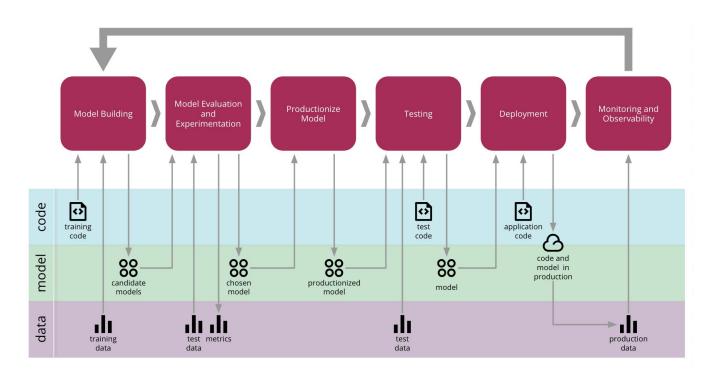
# Managing ML processes in the organization

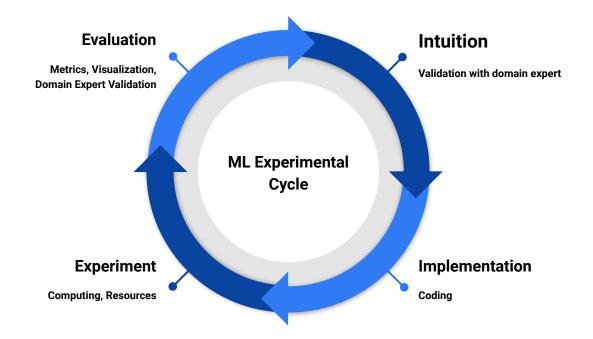
#### ML models life cycle

sources: https://martinfowler.com/articles/cd4ml.html and

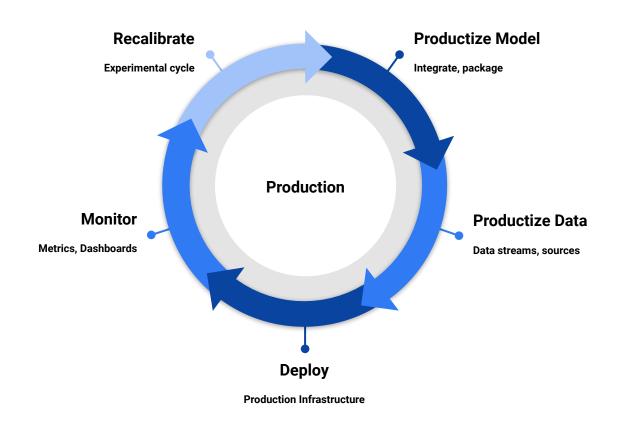
https://christophergs.com/machine%20learning/2020/03/14/how-to-monitor-machine-learning-models/



## Agility on the Experimental Cycle



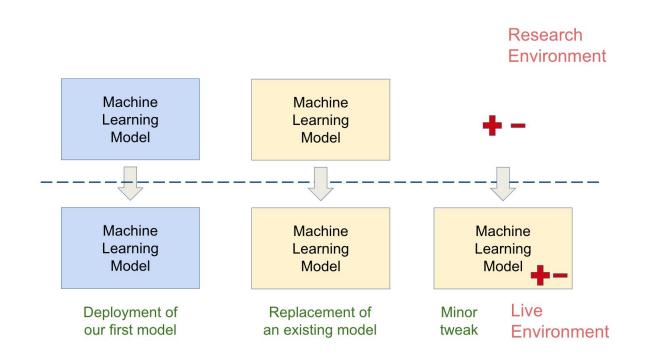
#### Robustness on the Production Cycle



#### Maintainability

sources: <a href="https://martinfowler.com/articles/cd4ml.html">https://martinfowler.com/articles/cd4ml.html</a> and

https://christophergs.com/machine%20learning/2020/03/14/how-to-monitor-machine-learning-models/

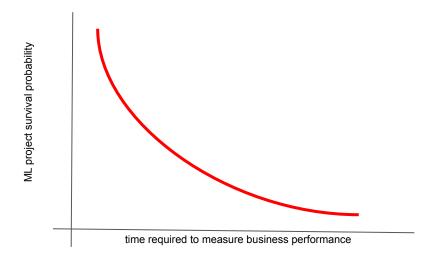


#### Model performance alerts

ML Metric (accuracy, RMSE, etc.) \_\_\_\_\_ Monitoring dashboards Business metric (PNL, Rol, etc.)

Performance measuring time vs. impact time

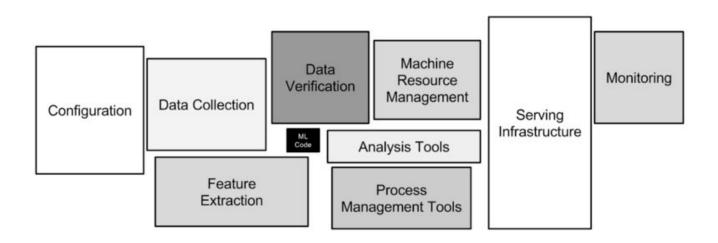
 What if I cannot measure a model has degraded business performance before the losses are too high?



#### Pipeline maintenance

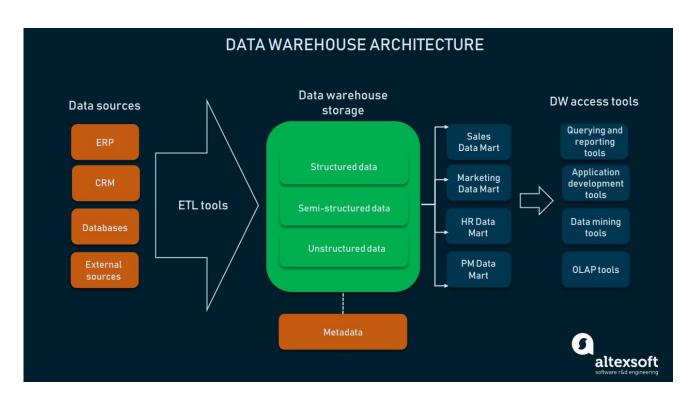
sources: <a href="https://martinfowler.com/articles/cd4ml.html">https://martinfowler.com/articles/cd4ml.html</a> and

https://christophergs.com/machine%20learning/2020/03/14/how-to-monitor-machine-learning-models/



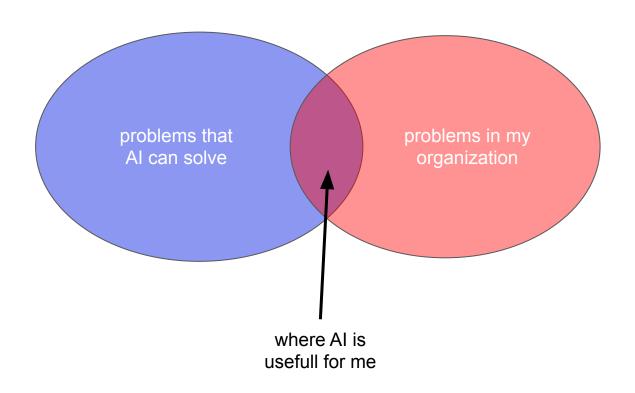
#### Data management

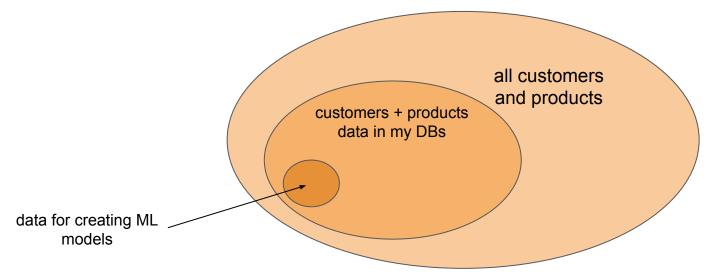
source: https://www.altexsoft.com/blog/datascience/what-is-data-engineering-explaining-data-pipeline-data-warehouse-and-data-engineer-role/

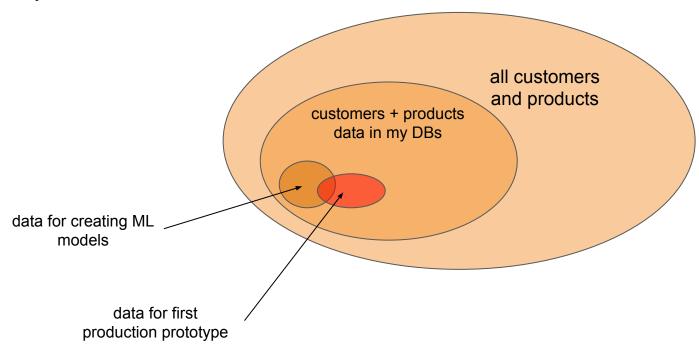


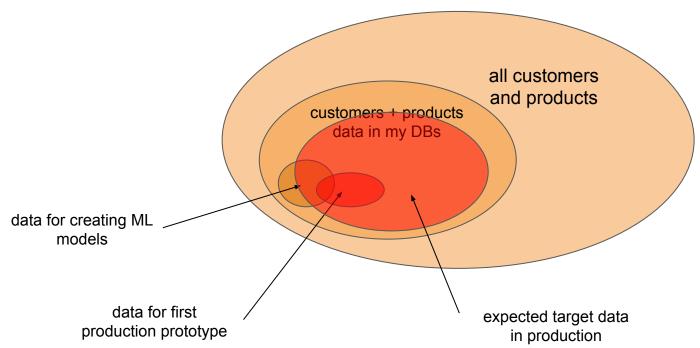
#### Is ML/Al really useful?

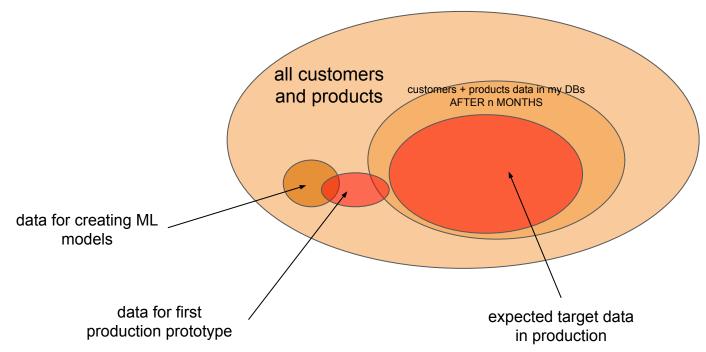
be smart selecting what ML really used for within my context







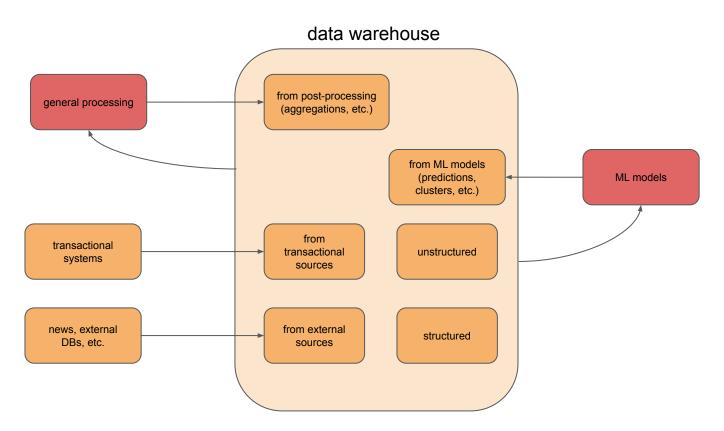




#### Examples

- 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: <a href="https://martinfowler.com/articles/cd4ml.html">https://martinfowler.com/articles/cd4ml.html</a>



