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Machine Learning Research in blibli

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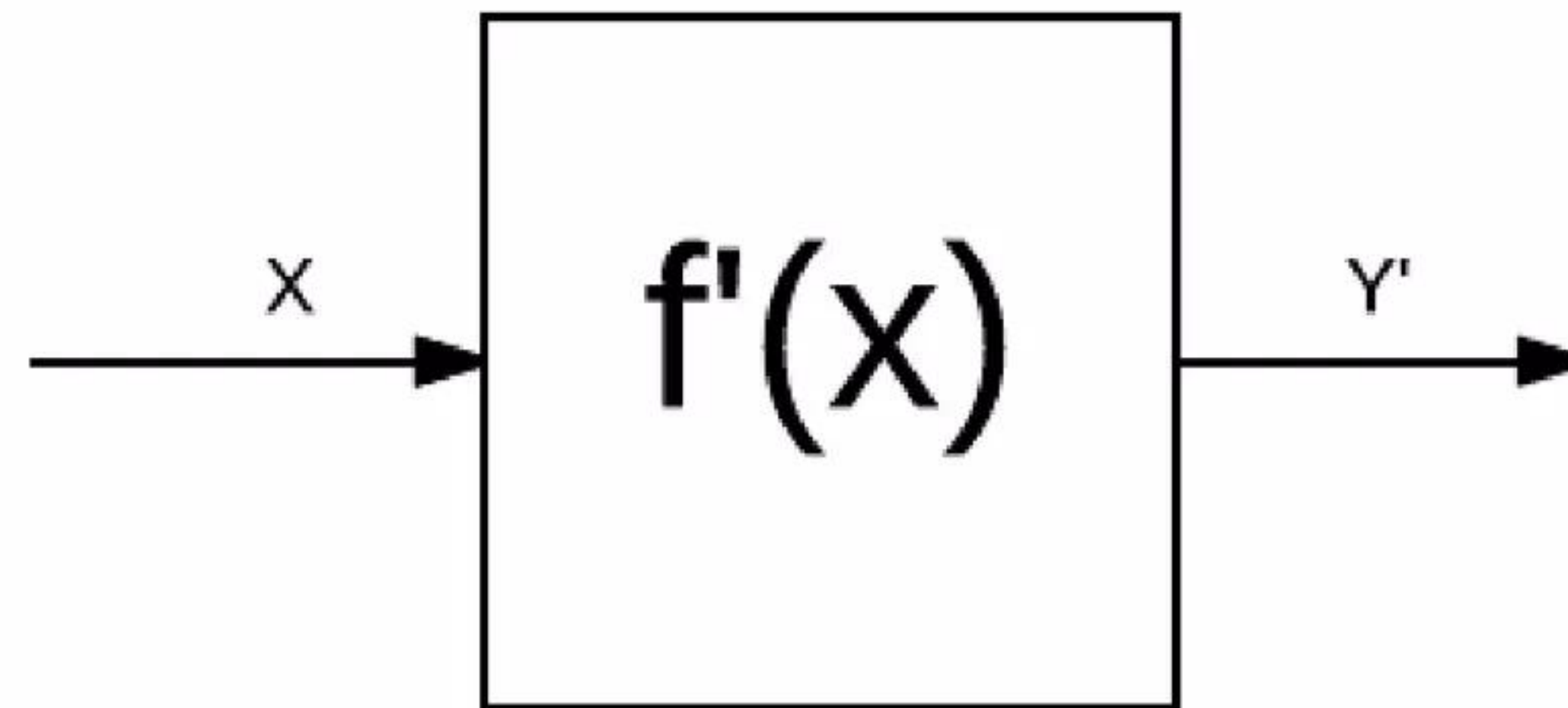
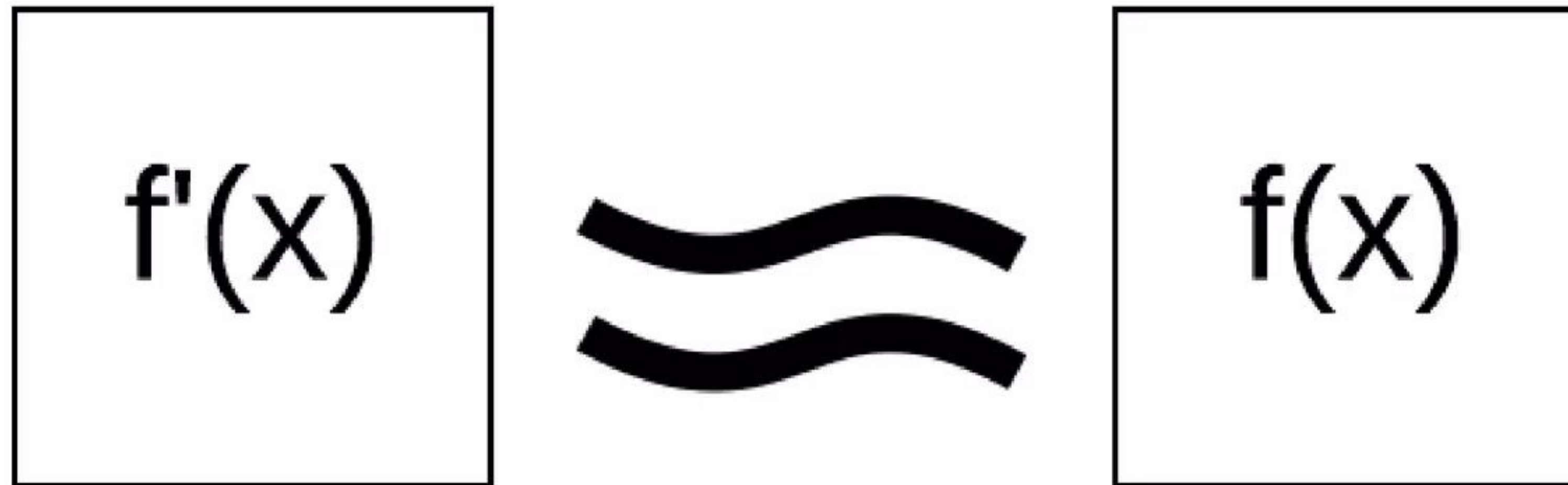
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- Sr. Research and Development Engineer at blibli.com (PT. Global Digital Niaga)
- Rnd Team for Data Science/intelligence system
- Working for Fraud Detection System. Current working in dynamic recommendation system project.

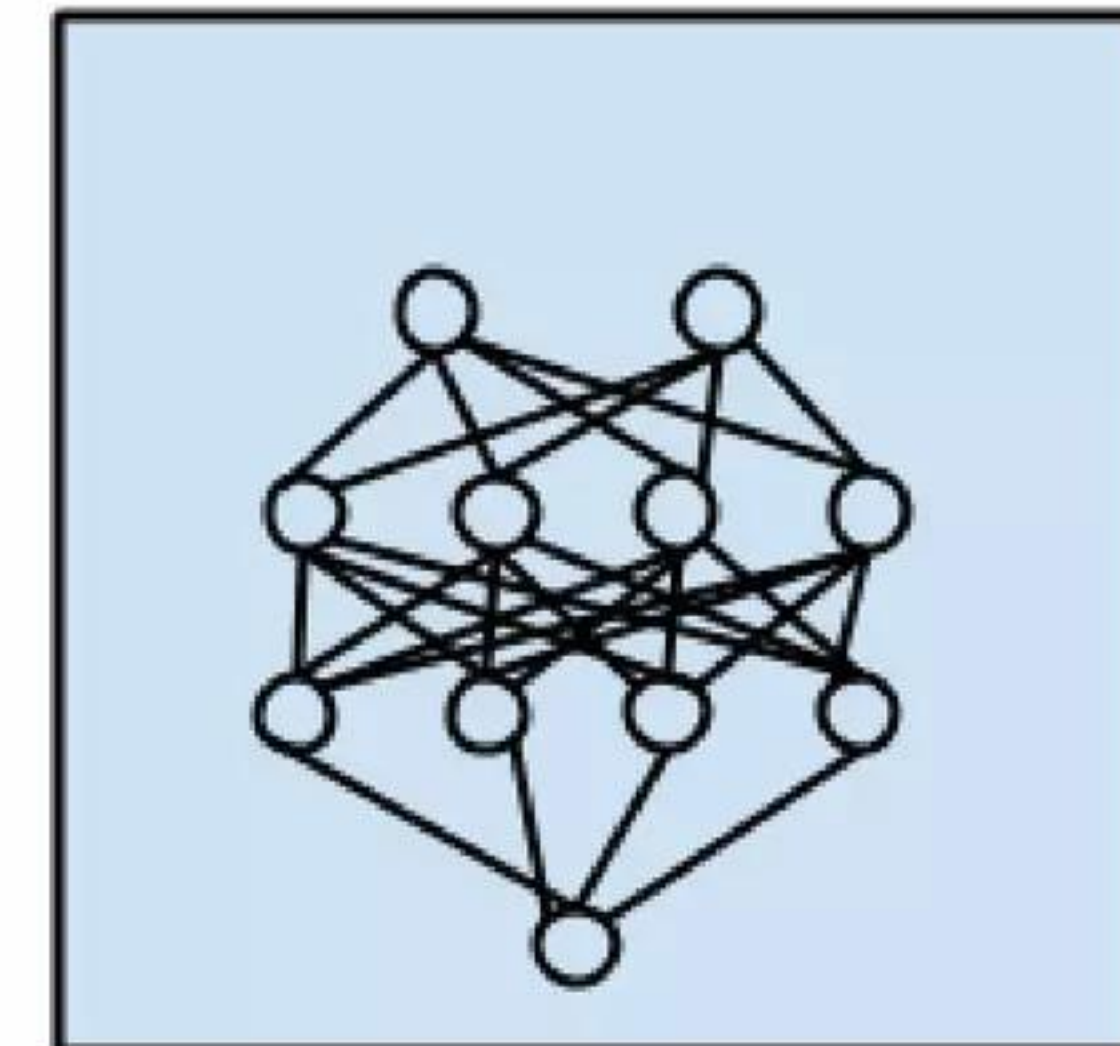
“Automation of Information” –
Prof. Dr. Ing. Iping Supriana

“A computer program is said to learn from experience E with respect to some class of tasks T and performance measure P , if its performance at tasks in T , as measured by P , improves with experience E .” – Prof. Tom Mitchel

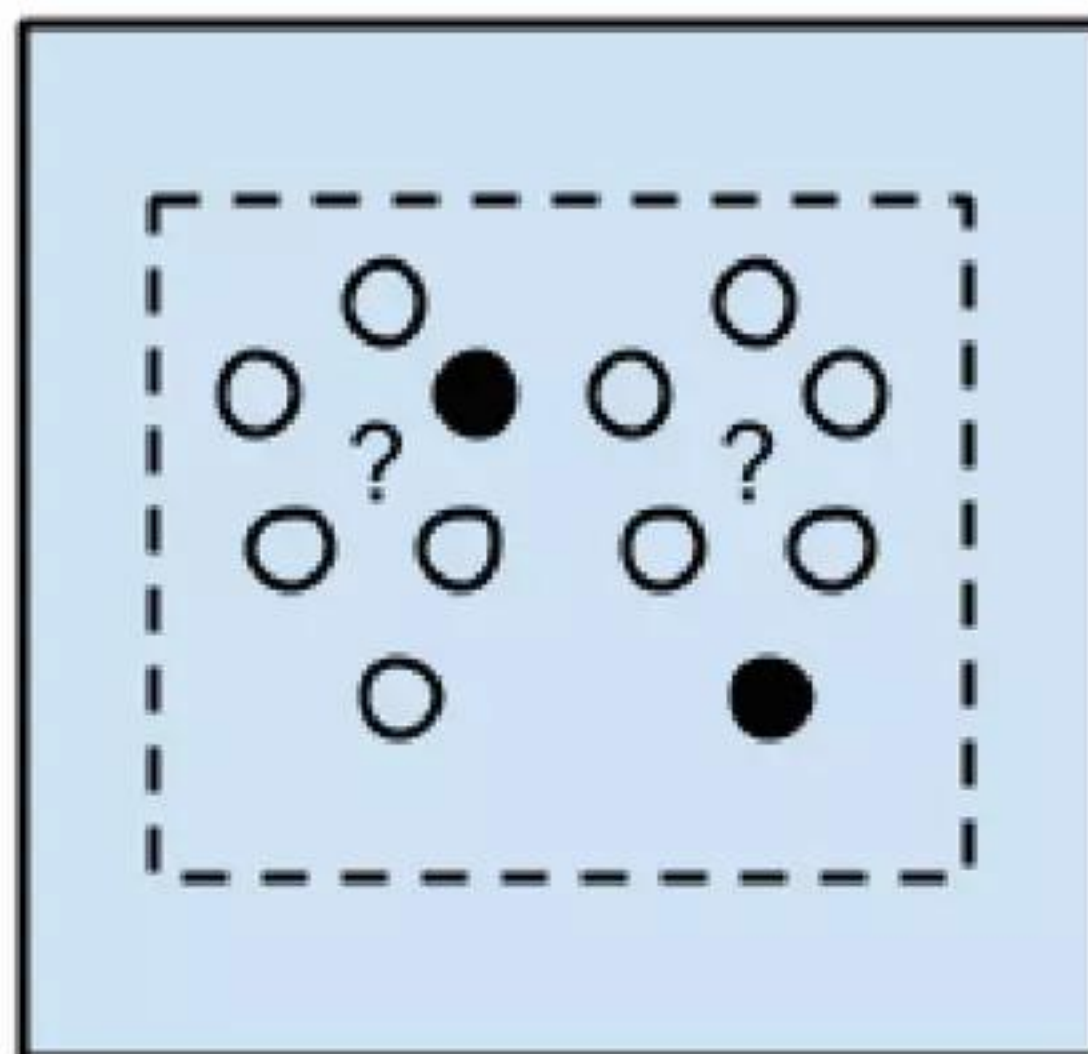


count the error $(y - y')$
Then minimize the error
or
maximize the likelihood

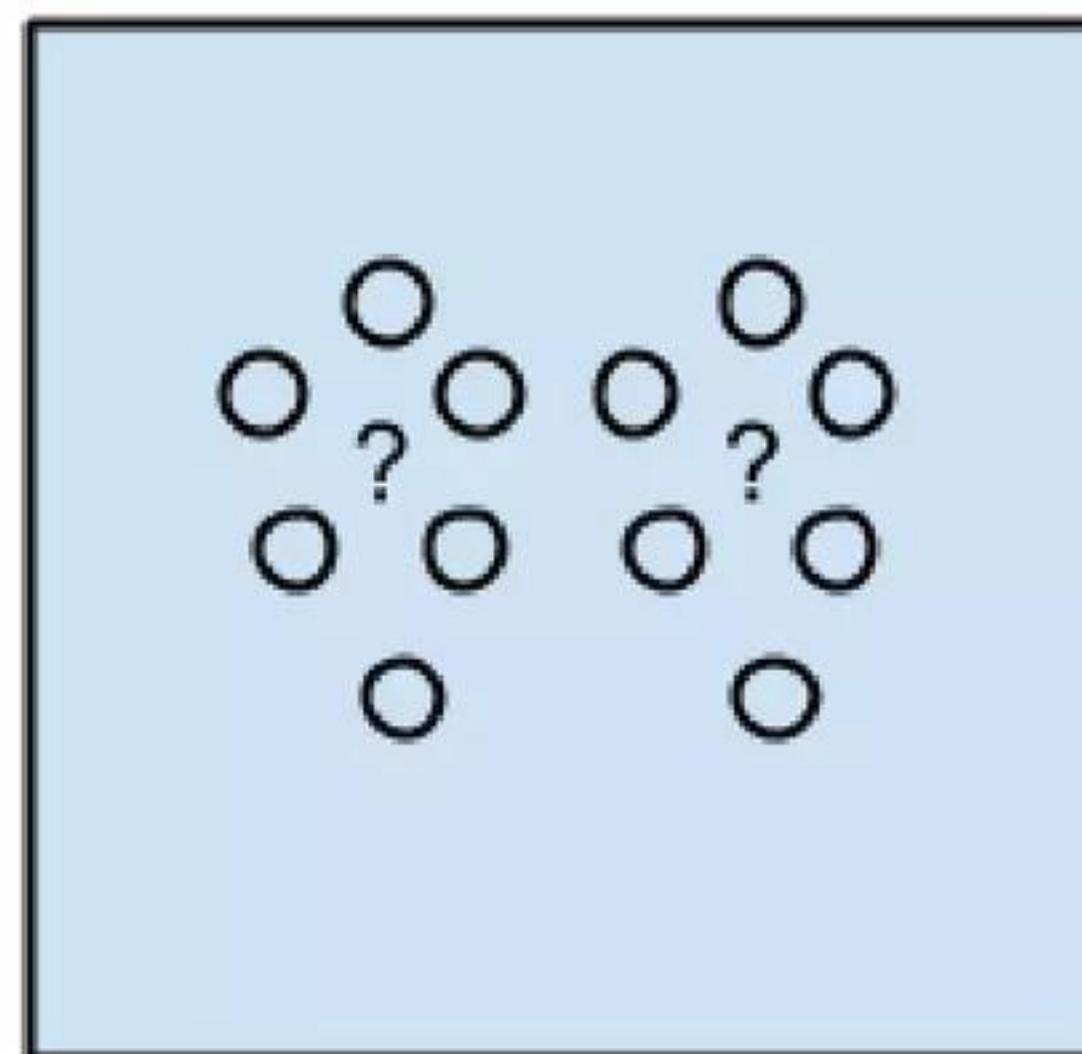
- Supervised
- Unsupervised
- Reinforcement Learning
- Semi-Supervised
- Deep Learning



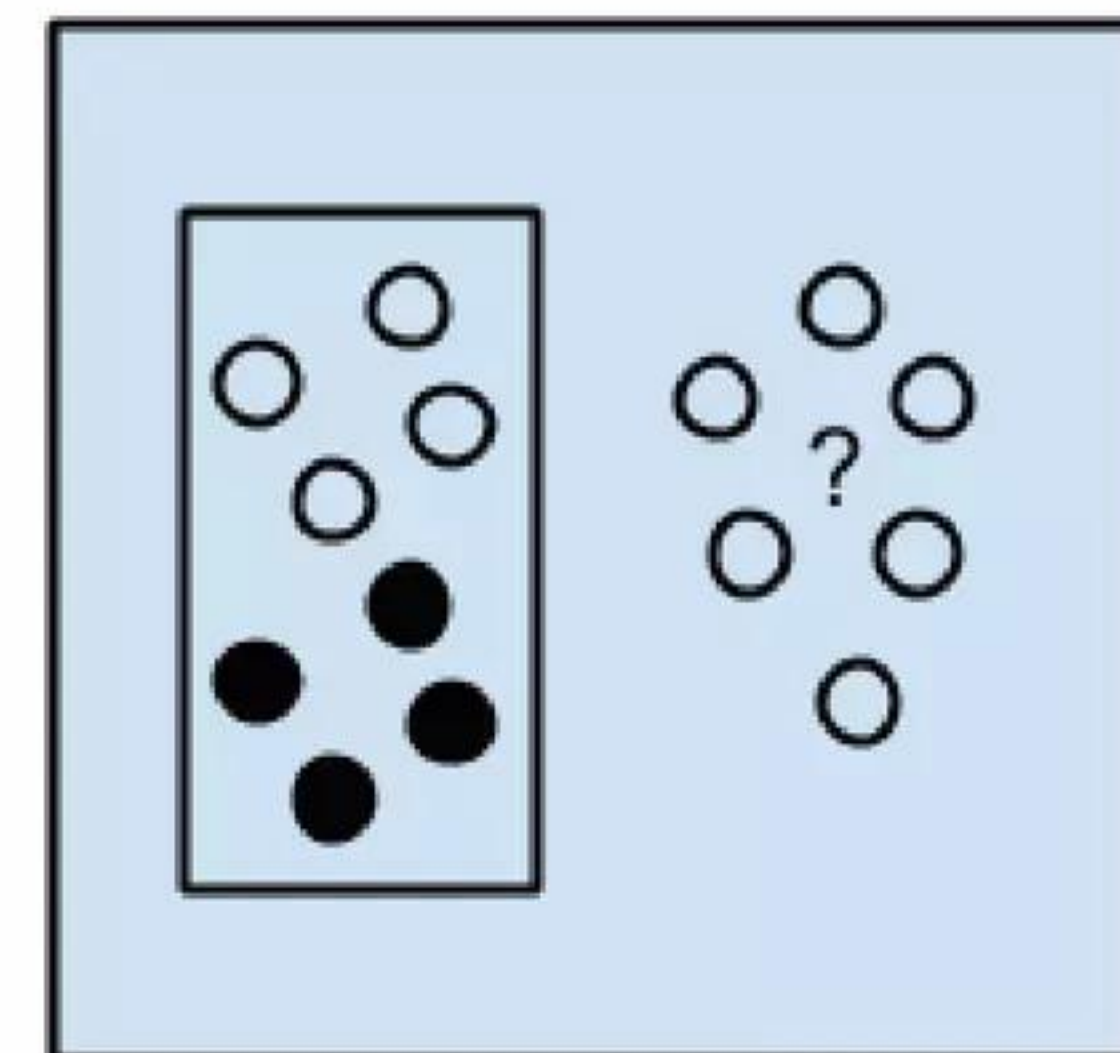
Deep Learning Algorithms



Semi-supervised Learning Algorithms

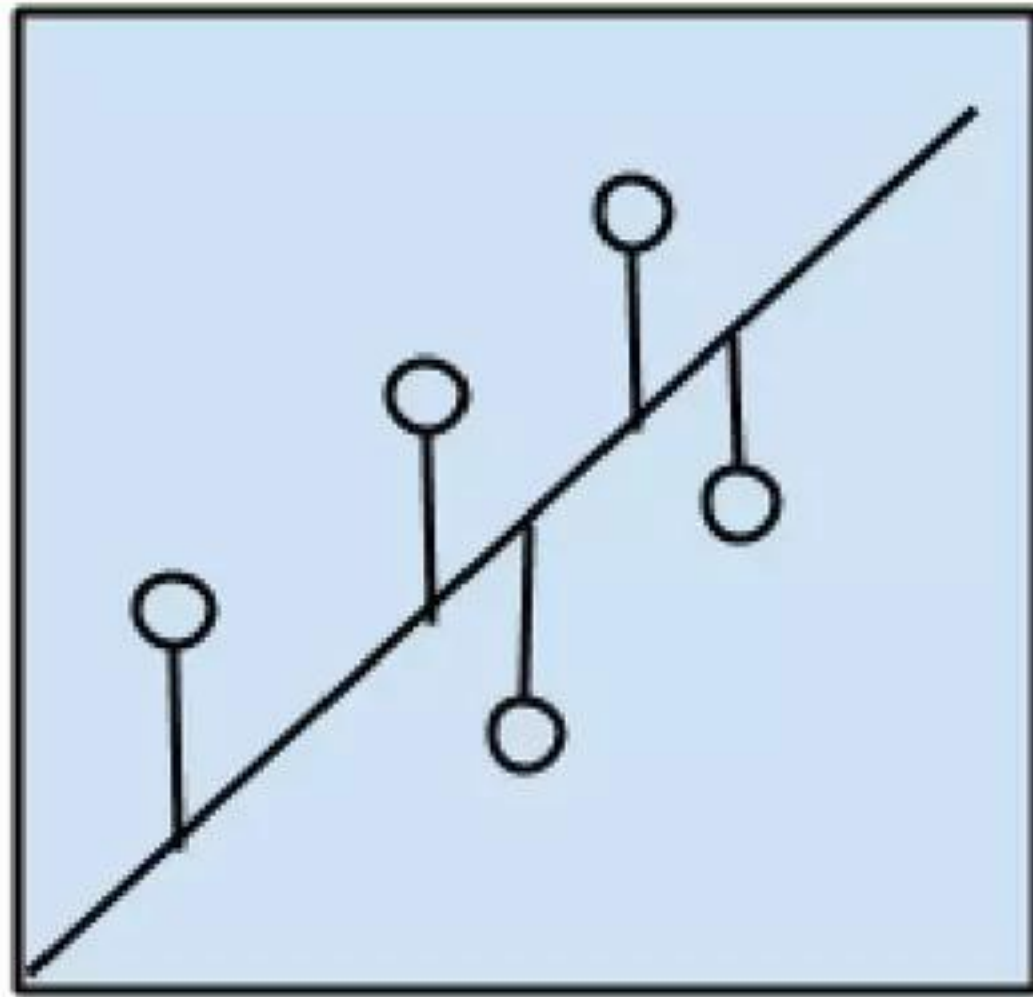


Unsupervised Learning Algorithms

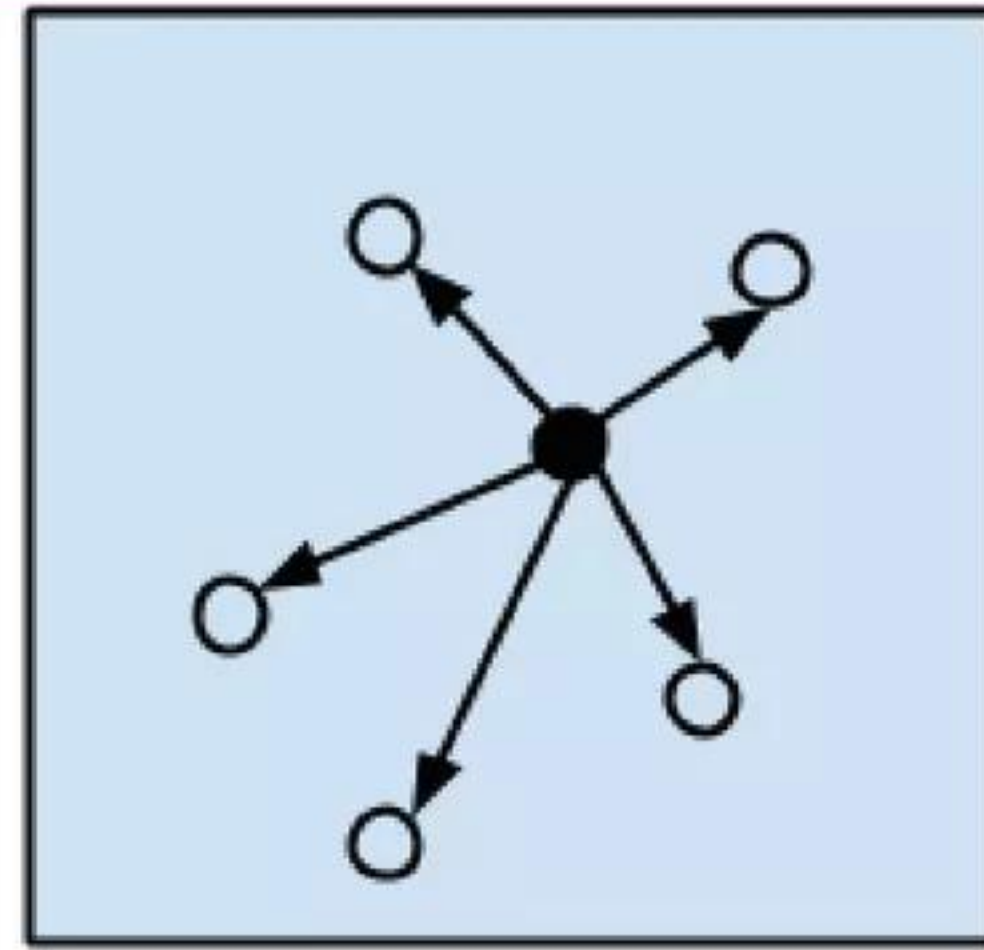


Supervised Learning Algorithms

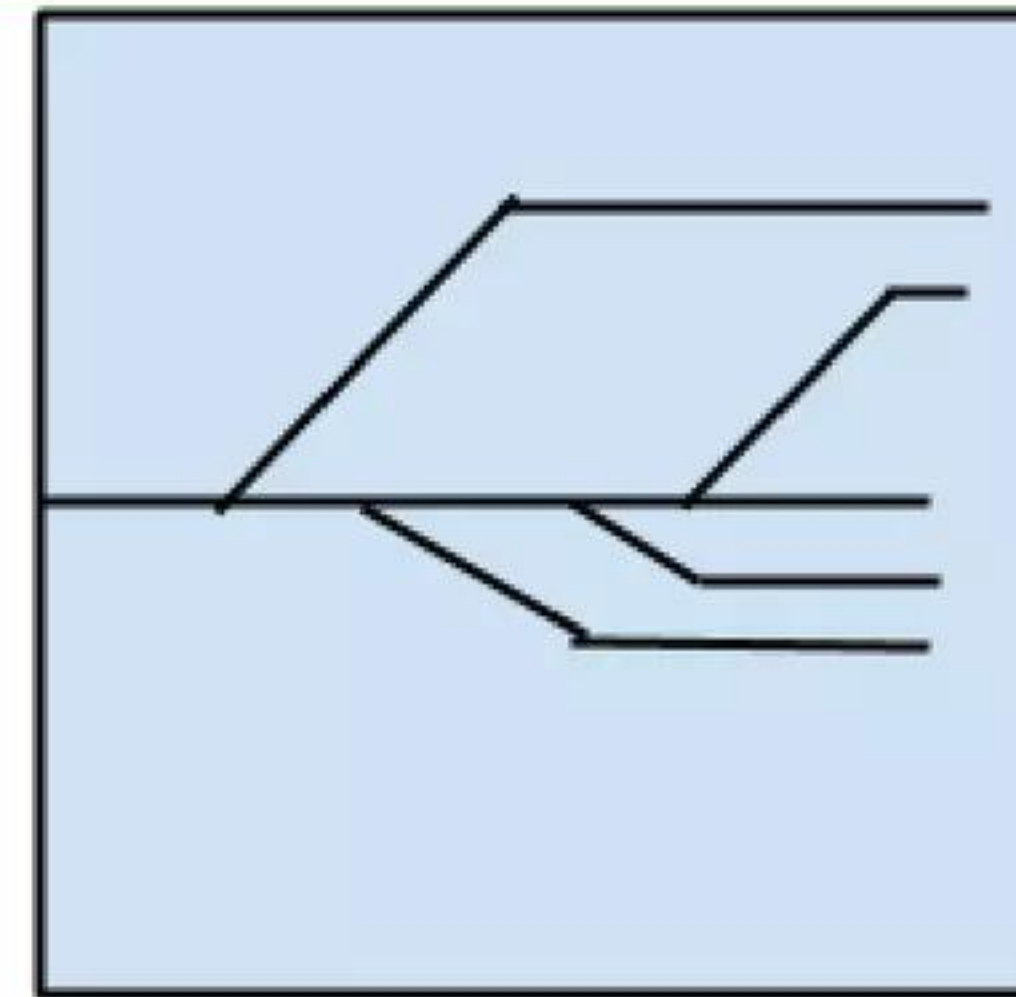
Machine Learning #2



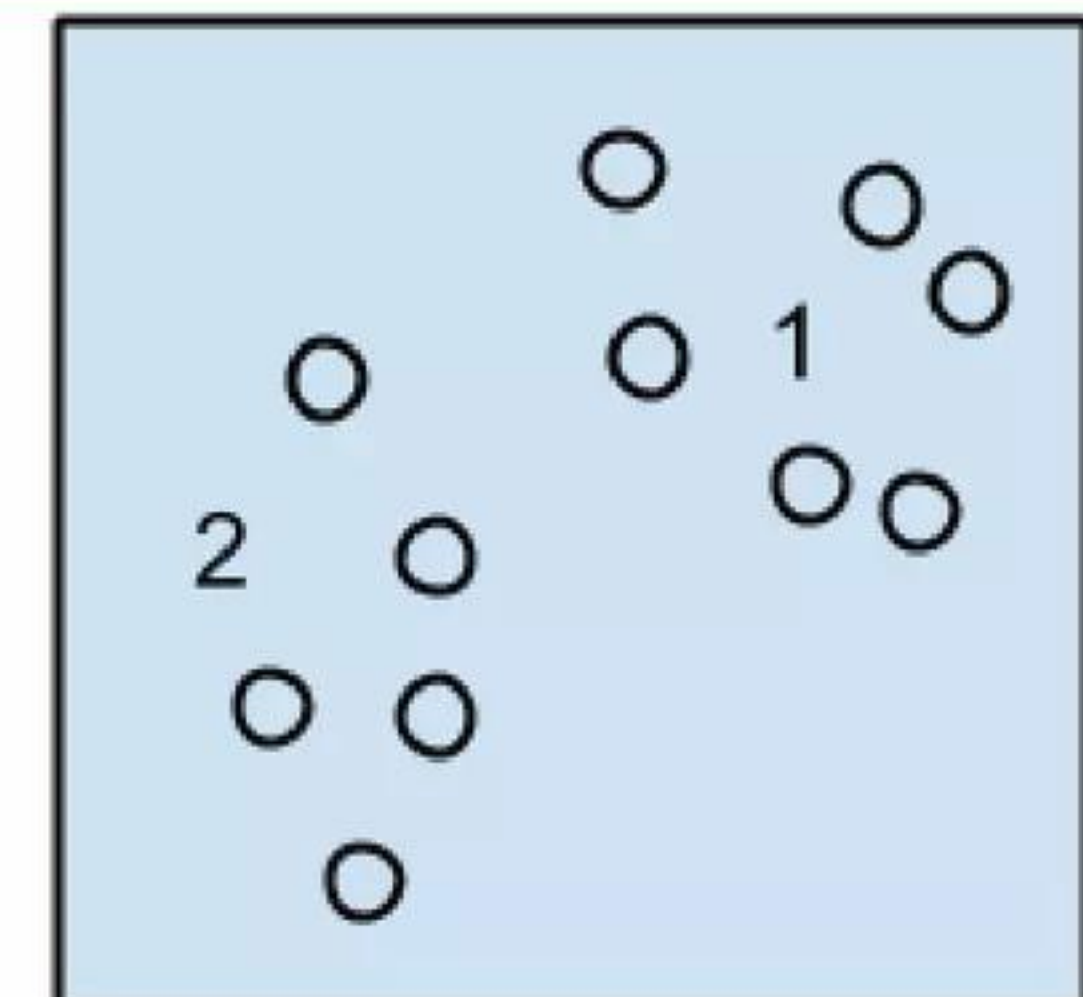
Regression Algorithms



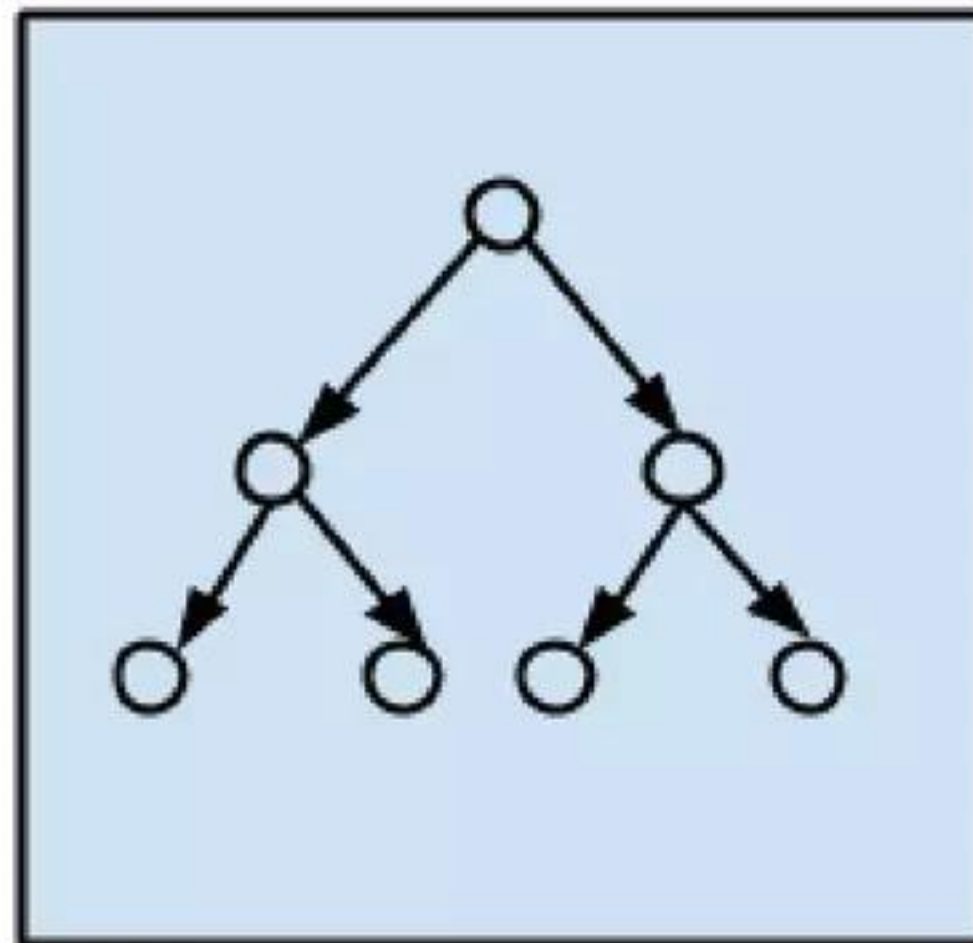
Instance-based Algorithms



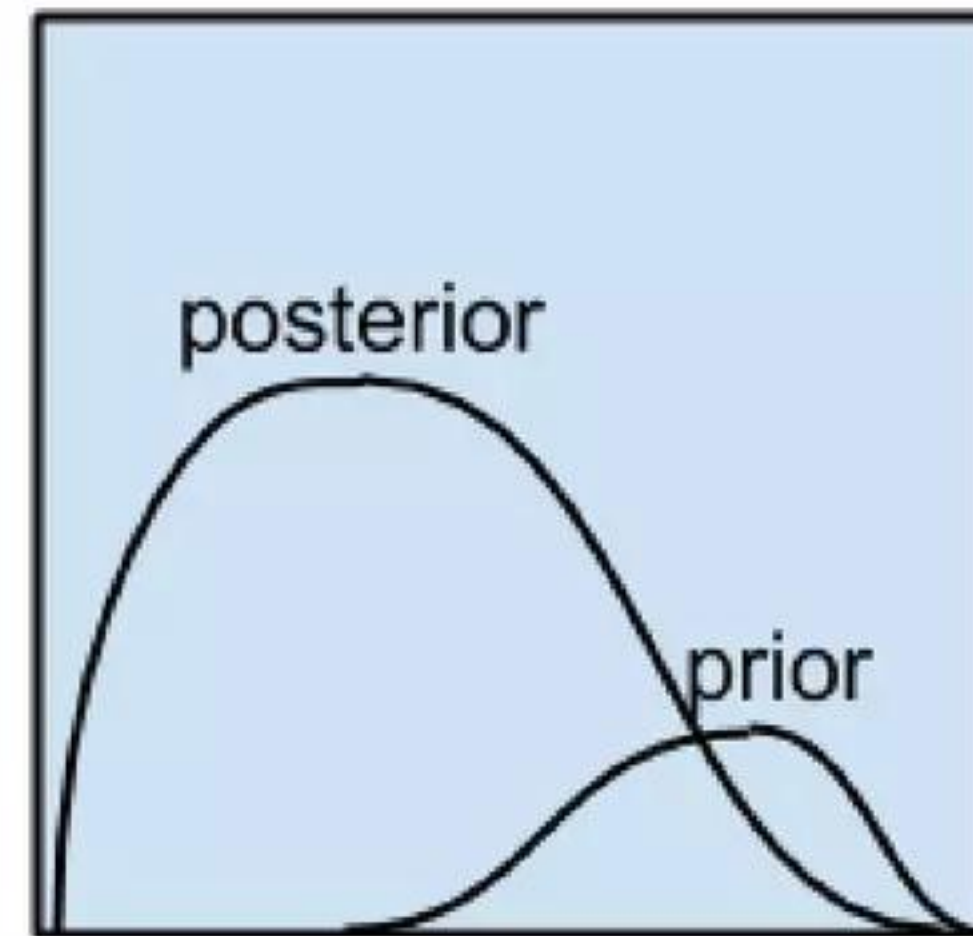
Regularization Algorithms



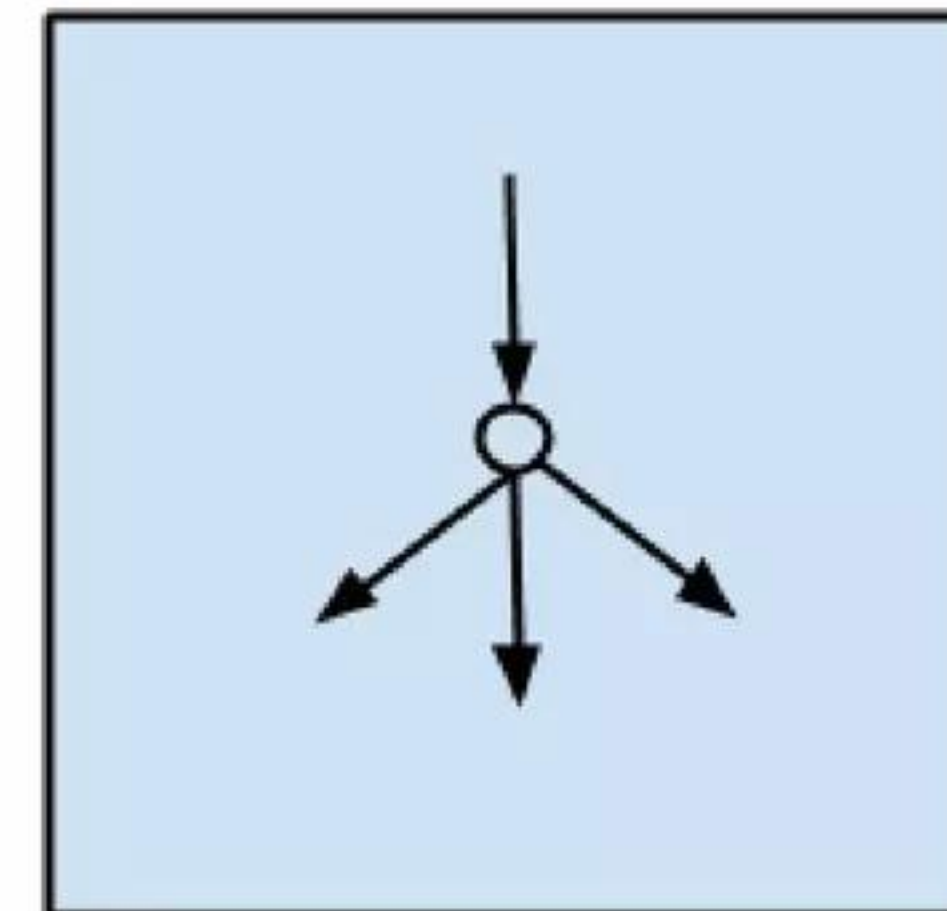
Clustering Algorithms



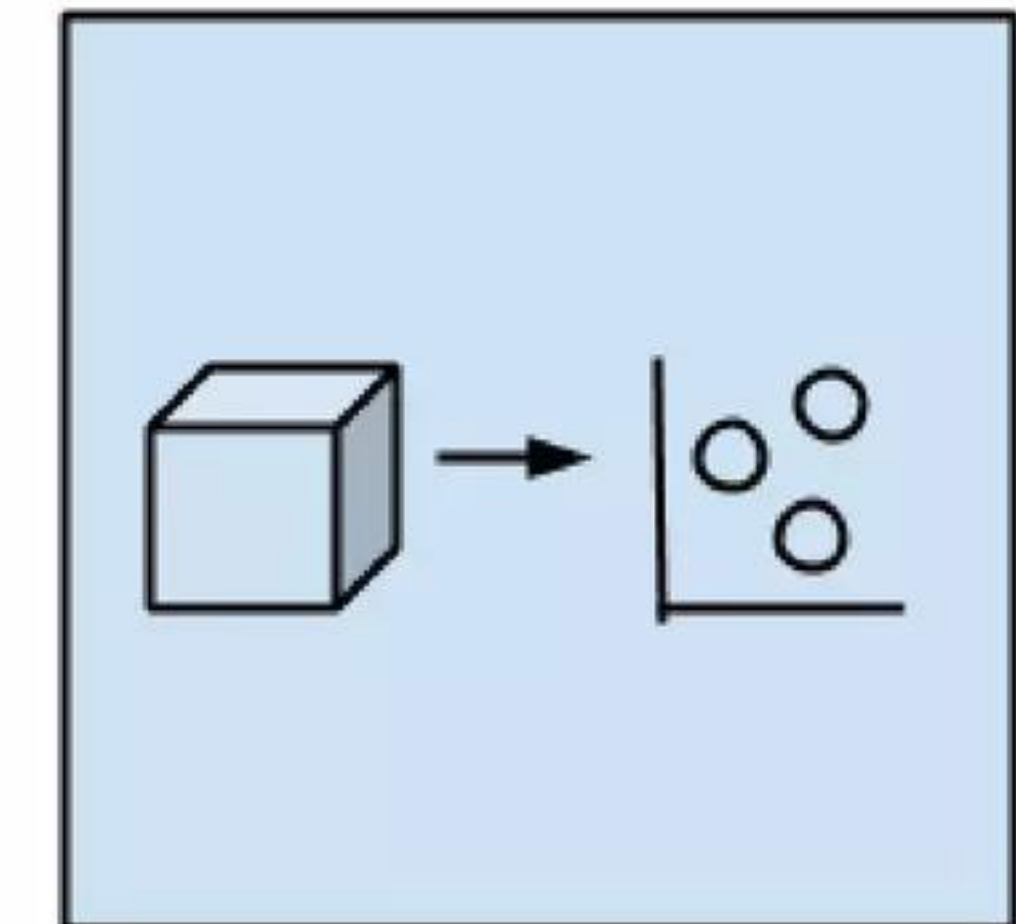
Decision Tree Algorithms



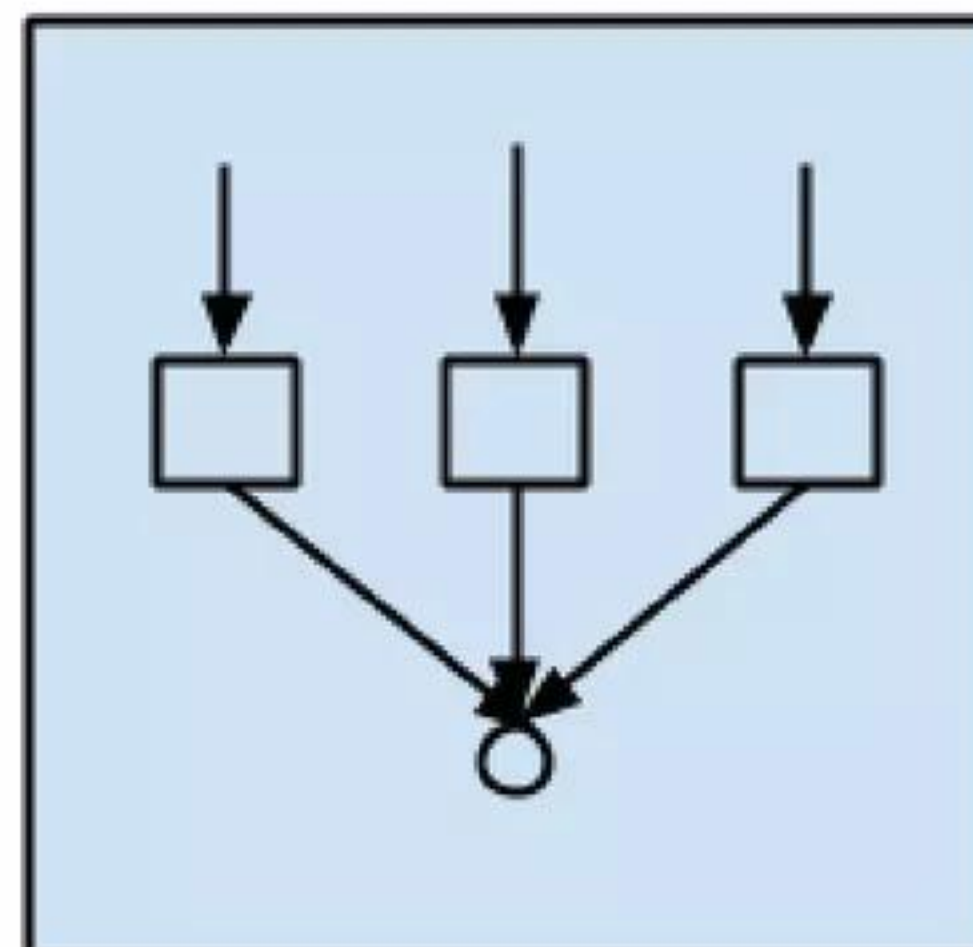
Bayesian Algorithms



Artificial Neural Network Algorithms



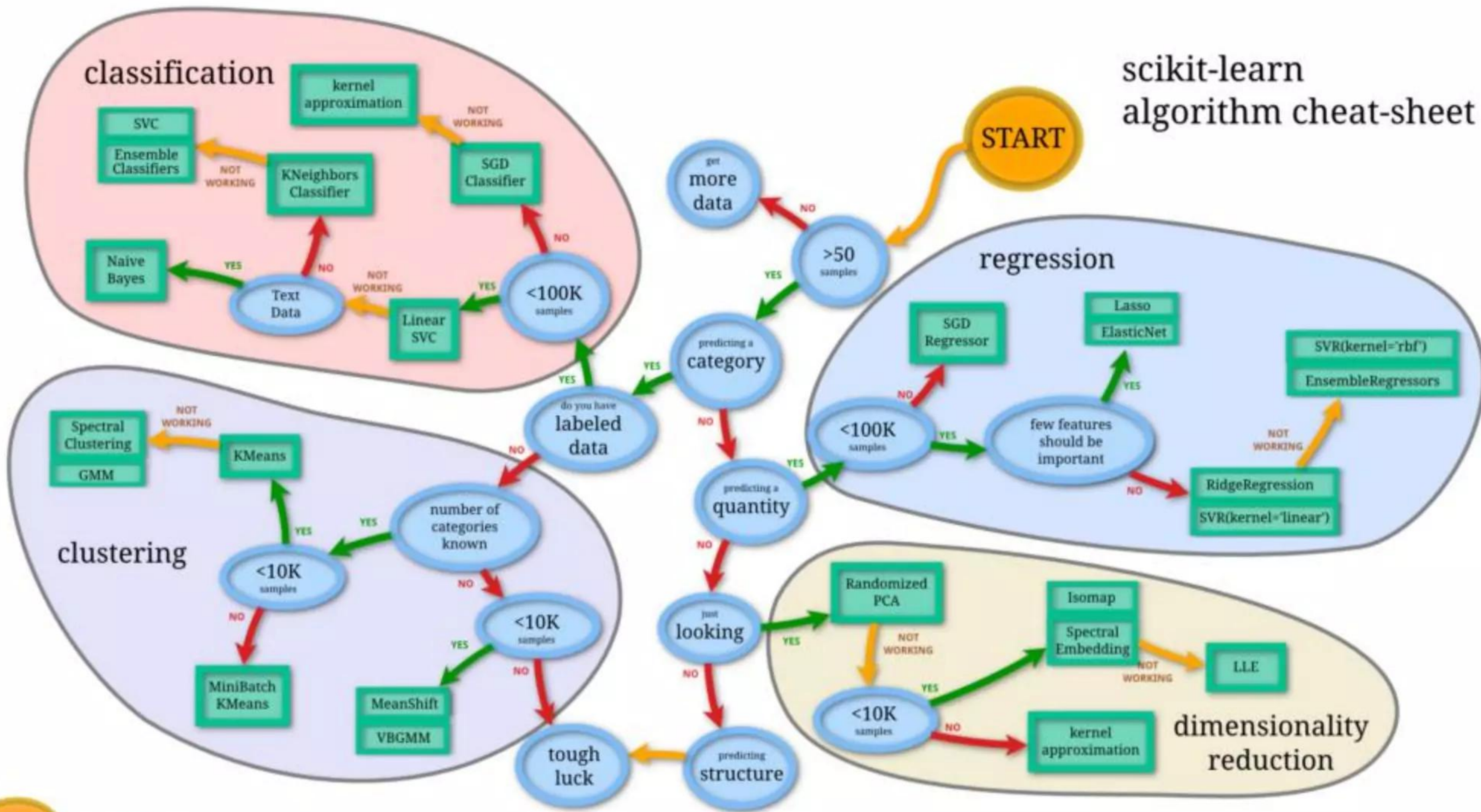
Dimensional Reduction Algorithms



Ensemble Algorithms

Machine Learning Taxonomy

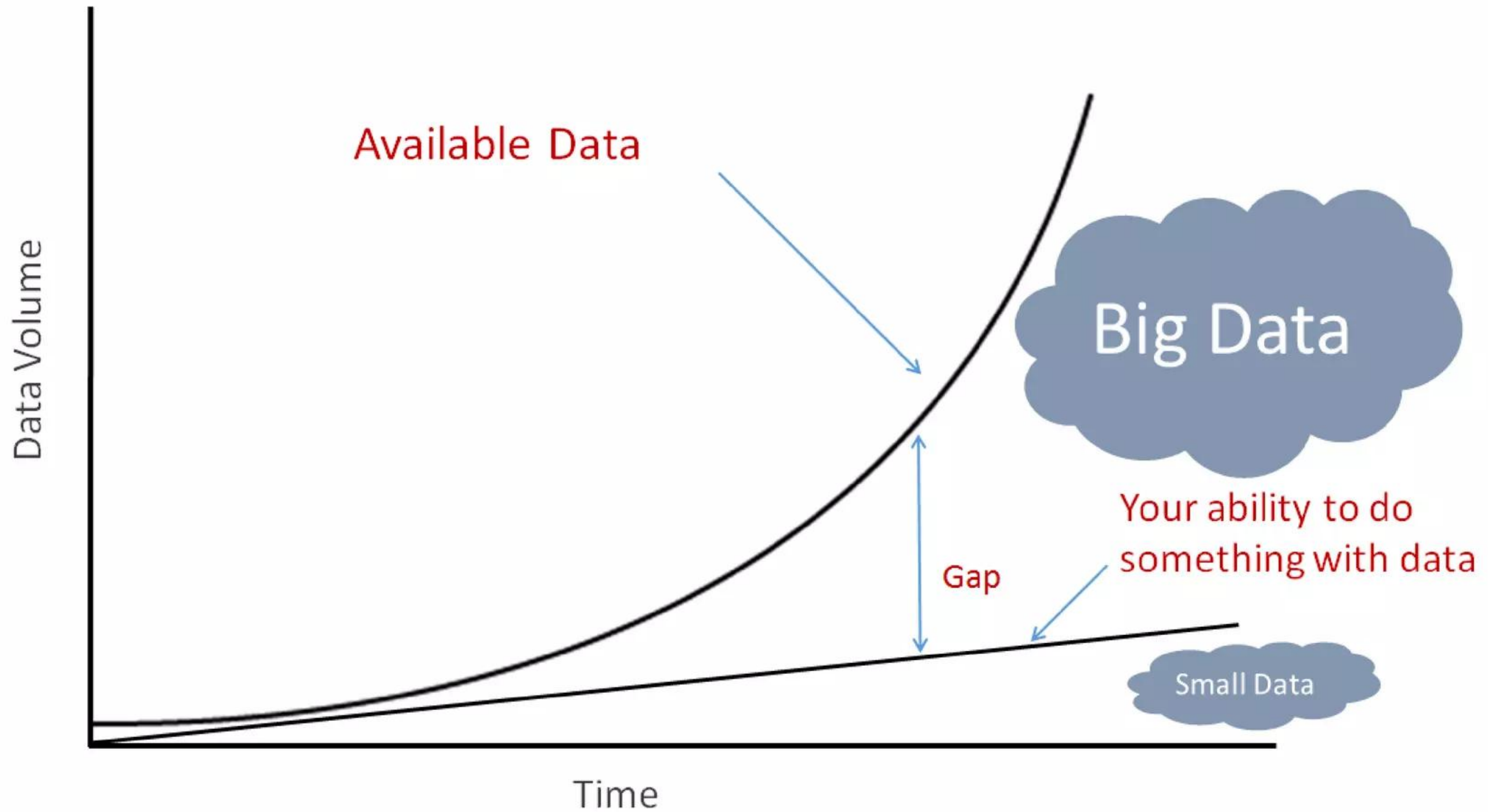
scikit-learn
algorithm cheat-sheet



- weka
- Deeplearning4j (working with spark and GPU)
- H2O (working with spark and GPU, support Tensorflow, MxNet, and cafe)
- jcuDNN (JNI for wrapping nVidia cuDNN)
- Mahout
- MLlib spark

- **Fraud Detection System**
- **Dynamic Recommendation System and User Profiling**
- **Traveling Salesman Problem and Binpacking Problem for better warehouse management**
- **Social Media Analysis**
- Chatbot
- Company condition forecasting
- Governance simulation

Data Explosion



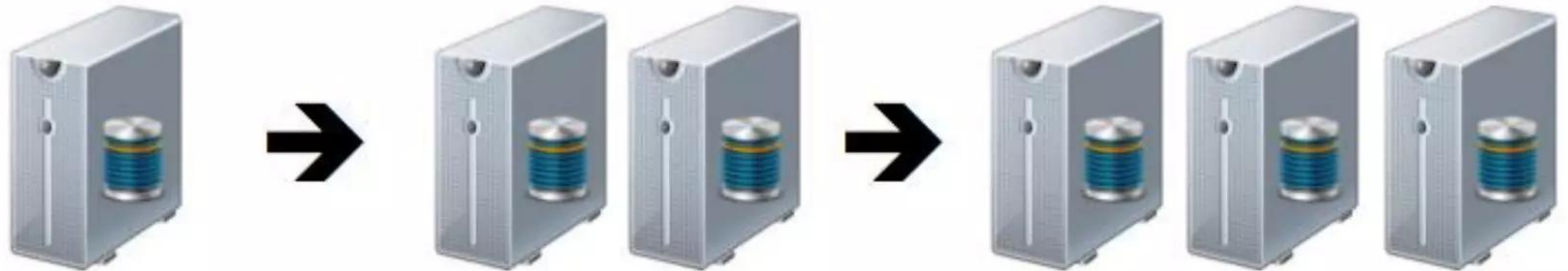
- Big data : volume, variety, velocity, and veracity. (You might consider a fifth V, value.)
- Knowledge representation or the architecture of the model
- Unimplemented methods/algorithms in any libraries
- **Stack of methods**
- Data mostly unlabeled data
- Data resources (microservices)
- Features Engineering (especially from unstructured data)
- Machines (Hardware)
- High Performance Computing

- More complex methods and models
- Methods characteristic & behavior
- Methods customization
- Ex. Semi-supervised, Deep learning, features engineering
- Sample cases : our research in FDS and dynamic recommendation system

Scale-Up



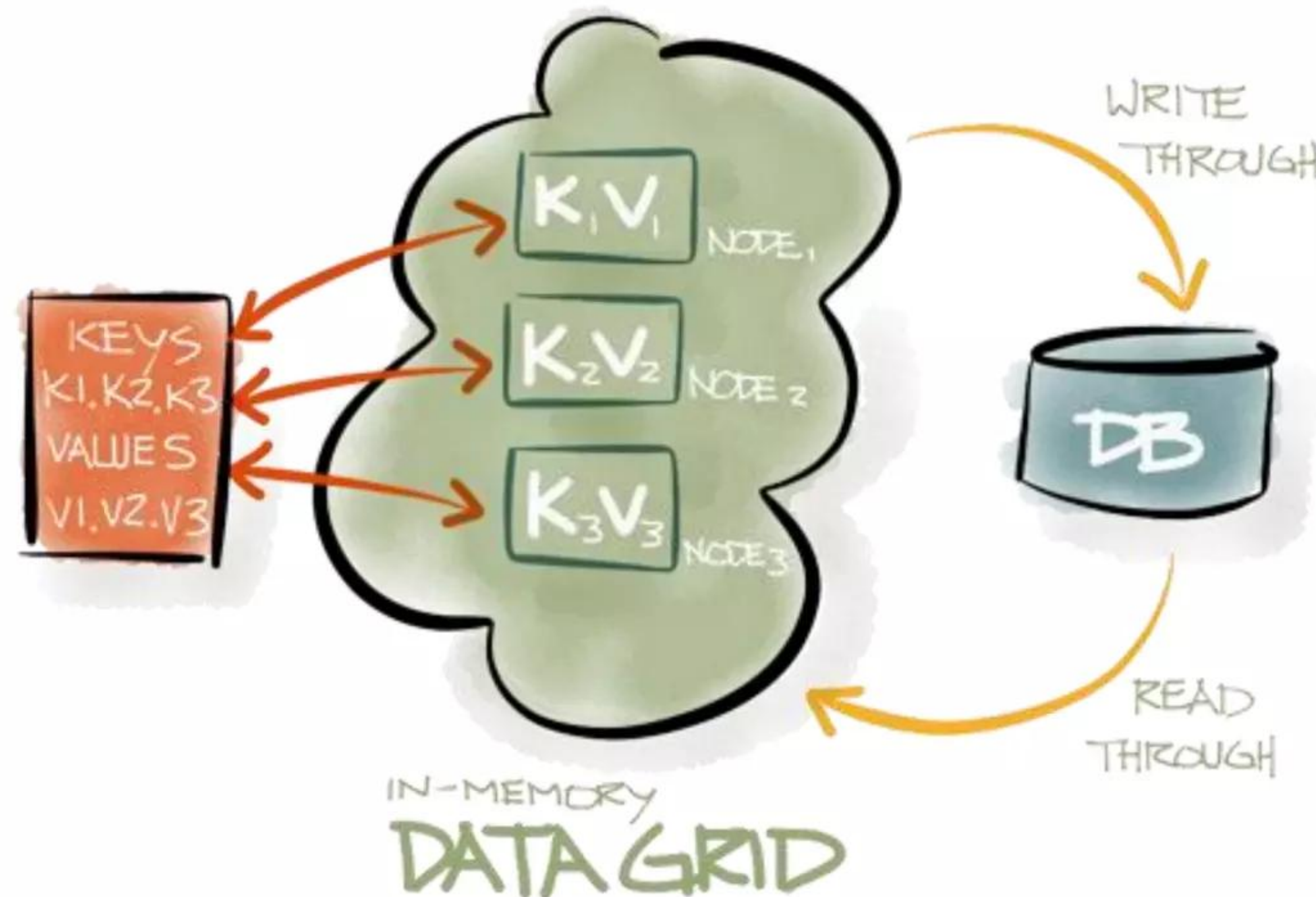
Scale-Out



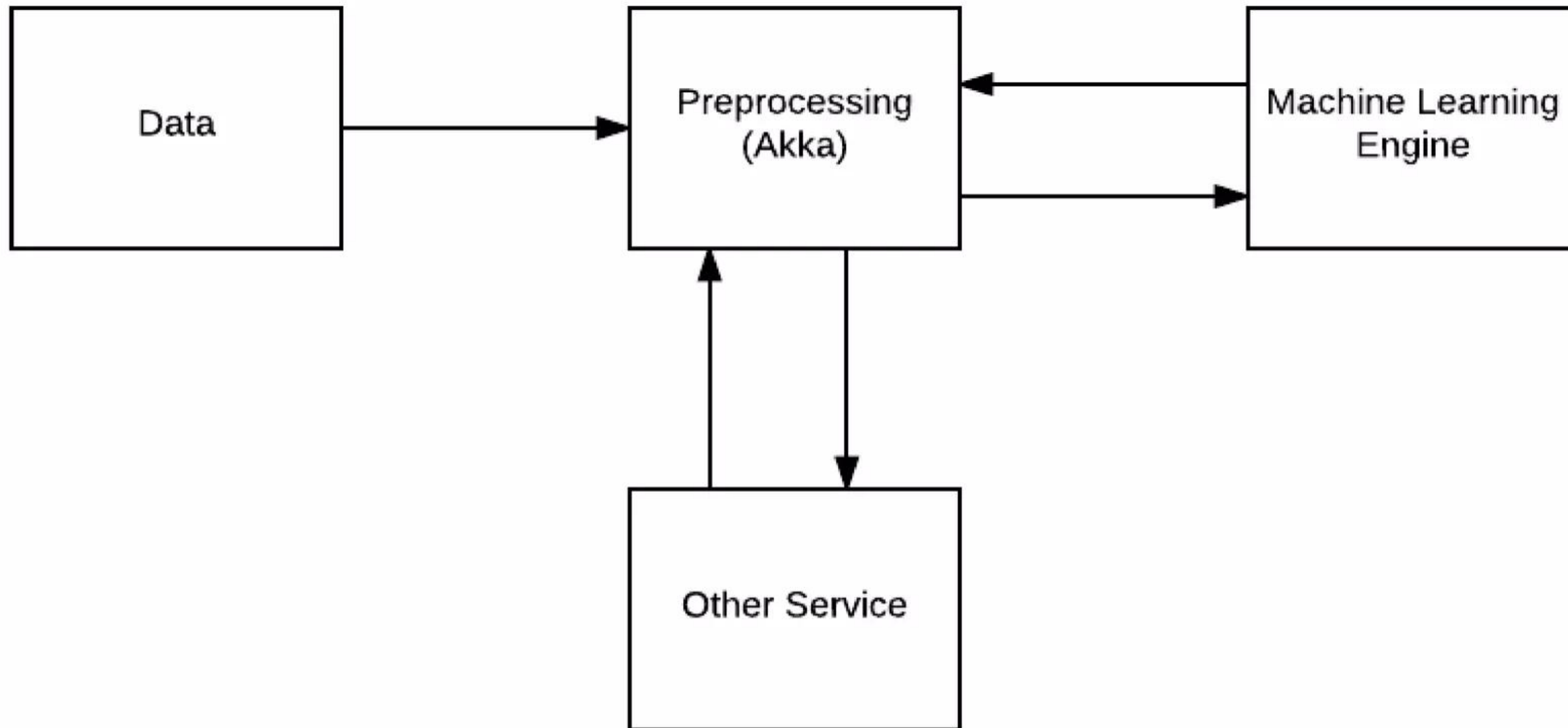
- **Cluster machine**
- **GPU machines (OpenCL and nvidia CUDA)**



- **In-memory data fabric:** provides low-latency access and processing of large quantities of data by distributing data across the dynamic random access memory (DRAM), Flash, or SSD of a distributed computer system.



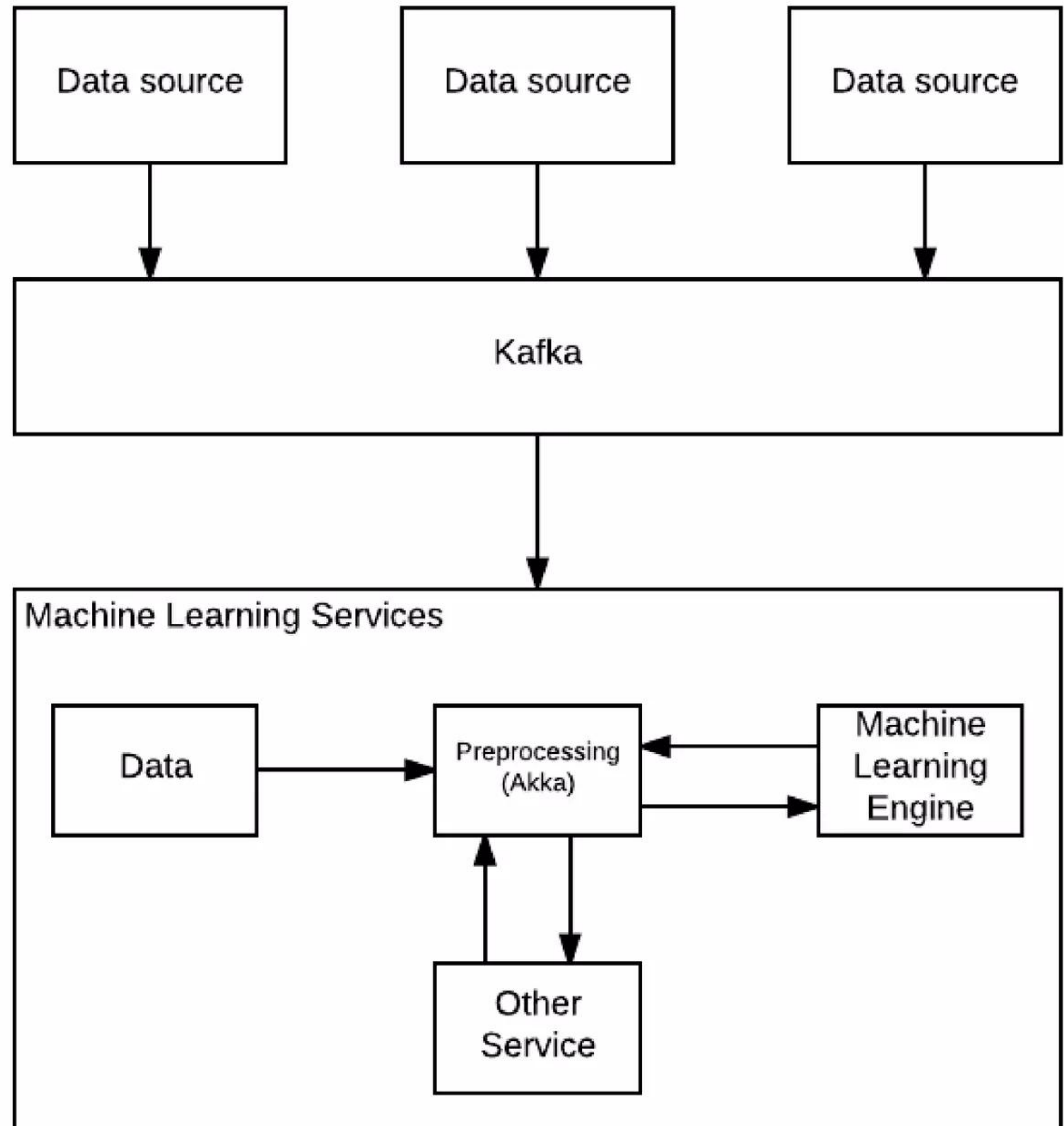
Scala + Play + Akka + Mongo



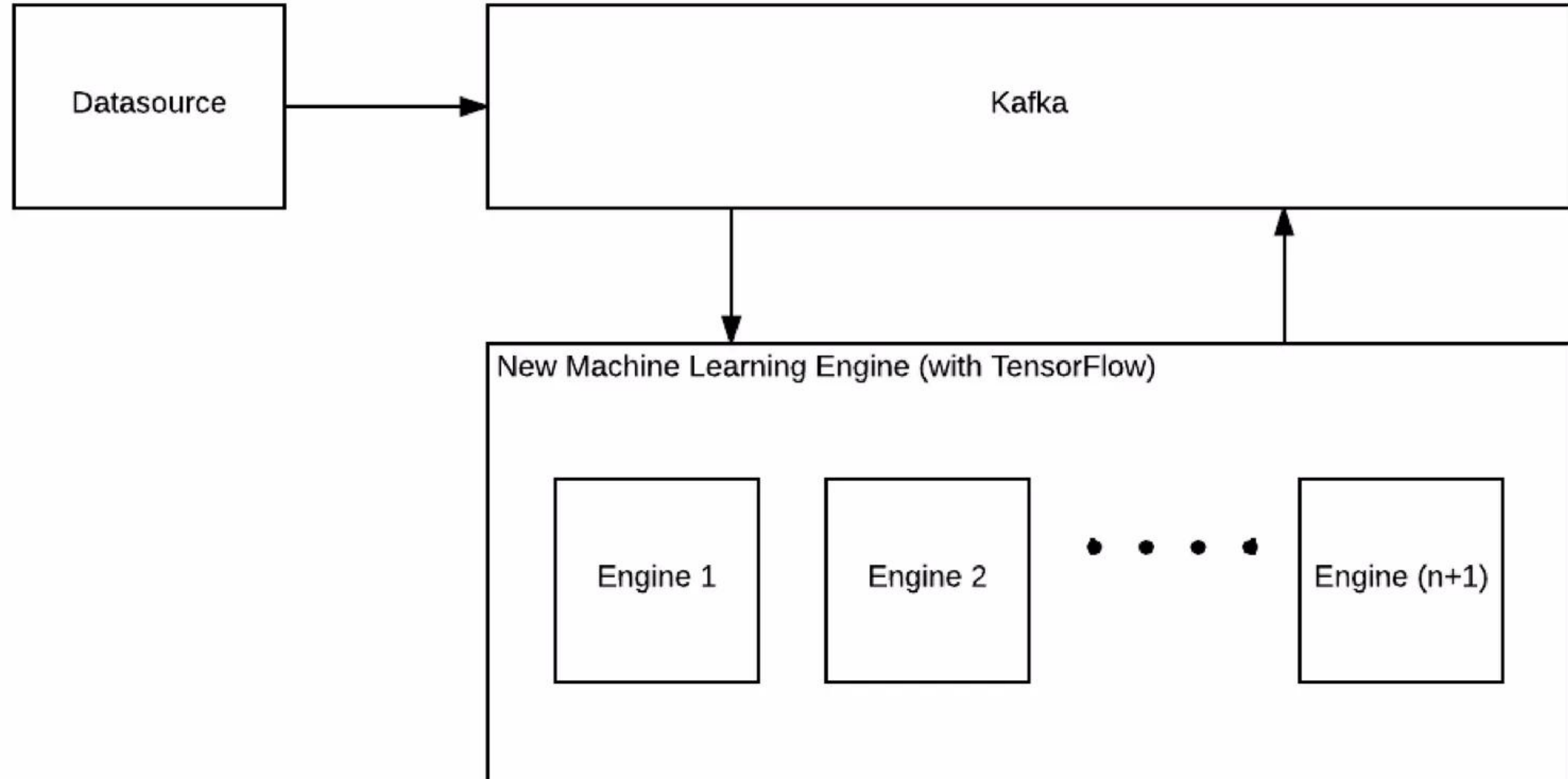
The problem is microservices

Next Machine Learning Flow

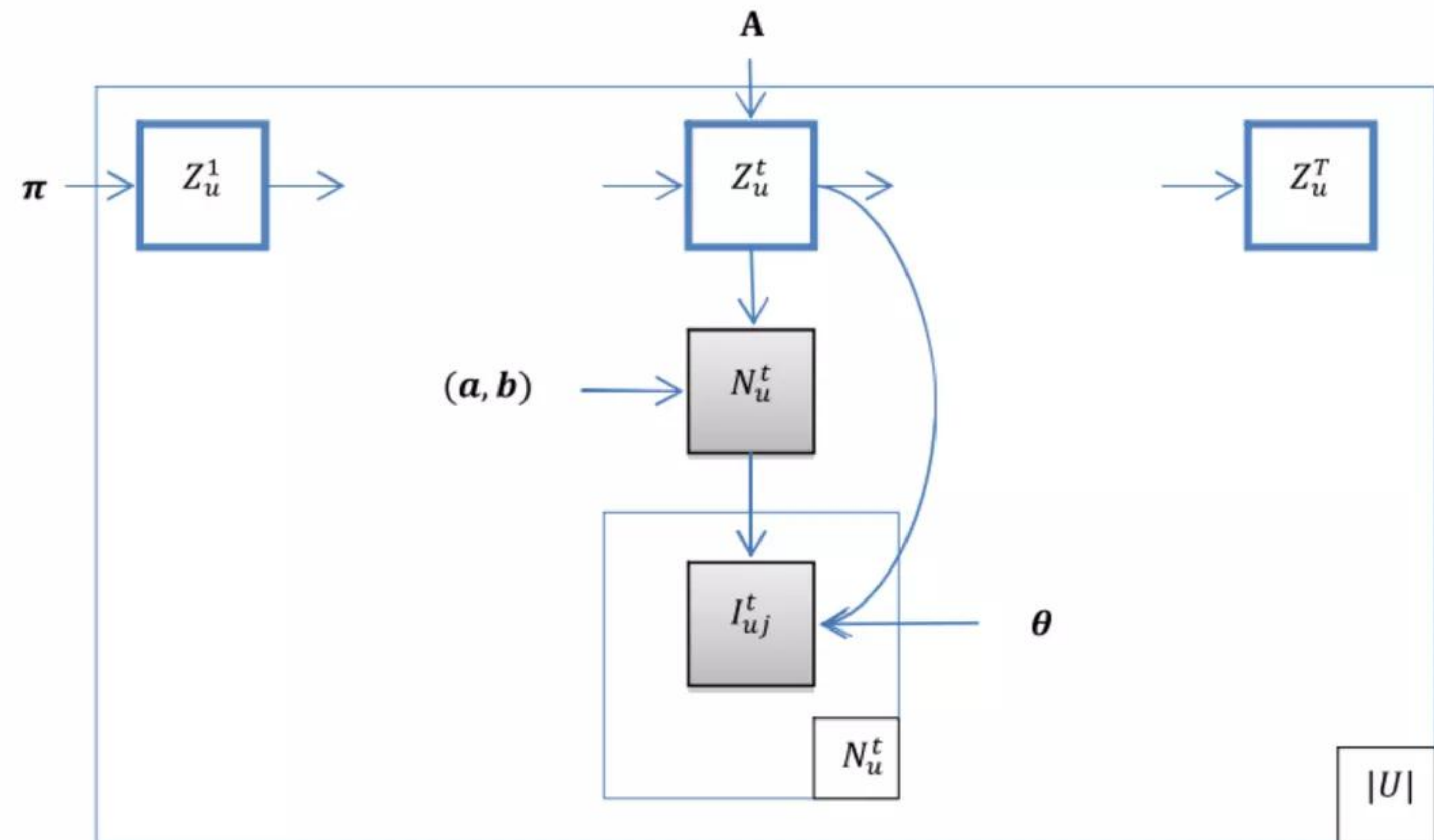
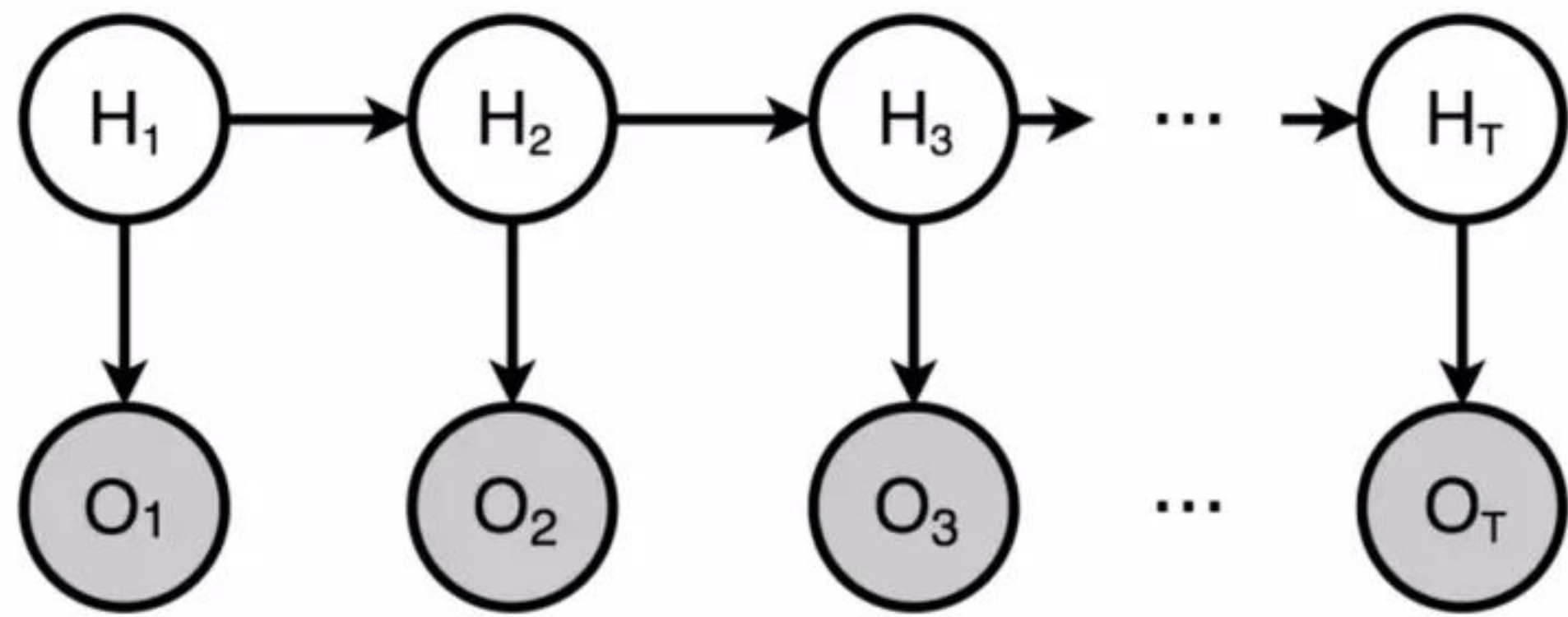
Scala + Play +
Akka + Mongo +
kafka



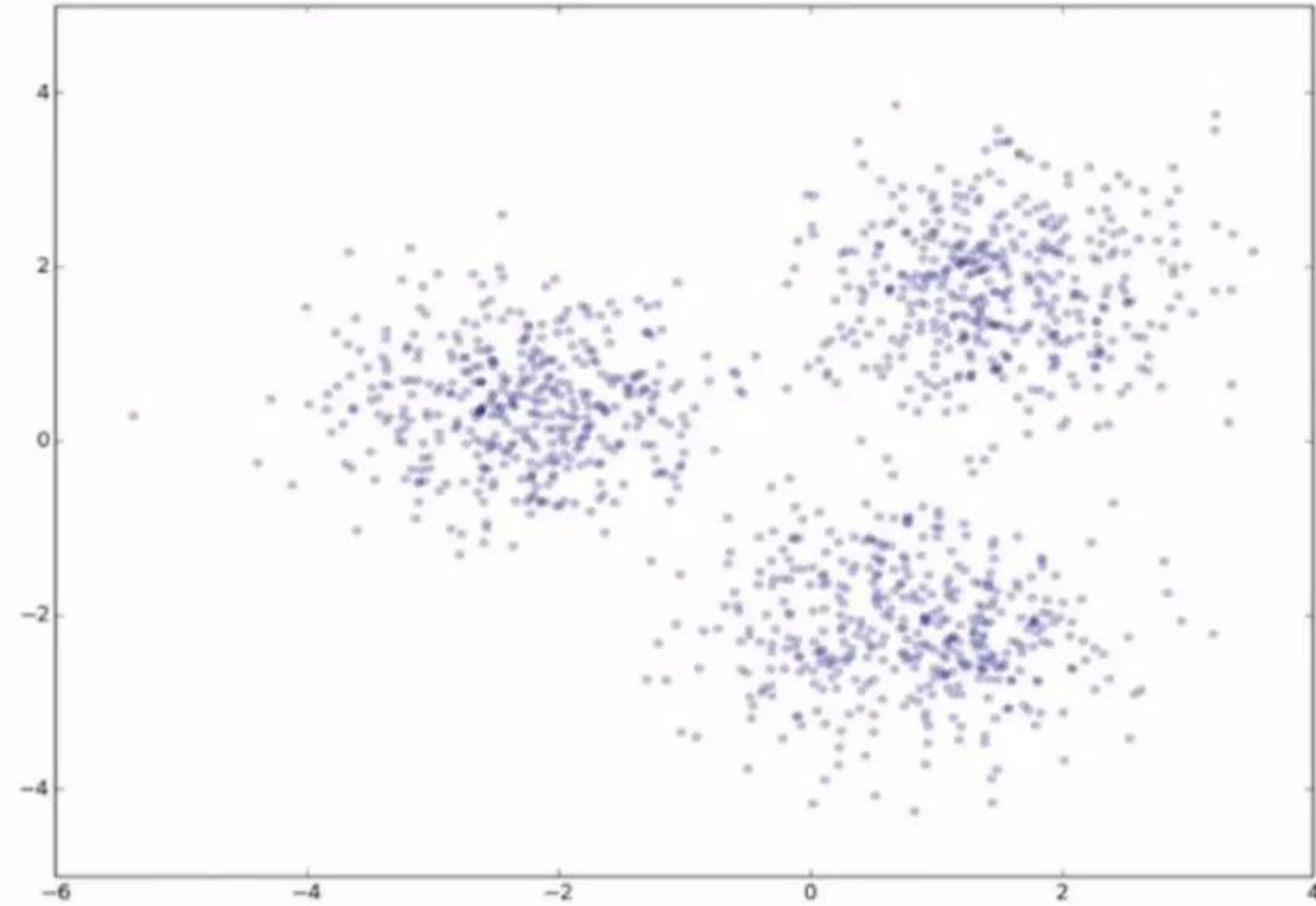
- Kafka
- Tensorflow (utilize GPU)
- Spark
- Java/Scala/python



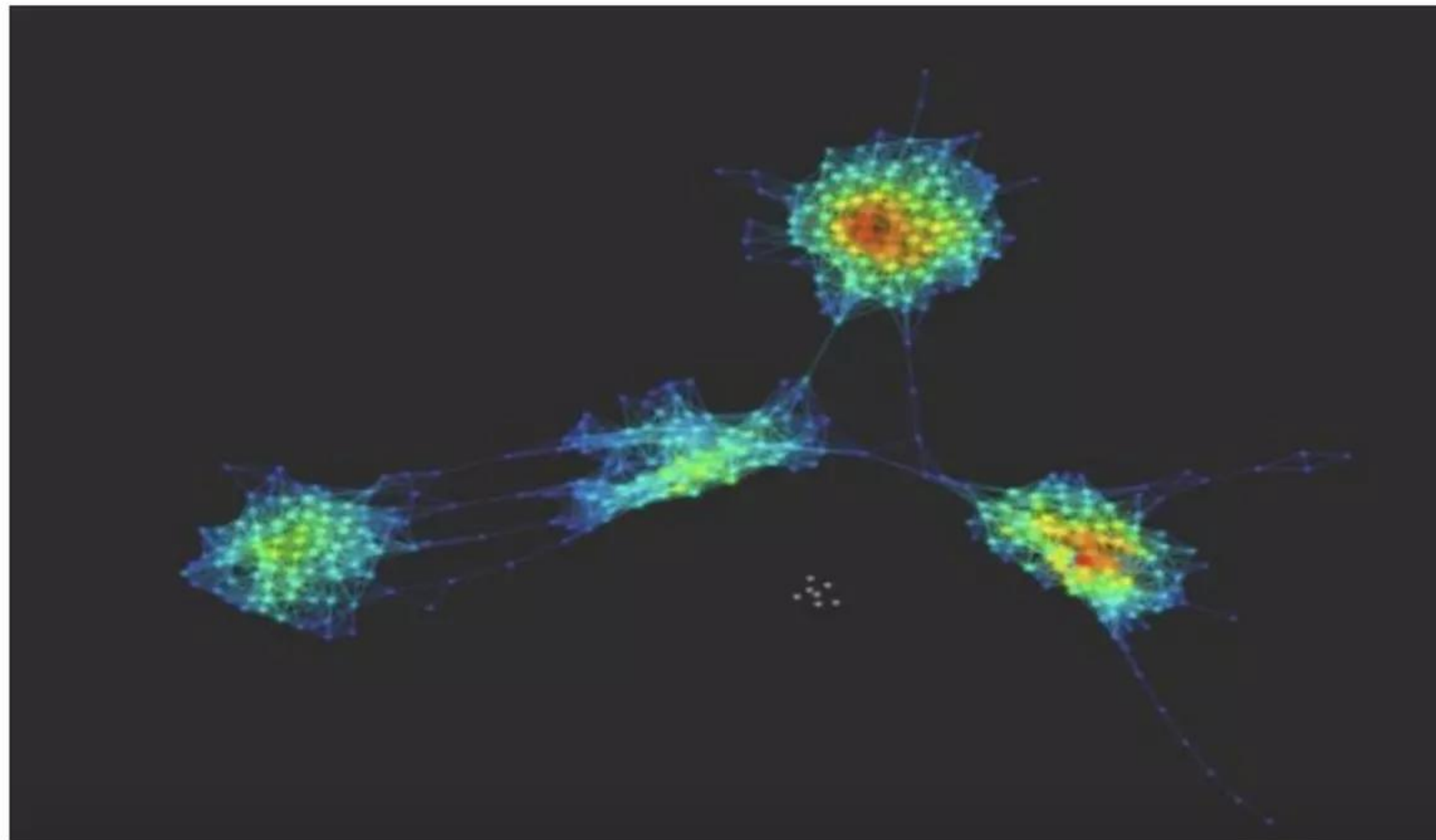
Dynamic Collaborative Filtering using HMM

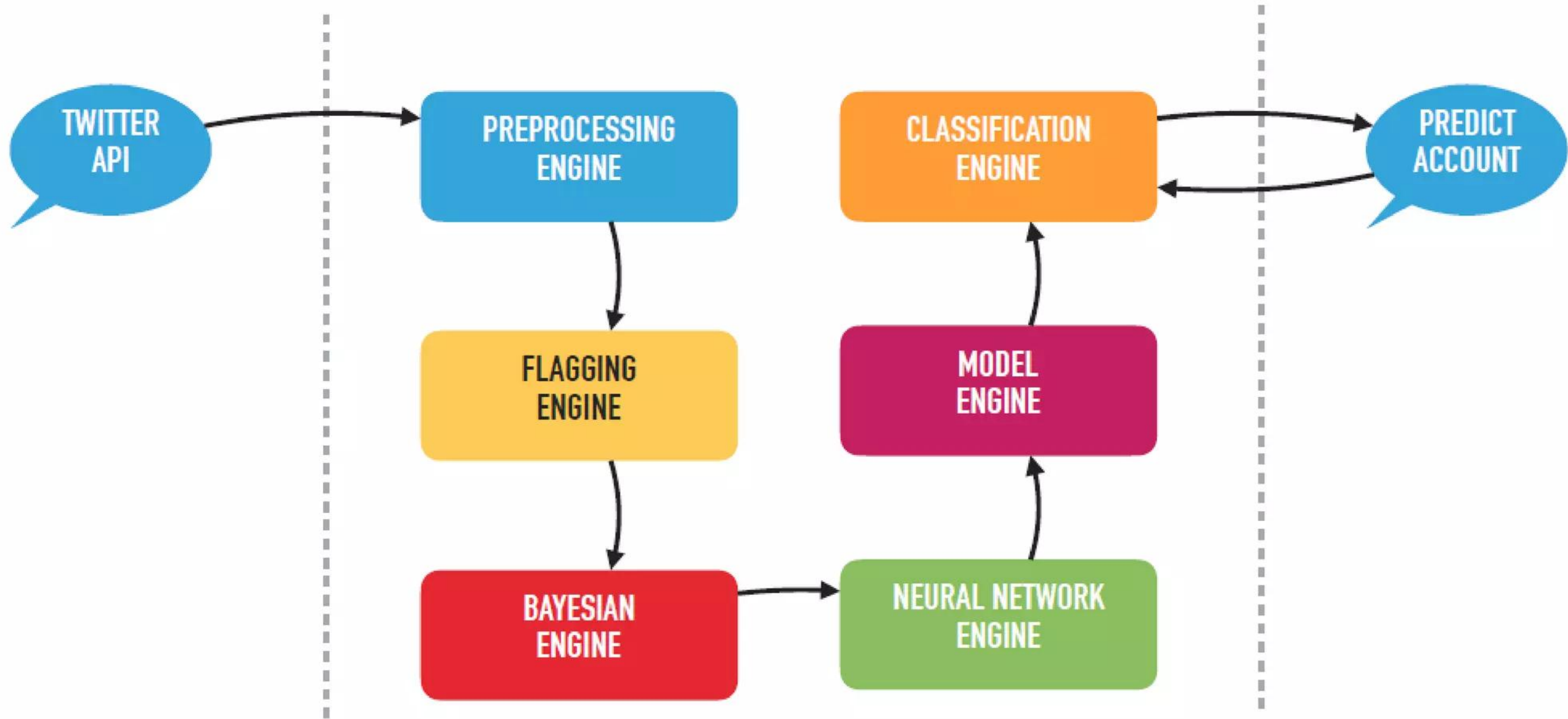


Algorithm	Control Parameters
<ol style="list-style-type: none"> 1. Use data collected over time period $1 \dots t_{trn}$ for training 2. Initialize $\pi, \mathbf{A}, (\mathbf{a}, \mathbf{b}), \theta$ to small random values 3. E-step: compute $P(Z_u^t I_u^{1:T})$ and $P(Z_u^{t-1}, Z_u^t I_u^{1:T})$ using Equations (2) and (3). 4. M-step: estimate π, \mathbf{A}, θ using Equations (7), (8) and (10). Estimate (\mathbf{a}, \mathbf{b}) using Section 2.1 of (Minka 2002) 5. If expected log likelihood has not converged go to step 2 6. For each user u compute $R(i, u)$ of each item i for time period $t_{trn} + 1$ using Equation (12) <ol style="list-style-type: none"> a. Recommend top N items with highest $R(i, u)$ 	<ol style="list-style-type: none"> 1. Set K to a value that maximizes the AIC score 2. Set length of the time period to 1 month <ol style="list-style-type: none"> a. Smaller if user preferences change quickly b. Larger if constrained by computing resource



- Coordinate invariance
- Deformation Invariance
- Compressed Representations





AVERI PROJECT

twitter prostitution account detection



THANK YOU

Any question?