

# AIM3 – Scalable Data Analysis and Data Mining

Machine Learning in Practice and Technical Debt  
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- Complex Models Erode Boundaries
- Entanglement
  - CACE principle: Changing Anything Changes Everything
- Correction Cascades
- Undeclared Consumers.

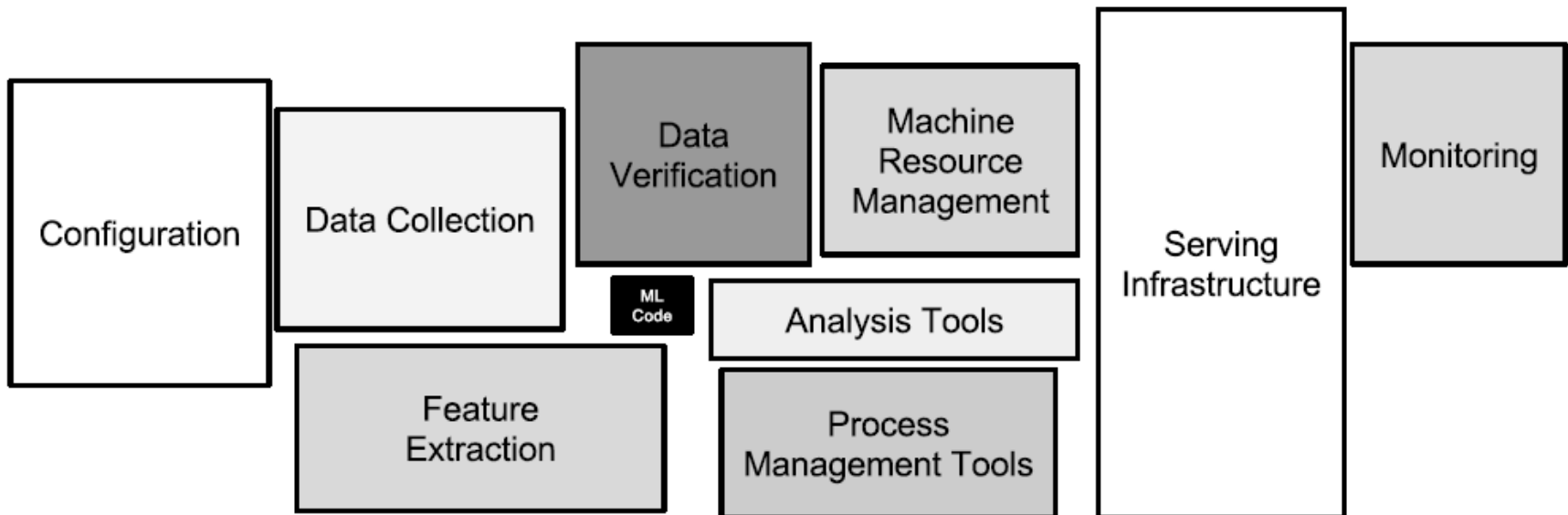
## ■ Unstable Data Dependencies

- -> Versioning

## ■ Underutilized Data Dependencies

- Legacy Features
- Bundled Features
- q-Features
- Correlated Features

- Only a small fraction of real-world ML systems is composed of the ML code
- The required surrounding infrastructure is vast and complex



- Direct Feedback Loops
- Hidden Feedback Loops

- Pipeline Jungles
- Glue Code
- Dead Experimental Codepaths
- Abstraction Debt

- Fixed Thresholds in Dynamic Systems
- Monitoring and Testing
  - Prediction Bias
  - Action Limits
  - Up-Stream Producers.

## ■ Hidden Technical Debt:

- D. Sculley, Gary Holt, Daniel Golovin, Eugene Davydov, Todd Phillips, Dietmar Ebner, Vinay Chaudhary, Michael Young, Jean-Francois Crespo, and Dan Dennison. 2015. Hidden technical debt in Machine learning systems. In *Proceedings of the 28th International Conference on Neural Information Processing Systems (NIPS'15)*
- Tom van der Weide, Dimitris Papadopoulos, Oleg Smirnov, Michal Zielinski, and Tim van Kasteren. 2017. Versioning for End-to-End Machine Learning Pipelines. In *Proceedings of the 1st Workshop on Data Management for End-to-End Machine Learning (DEEM'17)*
- The Anatomy of a Production-Scale Continuously-Training Machine Learning Platform KDD 2017 (forthcomming)
- Jimmy Lin and Dmitriy Ryaboy. 2013. Scaling big data mining infrastructure: the twitter experience. *SIGKDD Explor. Newsl.* 14, 2 (April 2013)
- [http://martin.zinkevich.org/rules\\_of\\_ml/rules\\_of\\_ml.pdf](http://martin.zinkevich.org/rules_of_ml/rules_of_ml.pdf)