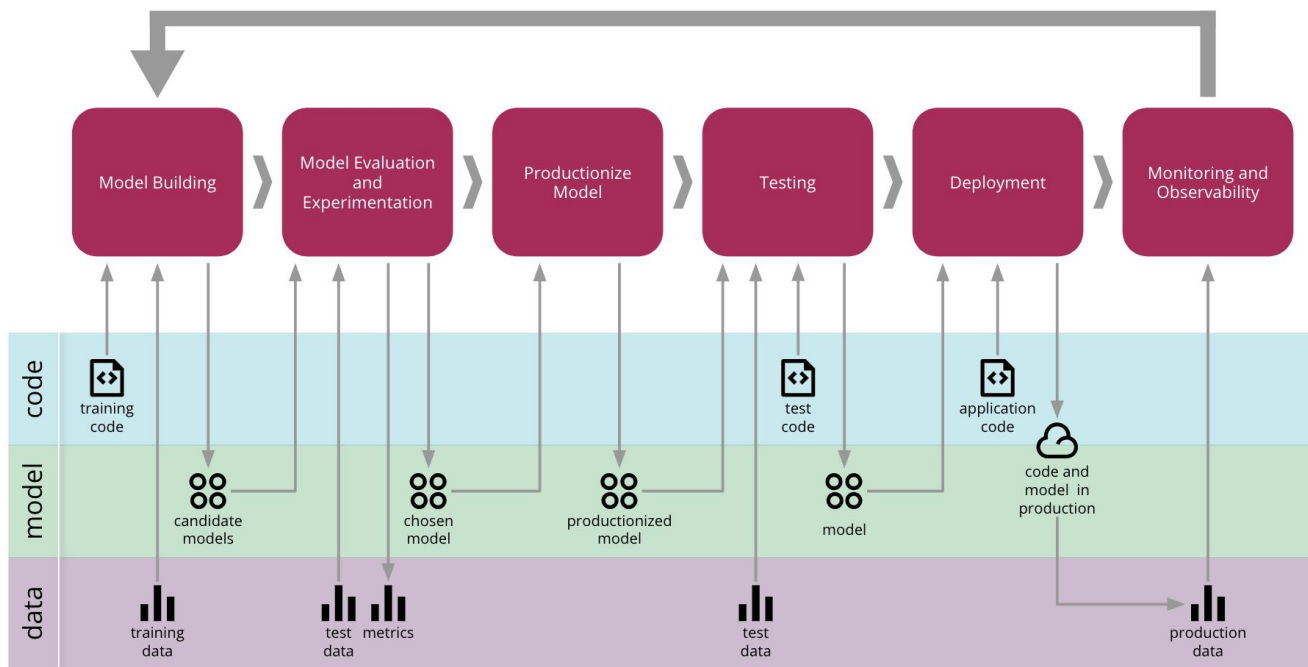


Managing ML processes in the organization

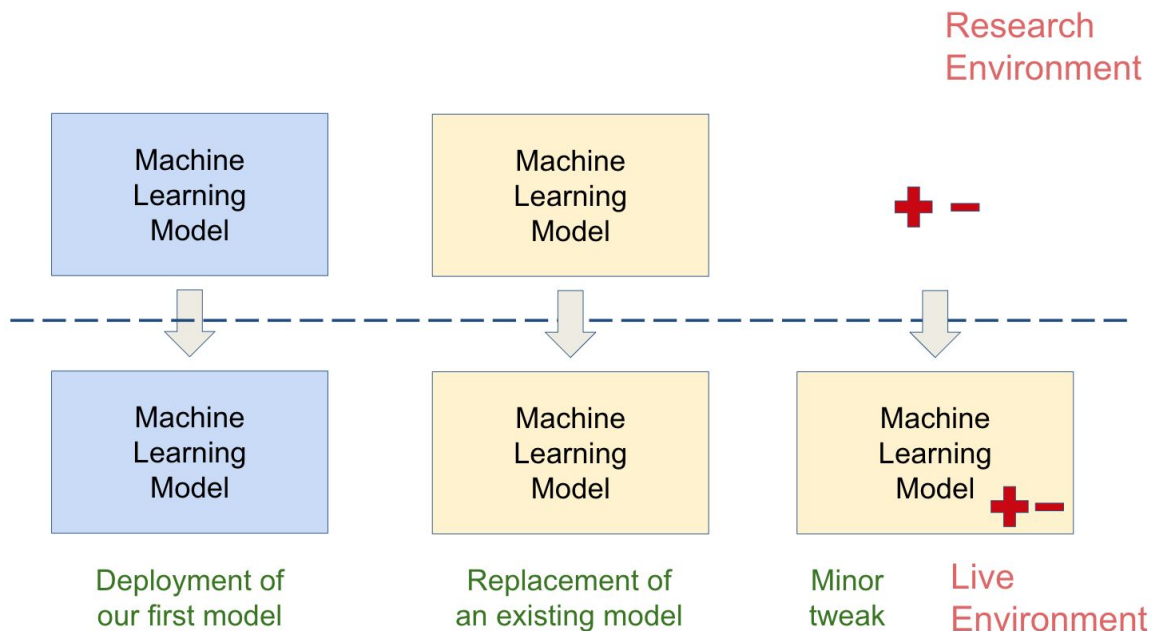
ML models life cycle

sources: <https://martinfowler.com/articles/cd4ml.html> and <https://christophergs.com/machine%20learning/2020/03/14/how-to-monitor-machine-learning-models/>



Maintainability

sources: <https://martinfowler.com/articles/cd4ml.html> and <https://christophergs.com/machine%20learning/2020/03/14/how-to-monitor-machine-learning-models/>

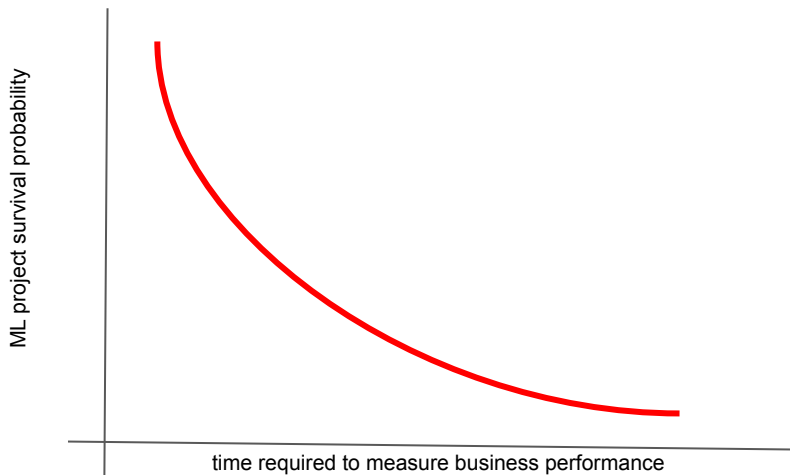


Model performance alerts

ML Metric (accuracy, RMSE, etc.)  Monitoring dashboards
Business metric (PNL, RoI, etc.)

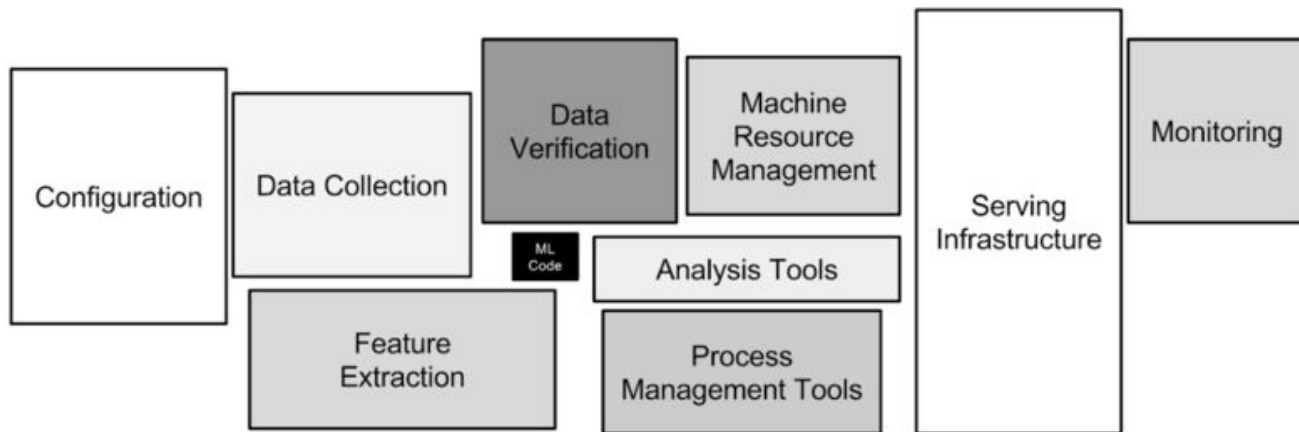
Performance measuring time vs. impact time

- What if I cannot measure a model has degraded business performance before the losses are too high?



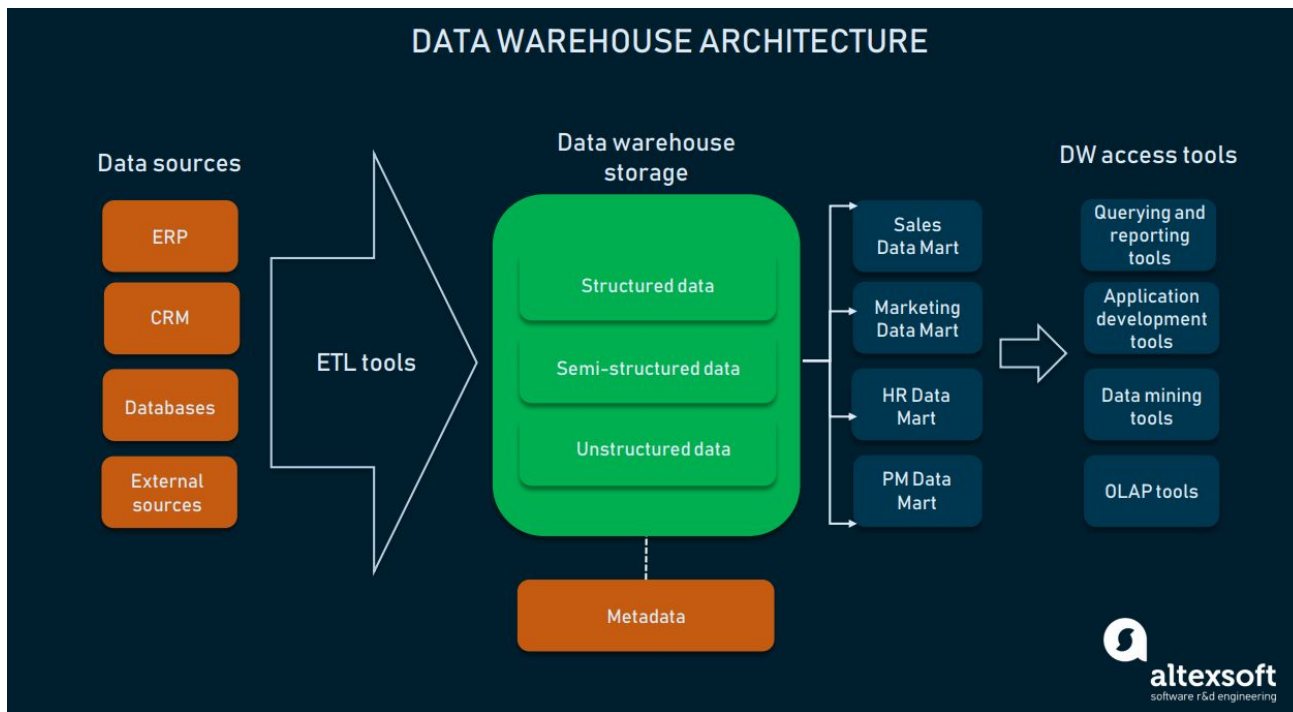
Pipeline maintenance

sources: <https://martinfowler.com/articles/cd4ml.html> and
<https://christophergs.com/machine%20learning/2020/03/14/how-to-monitor-machine-learning-models/>



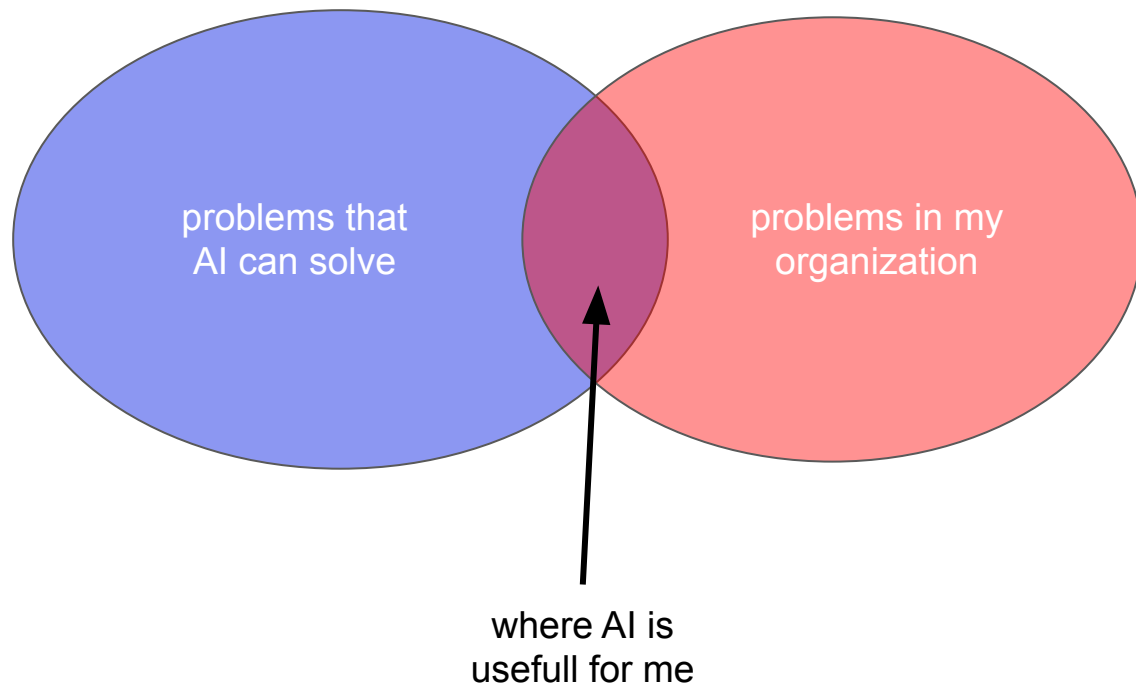
Data management

source: <https://www.altexsoft.com/blog/datascience/what-is-data-engineering-explaining-data-pipeline-data-warehouse-and-data-engineer-role/>



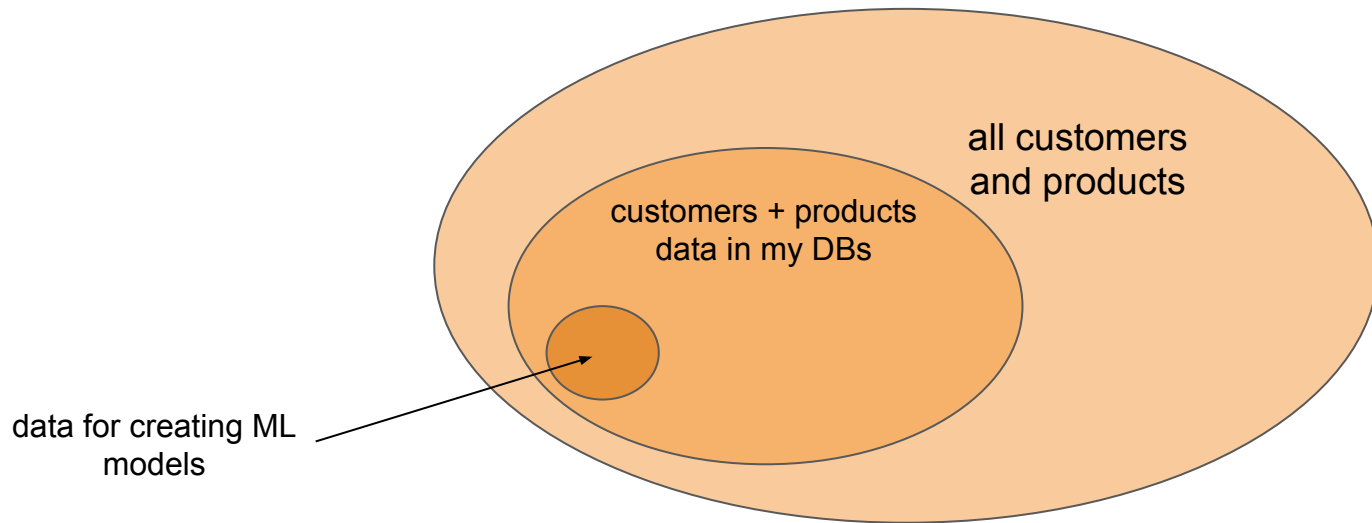
Is ML/AI really useful?

be smart selecting what ML really used for within my context



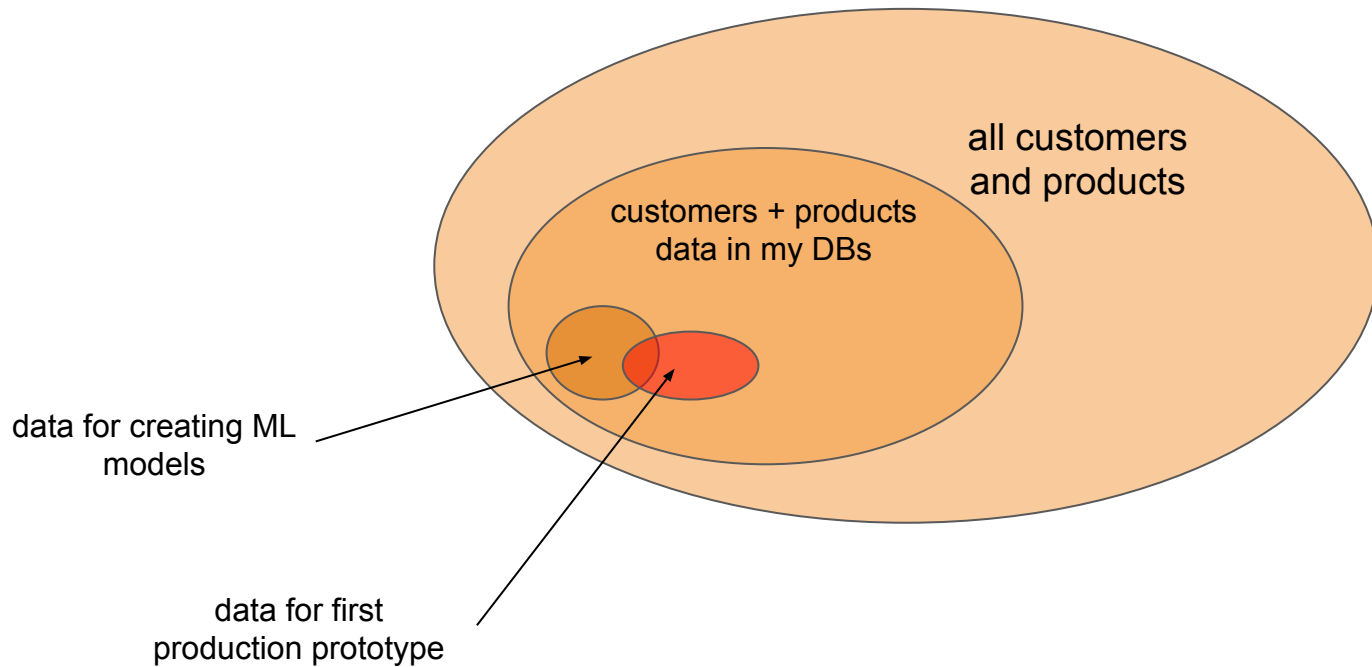
Data representativity

how **portable** are models?



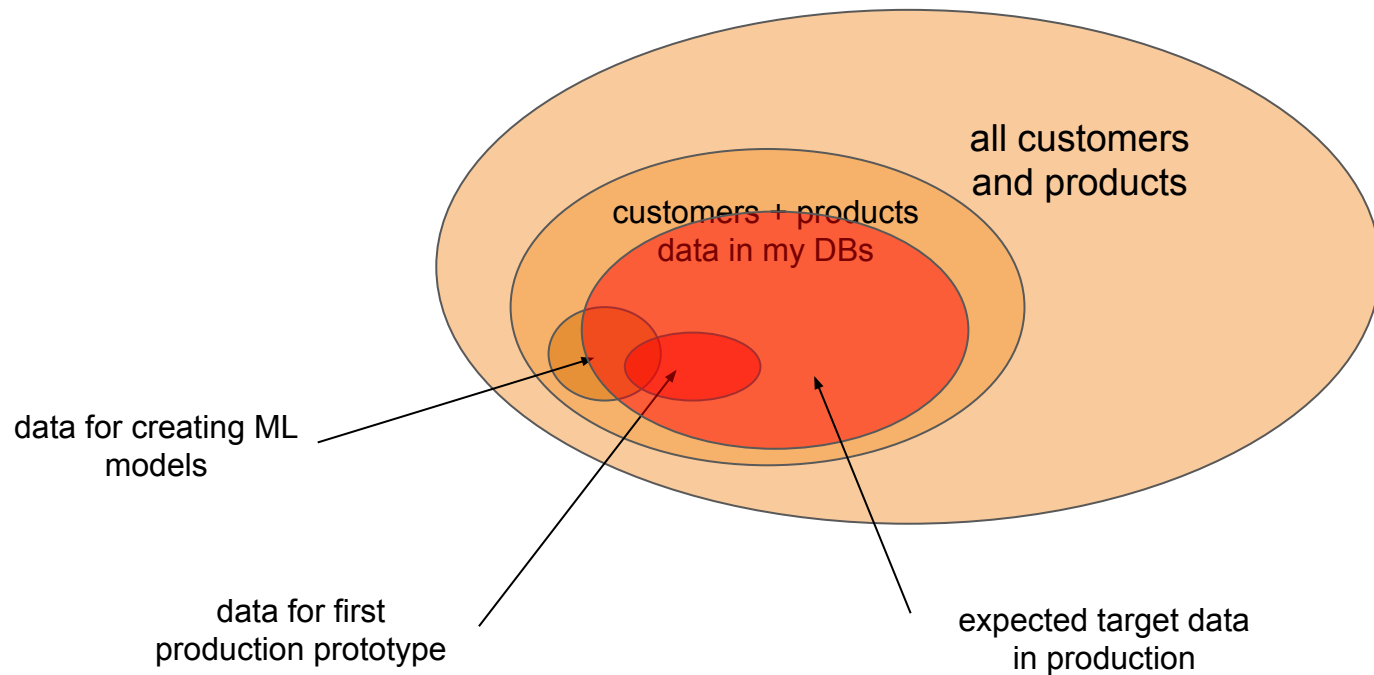
Data representativity

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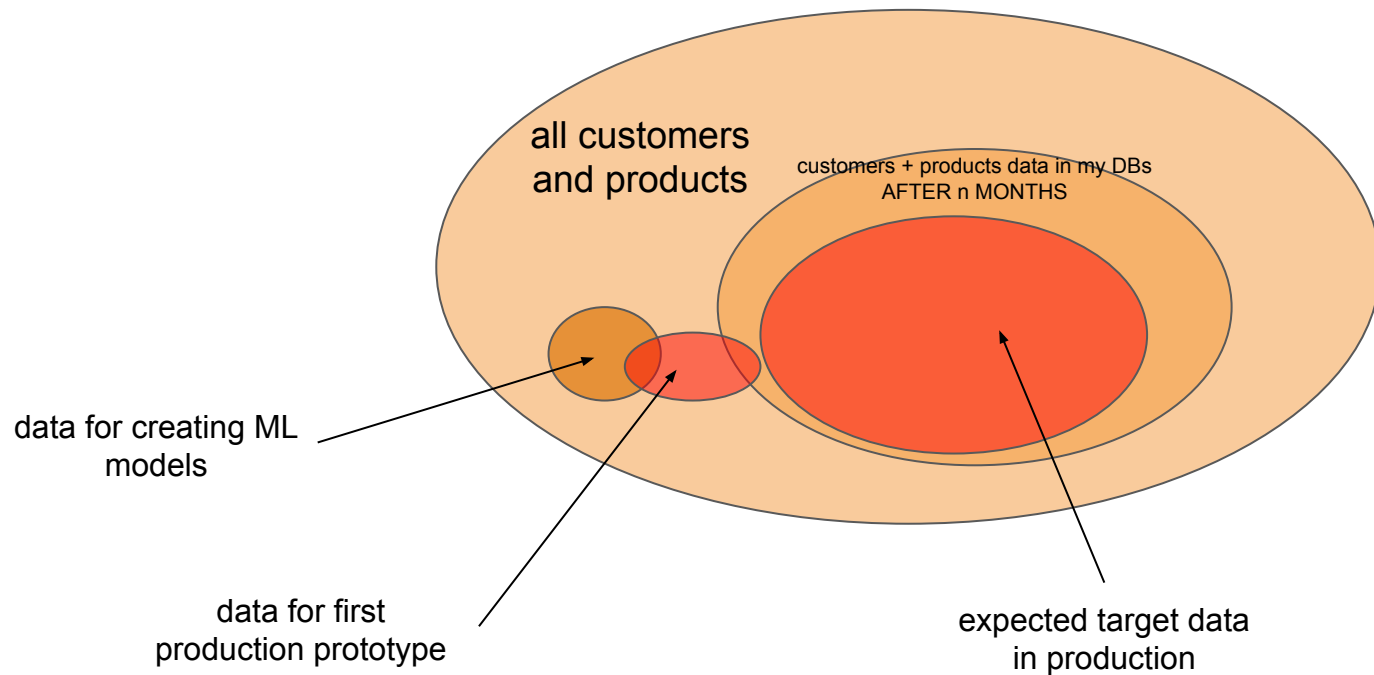
Data representativity

how **portable** are models?



Data representativity

how **portable** are models?

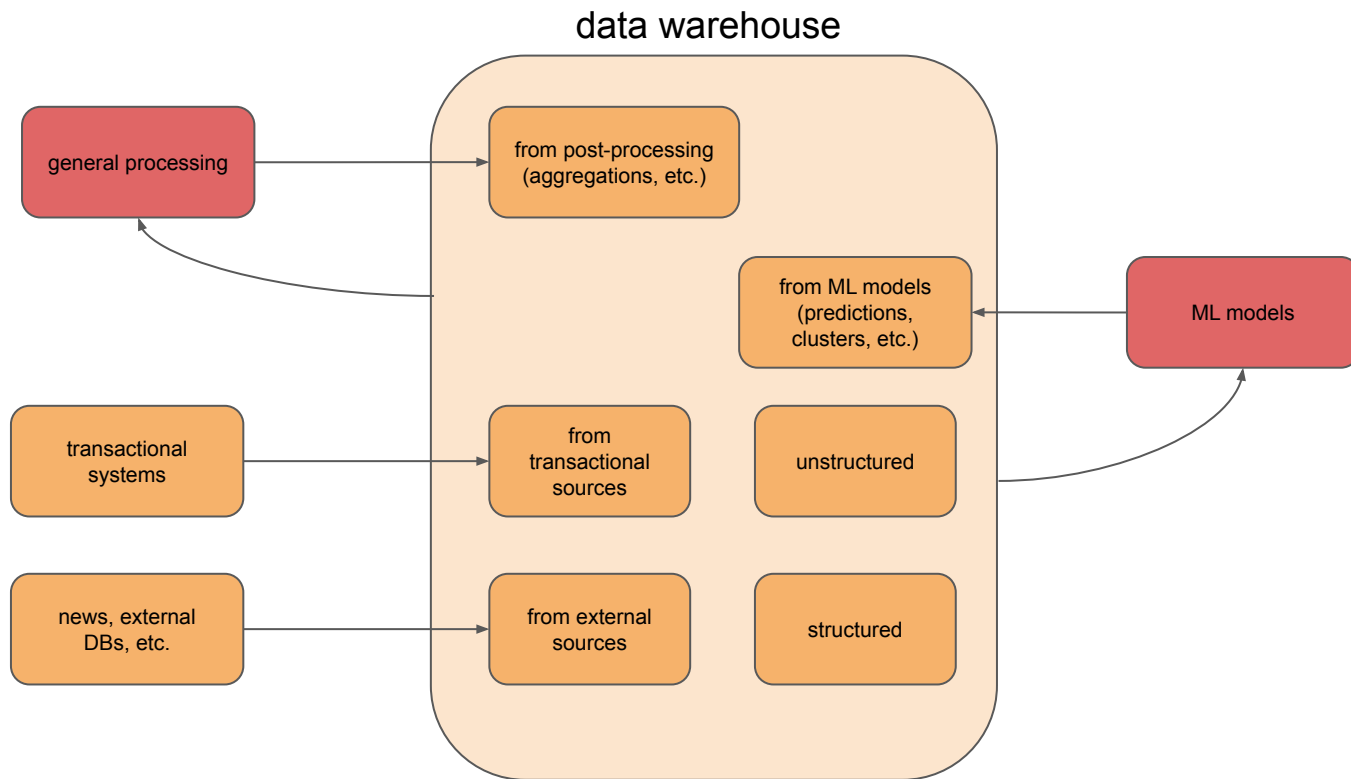


Data representativity

Examples

- Commodities time series, short range predictions too good due to sampling artifacts
- Defect detection deployment in factory: lighting conditions, requires periodic cleaning, etc.
- Energy time series: time to create predictive models
- Transportation: deployment of recalibrated to IoT devices with random connectivity.

Data lifecycle and maturity



Data lifecycle and maturity

How confident am I from a predictive model trained with data existing in the warehouse?

I run some analytics process (model, etc.) to make a report for management decision making, how can I be sure that:

- Source data was OK ?
- Analytics code is OK ?
- I can run the same process next month?

Datasets

What is it needed to build confidence in data products:

- Dataset Versioning
- Dataset Authoring
- Dataset Reproducibility
- Notebook/Script repository
- Libraries freeze (docker?)

ML Experiments reporting

- See MLFLOW demo

Team roles

adapted from: <https://martinfowler.com/articles/cd4ml.html>

