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Deploy and scale your machine learning models

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HOW TO DEPLOY AND SCALE YOUR ML/DL MODEL

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machine learning in Lab



Data scientist :

- Explore the data in local environments (a lab or even local machine with Jupyter notebooks).
- Experiment and test different approaches to understand the data (data and statistical analysis).
- "Train multiple" models on small datasets and compare them.
- Finetune the selected model and optimize it.
- Export the trained model to a model registry (parameters).

Assumptions:

- The business problem is well defined.
- The data is prepared and perfectly documented.
- The required work environment is prepared (labs, data access, tools configurations, etc).

DATA ENIGEERS



Data engineer :

- Collect data from various sources.
- Create data integration pipelines and ensure data quality.
- Store the data correctly in proper data storage services (Sql, NoSql).
- Maintain and improve databases continually.
- Work with data scientists to improve data pipelines.

machine learning in Prod & MLOPS

Machine Learning in prod :

Good production means :

- Model is reproducible.
- Code and models are version controlled.
- Continuous deployment.
- Collaborative (includes best practices).



Machine Learning in prod :



Kedro:

- "Kedro is a python framework for creating reproducible, maintainable and modular data science code by applying software engineering best practices to machine-learning code"
- It offers a UI to visualize the data pipeline.
- Integrates perfectly with existing tools like Jupyter, PySpark, etc.
- Kedro pipelines are easy to deploy as packages or containers:
 - As Airflow DAGs
 - With AWS batch
 - To Databricks, Kubeflow and many more

Machine Learning in prod :



Mlflow:

- Mlflow is an end-to-end machine learning lifecycle that allows to :
 - Track experiments and record/compare parameters
 - Packaging ML code in a reusable, reproducible form in order to share with other data scientists or transfer to production.
 - Managing and deploying models from a variety of ML libraries to a variety of model serving and inference platforms.
 - Model regitry , model versioning, stage transitions, and annotations.
 - Model monitoring.

