ML In Statistical Production Processes III
IT Infrastructure group

...or the "Everything but the actual ML use case"-group

Report from IT Infrastructure Group



Building capacity for machine learning

"Hidden Technical Debt in Machine Learning Systems," Google NIPS 2015

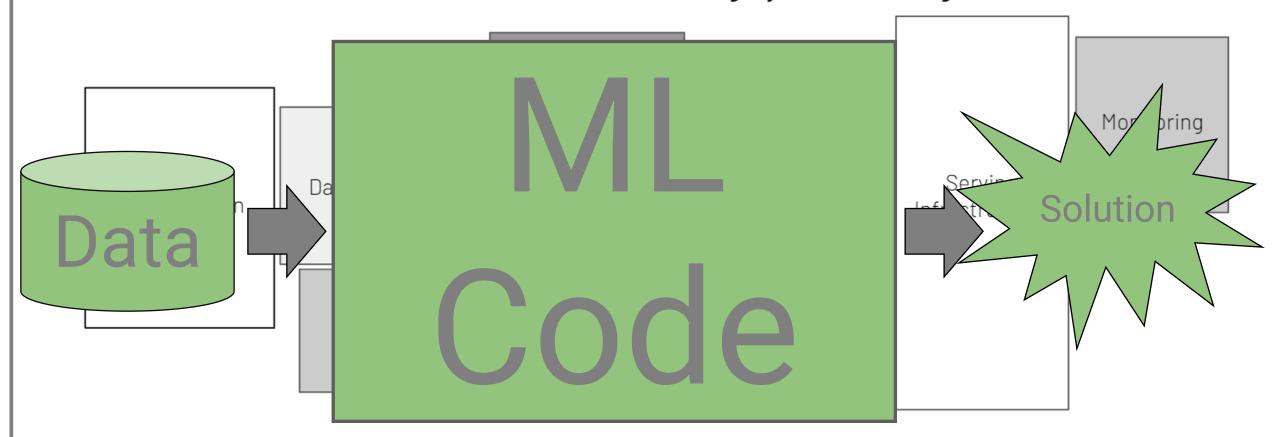


Figure 1: Only a small fraction of real-world ML systems is composed of the ML code, as shown by the small green box in the middle. The required surrounding infrastructure is vast and complex.

ML Maturity Model

Initial

A few individuals start getting interested in ML but experimenting in silos. There are difficulties in installing ML-related software as it is not supported by organization. Most business areas are not aware of potential of ML

Spreading

Groups of individuals become aware of others working on similar issues and form an informal network of knowledge sharing. Increasing awareness on the importance of providing model as a service, but individuals with ML skills have to work from beginning to end. Essential software is whitelisted but there is no "ML sandbox" and still difficult to work with these software

Widespread

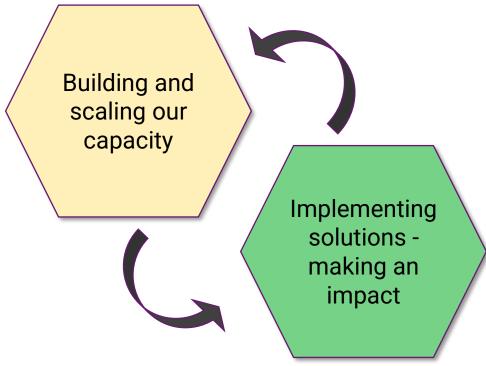
ML work become more centralized and formalized (e.g., restructuring organization or creation of new division). Supporting capabilities are identified and they start engaging in the ML development and operation. There exists regular training for staff. There is wide awareness about ML across organization and there exists a cooperate-wide strategy in place for the use of ML

Mature

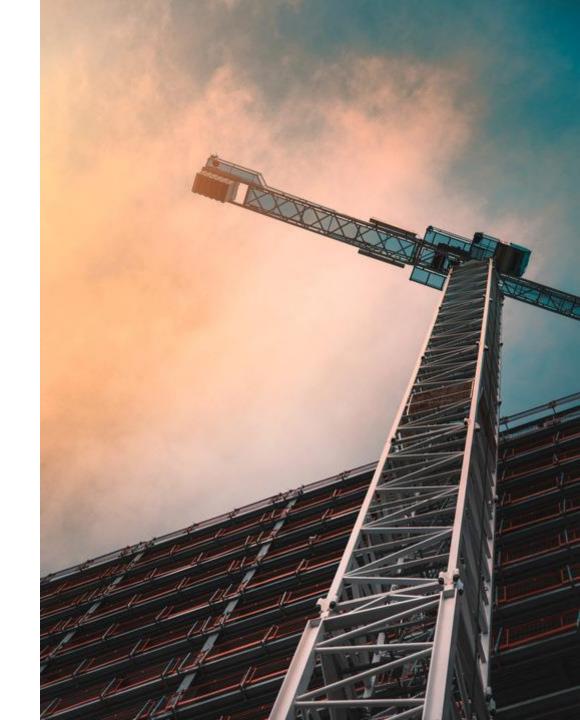
Essential capabilities are fully integrated in the organization providing ML infrastructure. ML development is streamlined and monitoring is automated



Building and using ML capacity



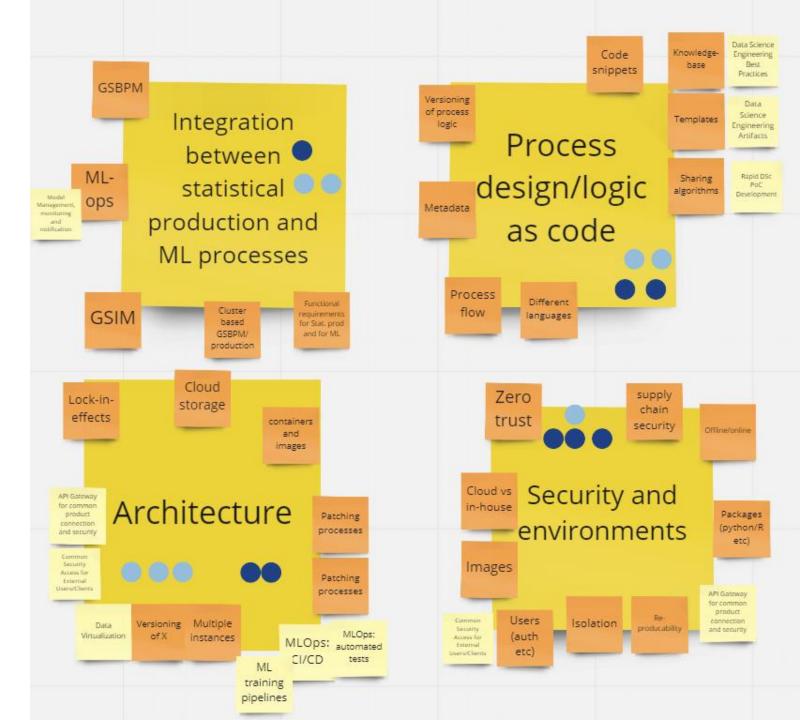




Topics for our group

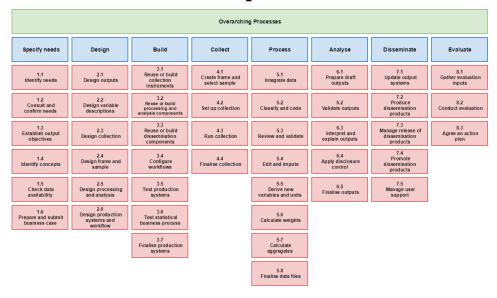
- Recurring themes in ML and ML-ops
 - Architecture, Technology, Infrastructure
 - Roles and capabilities
 - Connection between statistical production and ML processes



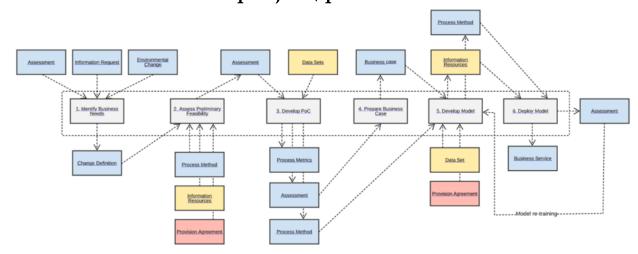


Connection between processes

Statistical production



ML project/process





Components for ML

Potential overlap with existing components

- (P4) (P5) Source Code Repository
- CI/CD P6) Component

- Workflow ^{©3}Orchestration
- **Model Training** Infrastructure

Component

ire

P7 ML Metadata

Monitoring

Component

Integration with

regular production

Environments for innovation and production

Collaboration and sharing solutions

P1) COLL

PRINCIPLES

P1 CI/CD automation

P2 Workflow orchestration

P3 Reproducibility

P4 Versioning of data, code, m

P5 Collaboration

P6 Continuous ML training & e

P7 ML metadata tracking

P8 Continuous monitoring

P9 Feedback loops

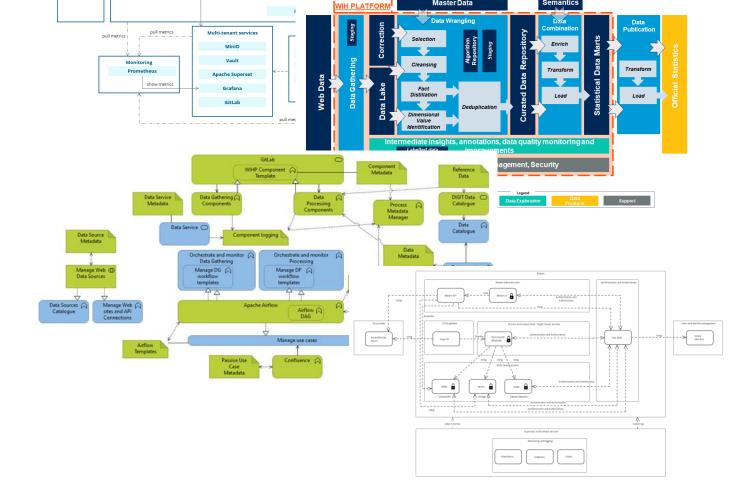
Stores

Cloud and on-prem options

COMPONENT

Collaboration around platforms for ML

- High degree of open source in ML
 - Easier to share platforms
 - Examples of platforms being developed and shared
 - Lots of similarities between shared solutions
- On-prem and Cloud
 - Use of container based solutions
- Most solutions aimed at supporting ML-processes

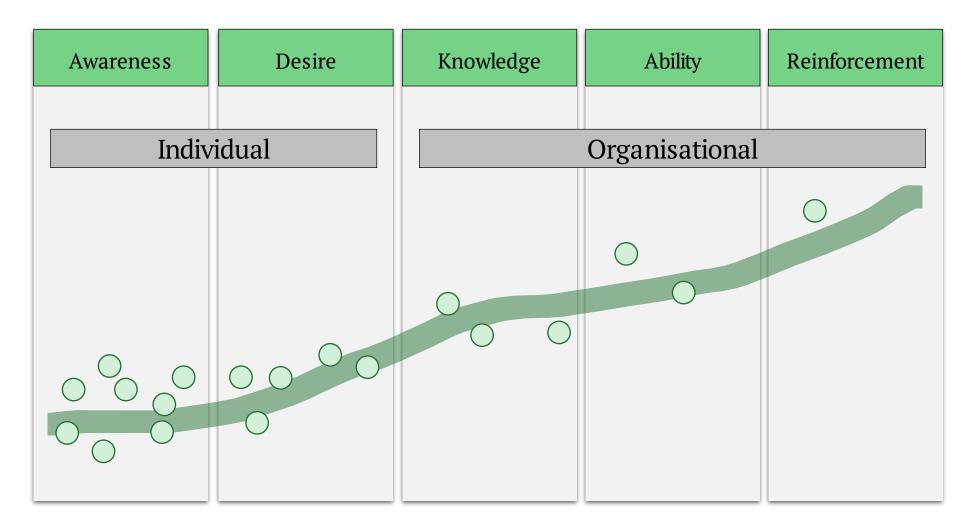




Different focus

Data Howmuch Production logic statisticians Metadata Versioning of data care Versioning of code Howmuch Security T care **Autometed production** Compute resources

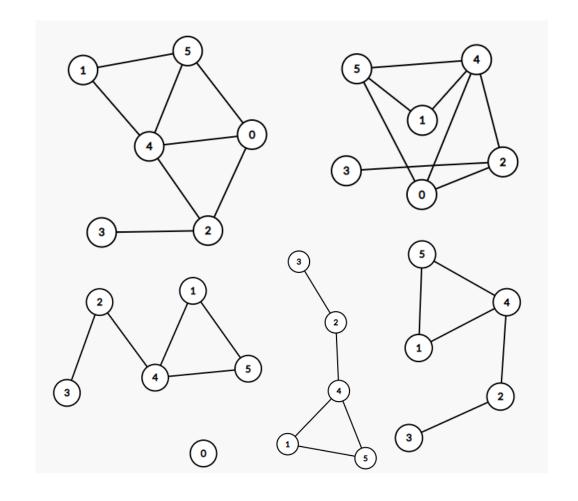
Building competence





Organisational structure

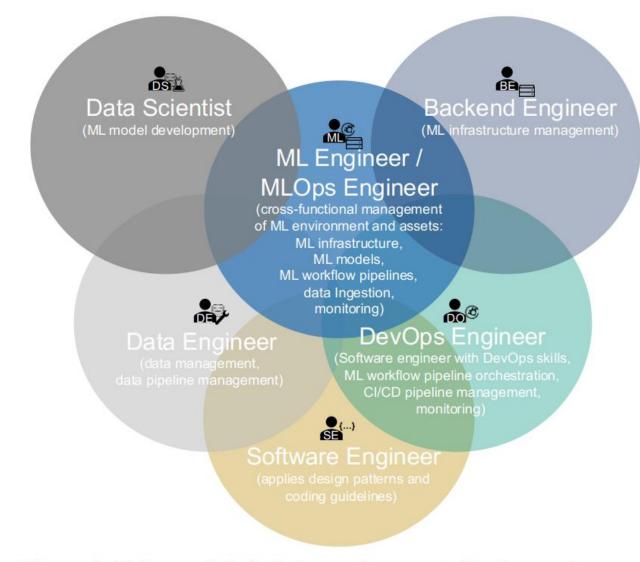
- Different operating structures
 - Hub & Spoke
 - CoE
 - Communities of practice
 - Informal networks
- Deliverables differ in structure
 - ML-solutions
 - Platforms
 - Skills building capacity





Roles and capabilities

- A combination of new and existing roles
- Cross-cutting topic
 - Business, IT, Methodology plus new roles
 - New forms of collaboration between roles are explored
- Often difficult to recruit and keep







Thank you!



