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Artificial Intelligence and The Complexity

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- Sr. Research and Development Engineer at blibli.com (PT. Global Digital Niaga)
- Rnd Team for Data Science
- Before working for Fraud Detection System. Current working in Dynamic Recommendation System.
- Research area : deep learning, recommendation system, fraud detection system, econometrics, stochastic methods, social media analysis, and natural language processing.

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

- S. Rusel and P. Norvig, Artificial Intelligence in Modern Approach

<p>Thinking Humanly</p> <p>“The exciting new effort to make computers think . . . <i>machines with minds</i>, in the full and literal sense.” (Haugeland, 1985)</p> <p>“[The automation of] activities that we associate with human thinking, activities such as decision-making, problem solving, learning . . .” (Bellman, 1978)</p>	<p>Thinking Rationally</p> <p>“The study of mental faculties through the use of computational models.” (Charniak and McDermott, 1985)</p> <p>“The study of the computations that make it possible to perceive, reason, and act.” (Winston, 1992)</p>
<p>Acting Humanly</p> <p>“The art of creating machines that perform functions that require intelligence when performed by people.” (Kurzweil, 1990)</p> <p>“The study of how to make computers do things at which, at the moment, people are better.” (Rich and Knight, 1991)</p>	<p>Acting Rationally</p> <p>“Computational Intelligence is the study of the design of intelligent agents.” (Poole <i>et al.</i>, 1998)</p> <p>“AI . . . is concerned with intelligent behavior in artifacts.” (Nilsson, 1998)</p>

- Computer would need to possess the following capabilities :
 - Natural Language Processing
 - Knowledge Representation
 - Automated Reasoning
 - Machine Learning
- Turing test :
 - Computer vision : perceive objects
 - Robotics : manipulate objects and move

- We need to get inside the actual workings of human minds :
 - **Introspection**, trying to catch our own thoughts as they go by
 - **Psychological experiments**, observing a person in action
 - **Brain imaging**, observing the brain in action

- By 1965, programs existed that could, in principle, solve any solvable problem described in logical notation. (Although if no solution exists, the program might loop forever.)
- Logician tradition within artificial intelligence hopes to build on such programs to create intelligent systems.
- There are two main obstacles to this approach :
 - First, it is not easy to take informal knowledge and state it in the formal terms required by logical notation, particularly when the knowledge is less than 100% certain.
 - Second, there is a big difference between solving a problem “in principle” and solving it in practice.

- An agent is just something that acts
- **Computer agents** are expected to do more:
 - operate autonomously
 - perceive their environment
 - persist over a prolonged time period
 - adapt to change, and create and pursue goals.
- Making correct inferences is sometimes part of being a rational agent, On the other hand, correct inference is not all of rationality; in some situations, there is no provably correct thing to do, but something must still be done.

- **Philosophy**
 - Can formal rules be used to draw valid conclusions?
 - How does the mind arise from a physical brain
 - Where does knowledge come from?
 - How does knowledge lead to action?
- **Mathematics**
 - What are the formal rules to draw valid conclusions?
 - What can be computed?
 - How do we reason with uncertain information?
- **Economics**
 - How should we make decisions so as to max payoff?
 - How should we do this when others may not go along
 - How should we do this when the payoff may be far in the future?
- **Neuroscience**
 - How do brains process information?
- **Psychology**
 - How do humans and animals think and act?
- **Computer engineering**
 - How can we build an efficient computer?
- **Linguistics**
 - How does language relate to thought?

- Searching for solution
- Knowledge Base and Planning
- Reasoning
- Learning

- Uninformed search strategies :
 - Breadth-first search
 - Deep-first search
 - Depth-limited search
 - Iterative Deepening depth-first search
 - Bidirectional search
- Informed (Heuristic) search strategies :
 - Greedy best-first search
 - A* search (minimizing the total estimated solution cost)
 - Memory-bounded heuristic search
- Local Search Algorithms and Optimization Problems
- Local Search in Continuous Spaces
- Searching with Non-deterministic Actions
- Searching with Partial Observation
- Online search agents and unknown environments
- Adversarial search (Mathematical Game Theory)
- Constraint Satisfaction Problems

- Planners that are used in the real world for planning and scheduling the operations of spacecraft, factories, and military campaigns are more complex.
- Three Approachment :
 - Searching based problem solving
 - Hybrid logical agent
 - Classical planning using planning graph

- Explain how to build network models to reason under uncertainty according to the laws of probability theory.
- Representing knowledge in an uncertain domain Bayesian or Probabilistic Reasoning.
- Probabilistic Reasoning over time
- Examples : Hidden Markov Models, Kalman filters, and dynamic Bayesian networks (which include hidden Markov models and Kalman filters as special cases)

- An agent is learning if it improves its performance on future tasks after making observations about the world.
- Why would we want an agent to learn?
 - The designers cannot anticipate all possible situations that the agent might find itself in
 - The designers cannot anticipate all changes over time; a program designed to predict tomorrow's stock market prices must learn to adapt when conditions change from boom to bust.
 - Sometimes human programmers have no idea how to program a solution themselves

- Inductive
 - Machine Learning
- Deductive
 - Expert system

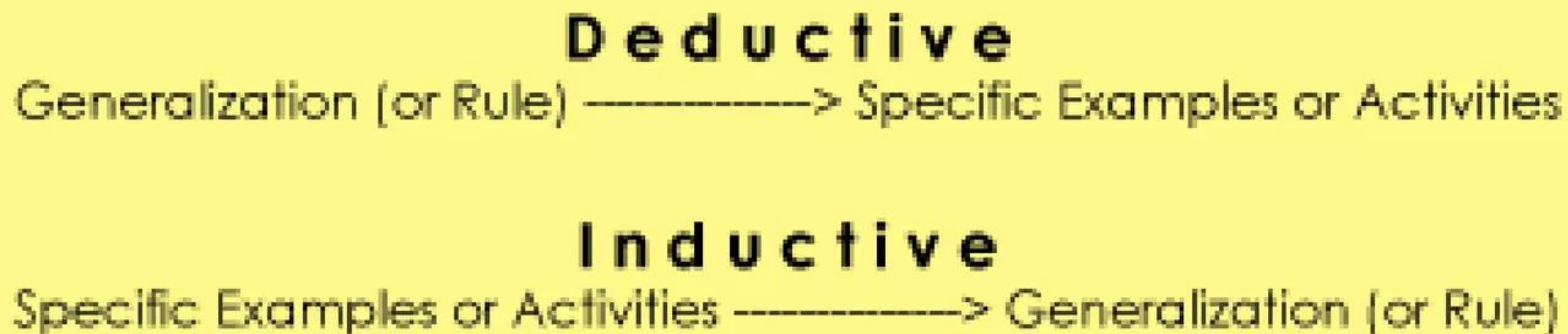
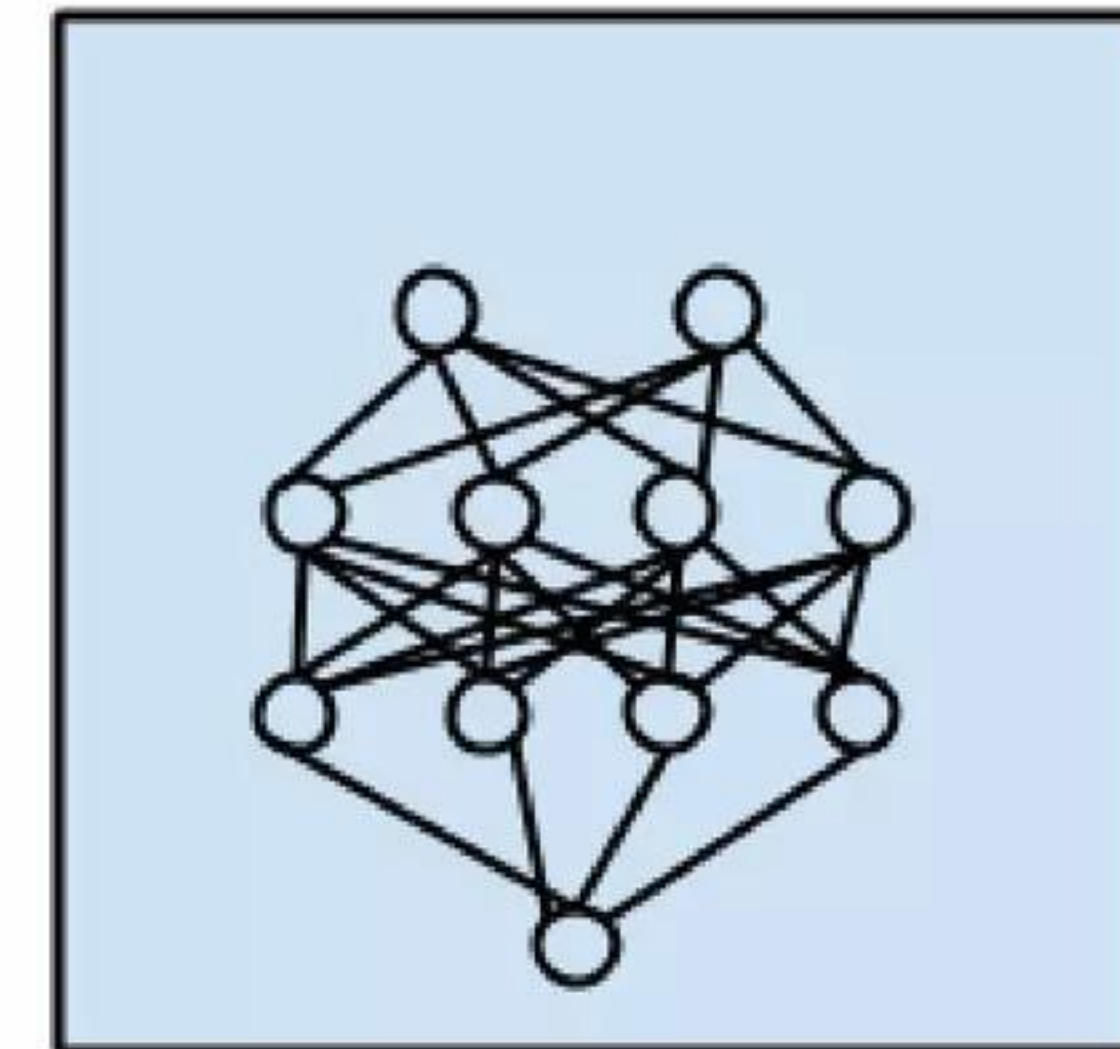


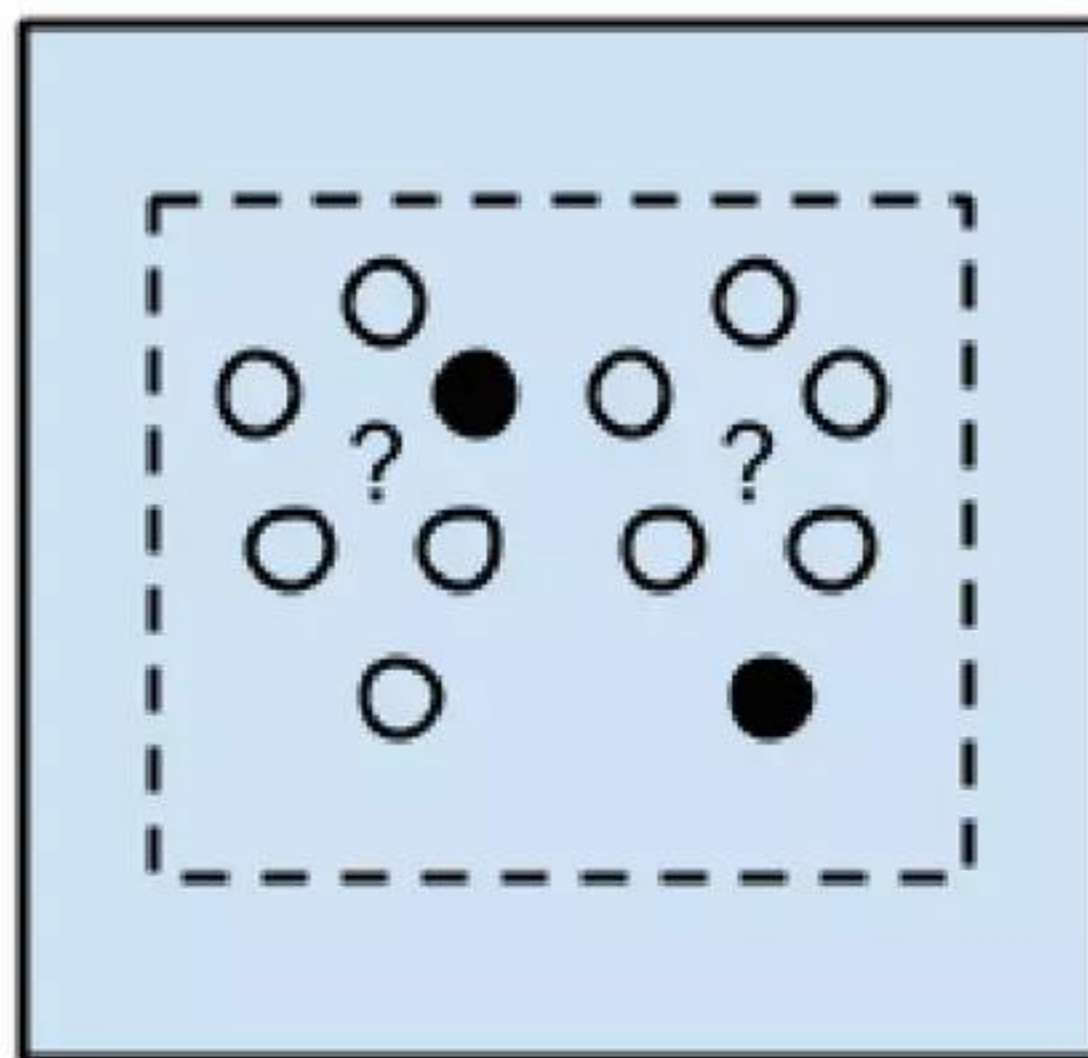
Figure 1, Deductive and Inductive Learning adapted from
<https://www.sasked.gov.sk.ca/docs/policy/approach/instrapp05.html>

“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

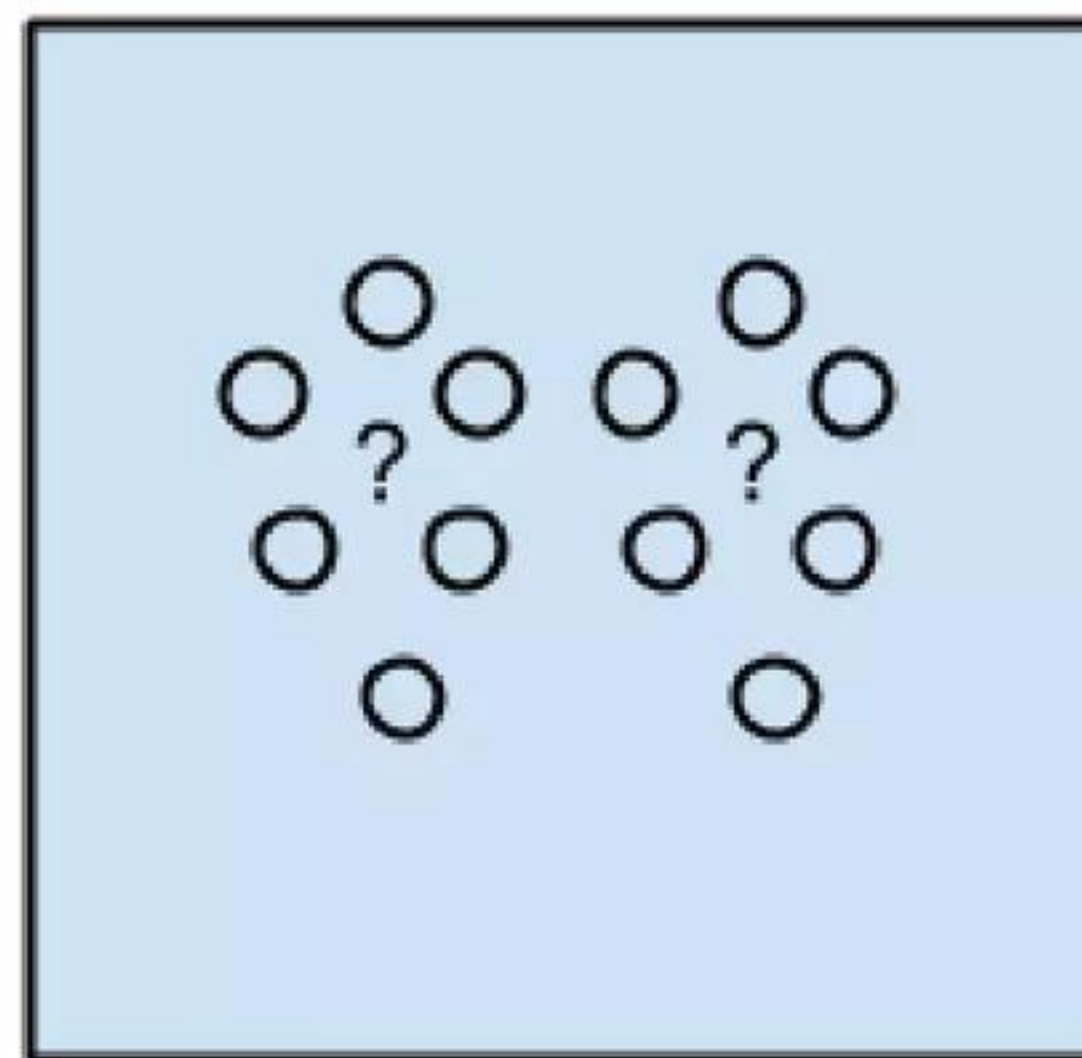
- Supervised
- Unsupervised
- Reinforcement Learning
- Semi-Supervised
- Deep Learning



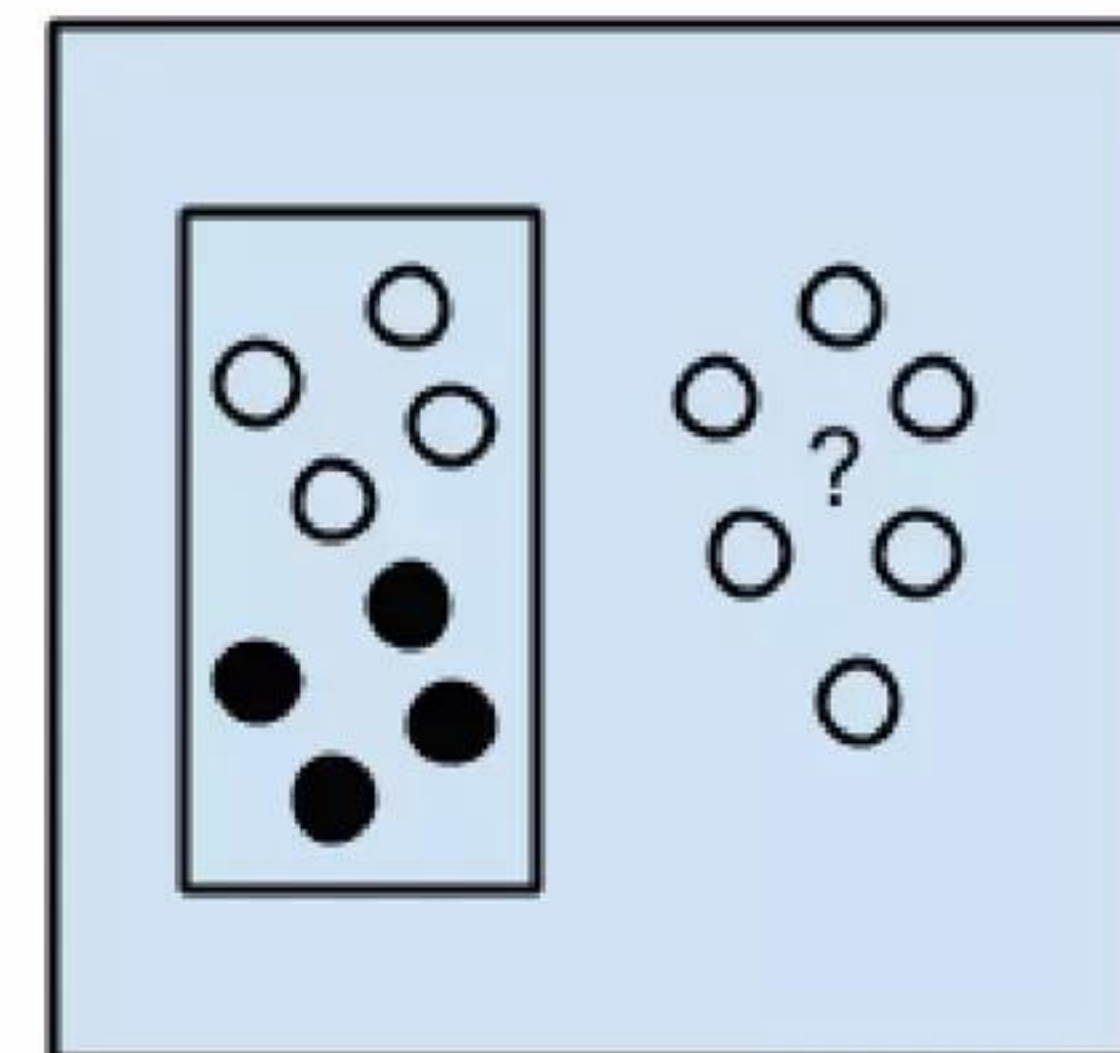
Deep Learning Algorithms



Semi-supervised Learning Algorithms

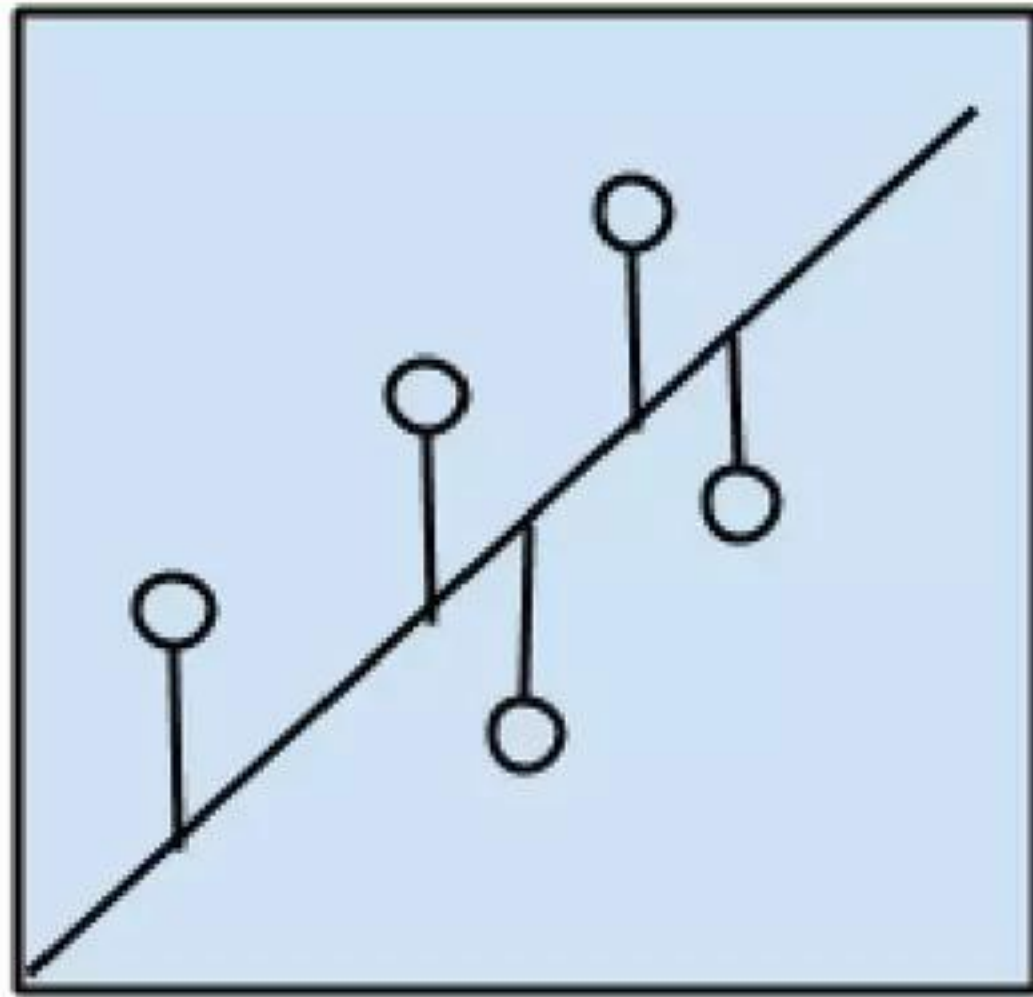


Unsupervised Learning Algorithms

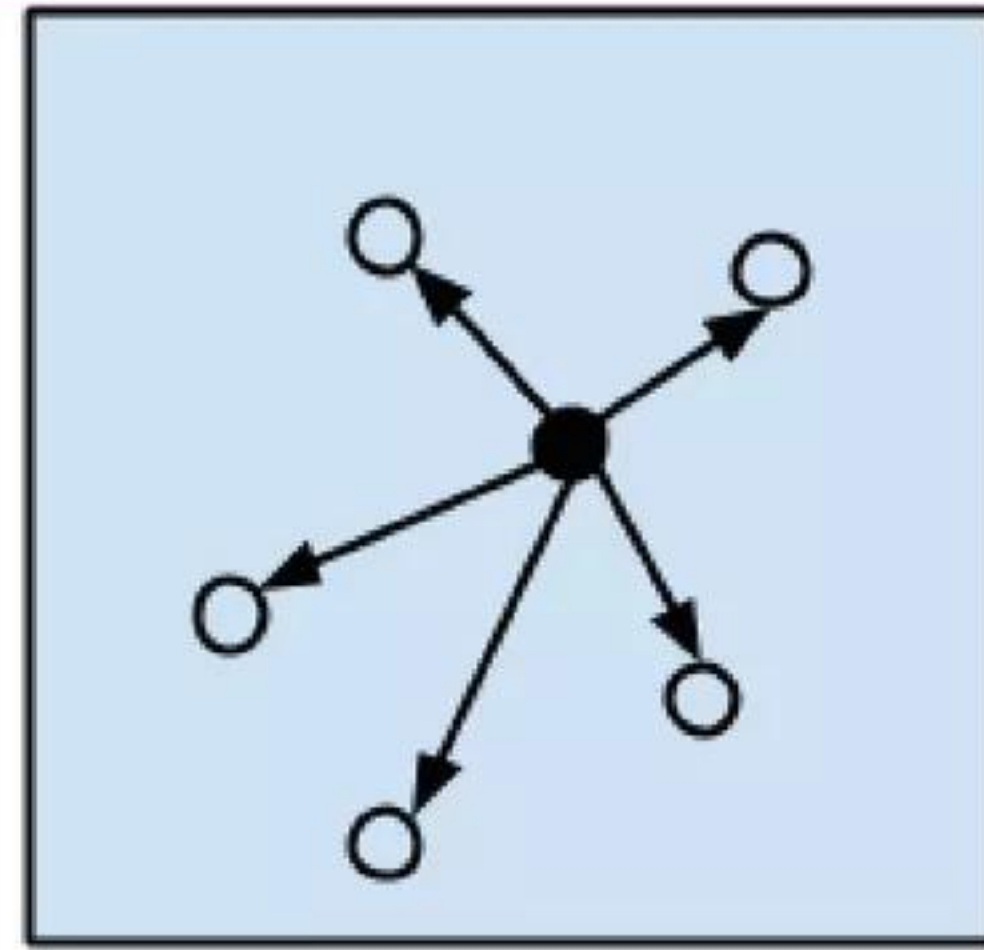


Supervised Learning Algorithms

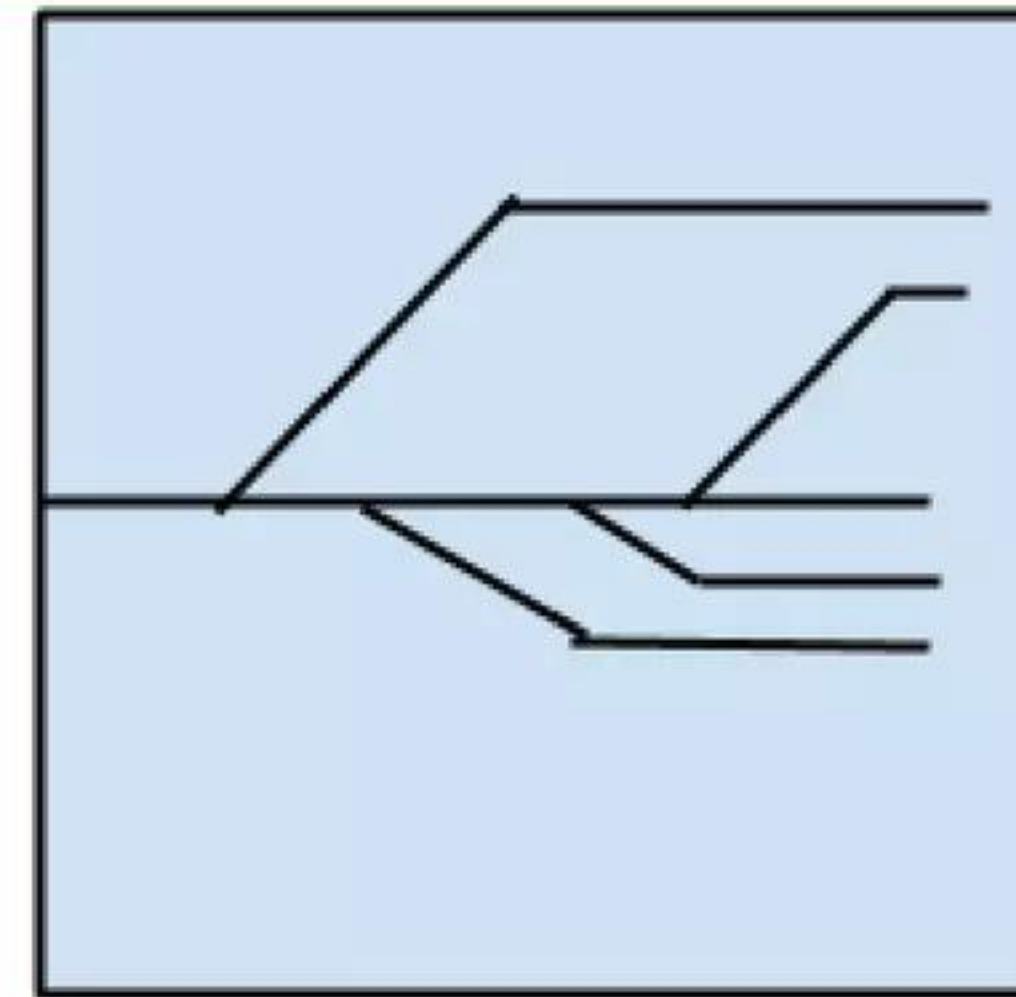
Machine Learning #2



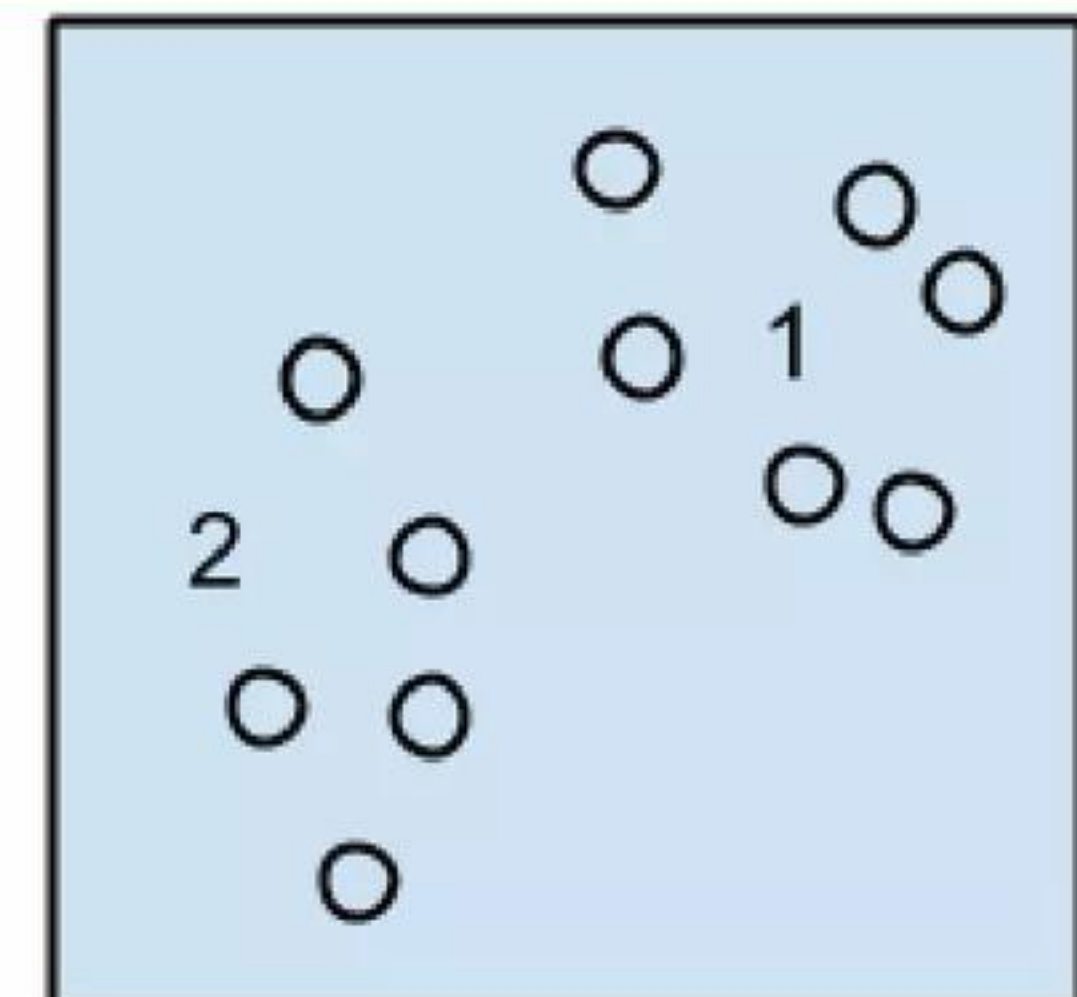
Regression Algorithms



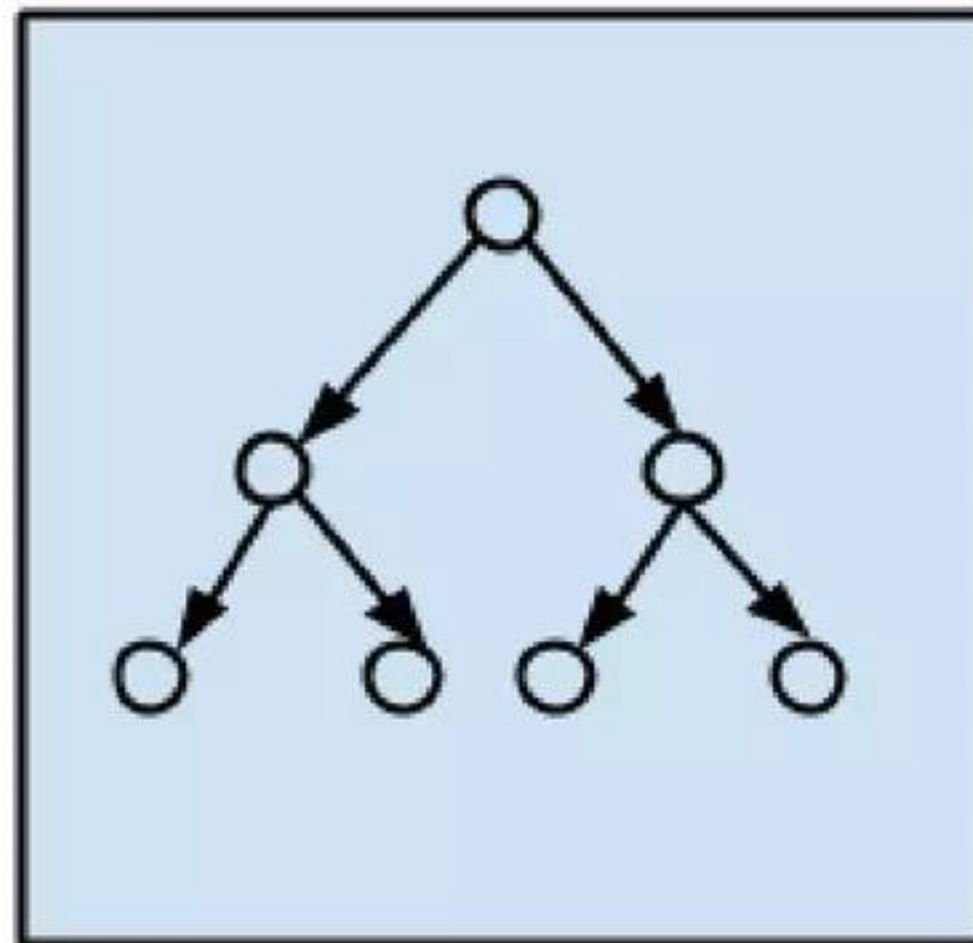
Instance-based Algorithms



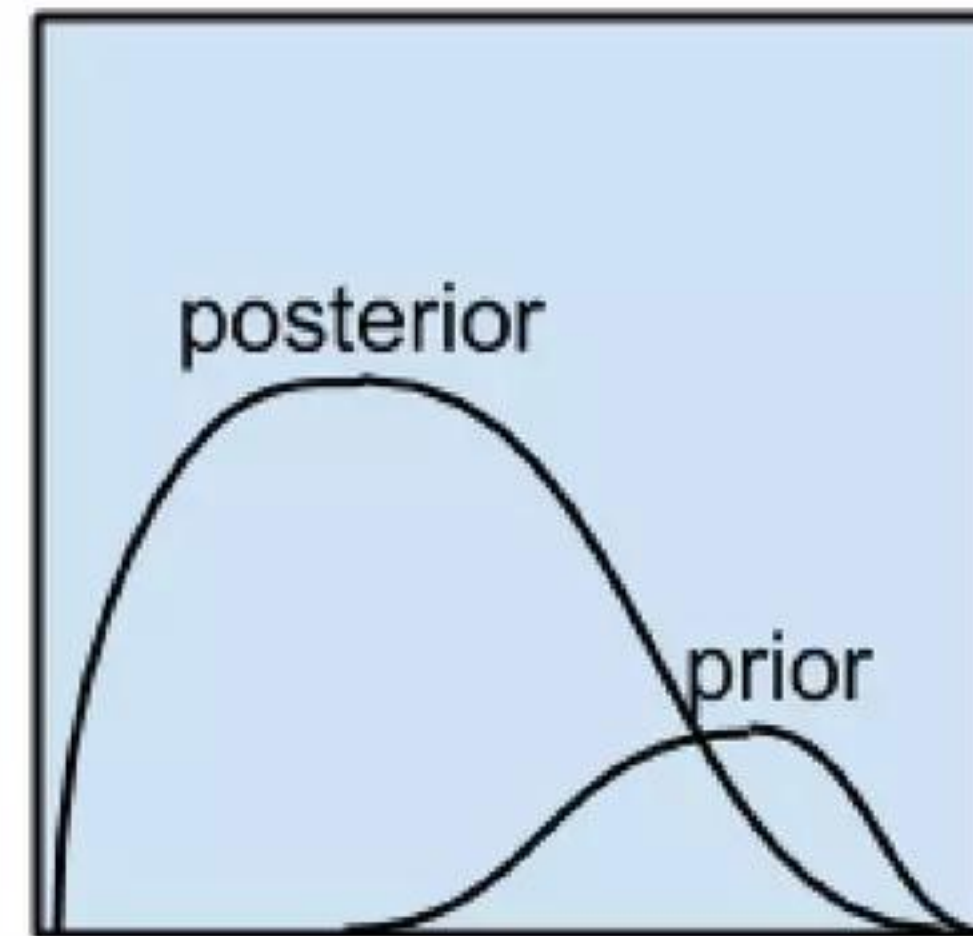
Regularization Algorithms



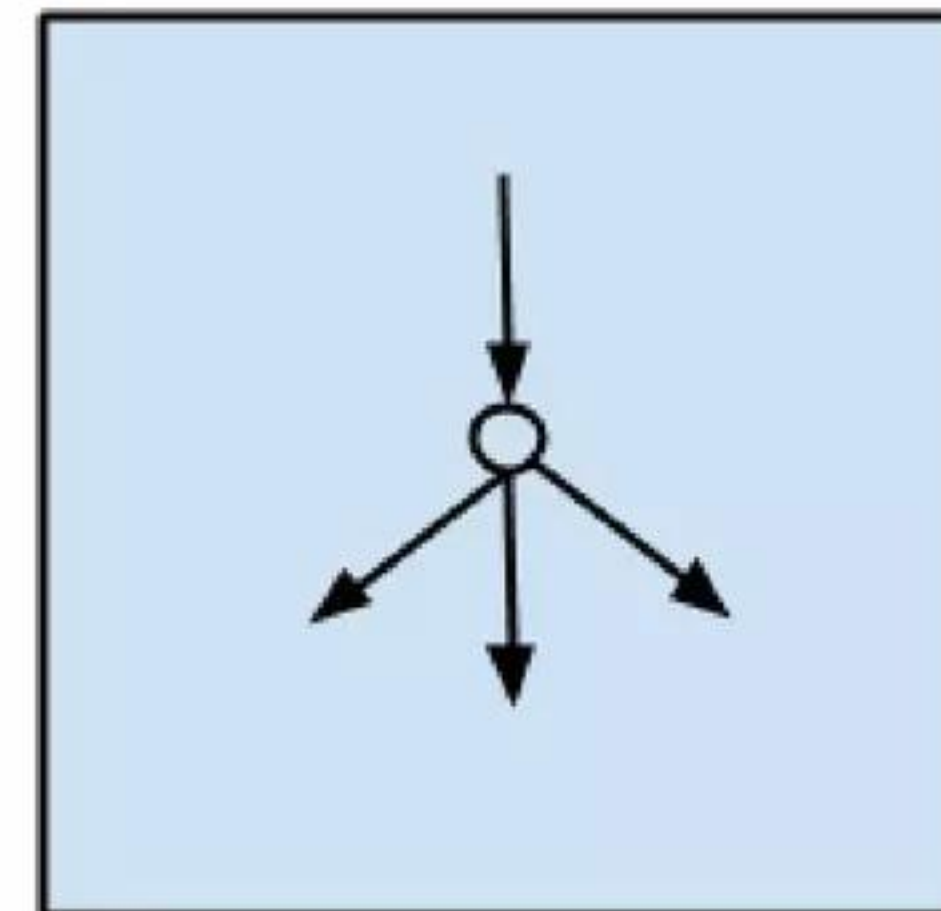
Clustering Algorithms



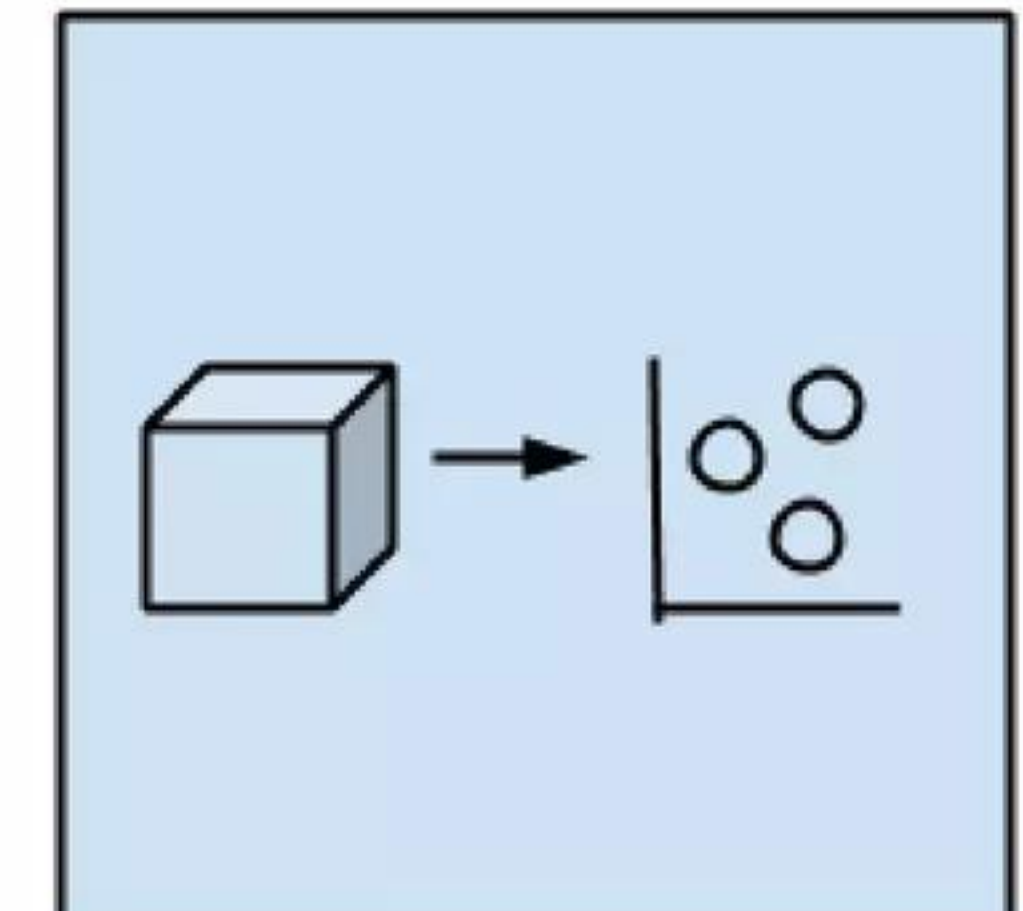
Decision Tree Algorithms



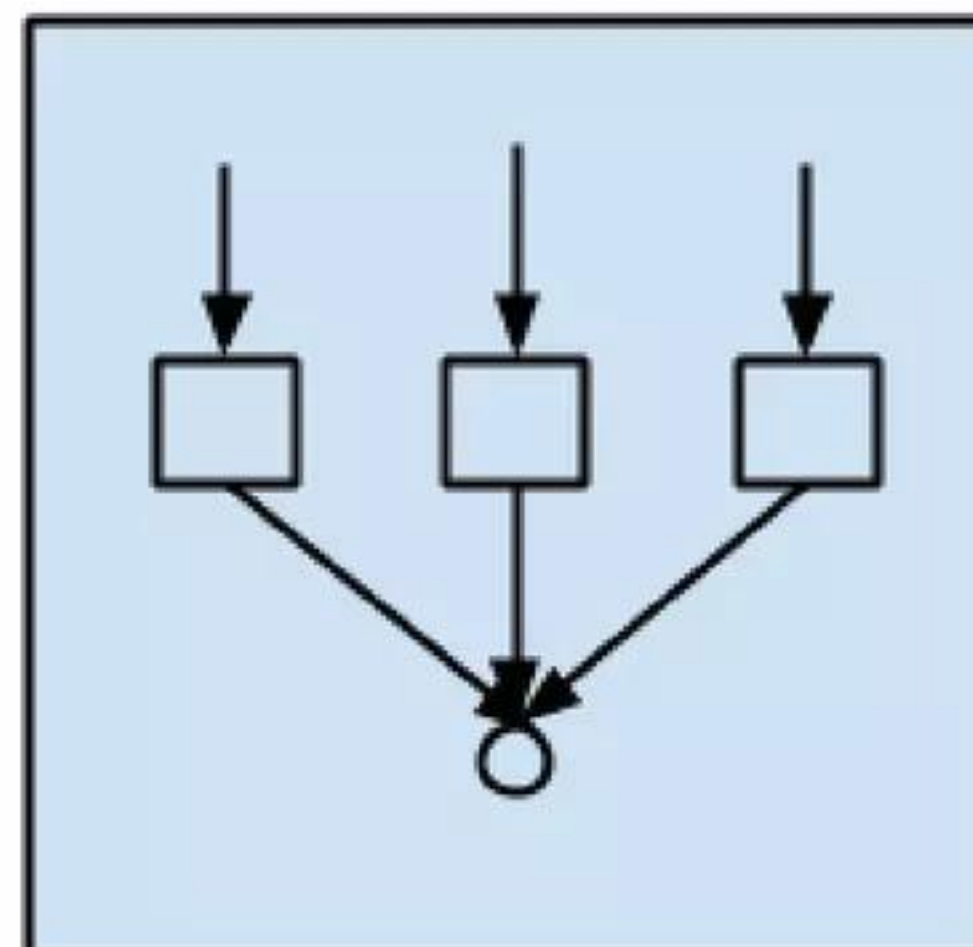
Bayesian Algorithms



Artificial Neural Network Algorithms



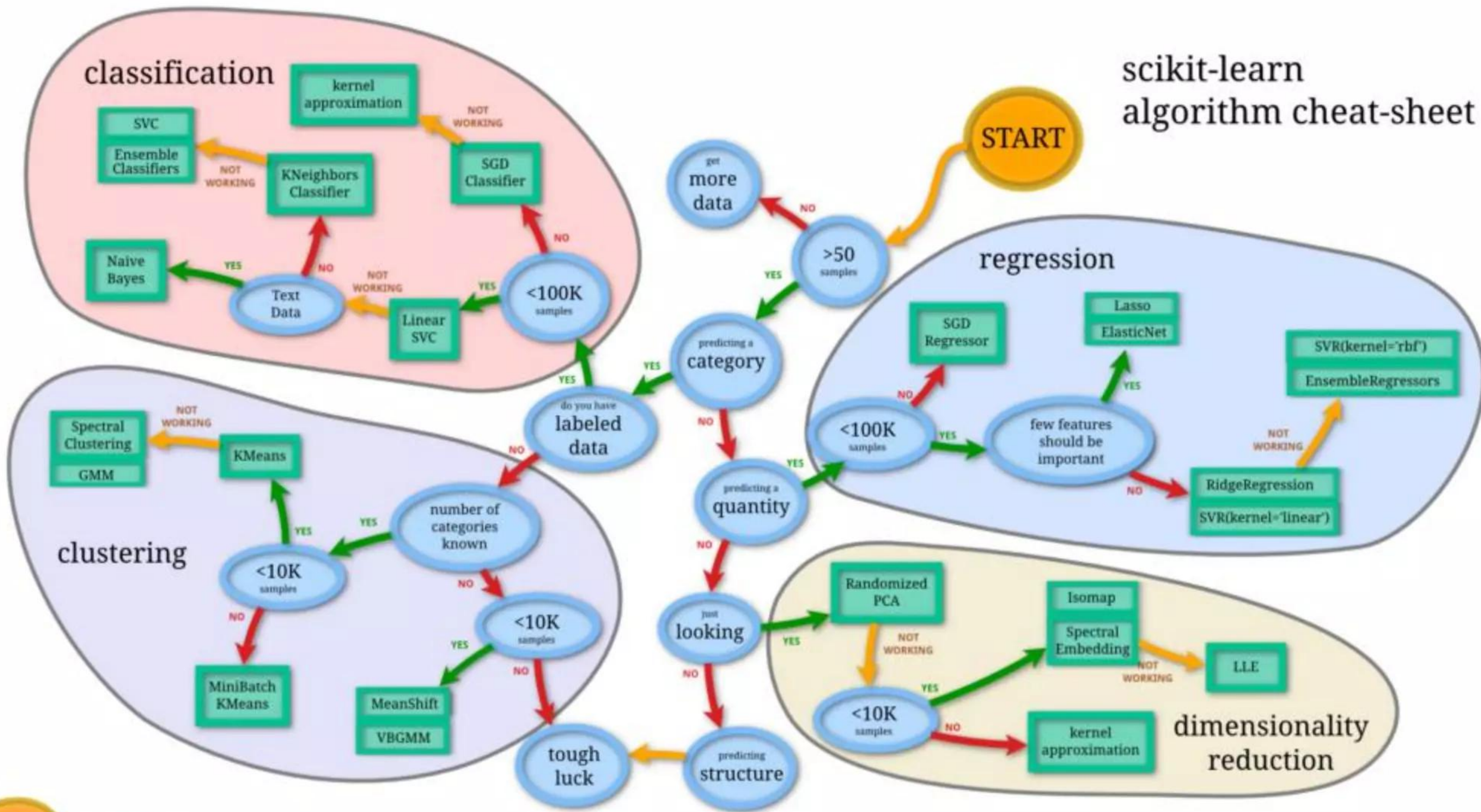
Dimensional Reduction Algorithms



Ensemble Algorithms

Machine Learning Taxonomy #1

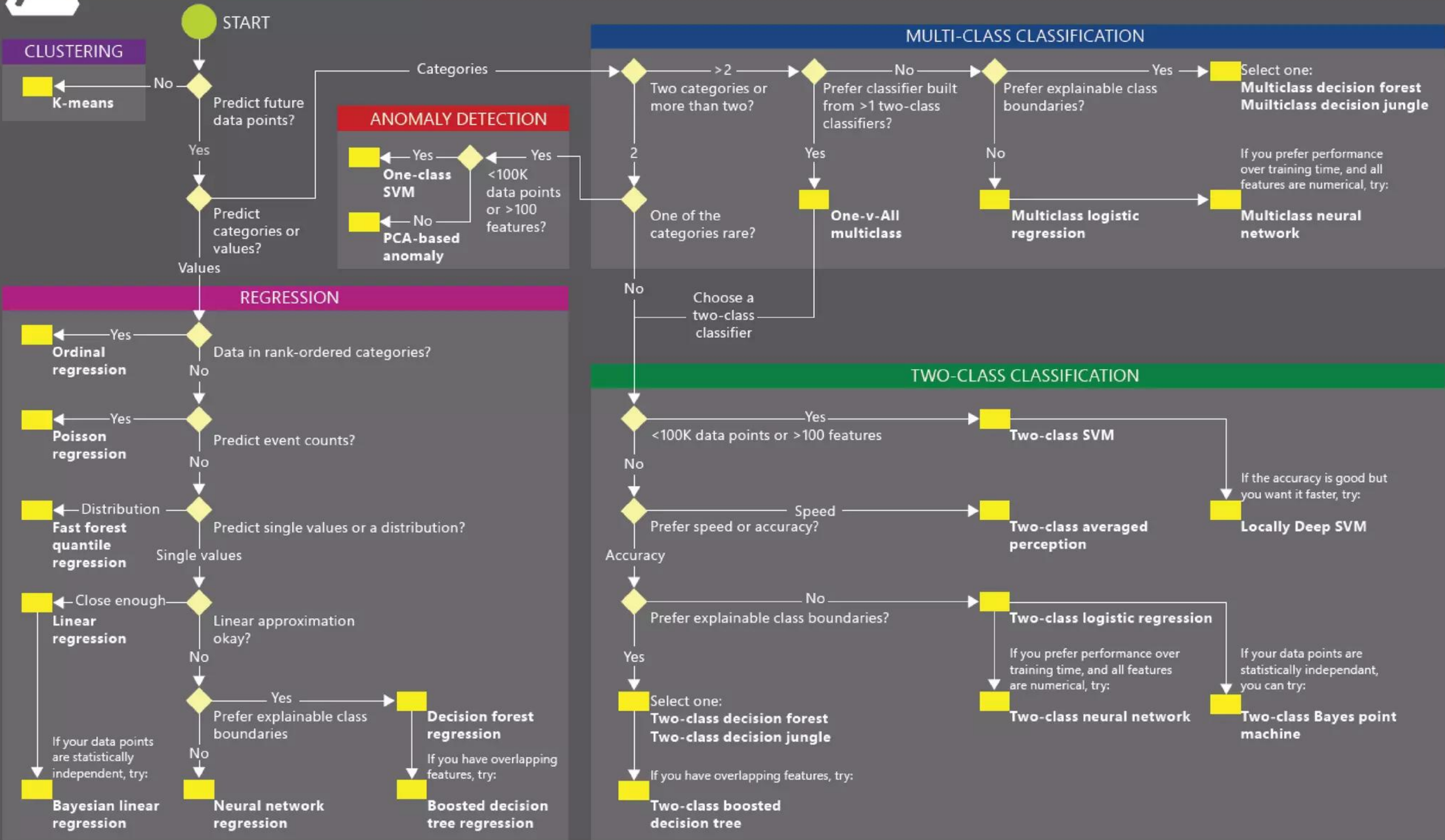
scikit-learn
algorithm cheat-sheet



Machine Learning Taxonomy #2



Machine Learning: Algorithm Cheat Sheet



Machine Learning Application in Business

MACHINE INTELLIGENCE 3.0

ENTERPRISE INTELLIGENCE

VISUAL Orbital Insight planet clarifai DEEP VISION cortica Igcocion SPACE_KNOW Captricity netra deepomatic	AUDIO Gridspace TalkIQ nexidia twilio CAPIO Expect Labs Clover Mobvoi Quirous.AI pop2P archive	SENSOR PREDIX G3IoT MAANA Sentenai PLANET OS UPTAKE IMUBIT thingworx KONUX Alluvium	INTERNAL DATA PRIMER IBM WATSON Cyncorp Palantir ARIMO Alation Sapho Outlier Digital Reasoning	MARKET mattermark Quid DataFox PREMISE Bottlenose MOTIVA enigma CB INSIGHTS Tracxn predata
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ENTERPRISE FUNCTIONS

CUSTOMER SUPPORT DigitalGenius Kasisto ELOQUENT Wiseio ACTIONIQ zendesk Proact CLARABRIDGE	SALES collective[i] sense fuse machines AVISO salesforce INSIDE SALES .COM clari Zensight	MARKETING MINTIGO Lattice RADIUS LiftIgniter (PERSADO) brightfunnel retention MOTIVA COGNICOR AIRPR msgcl	SECURITY CYLANCE DARKTRACE ZIMPERIUM deepinstinct Sentinel DEMISTO graphistry drawbridge SignalSense AppZen	RECRUITING textio entelo Wade & Wendy hiQ unilive SpringRole GIGSTER HireVue
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AUTONOMOUS SYSTEMS

GROUND NAVIGATION drive.ai AdasWorks ZOOX MOBILEYE UBER Google TESLA nuTonomy Auto Robotics	AERIAL SKYDIO SHIELD AI Airware DJI LILY DroneDeploy pilo.ai SKYCATCH	INDUSTRIAL JAYBRIDGE OSARO CLEARPATH fetch KINDRED HARVEST rethink robotics	PERSONAL amazon alexa Cortana Allo facebook Siri Replika	PROFESSIONAL butter.ai pogo SKIPFLAG clara x.ai slack talla Zoom sudo
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INDUSTRIES

AGRICULTURE BLUE RIVER MAVIX tile TRACE Pivot Bio Terraviva AGRI-DATA Descartes Labs udio abundant robotics	EDUCATION KNEWTON volley gradescope CTI coursera UDACITY alt school	INVESTMENT Bloomberg sentient ISENTIUM KENSHO alphasense Dataminr CEREBELLUM CAPITAL Quandl	LEGAL blue J BEAGLE Everlaw RAVEL seal ROSS LEGAL ROBOT	LOGISTICS NAUTO Acerta PRETECKT Routific clearmetal MARBLE PITSTOP
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INDUSTRIES CONT'D

MATERIALS zymergen Citrine Eigen Innovations SIGHT MACHINE GINKGO BIOWORKS nanotronics CALCULARIO	RETAIL FINANCE TALA zest finance Lendo earnest affirm MIRADOR wealthfront Betterment	PATIENT PULSE CareSkore ZEPHYR HEALTH EM Watson Health Oncoto SENTRIAN Atomwise Numerate	IMAGE BUTTERFLY 3SCAN ARTERYS enlitic BAYLABS imagia Google DeepMind	BIOLOGICAL iCarbonX color GRAIL deep genomics RECURSION LUMINIST Numerate Atomwise verily WHOLE BIOME
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TECHNOLOGY STACK

AGENT ENABLERS
 OCTANE.AI howdy. Maluuba KITT.AI
 OpenAI Gym Kasisto AUTOMAT
 semanticmachines

DATA SCIENCE
 DOMINO SPARKBEYOND rapidminer
 kaggle DataRobot yhat AYASDI
 data iku seldon yseop bigml

MACHINE LEARNING
 CognitiveScale GoogleML context relevant
 Cyncorp HyperScience nara logics minds.ai H2O.ai
 SCALED INFERENCE sparkcognition loop GEOMETRIC INTELLIGENCE
 deepsense.io reactive skymind bonsai

NATURAL LANGUAGE
 agolo AYLIEN LEXALYTICS
 Narrative Science loop spaCy LUMINOSO
 cortical.io MonkeyLearn

DEVELOPMENT
 SIGOPT HyperOpt fuzzyio okite
 rainforest lobe Anodot
 Signifai LAYER6 bonsai

DATA CAPTURE
 CrowdFlower diffbot CrowdAI import io
 Paxata DATASIFT amazon mechanicalturk enigma
 WorkFusion DATALOGUE TRIFACTA parsehub

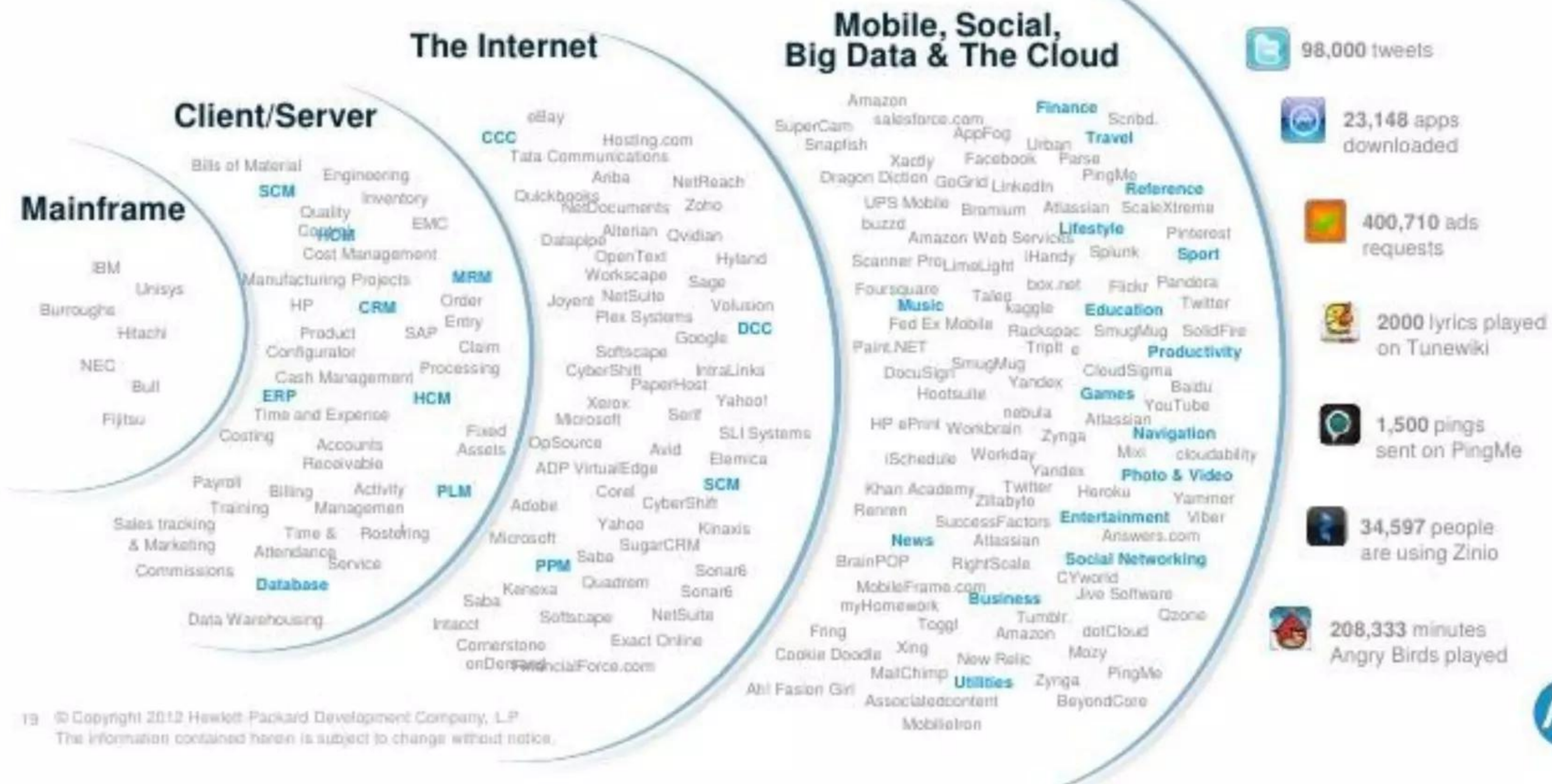
OPEN SOURCE LIBRARIES
 Keras Chainer CNTK TensorFlow Caffe
 H2O DEEPLARNING4J theano torch
 DSSTNE Scikit-learn AzureML neon
 MXNet DMTK Spark PaddlePaddle WEKA

HARDWARE
 KNUPATH TENSTORRENT Cirrascale
 NVIDIA intel nervana Movidius
 tensilica GoogleTPU 10²⁶ Labs Qualcomm
 Cerebras Isosemi

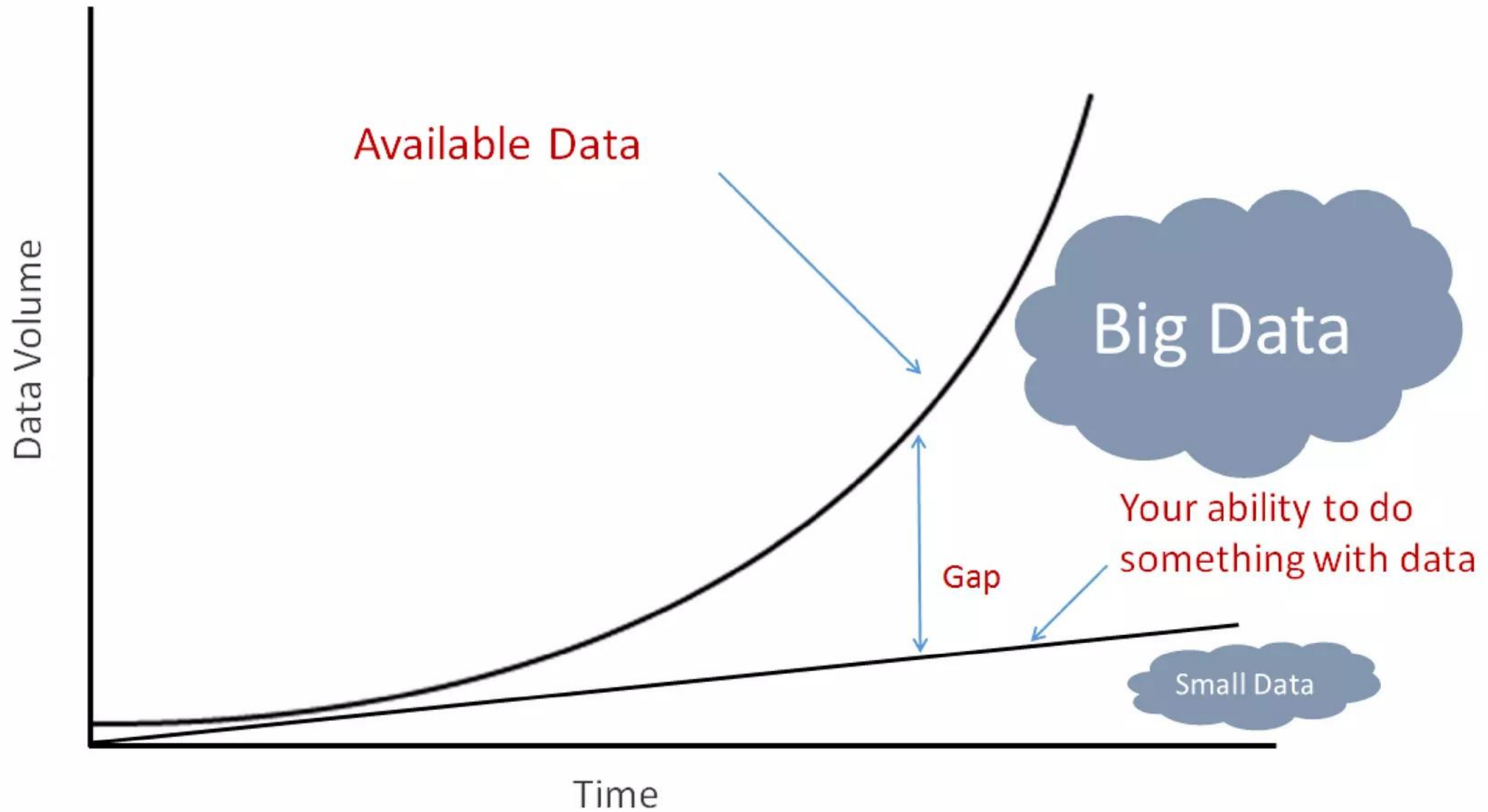
RESEARCH
 OpenAI KNOGGIN nraisense ELEMENT^{AI} vicarious
 Numenta Kimera Systems Cogitai

- **Fraud Detection System**
- **Dynamic Recommendation System and User Profiling**
- **Traveling Salesman Problem and Binpacking Problem for better warehouse management**
- Chatbot
- **Social Media Analysis**
 - Vocal users
 - Preprocessing for another approachment (add more features)
- Company condition forecasting
- Governance simulation

Accelerating Innovation & Change – every 60 seconds



Data Explosion

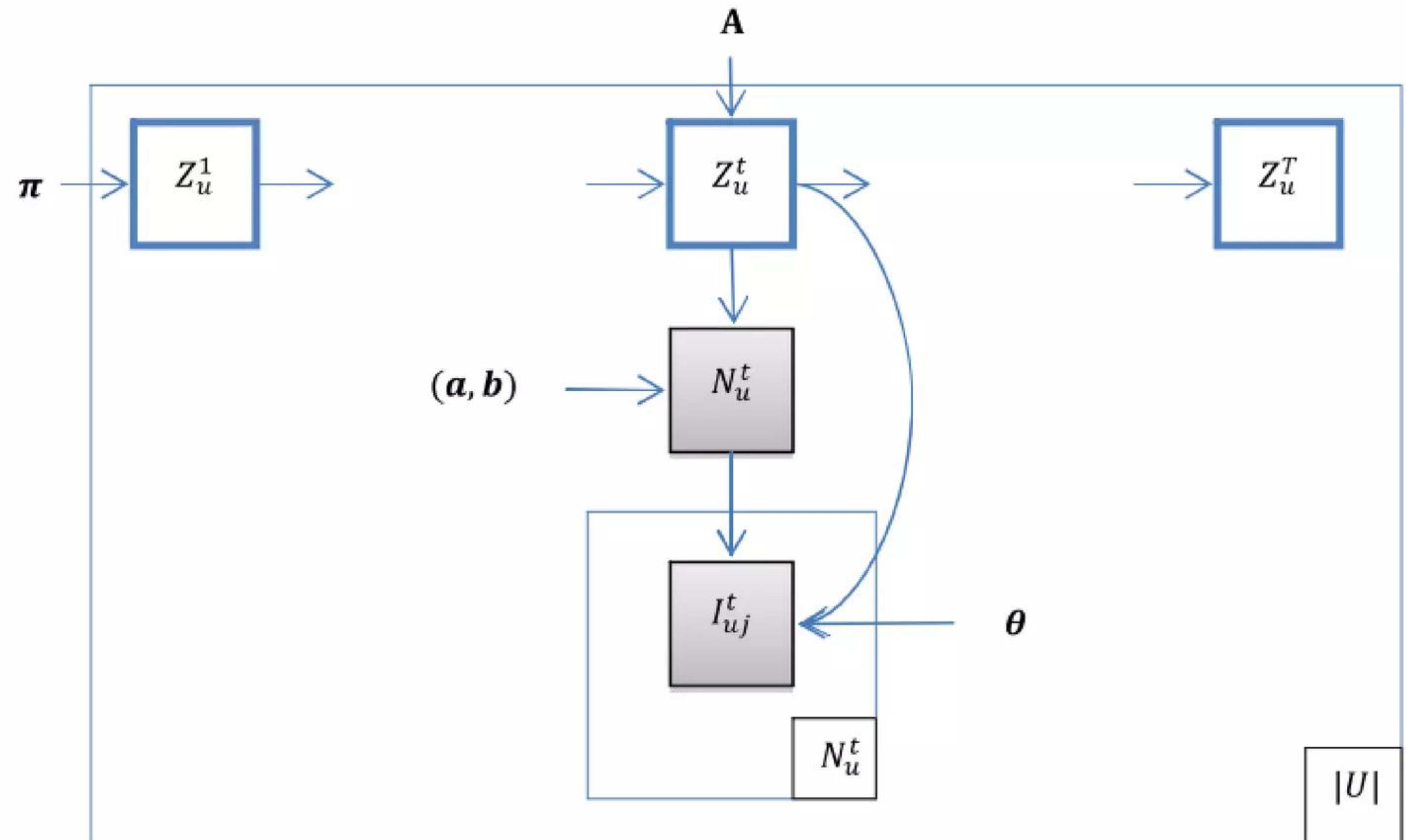
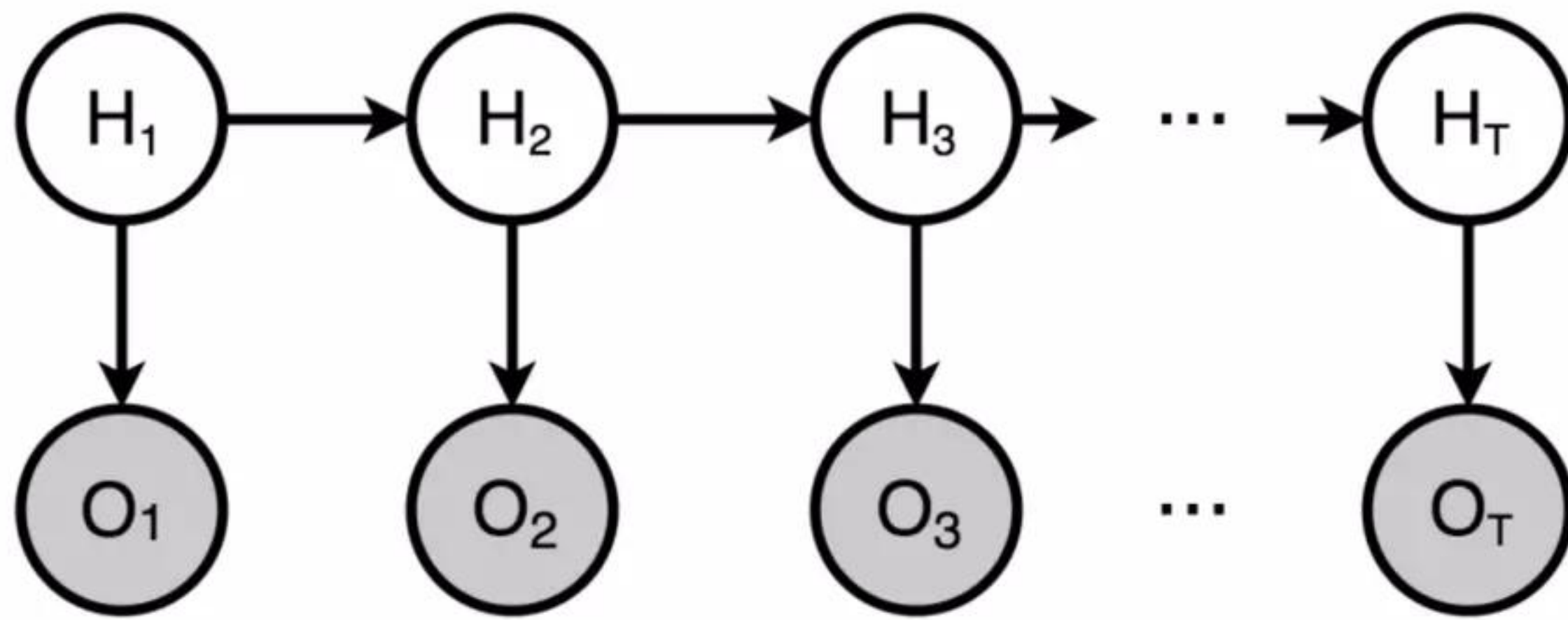


- Big data : volume, variety, velocity, and veracity. (You might consider a fifth V, value.)
- Knowledge representation or the architecture of the model
- Unimplemented methods/algorithms in any libraries
- Stack of methods
- Data mostly unlabeled data
- Features Engineering (especially from unstructured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

- **Data retrieving:** Big Query
- **Stream analytics:** software that can filter, aggregate, enrich, and analyze a high throughput of data from multiple disparate live data sources and in any data format.
- **Data Sources:** operational and functional systems, machine logs and sensors, Web and social and many other sources
- **Data Platforms, Warehouses and Discovery Platforms:** that enable the capture and management of data, and then – critically – its conversion into customer insights and, ultimately, action

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

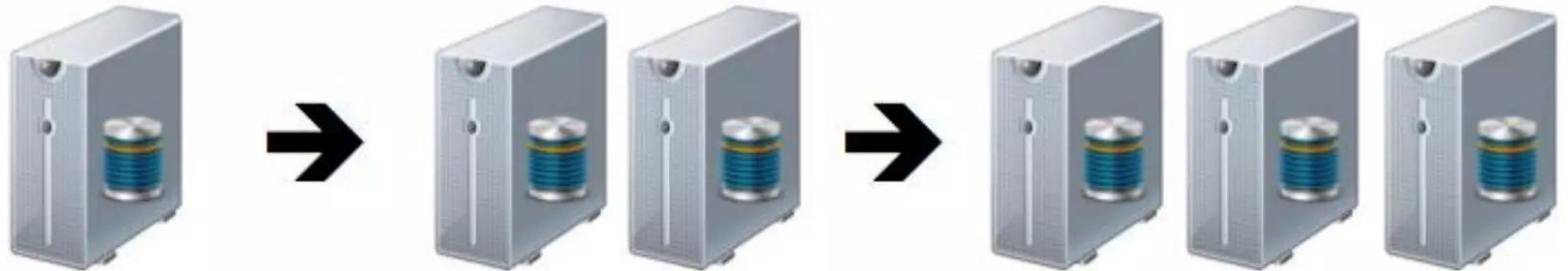
Dynamic Collaborative Filtering using HMM



Scale-Up

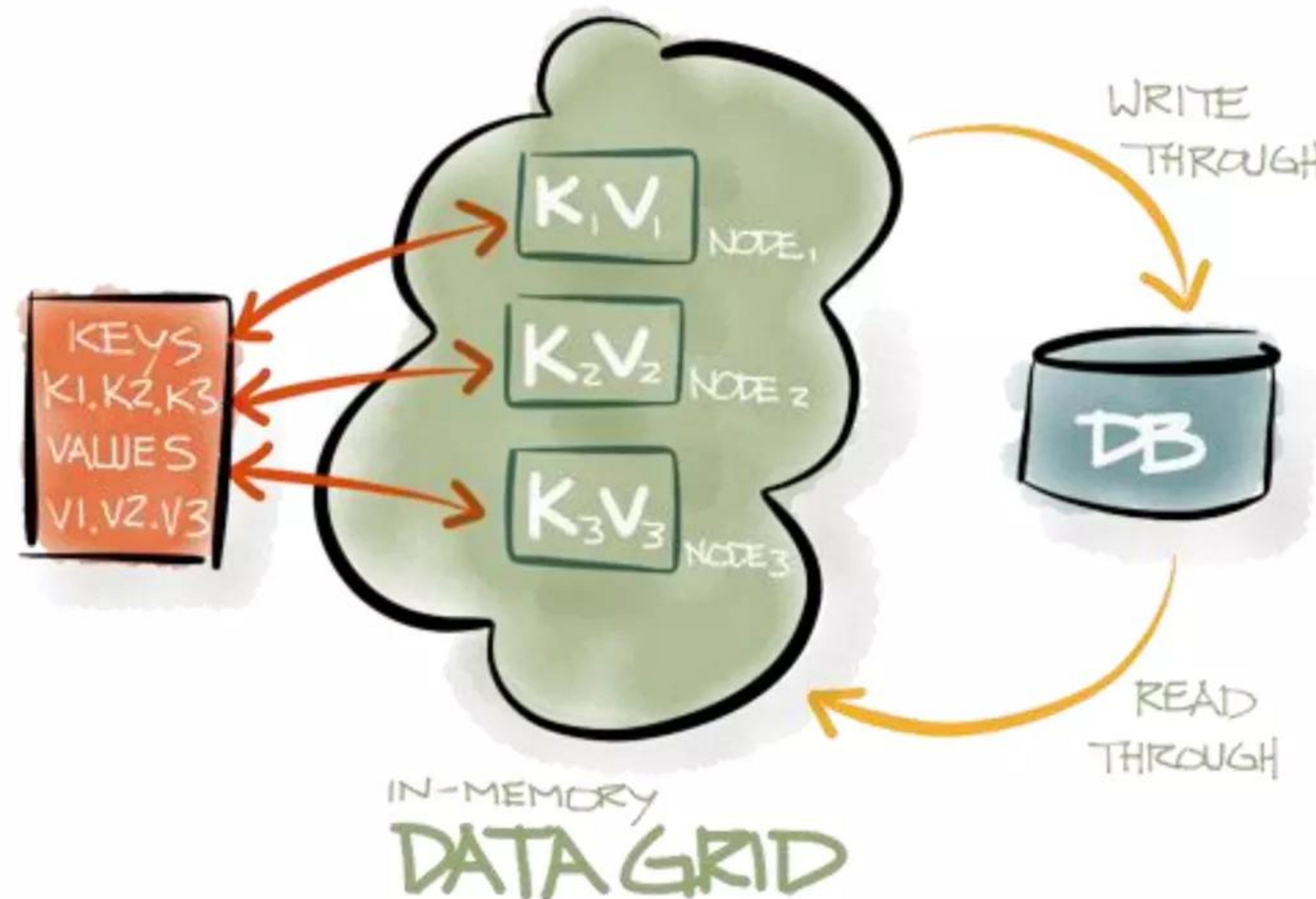


Scale-Out



<https://hadoop4usa.wordpress.com/2012/04/13/scale-out-up/>

- **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.



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



Stack of Technologies

Microsoft Azure Machine Learning | Home Studio Gallery

Binary Classification: Direct marketing

In draft

Search experiment items

- ▶ Saved Datasets
- ▶ Data Format Conversions
- ▶ Data Input and Output
- ▶ Data Transformation
- ▶ Feature Selection
- ▶ Machine Learning
- ▶ OpenCV Library Modules
- ▶ Python Language Modules
- ▶ R Language Modules
- ▶ Statistical Functions
- ▶ Text Analytics
- ▶ Web Service
- ▶ Deprecated

```
graph TD; Reader[Reader] --> Metadata[Metadata Editor]; Metadata --> Project[Project Columns  
remove columns that are part of the label]; Project --> Split1[Split]; Split1 --> Boosted[Two-Class Boosted Decision T...]; Split1 --> SVM[Two-Class Support Vector Ma...]; Boosted --> Split2[Split]; SVM --> Split2; Split2 --> Sweep1[Sweep Parameters]; Split2 --> Sweep2[Sweep Parameters]; Sweep1 --> Score1[Score Model]; Sweep2 --> Score2[Score Model]; Score1 --> Evaluate[Evaluate Model]; Score2 --> Evaluate;
```

Properties

Two-Class Boosted Decision Tree

Create trainer mode: Single Parameter

Maximum number of leav...: 20

Minimum number of sam...: 10

Learning rate: 0.2

Number of trees construct...: 100

Random number seed: 0

Allow unknown categ...

Quick Help

Creates a binary classifier using a boosted decision tree algorithm (more help...)

Stack of Technologies #2

Microsoft Azure Machine Learning | Home Studio Gallery

Binary Classification: Direct marketing

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```

Properties

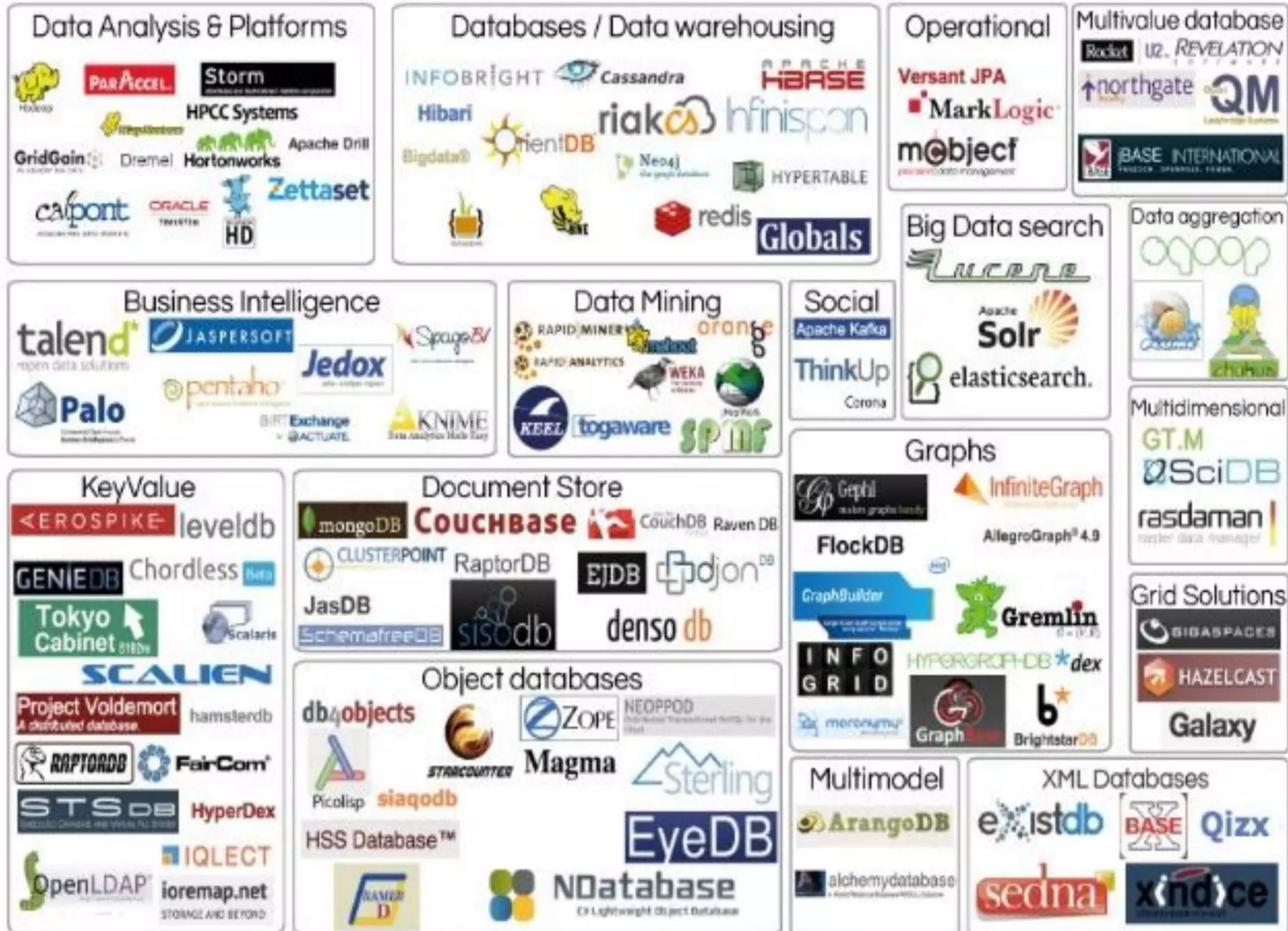
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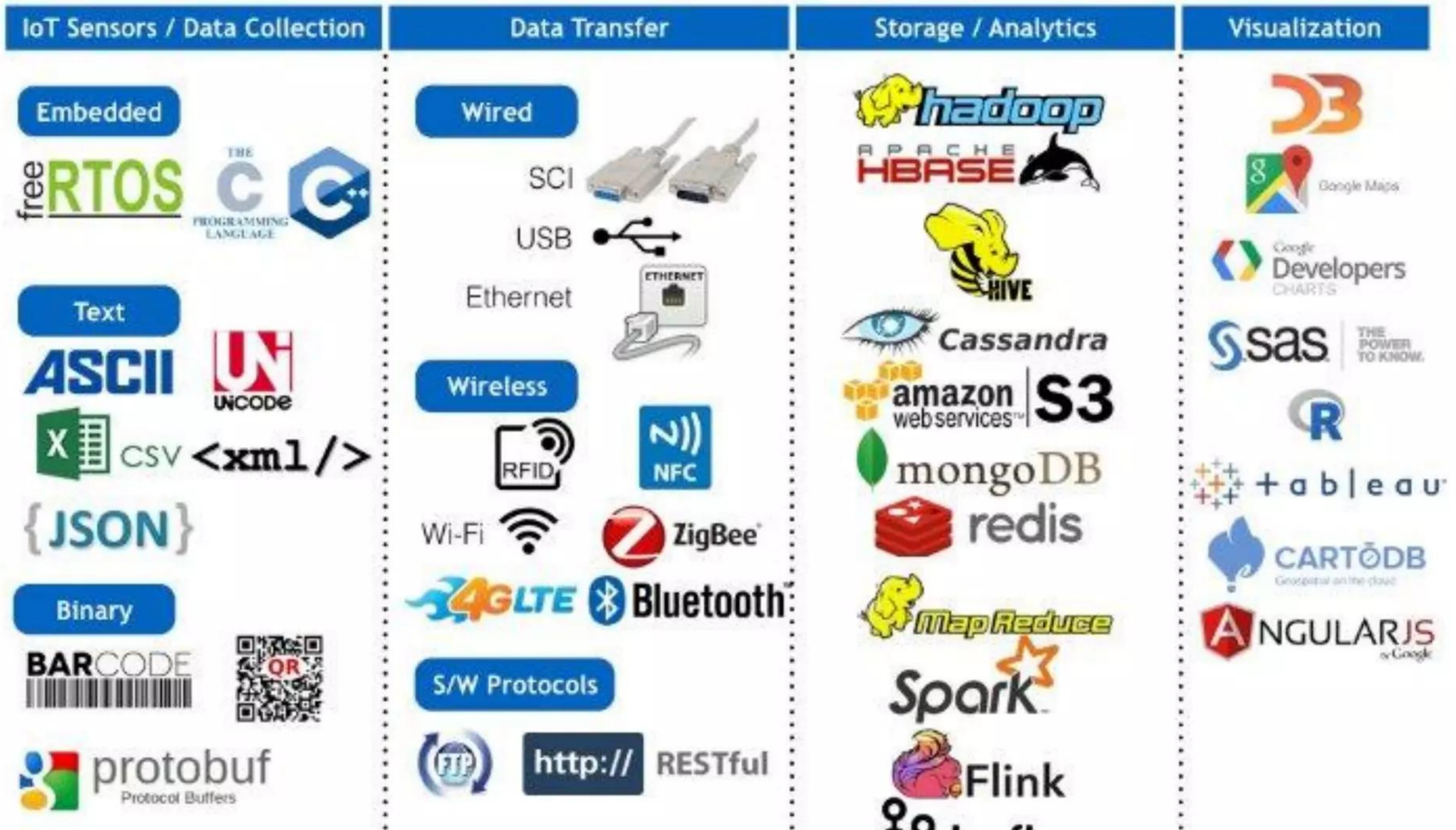
Quick Help

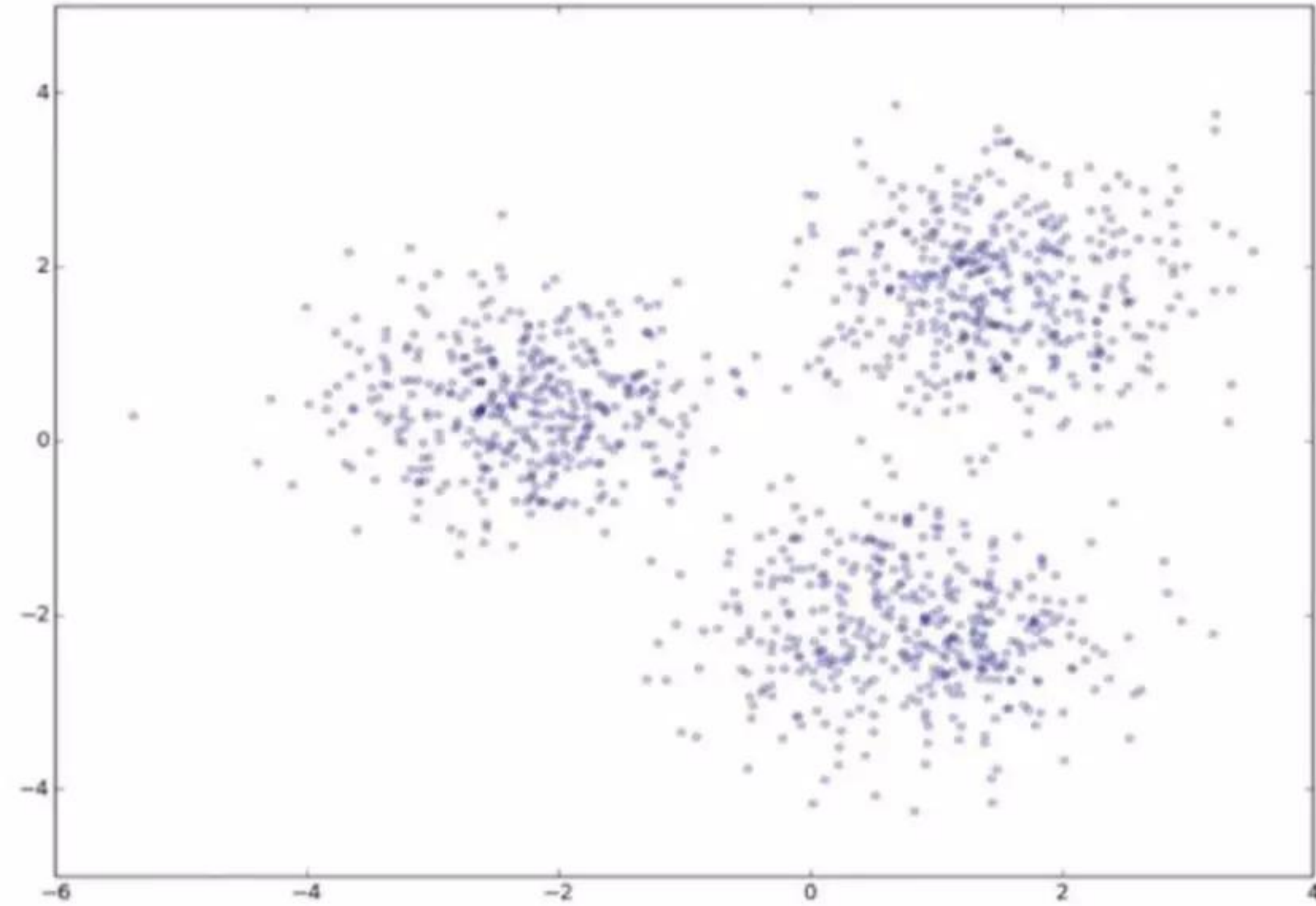
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Stack of Technologies #4

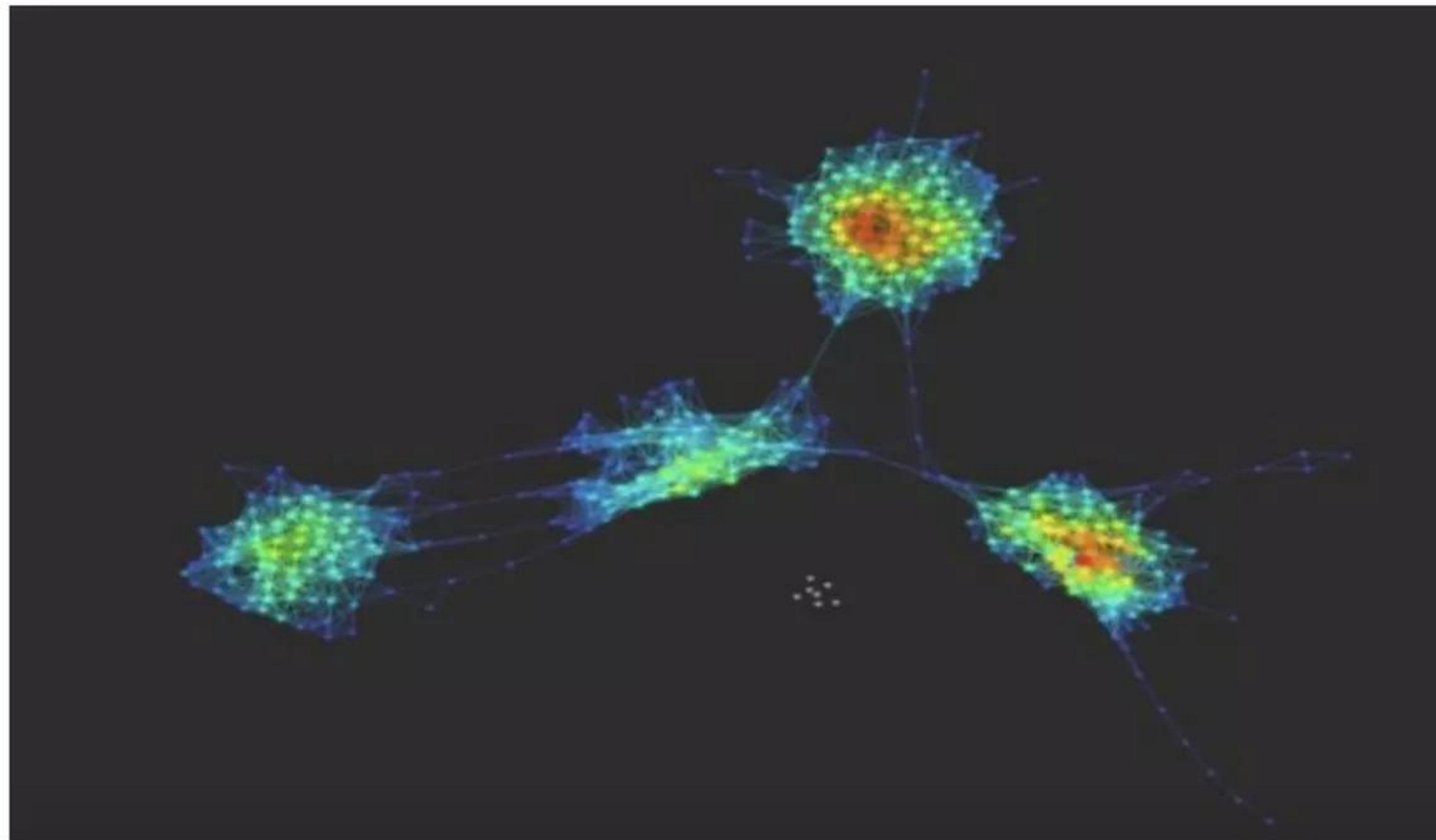


Stack of Technologies #5





- Coordinate invariance
- Deformation Invariance
- Compressed Representations





THANK YOU

Any question?