



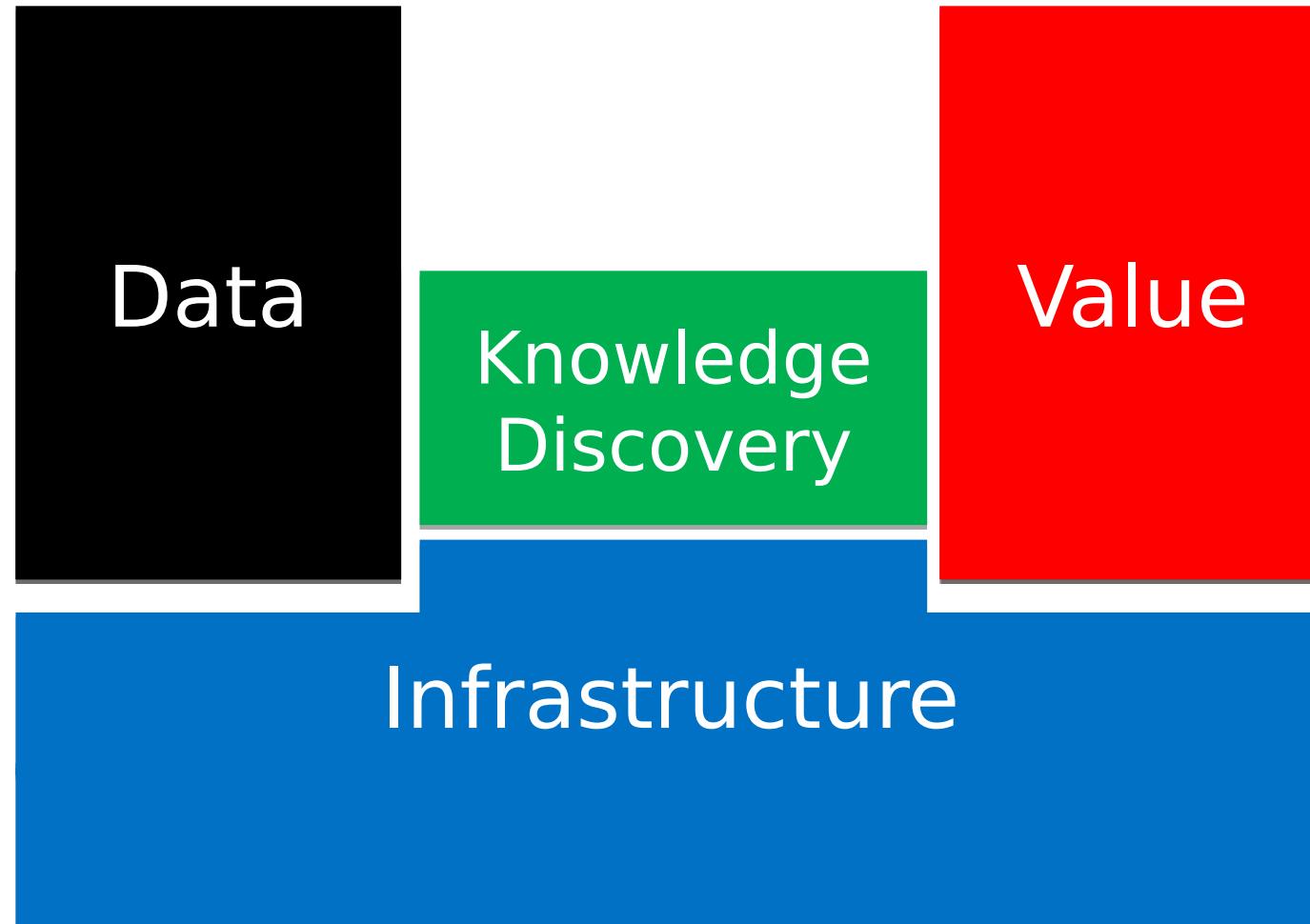
大数据分析师的卓越之道

吴甘沙

英特尔中国研究院



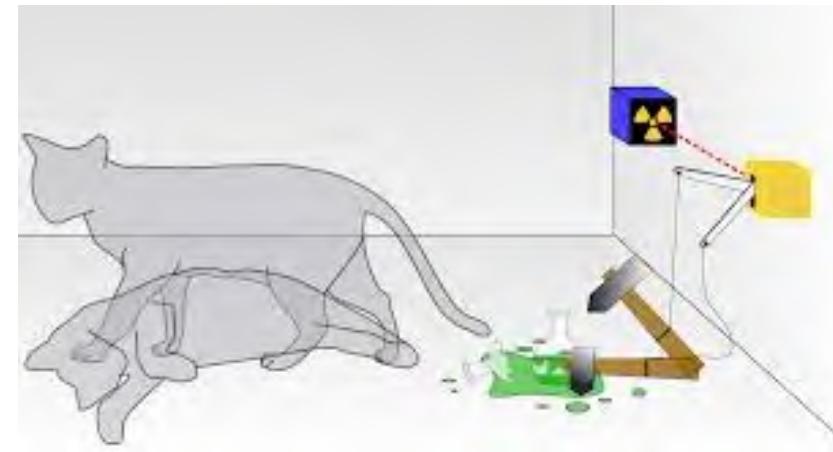
数据分析的典型场景



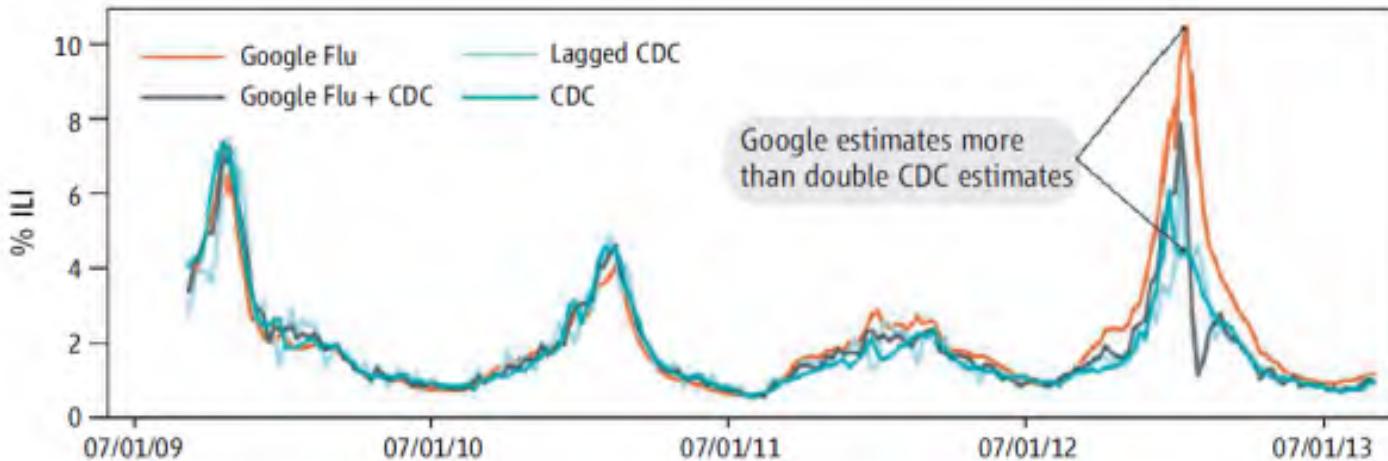
基础设施已经改朝换代 分析师也需要与时俱进

改变思维方式
提高技术素养
丰富分析能力

新的世界观：不确定的世界



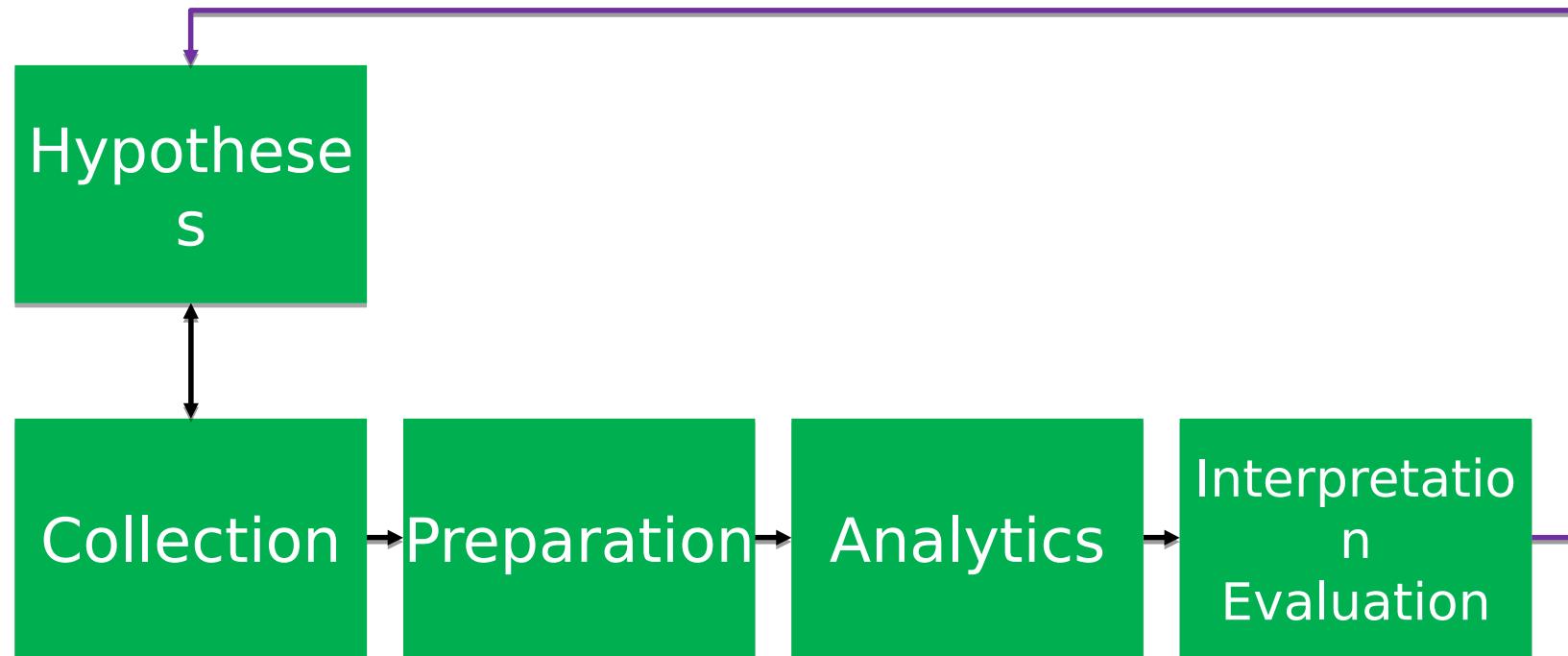
大数据的测不准



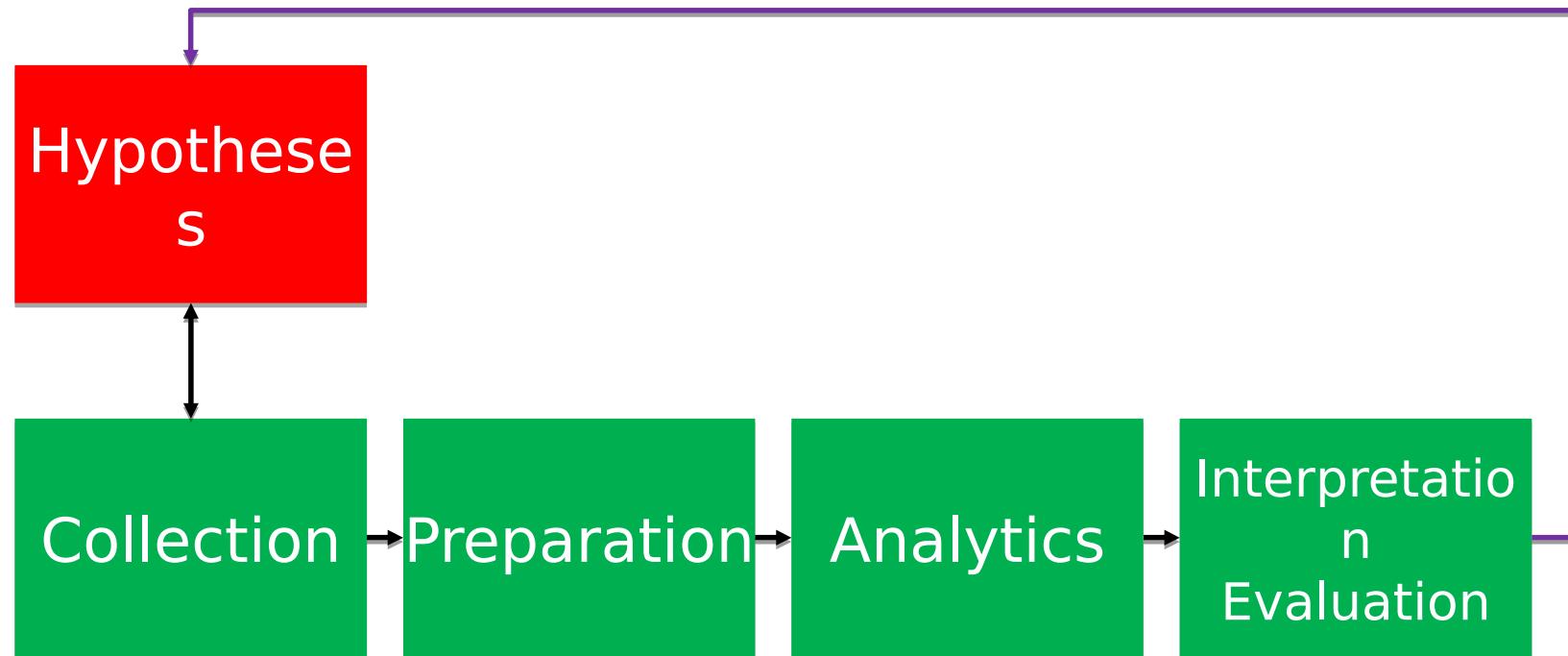
《自然》：测不准

《科学》：大数据傲慢

数据分析方法论的升级



数据分析方法论的升级



Hypotheses

机械地发掘相关性和假设

直觉，拿侦探小说练手

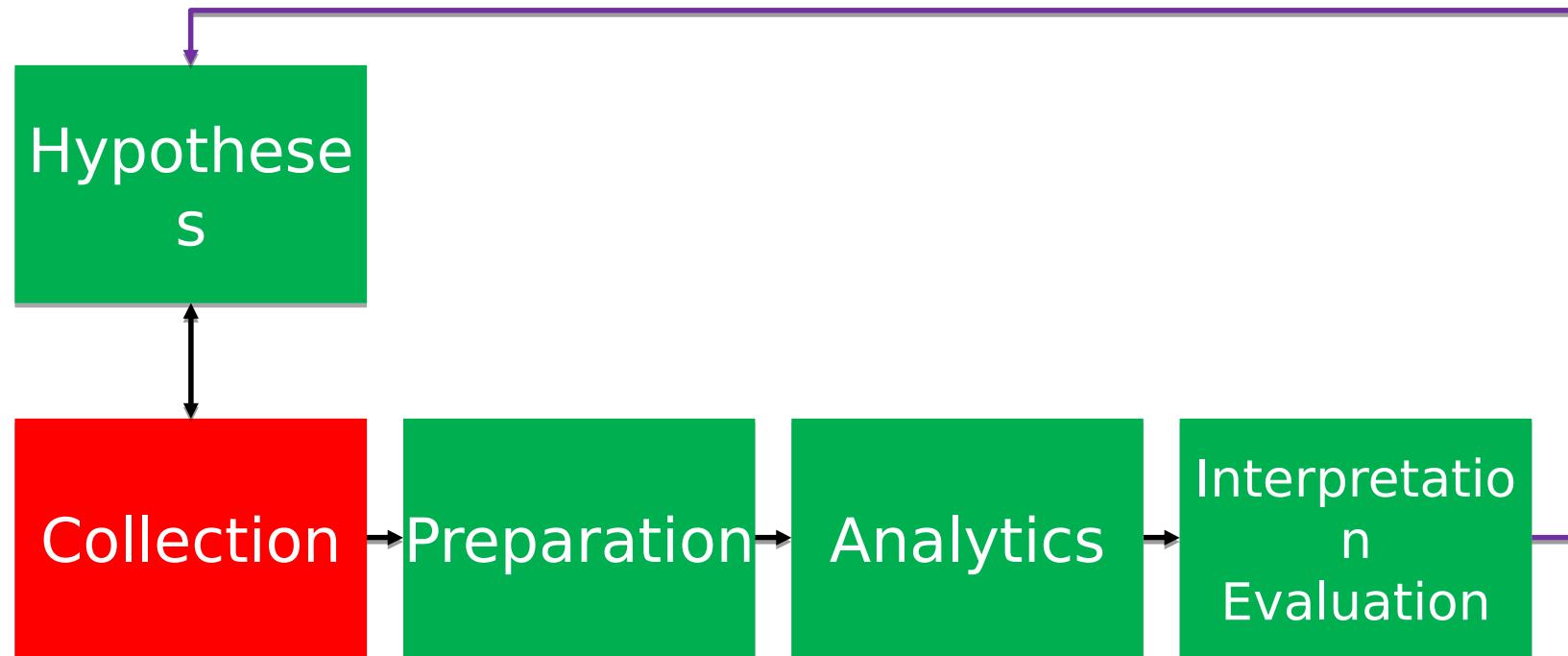
阅读广泛涉猎

跨界思维碰撞

融入业务部门

- 防止数据采集与分析、业务与数据分析的脱节

数据分析方法论的升级



数据！ 数据！ 数据！

n=All !

Enterprise Data Warehouse □ Enterprise Data Hub/Data Lake

External data sources

Structured □ semi-structured □ unstructured

- Log analysis
- Text analysis
- Image/video
- Data with geo and temporal tags
- Networks and graphs

数据? 数据? 数据?

n=All ?

- More data vs. sampling
- “Raw data” is an oxymoron
- Signals and noises
- Sampling bias

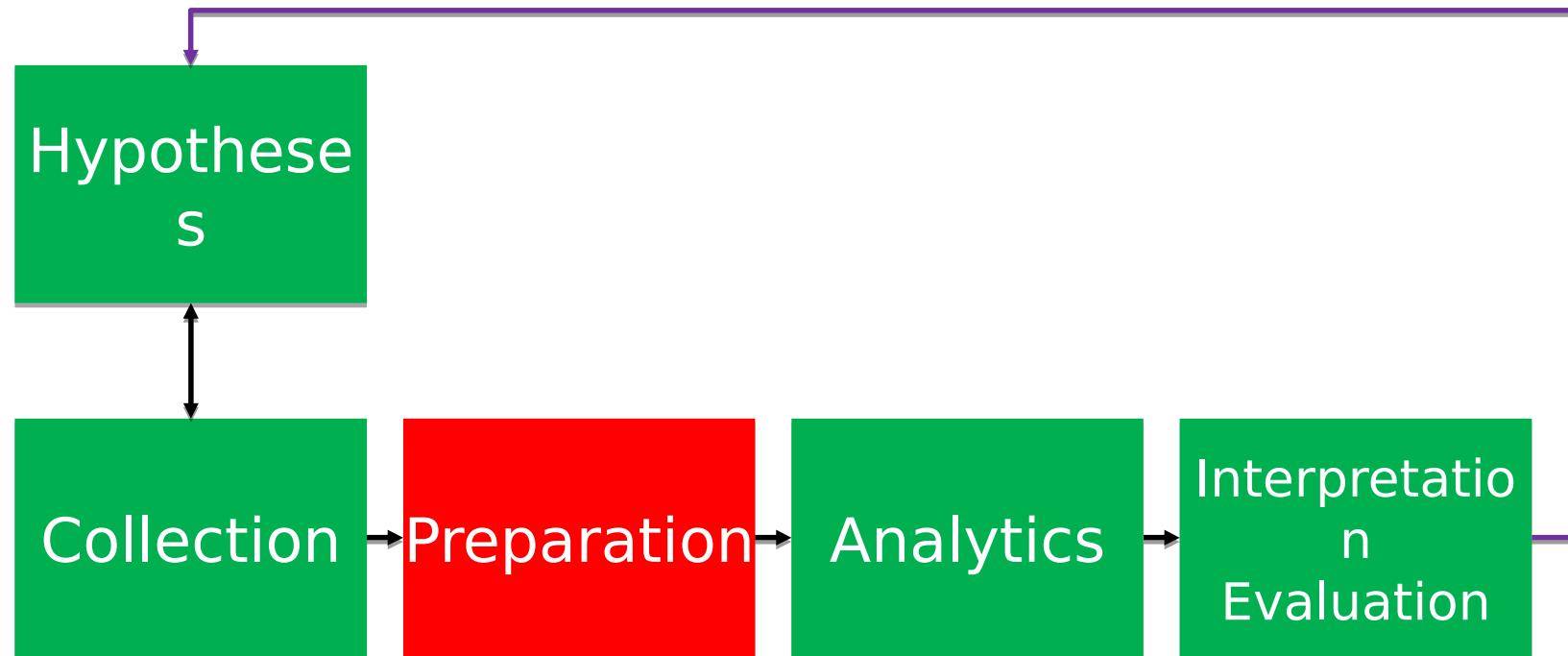
Data exchange and sharing

- Data rights, data pricing

Data lifecycle management

- Provenance capture, representation, and querying
- Sometimes data are not assets, but costs

数据分析方法论的升级



数据质量：重中之重

Noisy, biased and polluted data are unavoidable

- Goal: models = components for noise + relatively complex models for signal

Cleansing, validation, ...

- Can it start with a small subset? Can the process be automated?
- Work together with visualization, machine learning

Curation, Wrangling, ...

- Automated learning to discover structure, resolve entities, and transform data

数据表示

Reduce compute and communication complexity

- Sparse, compressed data structure
- Approximate computation

Reduce statistical complexity

- Dimensionality reduction, clustering

Sampling

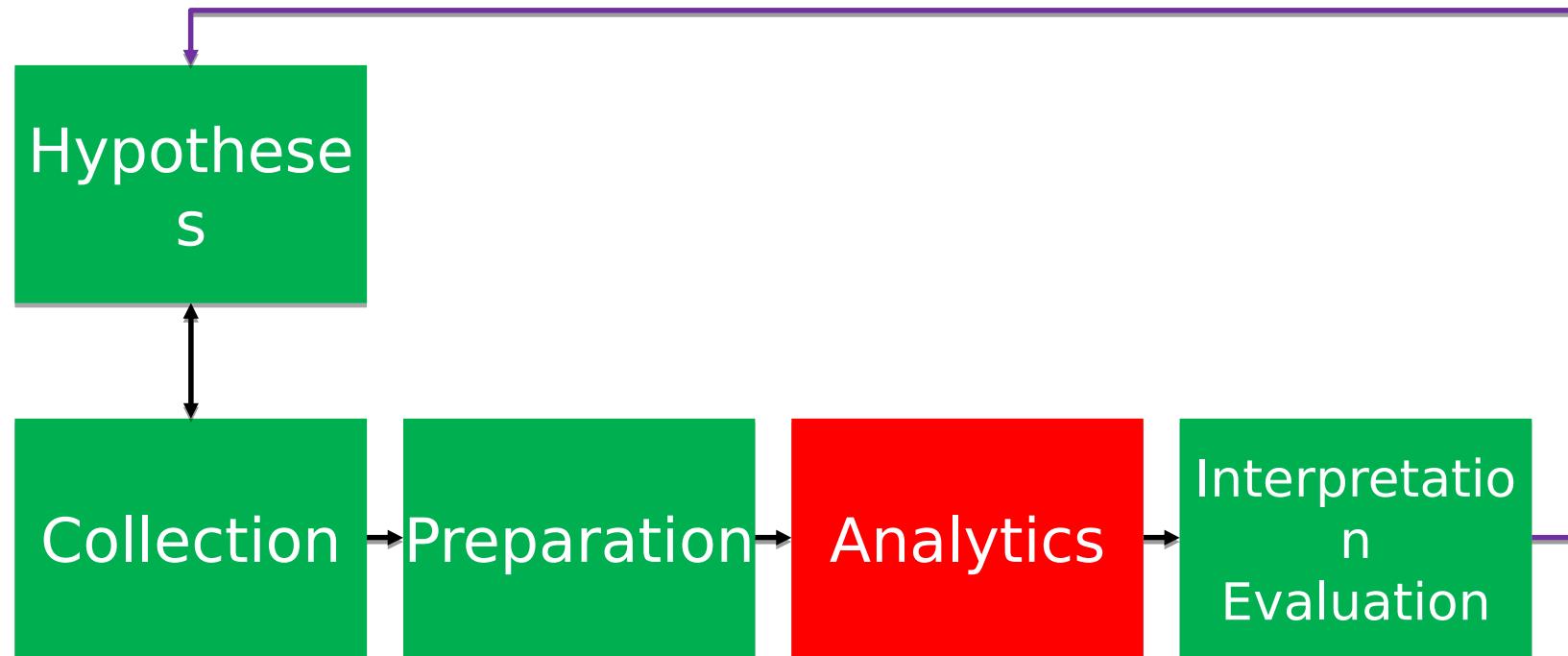
- Non-random sampling, compressive sensing,

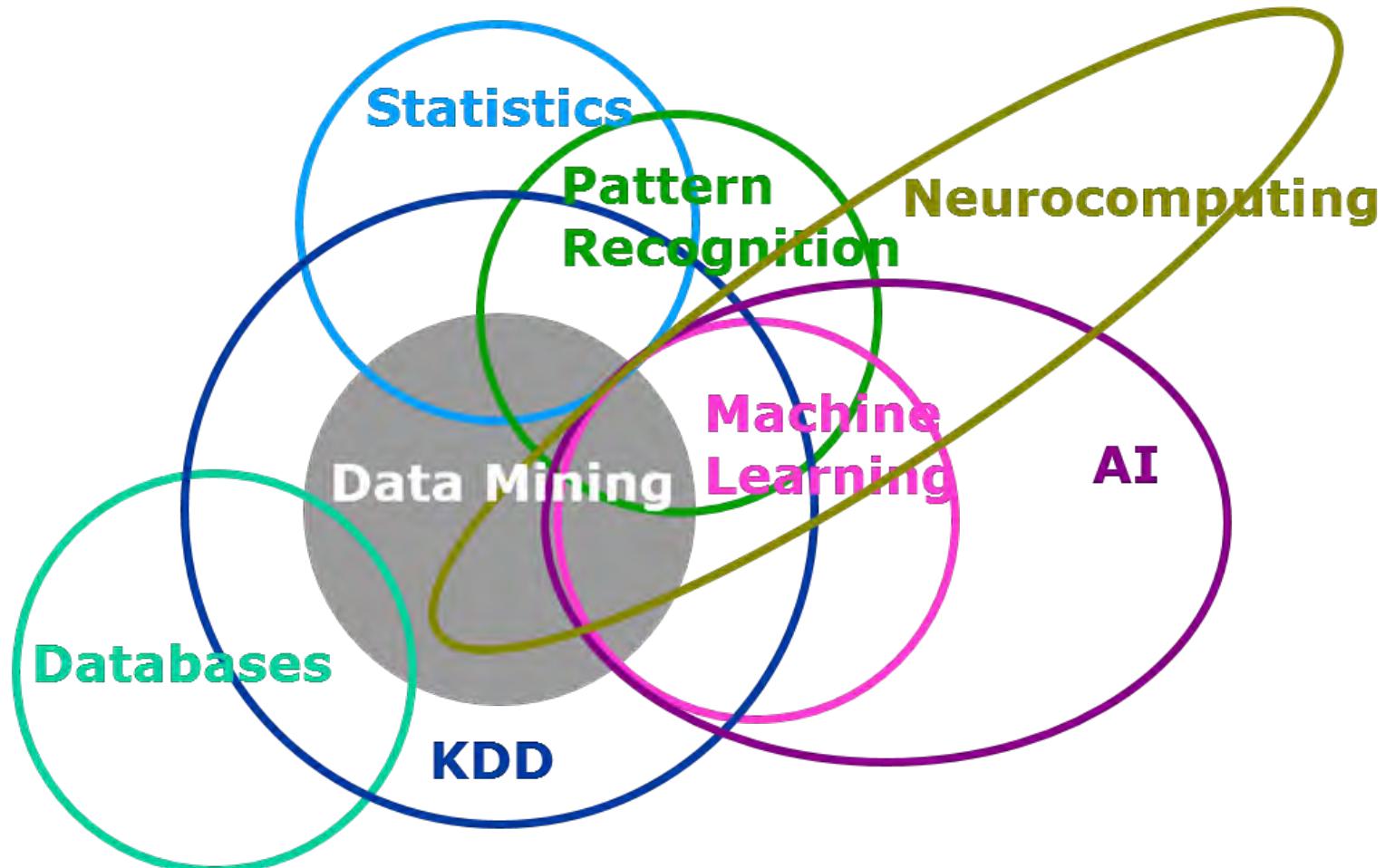
UIMA: Unstructured
Information Management
Architecture

Choose best representation for specific computational methods

- E.g. tables for data parallelism, networks/graphs for graph parallelism

数据分析方法论的升级





Computational Science

Source: blogs.sas.com

检查自身装备



JavaScript

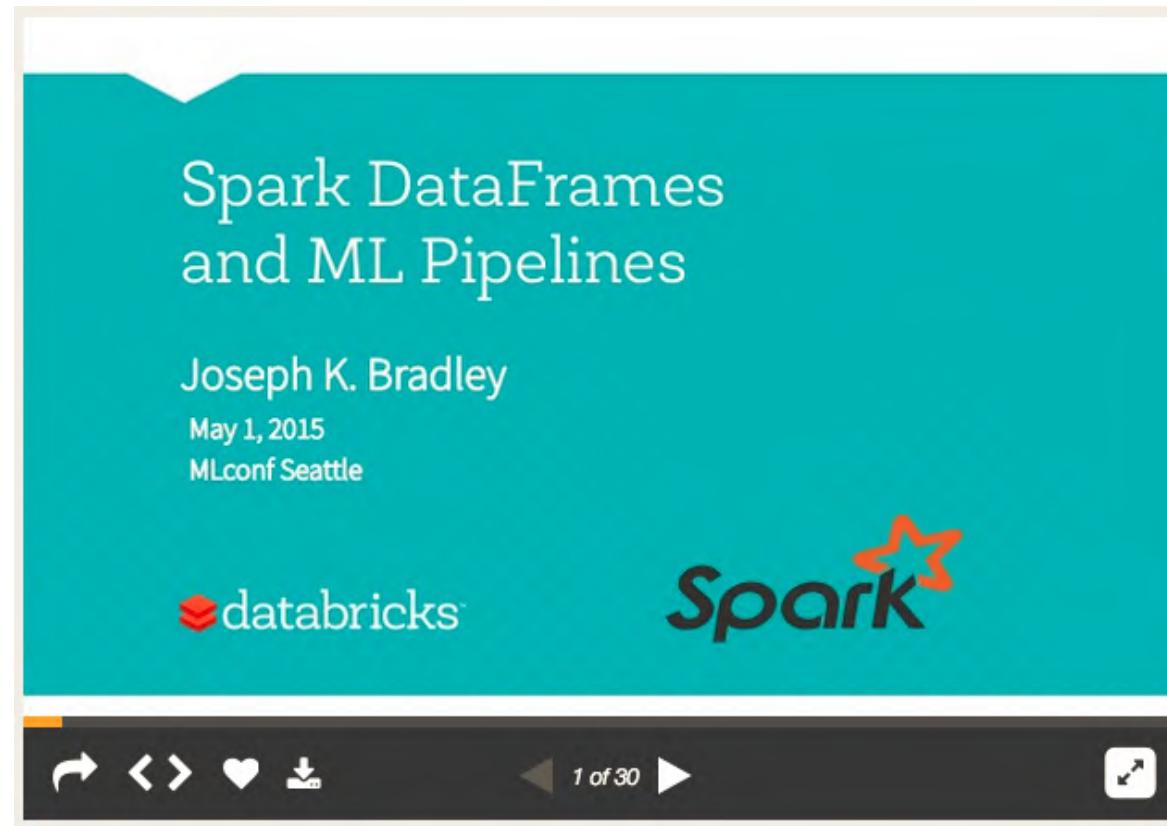


检查自身装备

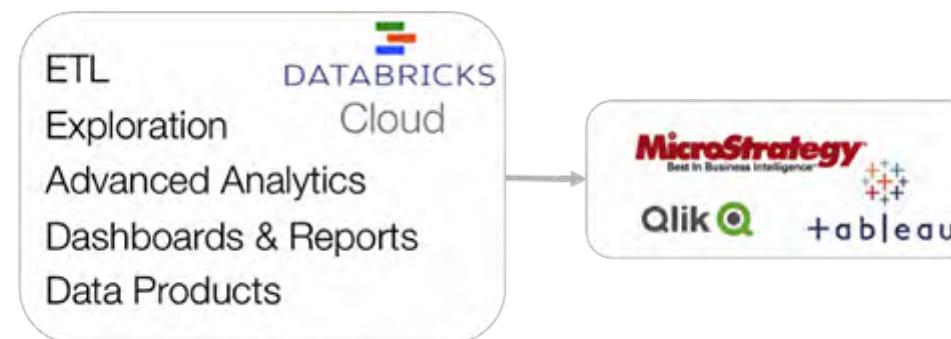


ML Pipeline

Scikit-learn style pipelines



拥抱云的世界



all models are wrong, but some are useful

刺猬（一招鲜吃遍天） vs. 狐狸（一把钥匙开一把锁）

模型的复杂度与问题匹配：奥卡姆剃刀原理

如何做到数据越多、边际收益越大？

- 数据不可名状的功效：简单模型 + 大数据 > 复杂模型 + 小数据 ?
- Ensemble
- 混合模型
 - Non-parametric vs. Parametric
 - Linear vs. Non-Linear
 - Discriminative vs. Generative

感知长尾信号 Exponential assumption vs. long tail

- PCA/LDA/pLSA vs. 分级训练、模型组合、概率图模型、深度神经网络

Velocity

Interactive query

流计算

- 如时空数据

在线 / 流式 / 增量学习

- 增量训练，模型异步更新，快速部署

当又大又快时，你必须懂系统

- Big Learning System：并行化 / 分布式化，系统调优

Deep Learning

从语音识别，到图像理解，到自然语言理解

- 领域特化：如医学图像分析

进军“非认知”任务

- 搜索，广告，推荐，… …， visuo-motor skills, drug discovery
- Automated laboratory

Open Source ▾ Collaborative Open Computer Science

- Pylearn2, Theano, Caffe
- GitXiv = arXiv + Github + Links + Discussion

Sparse Coding

A strong tool to deal with low SNR and poor veracity

We Have Successfully Applied Sparse Coding Based Prediction to a Number of Applications

2015:

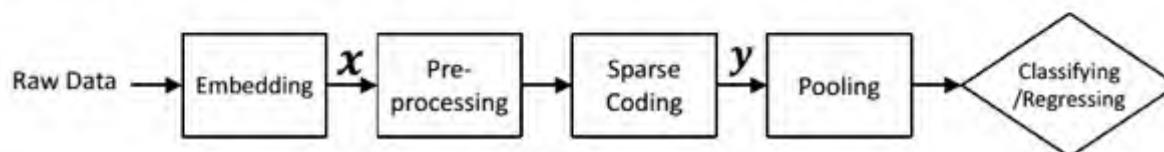
Twitter geolocation, facial emotion classification, wearable pulse sensing, link-layer adaptation to wireless fluctuations

2014:

Image object classification, chip power management, Wi-Fi traffic pattern inference, gait recognition, distributed spectrum sensing

(See <http://www.eecs.harvard.edu/htk/publications/>)

All use **sparse coding based** prediction pipelines:



Source: HT Kung

缺乏标注数据下的学习



Supervised Classification



Semi-supervised Learning



Transfer Learning



Self-taught Learning

Source: Andrew Ng

人的角色 Human Machine Intelligence

人的工作不断被取代

- 特征工程
- 比如以 MLBase 和 VizDeck 为代表的自动化分析和可视化

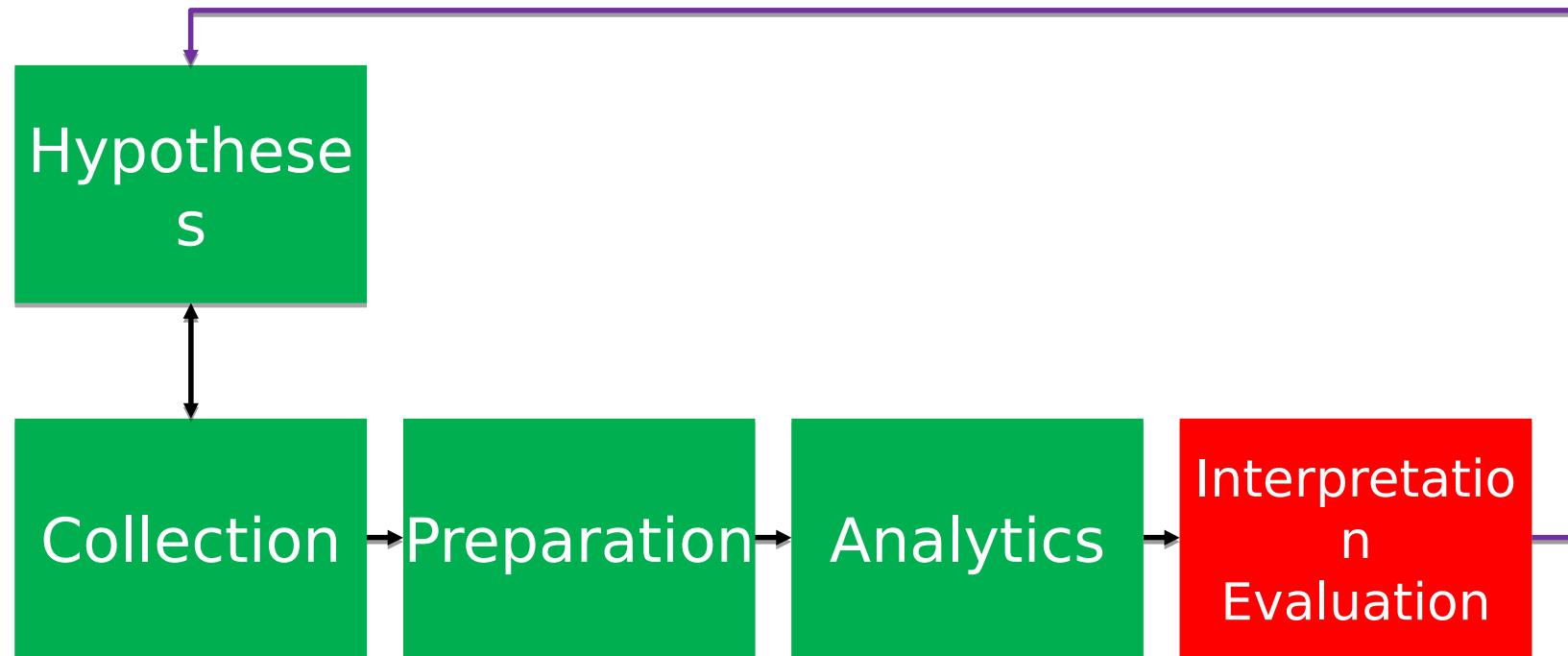
人与机器搭配获得最佳性能

- 比如 Exploratory analytics/visualization

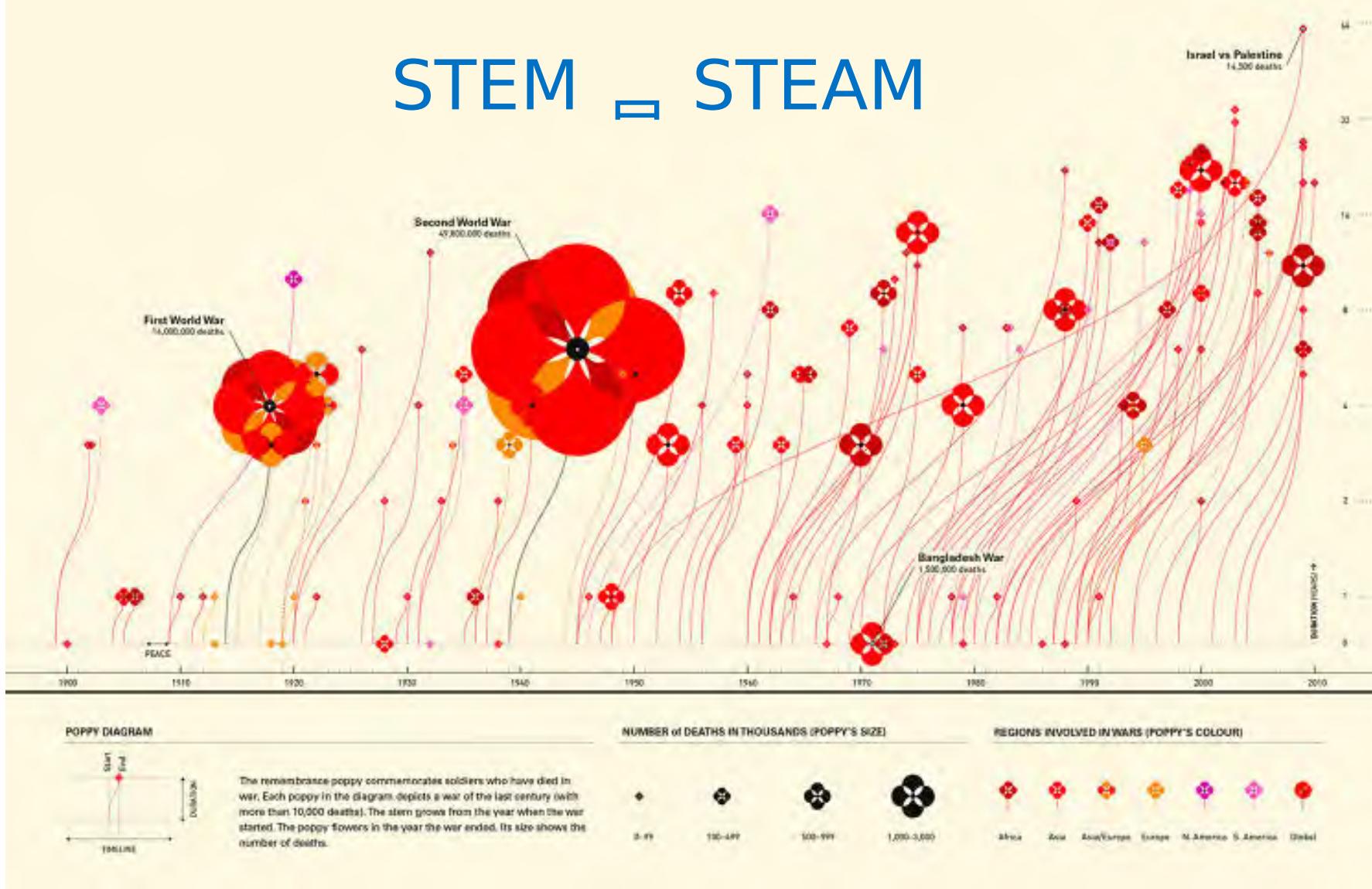
大规模人人、人机协作分析

- 外包 Kaggle，众包 CrowdDB，协作 DataHub，人类计算 Duolingo

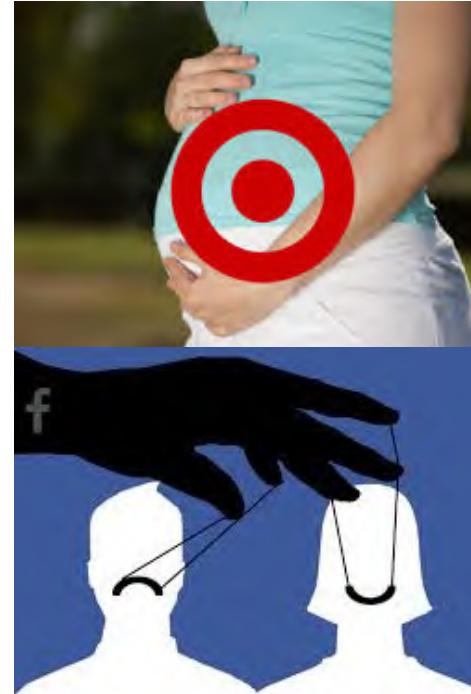
数据分析方法论的升级



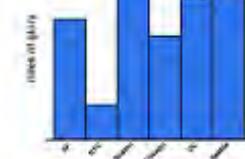
STEM □ STEAM

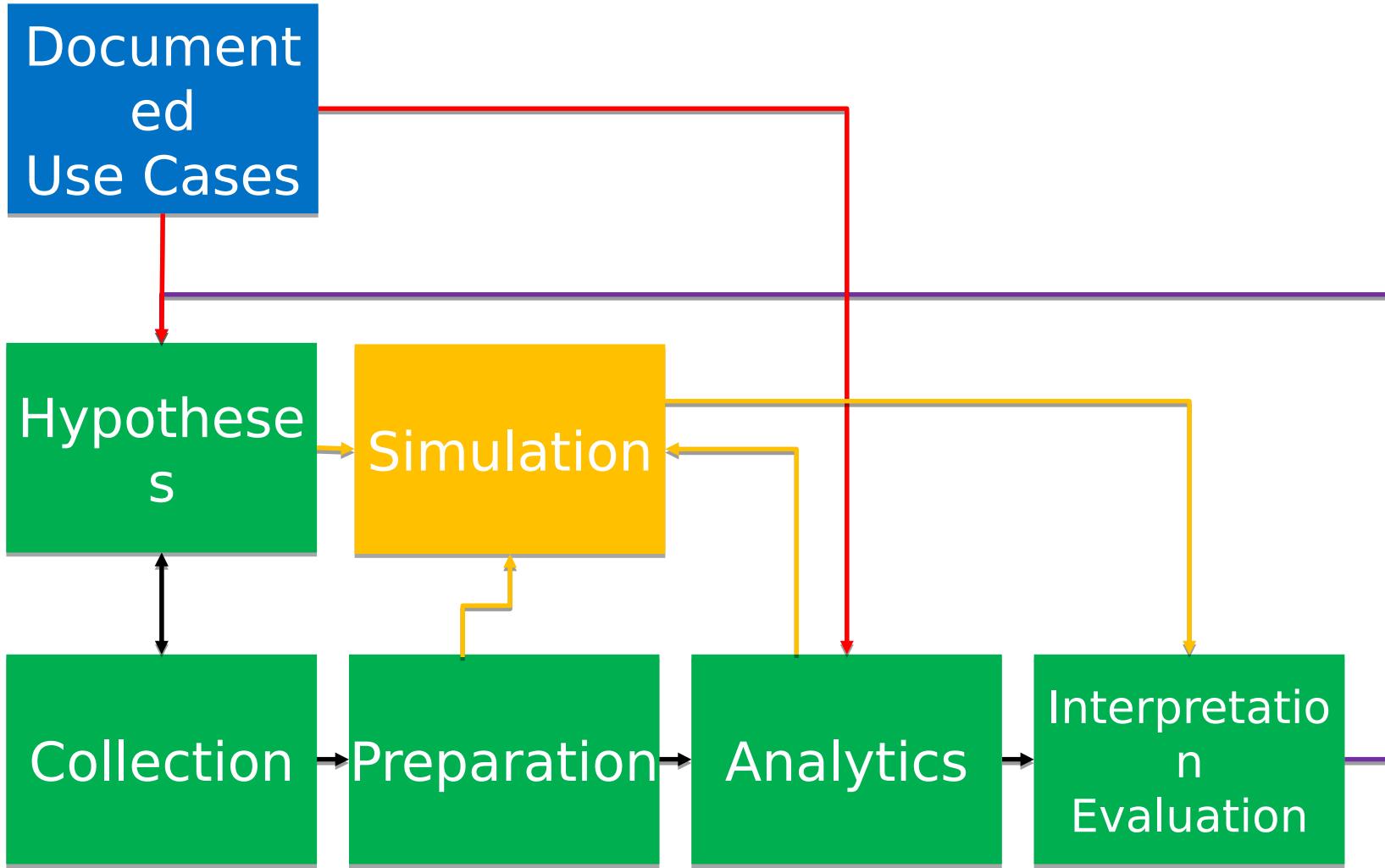


Story telling and “ideas worth spreading”



"RIDES OF GLORY"





基础设施已经改朝换代 分析师也需要与时俱进

改变思维方式
提高技术素养
丰富分析能力

敬谢聆听