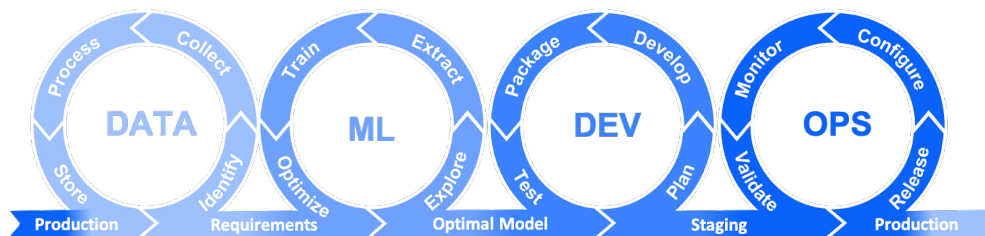
Hardware
Abstraction
Layer
is needed!



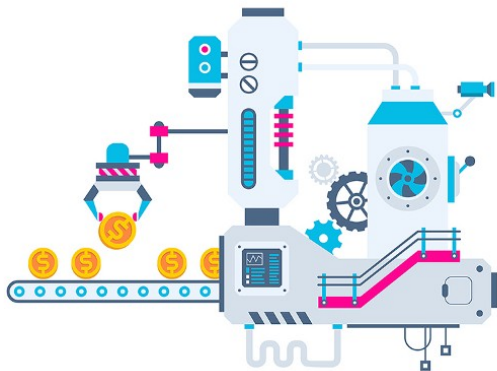
A Modern Automation Platform



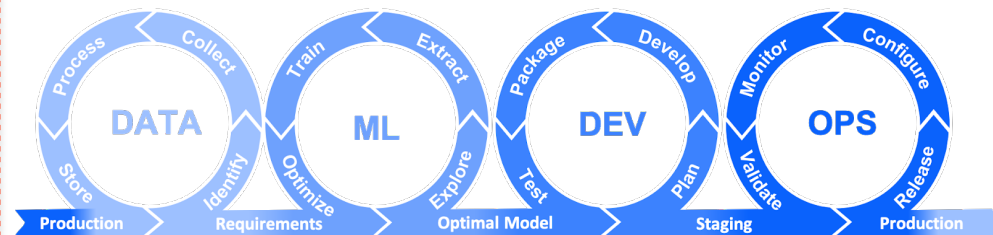
A Modern Methodology



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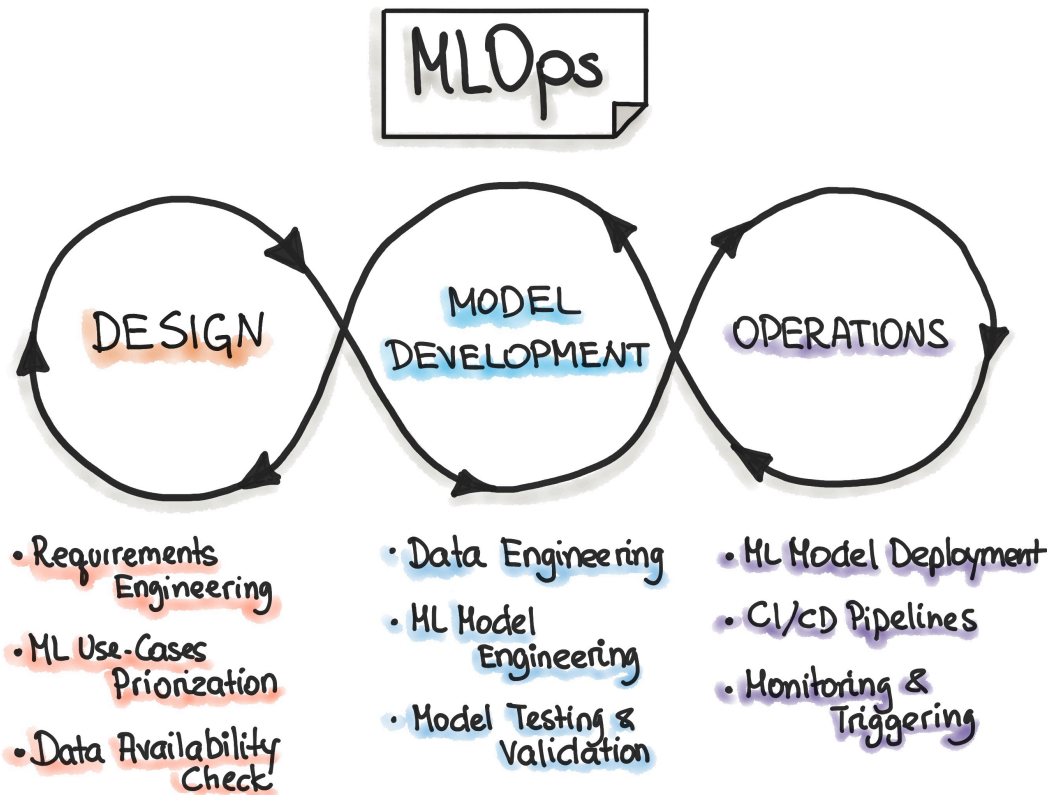


A Modern Methodology

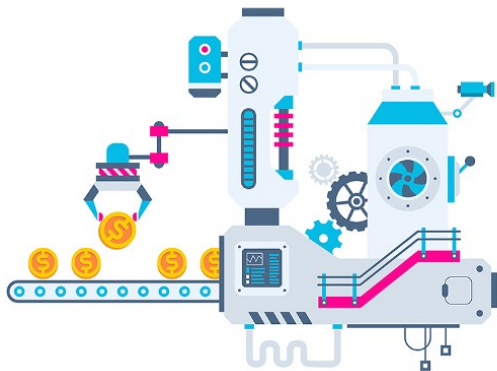


MLOps is the DevOps for data science projects

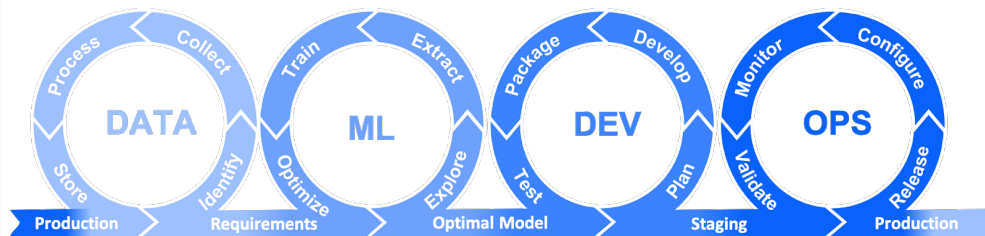
- **MLOps** aims to unify the release cycle for machine learning and software application release.
- **MLOps** enables automated testing of machine learning artifacts (e.g. data validation, ML model testing, and ML model integration testing)
- **MLOps** enables the application of agile principles to 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.

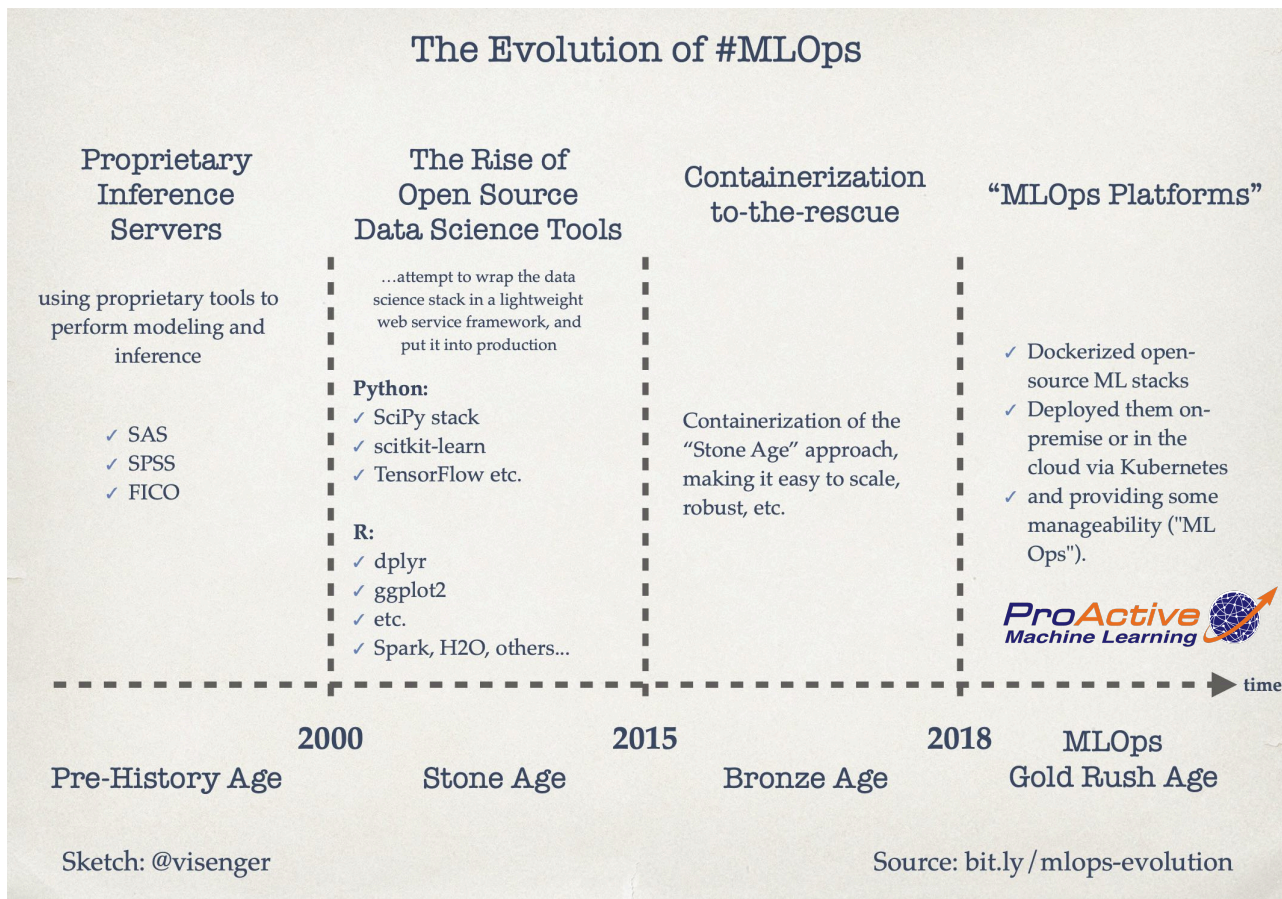


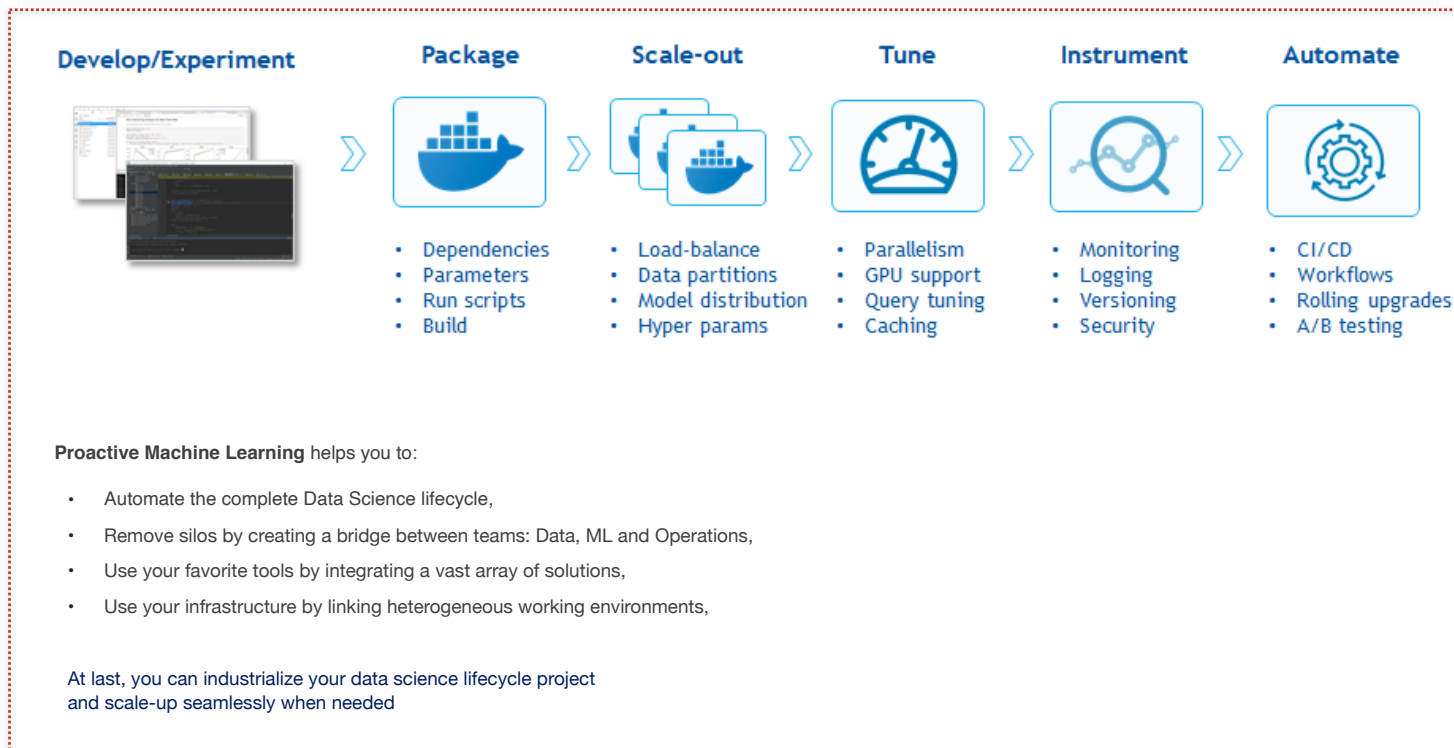
A Modern Automation Platform



A Modern Methodology









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!



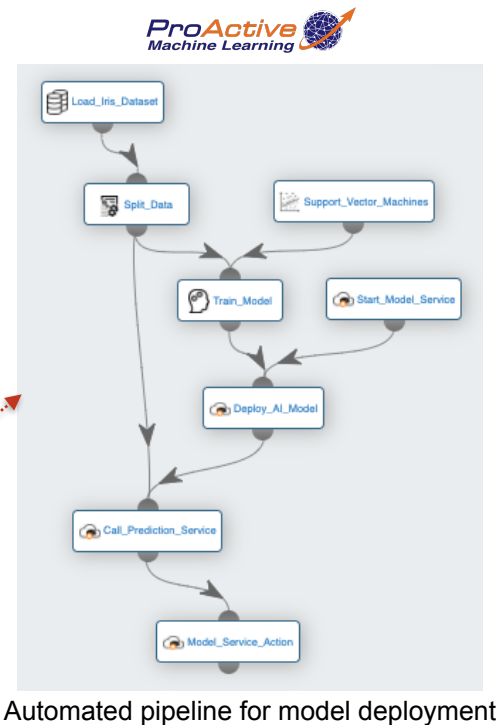
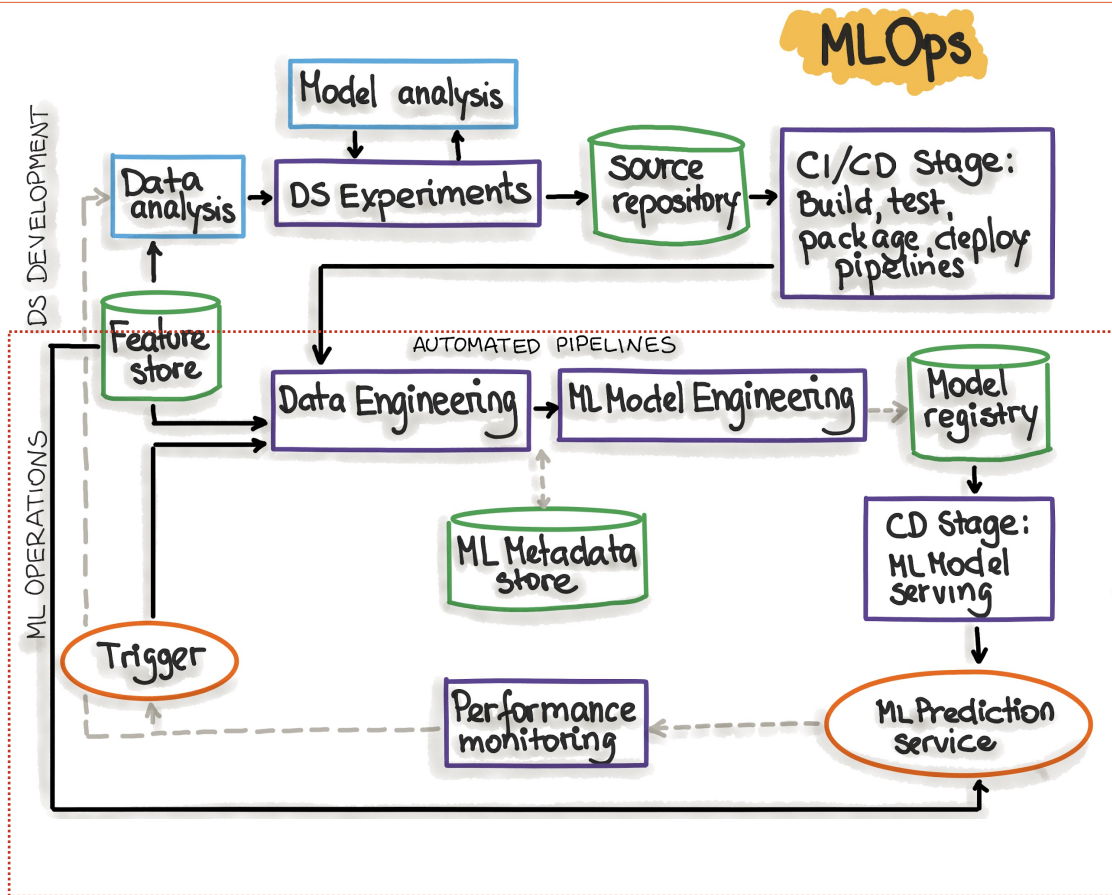
Stage 2 ML pipeline automation. The next level includes the execution of model training automatically. We introduce here the continuous training of the model. Whenever new data is available, the process of model retraining is triggered. This level of automation also includes data and model validation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

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

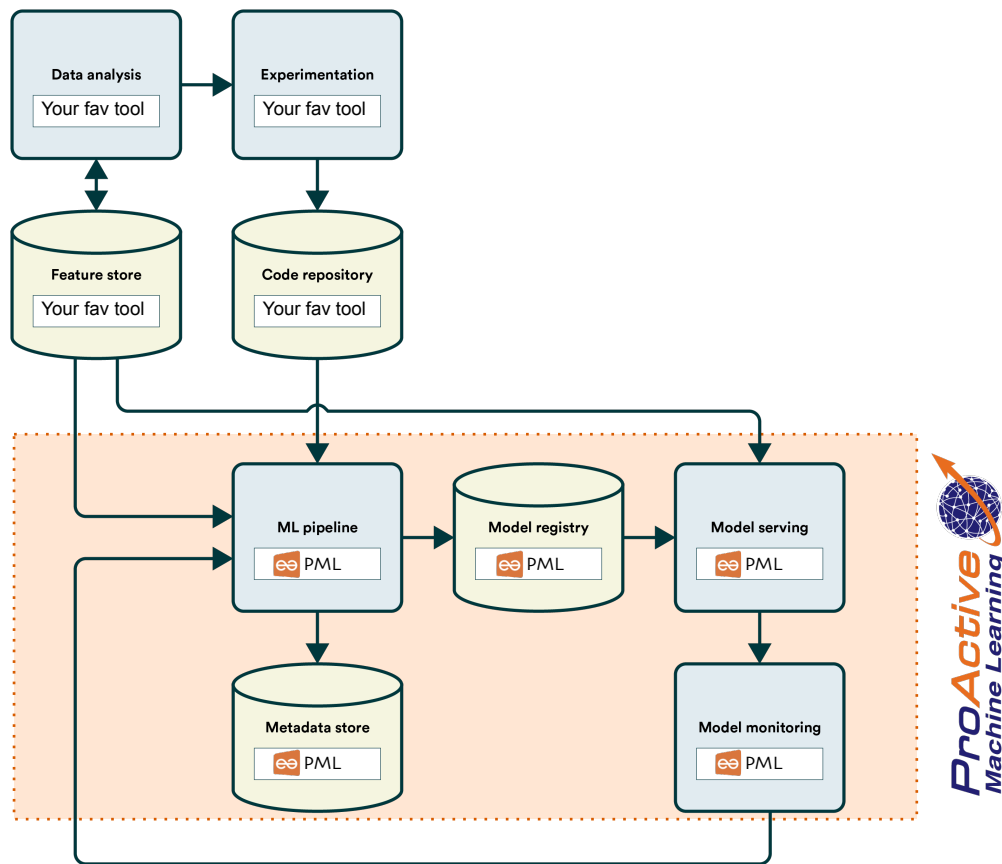


Stage 3 - CI/CD pipeline automation. In the final stage, we introduce a CI/CD system to perform fast and reliable ML model deployments in production. The core difference from the previous step is that we now automatically build, test, and deploy the Data, ML Model, and the ML training pipeline components.

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



The MLOps stack with PML



Thank you for watching

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