

# Financial Security & Machine Learning

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# • Outline

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- Introduction
- Immune System
- Machine Learning
- Solutions

# • Introduction

## Motive

신제윤 금융위원장은 금융보안을 위해 모든 금융권이 이상거래탐지시스템(**FDS**) 구축을 완료해야 한다고 촉구했다.

“핀테크 활성화 방안을 추진하기 위해서 반드시 전제돼야 할 사항은 보안의 중요성”이라며 “정보보안이 확보되지 않은 서비스는 결국 사상누각이 될 것”이라고 우려했다.

그는 핀테크(**Fintech**) 추진 방안과 관련해서는 “오프라인 위주의 금융제도 개편을 통해 핀테크 기술이 금융에 자연스럽게 접목될 수 있도록 지원할 것”이라며 “전자금융업종 규율을 재설계토록 하겠다”고 밝혔다.

# • Introduction

## FinTech



Venture Scanner

# ● Introduction

## Fraud Detection Basics

### ● Outlier Detection

- detecting data points that don't follow the trends and patterns in the data
- rule base detection
- anomaly detection

### ● Two approaches for treating input

- focus on **instance** of data point
- focus on **sequence** of data points

### ● Three kinds of algorithms

- building a model out of data
- using data directly.
- immune system base on temporal data

### ● Real time fraud detection

- feasible with model based approach
- A model is built with **batch processing** of training data
- A real time stream processor *uses the model* and makes **predictions in real time**

# ● Introduction

## Economy Imperative

- Not worth spending \$200m to stop \$20m fraud
- The Pareto principle
  - the first 50% of fraud is easy to stop
  - next 25% takes the same effort
  - next 12.5% takes the same effort
- Resources available for fraud detection are always limited
  - around 3% of police resources go on fraud ?
  - this will not significantly increase
- If we cannot outspend the fraudsters we must out-think them

# • Introduction

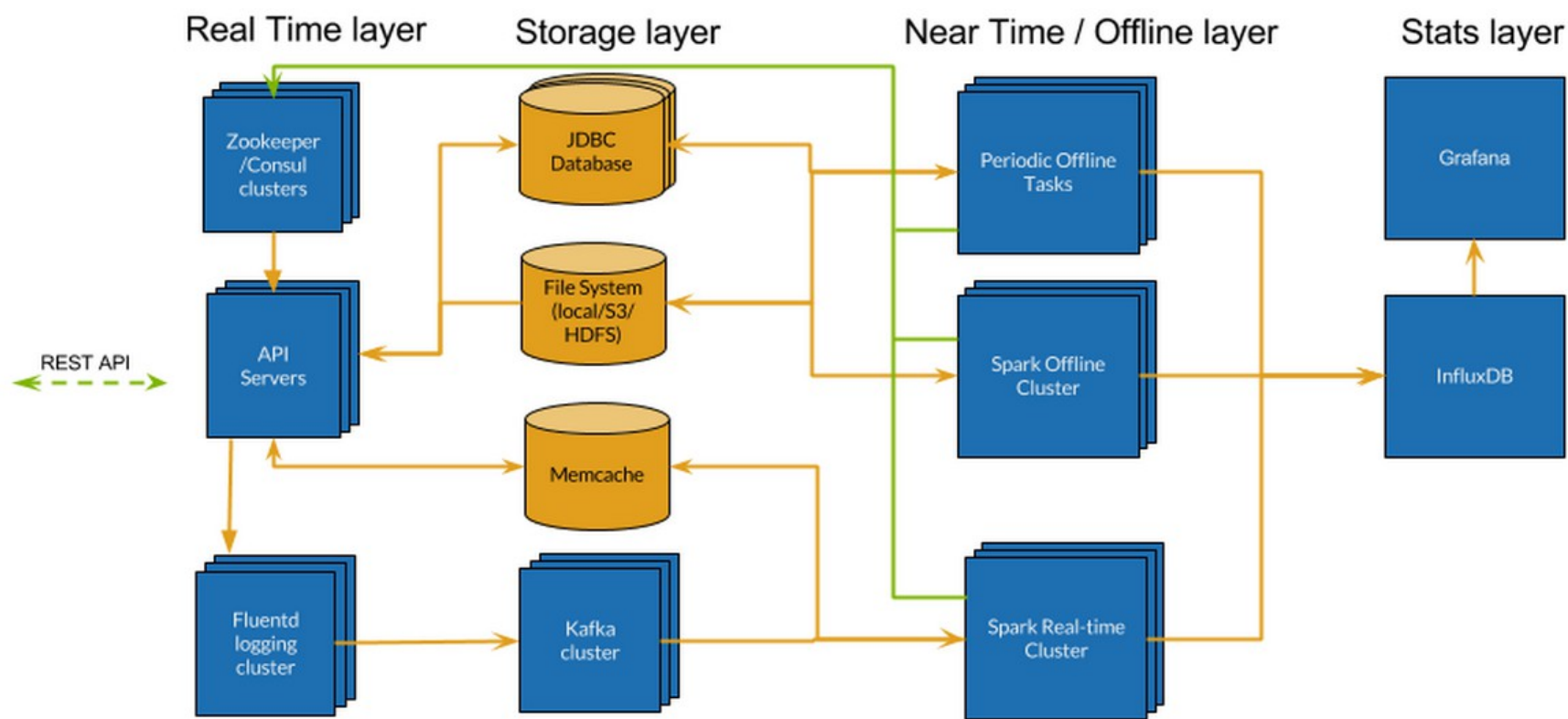
## Bigdata Ecosystem

### Open Source Bigdata Ecosystem

- Query (NOSQL) : Cassandra, HBase, MongoDB and more
- Query (SQL) : Hive, Stinger, Impala, Presto, Shark
- Advanced Analytic : Hadoop, Spark, H2O
- Real time : Storm, Samza, S4, Spark Streaming

# • Introduction

## Bigdata Ecosystem



### Seldon infrastructure

- Real-Time Layer : responsible for handling the live predictive API requests.
- Storage Layer : various types of storage used by other components.
- Near time / Offline Layer : components that run compute intensive or otherwise non-realtime jobs.
- Stats layer : components to monitor and analyze the running system.



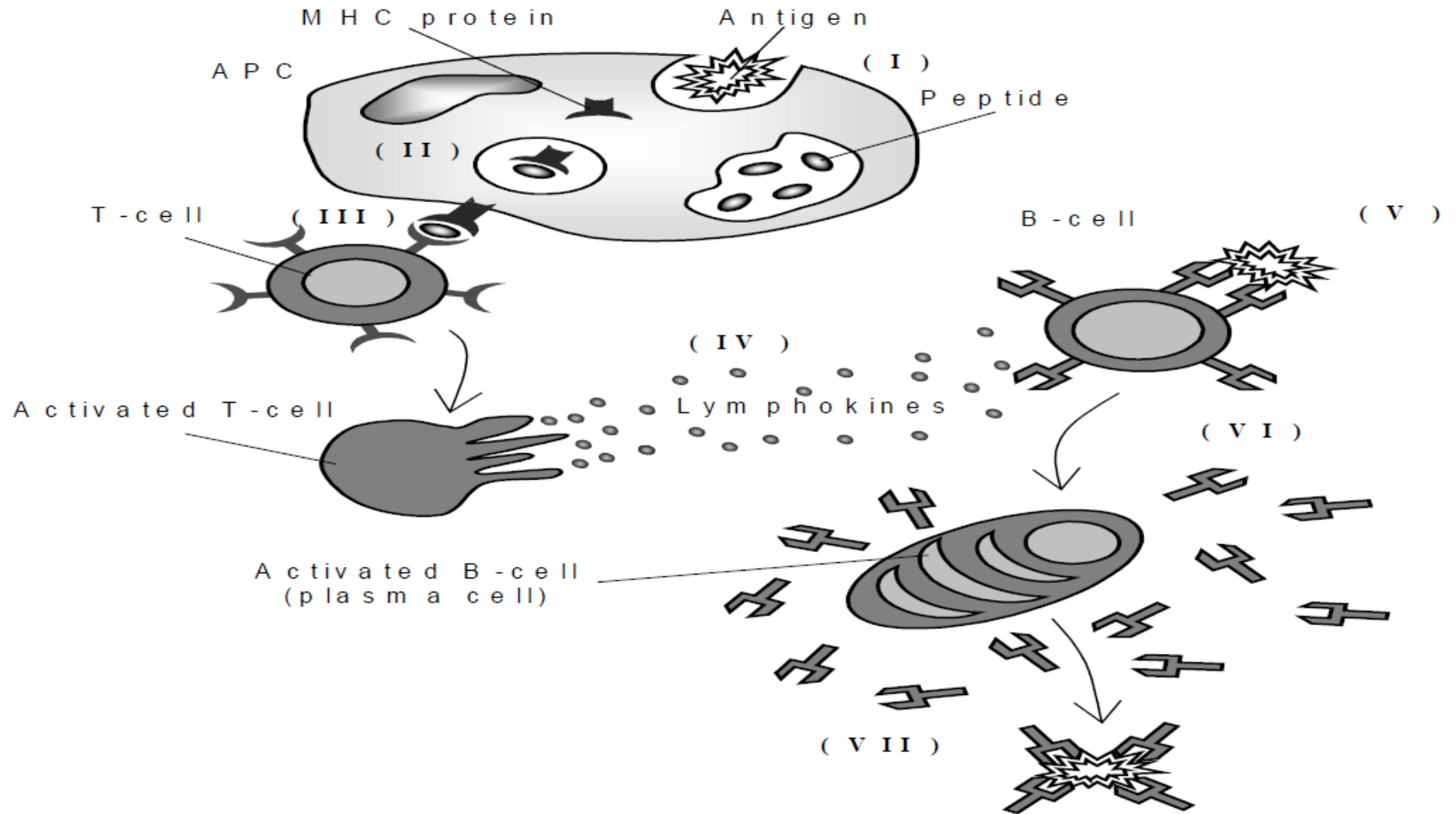
# • Immune Systems     Artificial Immune Systems

AISS are adaptive systems inspired by theoretical immunology and observed immune functions, principles and models, which are applied to complex problem domains

- Immune system needs to be able to differentiate between self and non-self cells
- may result in cell death therefore
  - Some kind of positive selection (Clonal Selection)
  - Some kind of negative selection

# • Immune Systems

## Simple View



# ● Immune Systems

## Lymphocyte(림프구)

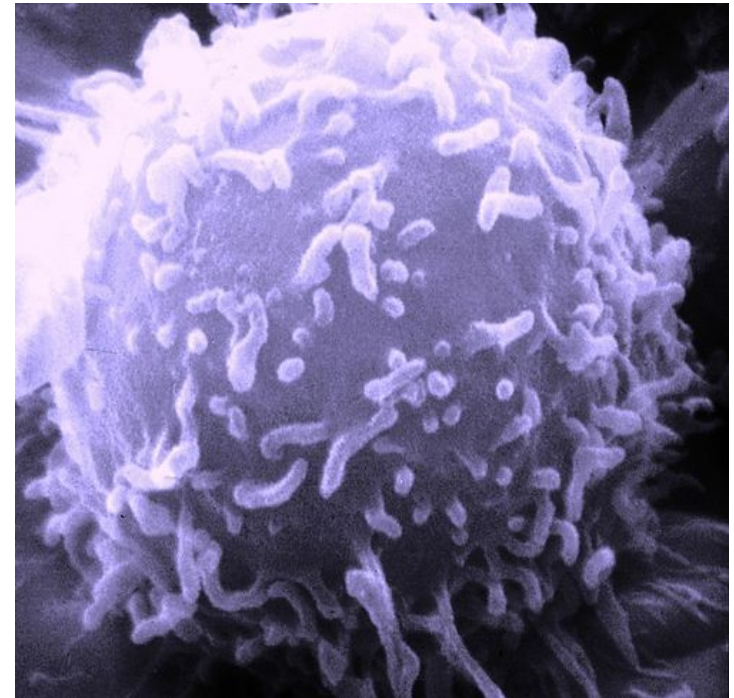
무과립성 백혈구(無顆粒性 白血球, **agranulocyte**)의 일종으로 면역 기능 관여하며 전체 백혈구 중에서도 **30%**를 차지한다.

- T세포(T cell)

- 보조 T세포(Helper T cell)
- 세포독성 T세포(killer T cell)
- 억제 T세포(suppressor T cell)

- B세포(B cell)

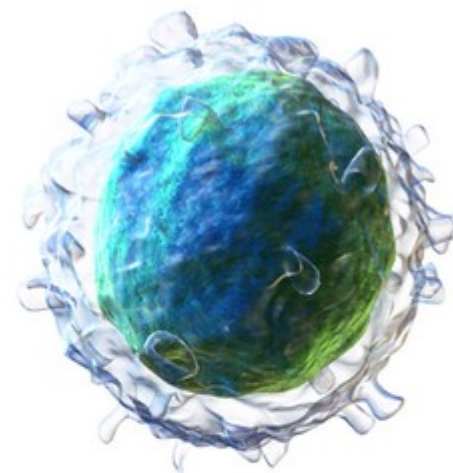
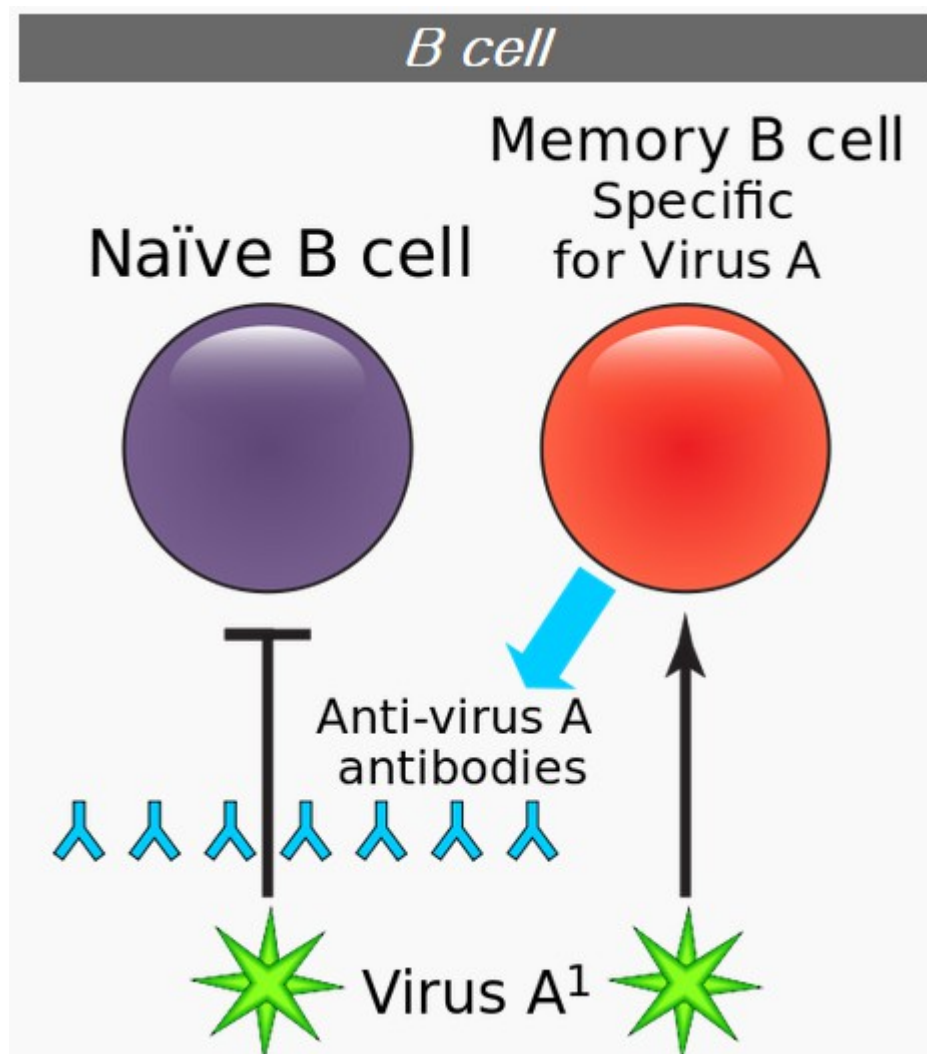
- NK세포(Natural killer cell, NK cell)



# • Immune Systems

## B cell

B 세포(B細胞, B cell)는 림프구 중 항체를 생산하는 세포



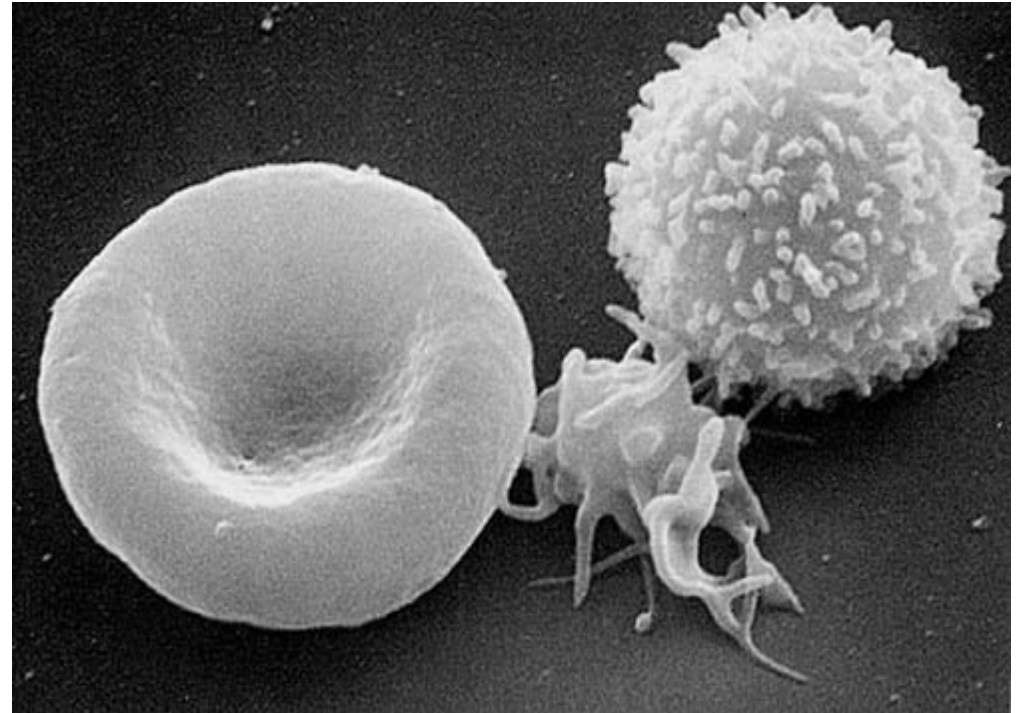
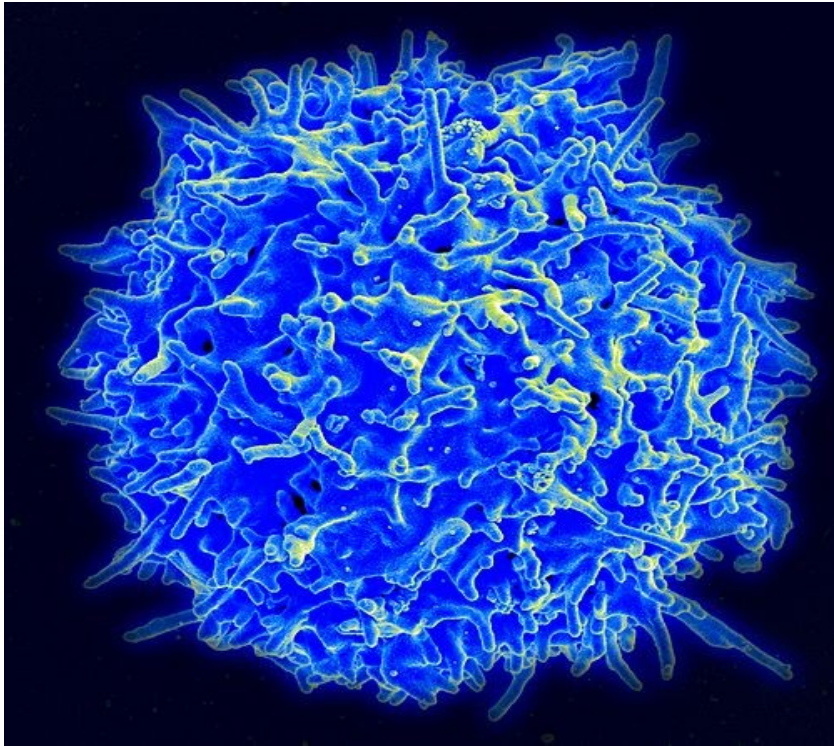
Lymphocyte  
*B cell*



# • Immune Systems

## T cell

**T세포(T細胞, T cell)** 또는 **T림프구(T lymphocyte)**는 항원 특이적인 적응 면역을 주관하는 림프구의 하나이다. 가슴샘(**Thymus**)에서 성숙되기 때문에 첫글자를 따서 **T세포**라는 이름이 붙었다. 전체 림프구 중 약 **4분의 3**이 **T세포**

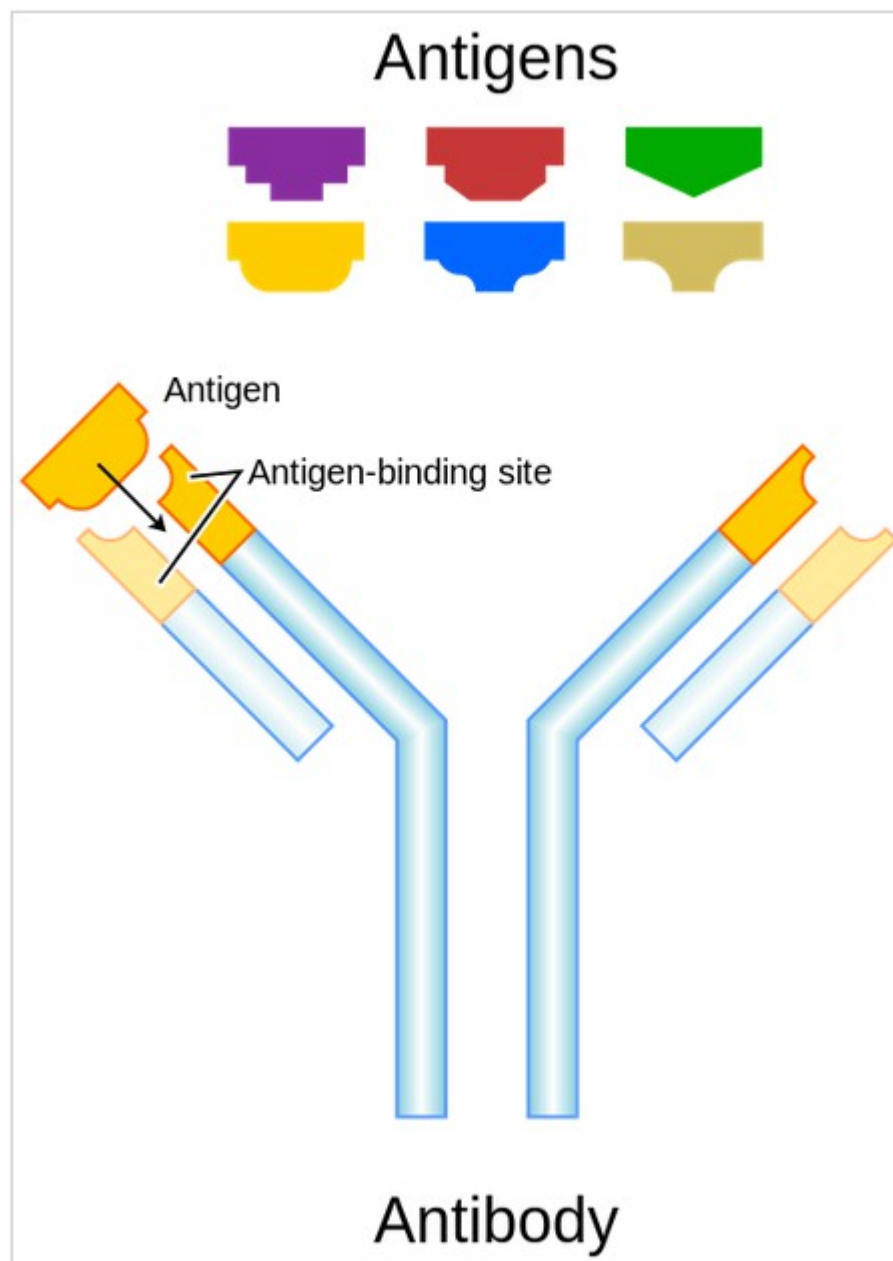
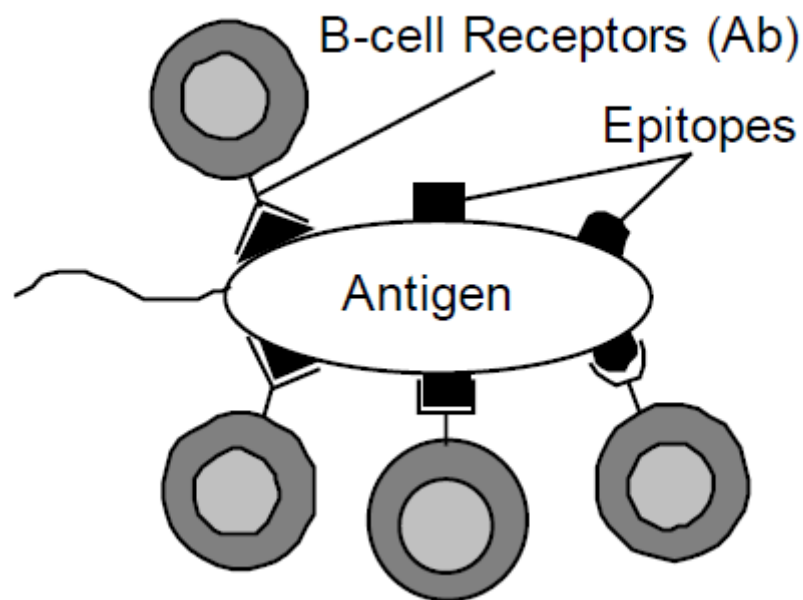


**T세포**는 아직 항원을 만나지 못한 미접촉 T세포와, 항원을 만나 성숙한 효과 T세포(보조 **T세포**, 세포독성 **T세포**, 자연살상 **T세포**), 그리고 기억 T세포로 분류

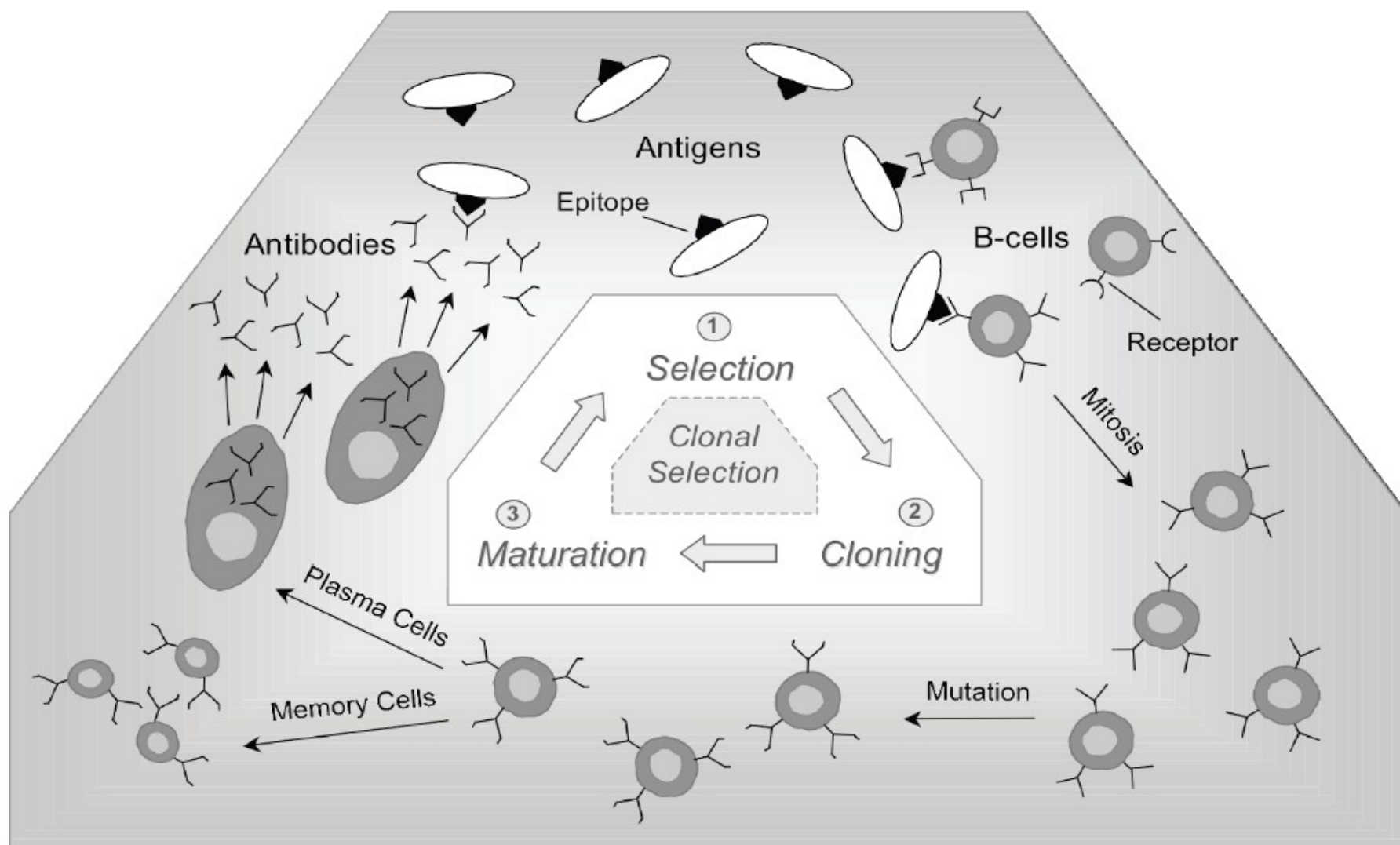
# • Immune Systems

## Antibody, Antigen

each antibody can recognize a single antigen



# • Immune Systems      Biological Immune System



# • Immune Systems

## Danger Theory

- Proposed by Polly Matzinger, around 1995
- Traditional self/non-self theory doesn't always match observations
  - Immune system always responds to non-self
  - Immune system always tolerates self
- Antigen-presenting cell(APC): T-cell activation by APCs
- Danger theory relates innate and adaptive immune systems
  - Tissues induce tolerance towards themselves
  - Tissues protect themselves and select class of response



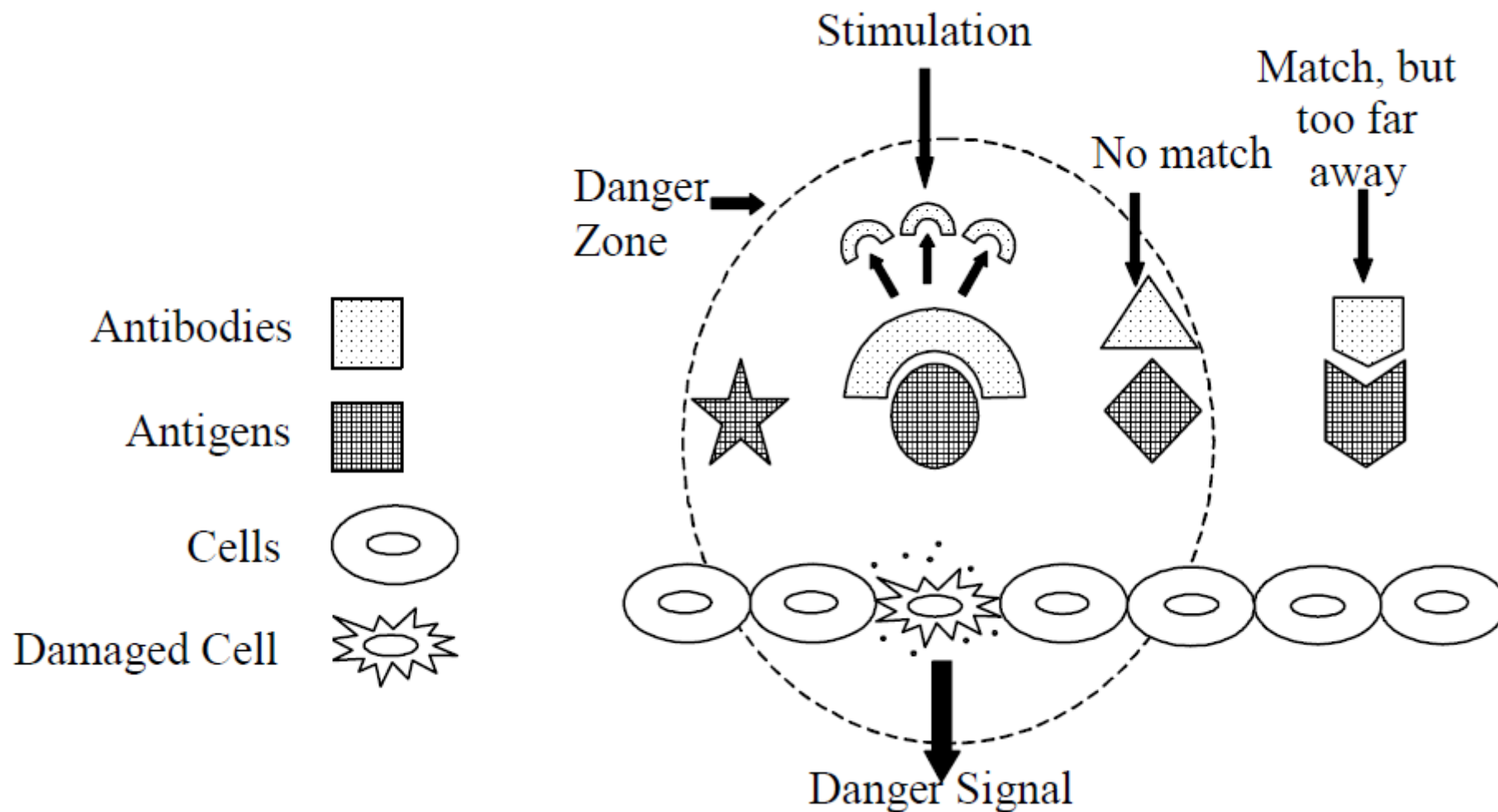
# • Immune Systems

## Danger Theory

- Tissues induce tolerance by
  - Lymphocytes receive 2 signals
    - antigen/lymphocyte binding
    - antigen is properly presented by APC
  - Signal 1 WITHOUT signal 2 : lymphocyte death
- Tissues protect themselves
  - Alarm Signals activate APCs
    - Alarm signals come from
      - Cells that die unnaturally
      - Cells under stress
  - APCs activate lymphocytes
- Tissues dictate response type
  - Alarm signals may convey information

# • Immune Systems

## Danger Theory



# • Immune Systems

## Artificial Immune System

### Artificial Immune Systems

- Vectors

$$\mathbf{Ab} = \{Ab_1, Ab_2, \dots, Ab_L\}$$

$$\mathbf{Ag} = \{Ag_1, Ag_2, \dots, Ag_L\}$$

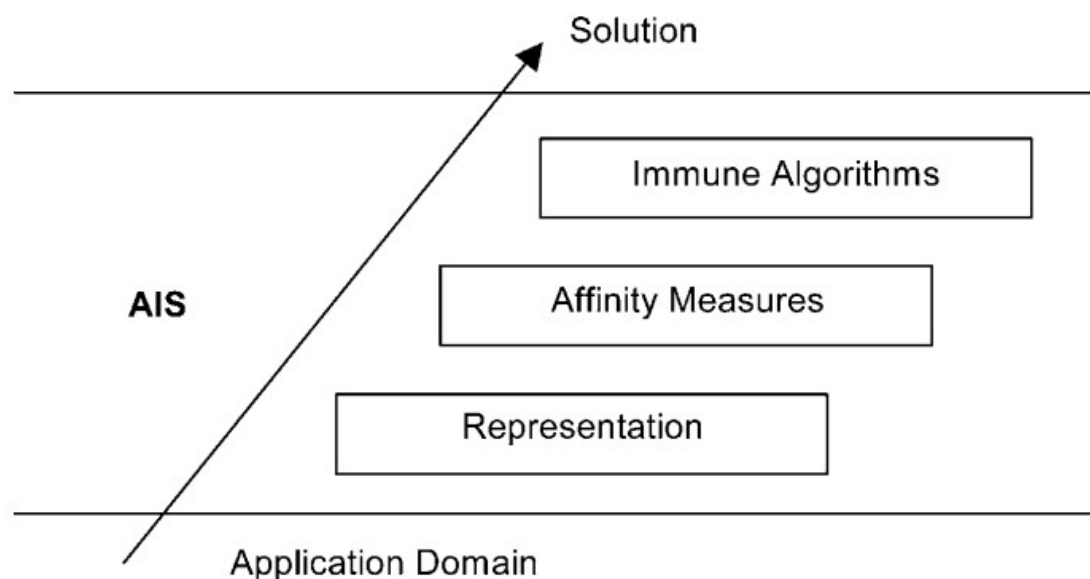
$$D = \sqrt{\sum_{i=1}^L (Ab_i - Ag_i)^2}$$

- Real-valued shape-space

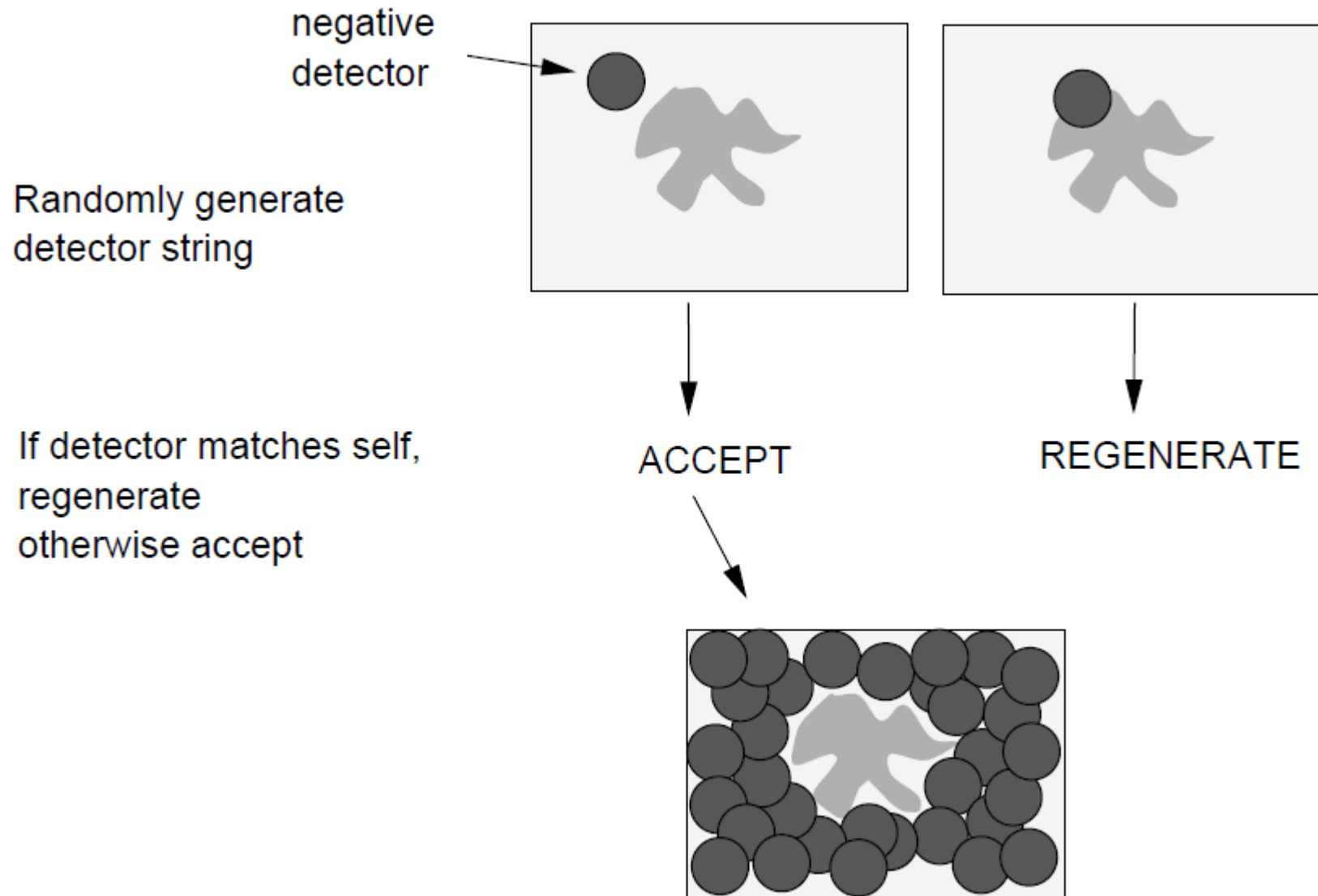
- Integer shape-space

- Binary shape-space

- Symbolic shape-space

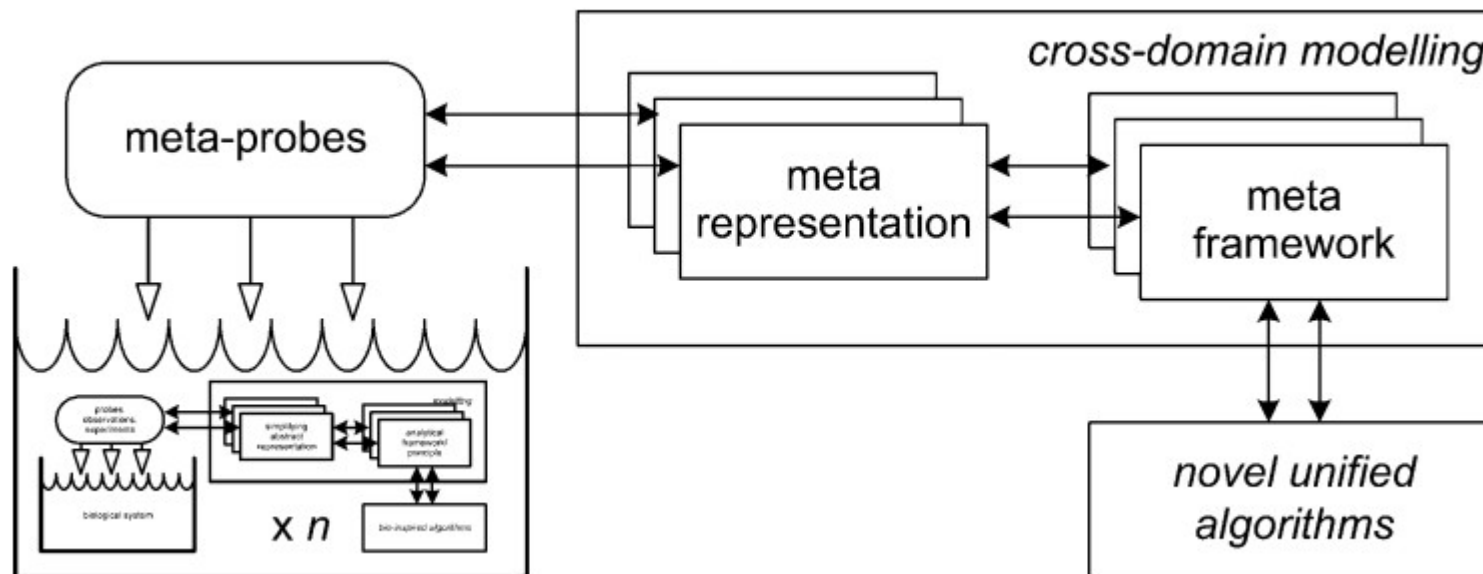


# • Immune Systems      Artificial Immune System

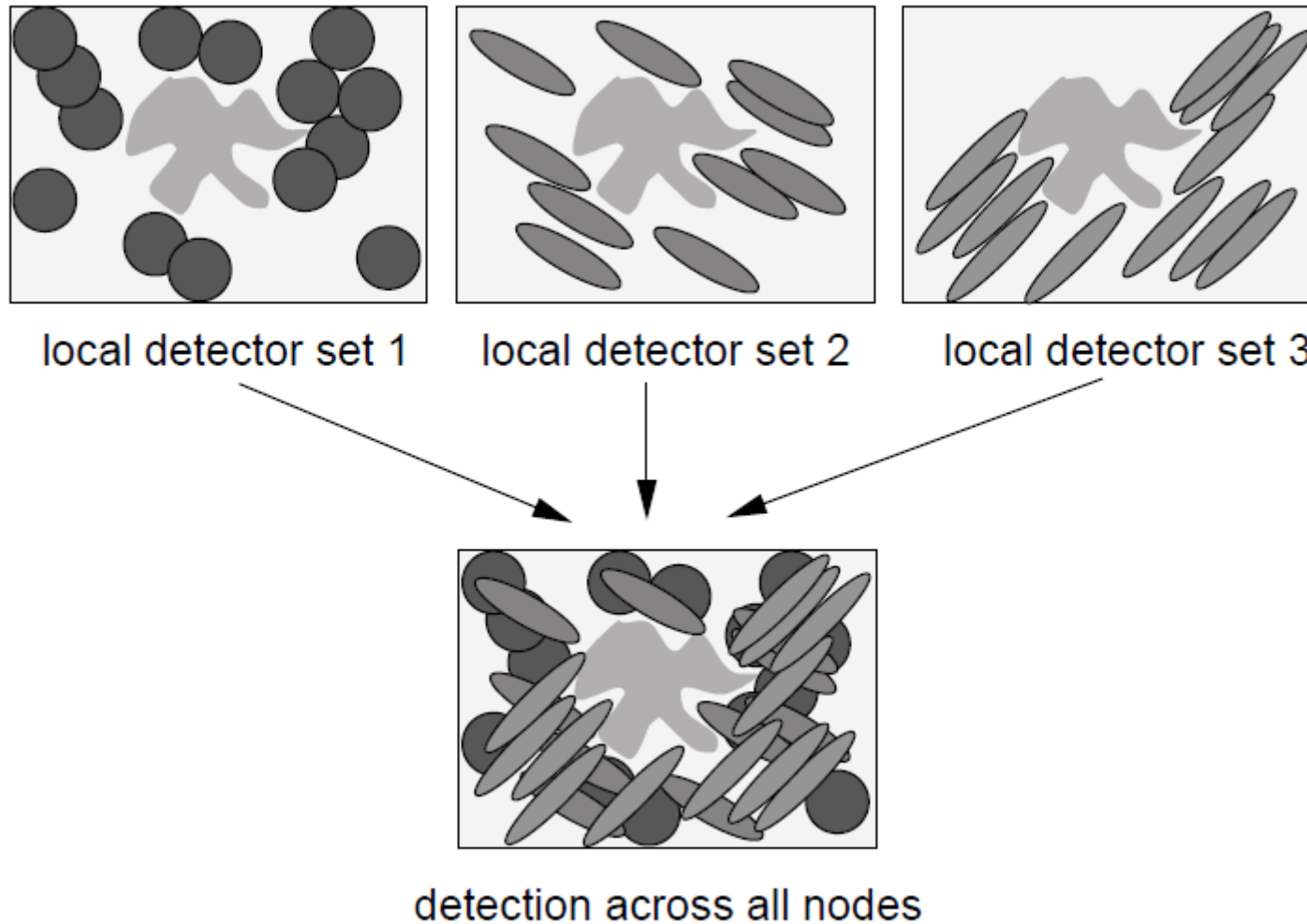


# • Immune Systems      Artificial Immune System

## Meta-Frameworks

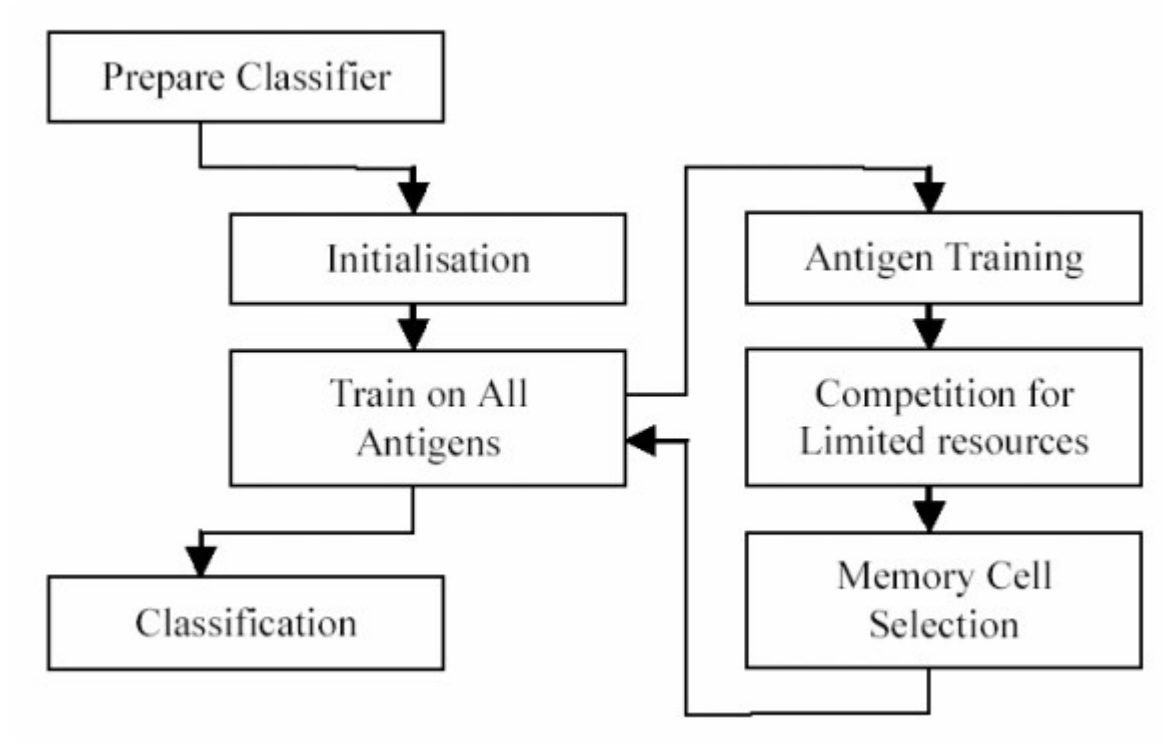


# • Immune Systems      Artificial Immune System



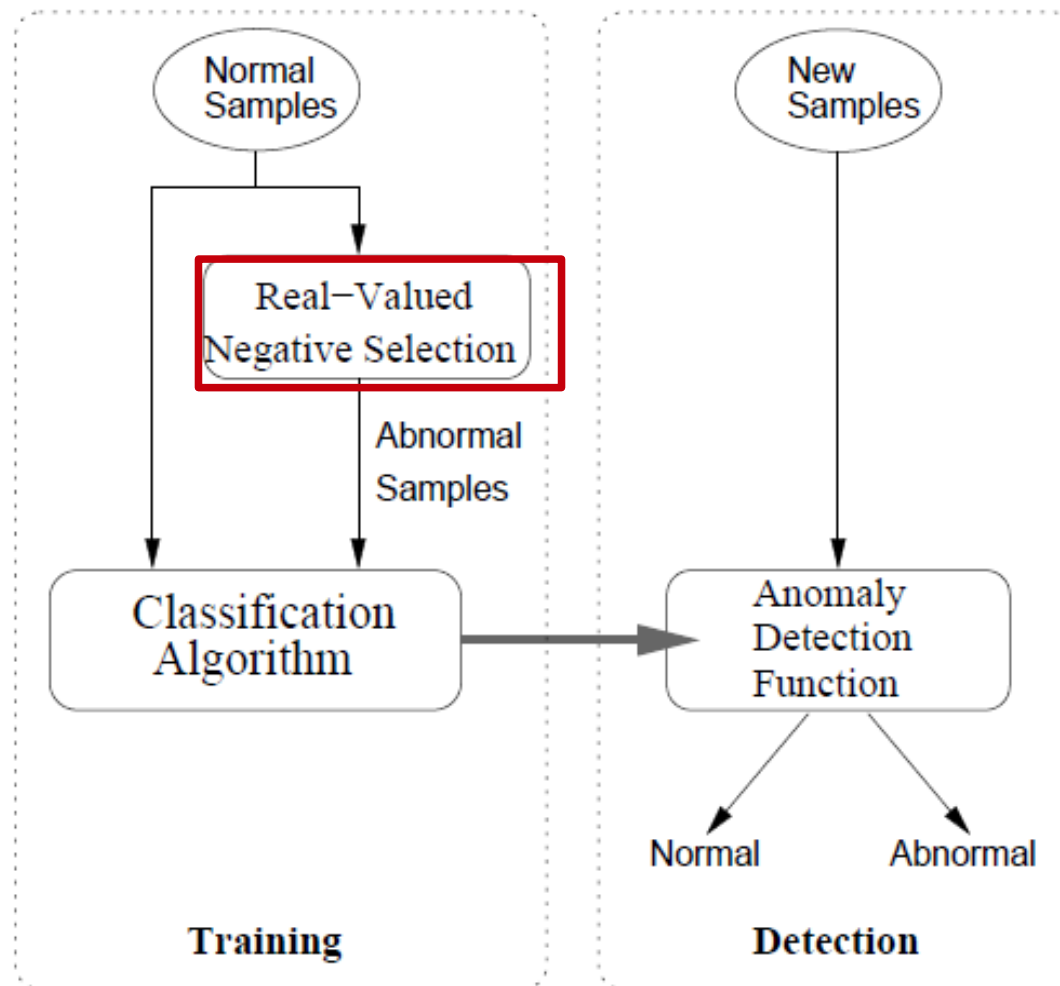
# • Immune Systems

## Artificial Immune Recognition System



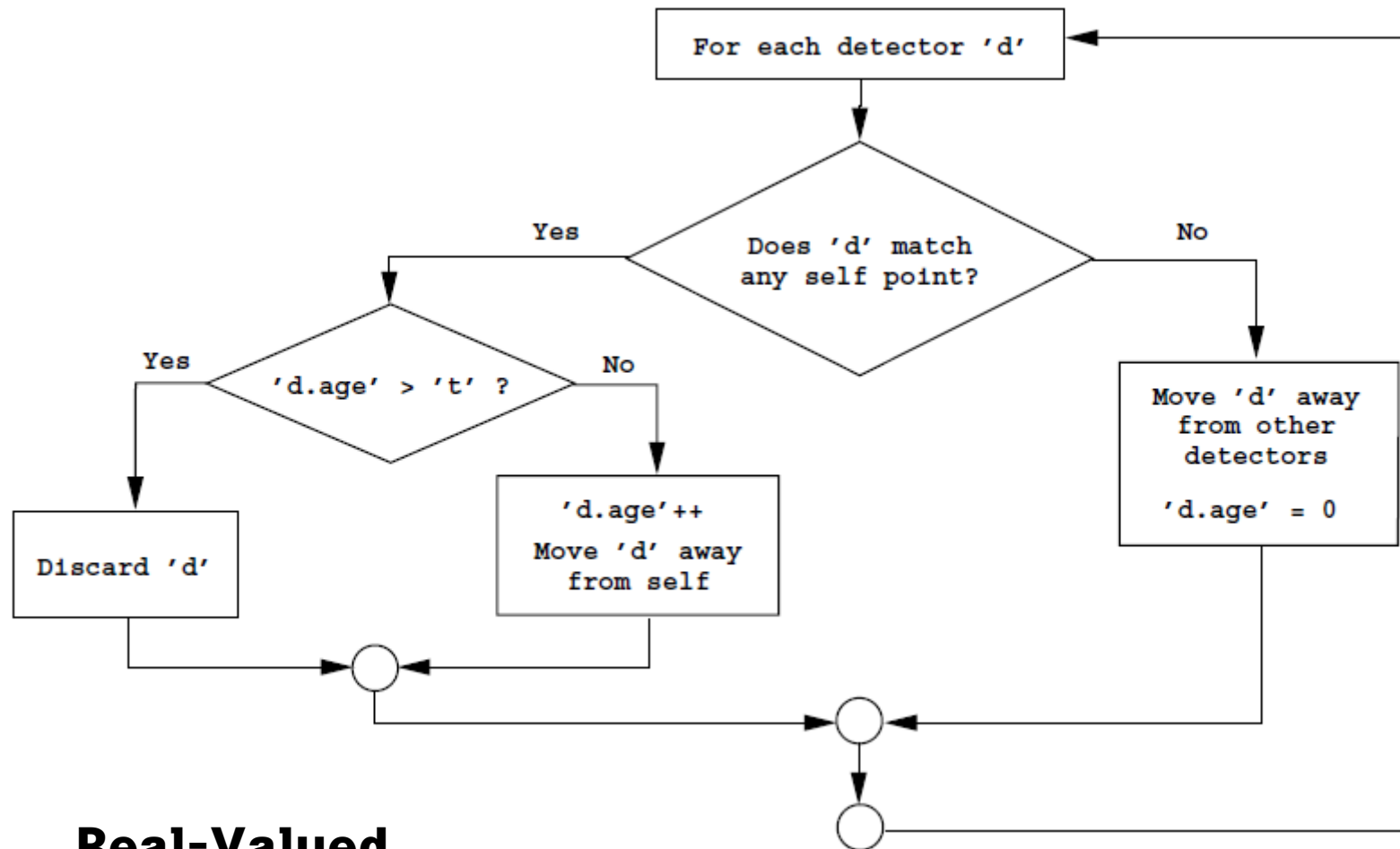
# • Immune Systems

## Hybrid Immune Learning





# • Immune Systems Hybrid Immune Learning

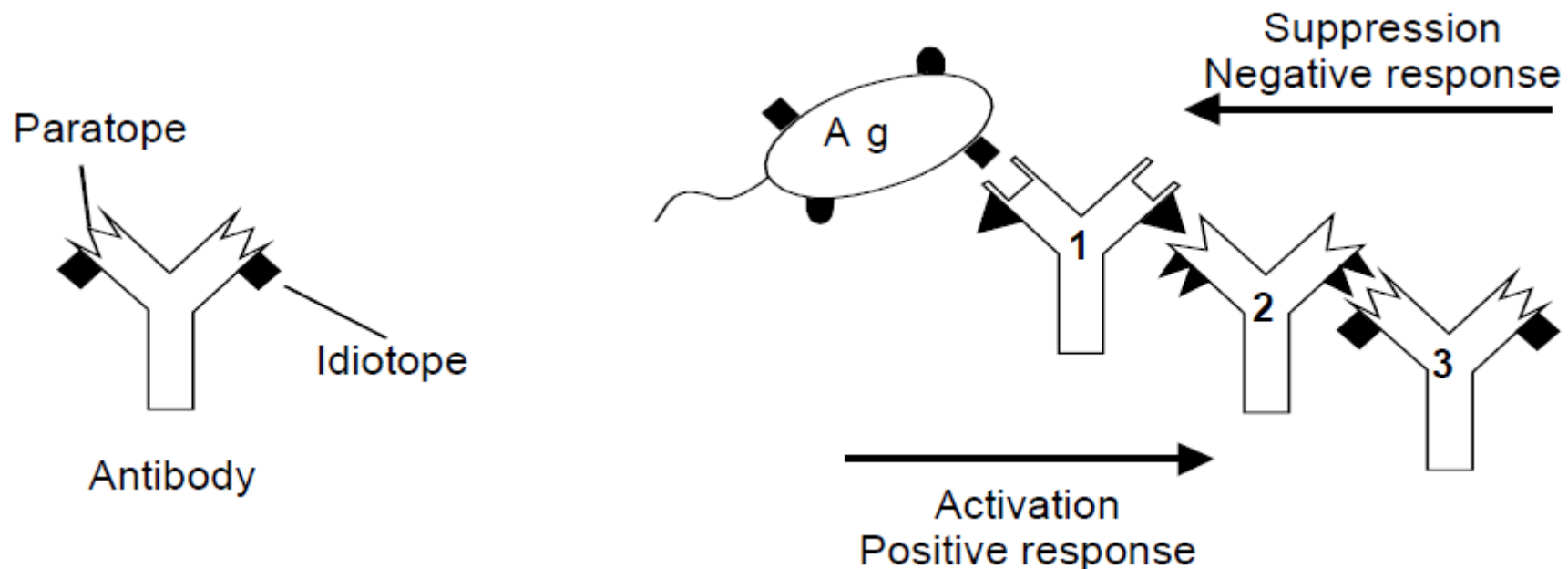


**Real-Valued  
Negative Selection**

# • Immune Systems

## Immune Network Theory

- Idiotypic network (Jerne, 1974)
- B cells co-stimulate each other
  - Treat each other a bit like antigens
- Creates an immunological memory



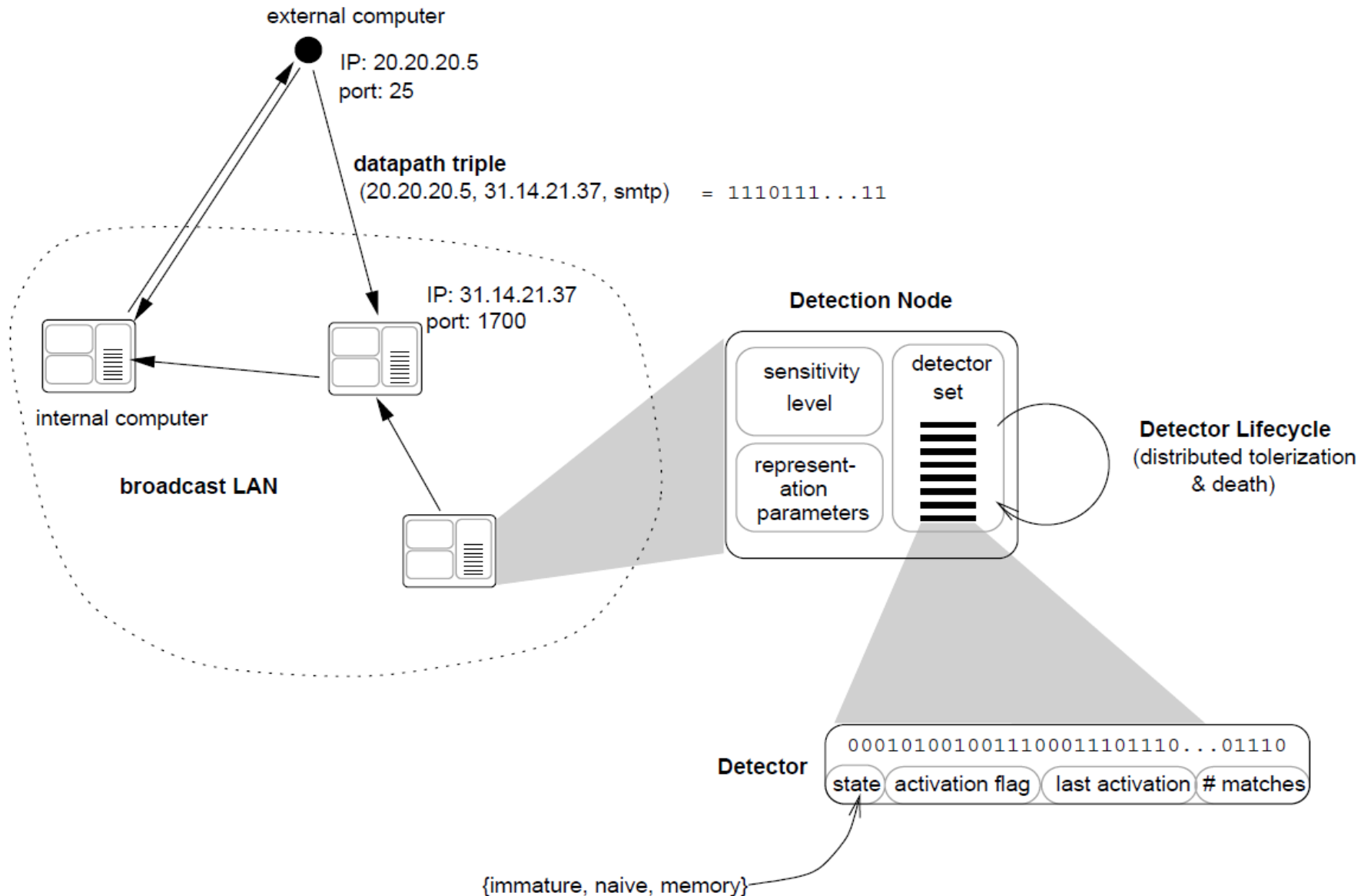
# • Immune Systems

## Non-self Detection Principle

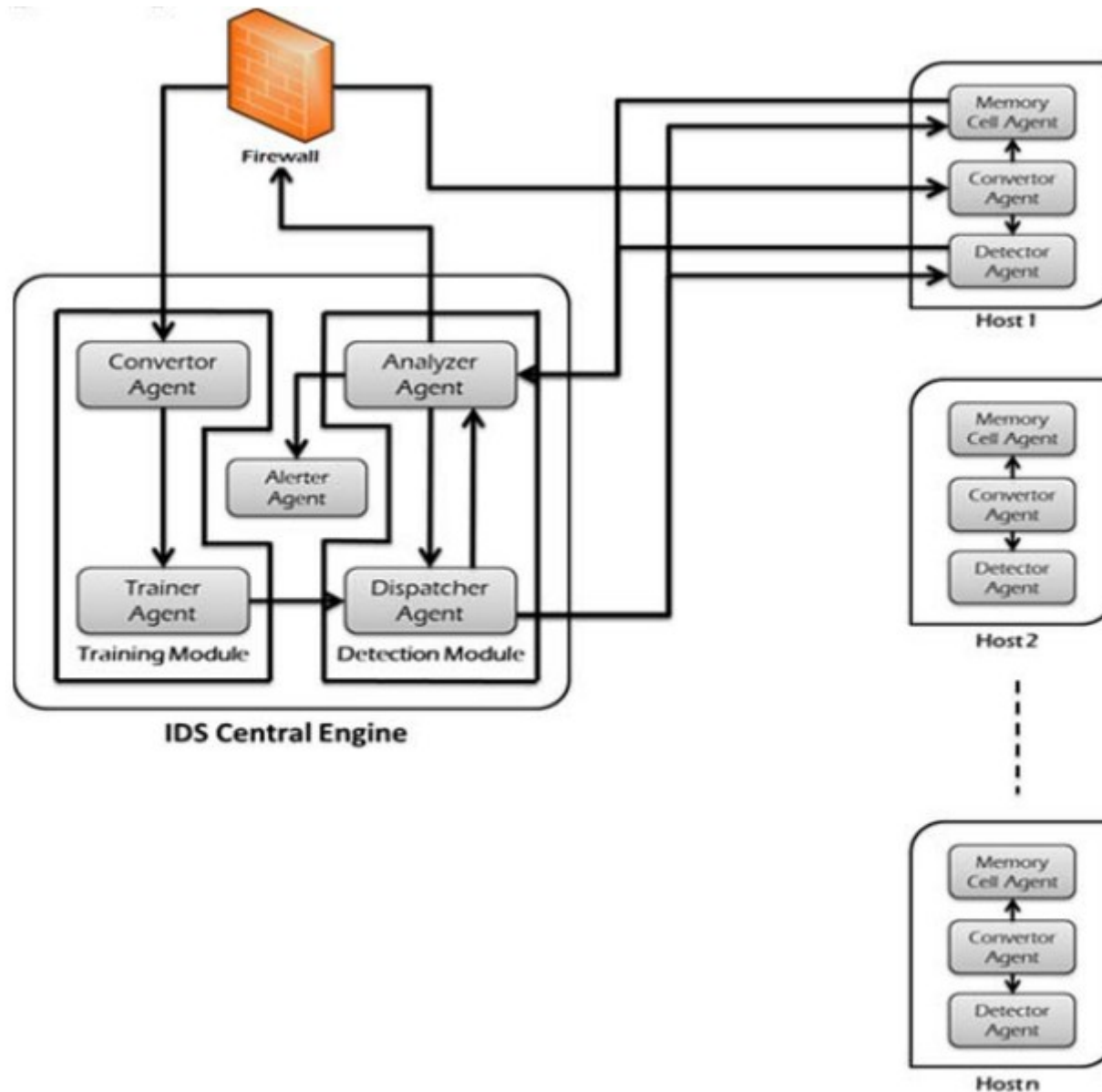
For natural immune system, all cells of body are categorized as two types of **self** and **non-self**. The immune process is to **detect non-self** from cells.

use the Positive Selection Algorithm (**PSA**) to perform the **non-self detection** for recognizing the malicious executable.

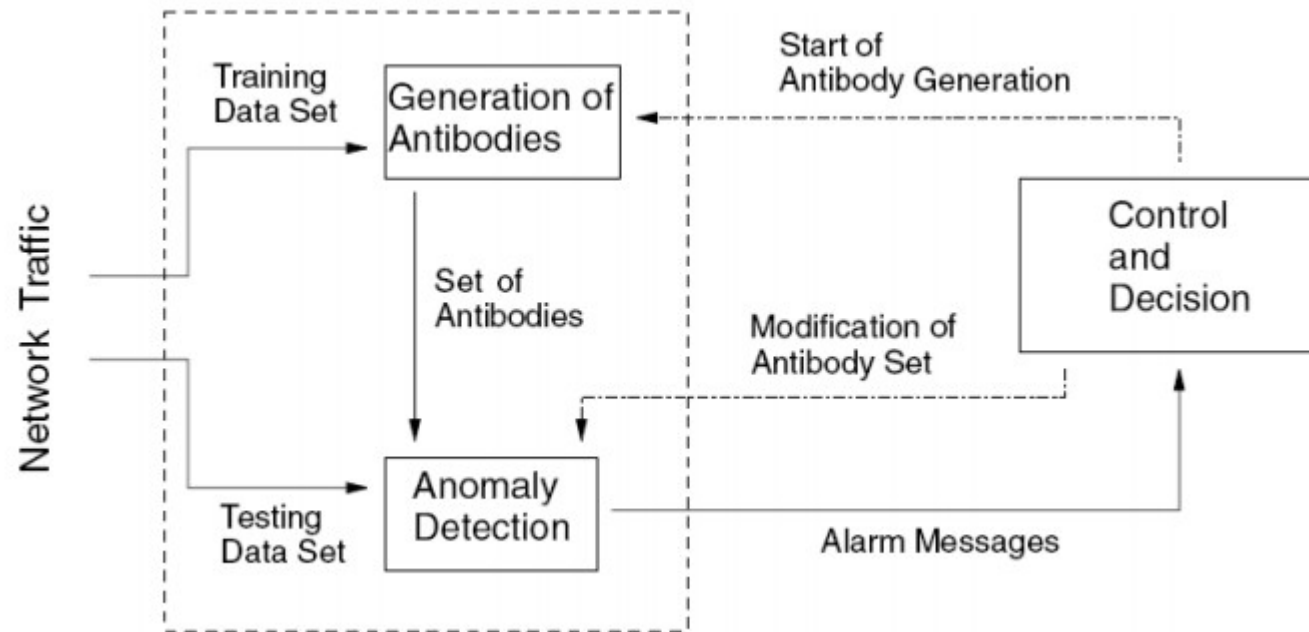
# Immune Systems Network Security



# • Immune Systems Intrusion Detection Systems

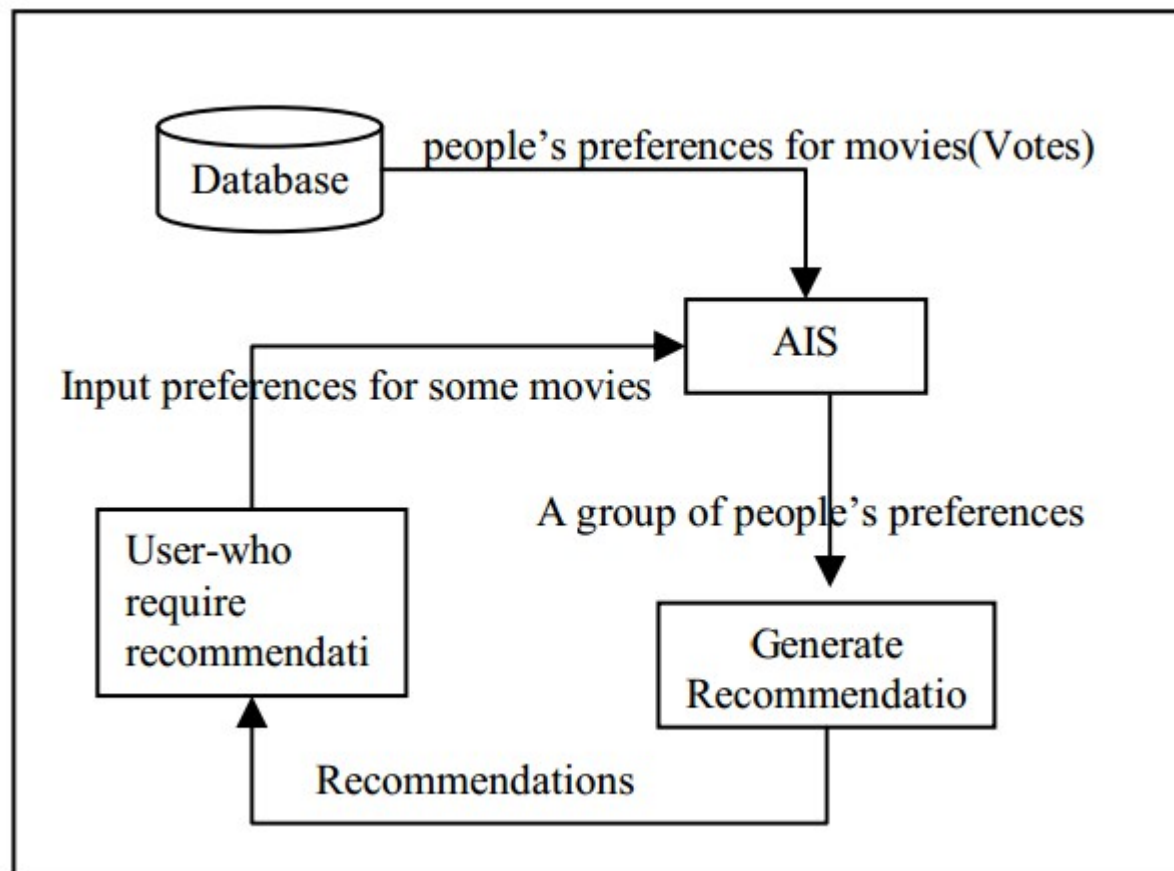


# • Immune Systems Network Security



**Architecture of anomaly detection system.**

# • Immune Systems      Movie Recommendation Systems



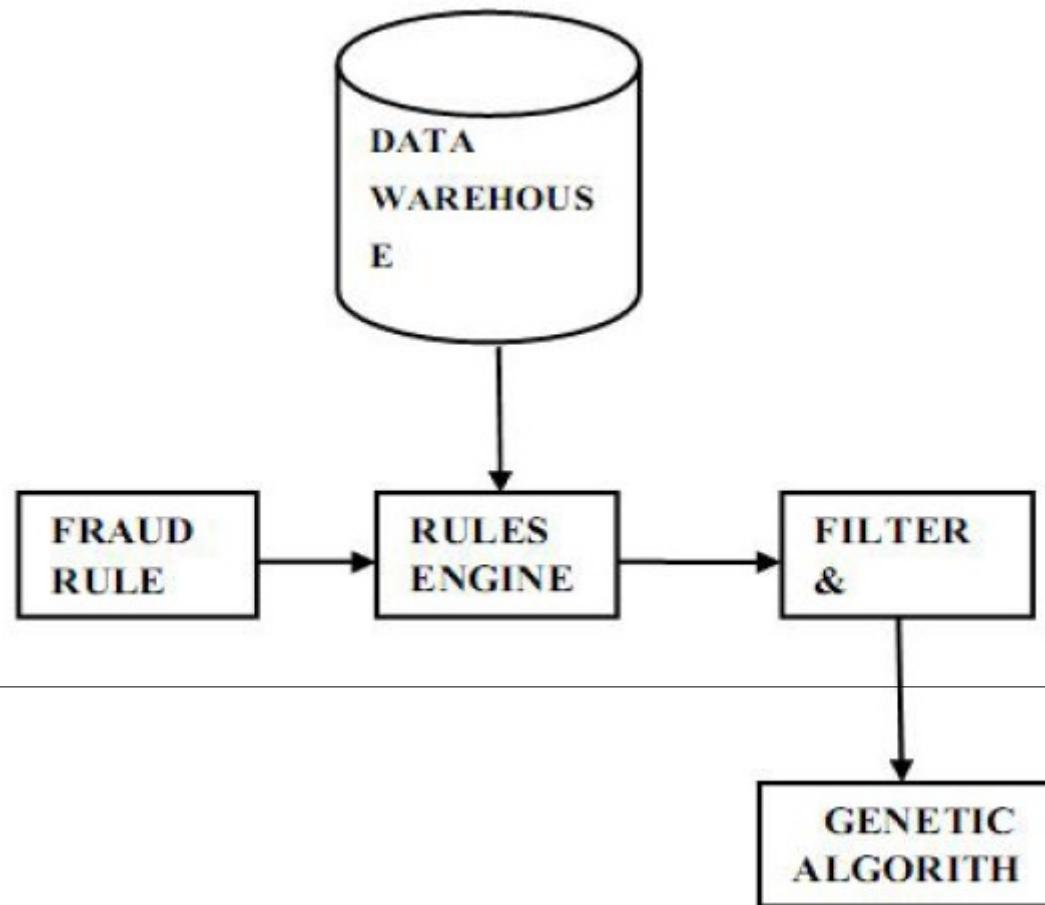
# • Machine Learning | Types

- Supervised learning : 지도학습
  - Data의 종류를 알고 있을 때(Category, Labeled)
  - ex: spam mail
- Unsupervised : 비지도학습
  - Data의 종류는 모르지만 패턴을 알고 싶을 때
  - SNS, Twitter
- Semi-supervised learning : 지도학습 + 비지도학습
- Reinforcement learning : 강화학습
  - 잘못된 것을 다시 피드백
- Evolutionary learning : 진화학습(GA, AIS)
- Meta Learning : Landmark of data for classifier



# • Machine Learning

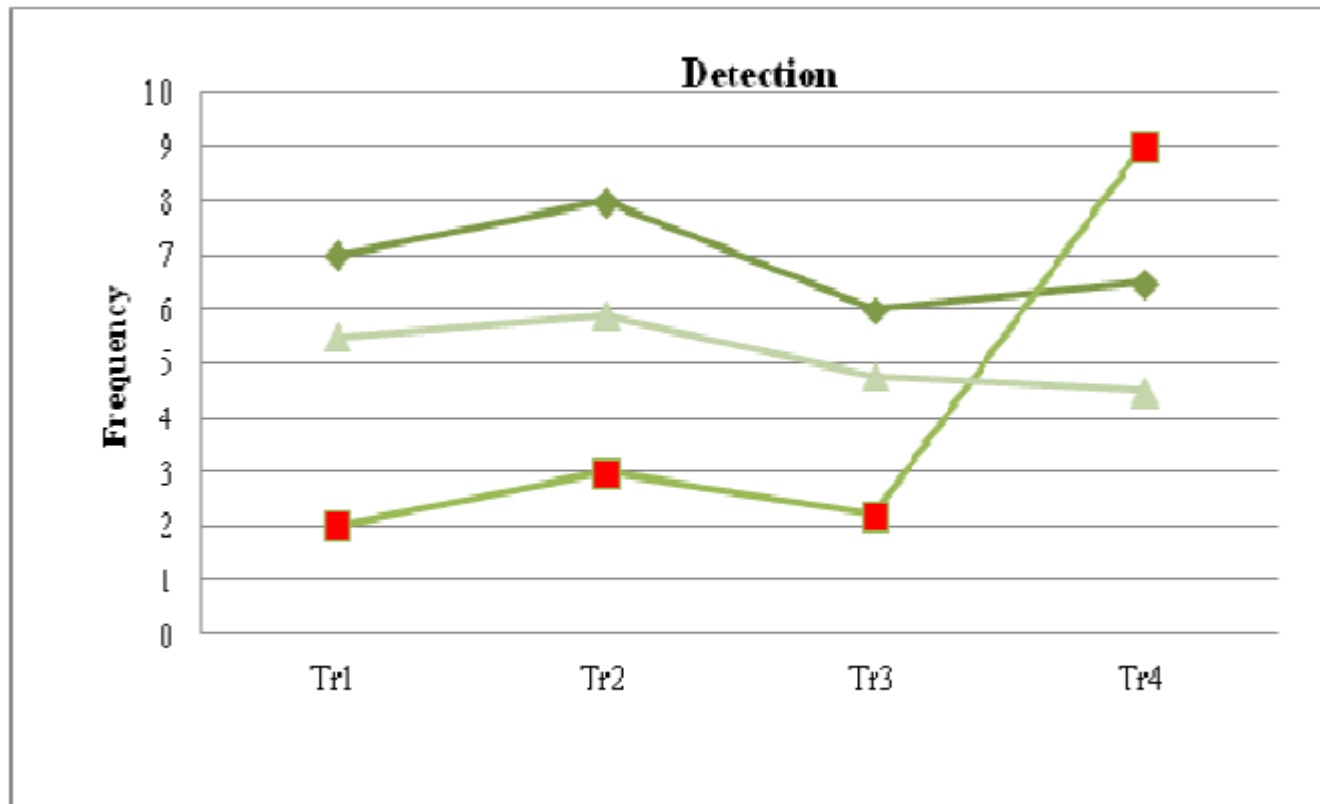
## Genetic algorithm



# • Machine Learning

## Genetic algorithm

### Abnormal Behavior

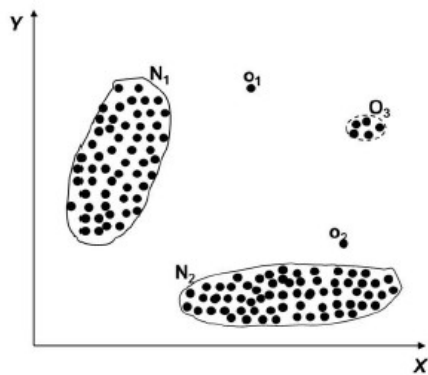


# Machine Learning

## Types of Anomaly

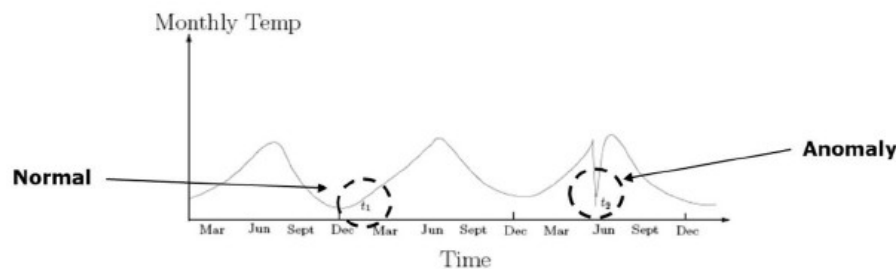
### Point Anomalies

- An individual data instance is anomalous w.r.t. the data



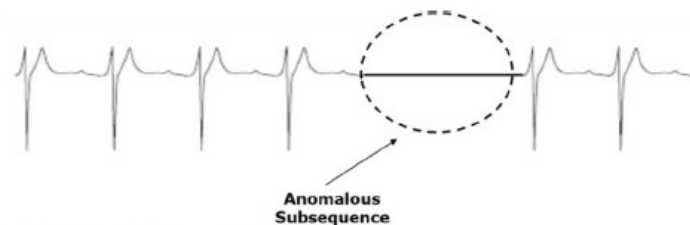
### Contextual Anomalies

- An individual data instance is anomalous within a context
- Requires a notion of context
- Also referred to as conditional anomalies\*



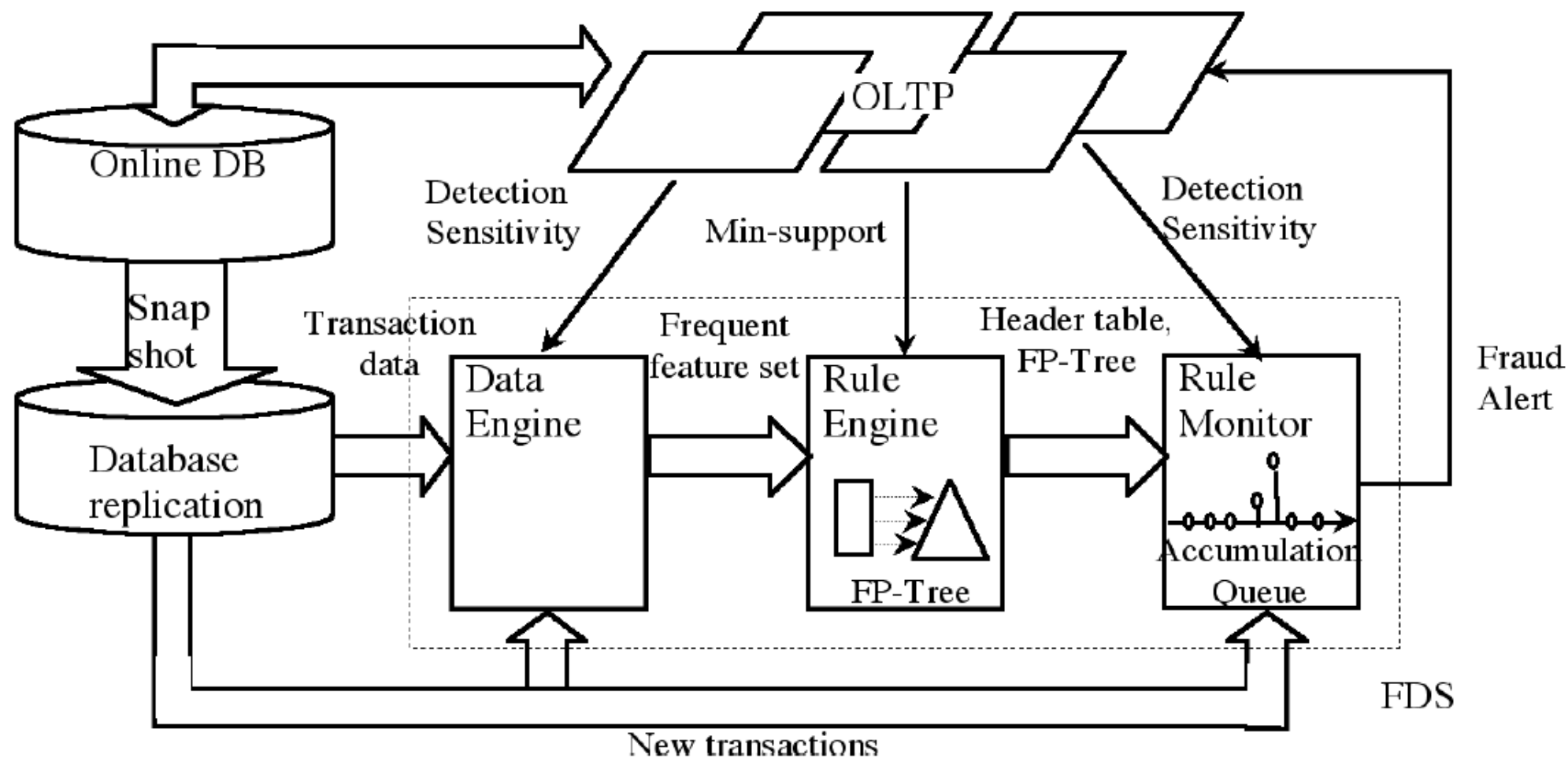
### Collective Anomalies

- A collection of related data instances is anomalous
- Requires a relationship among data instances
  - Sequential Data
  - Spatial Data
  - Graph Data
- The individual instances within a collective anomaly are not anomalous by themselves



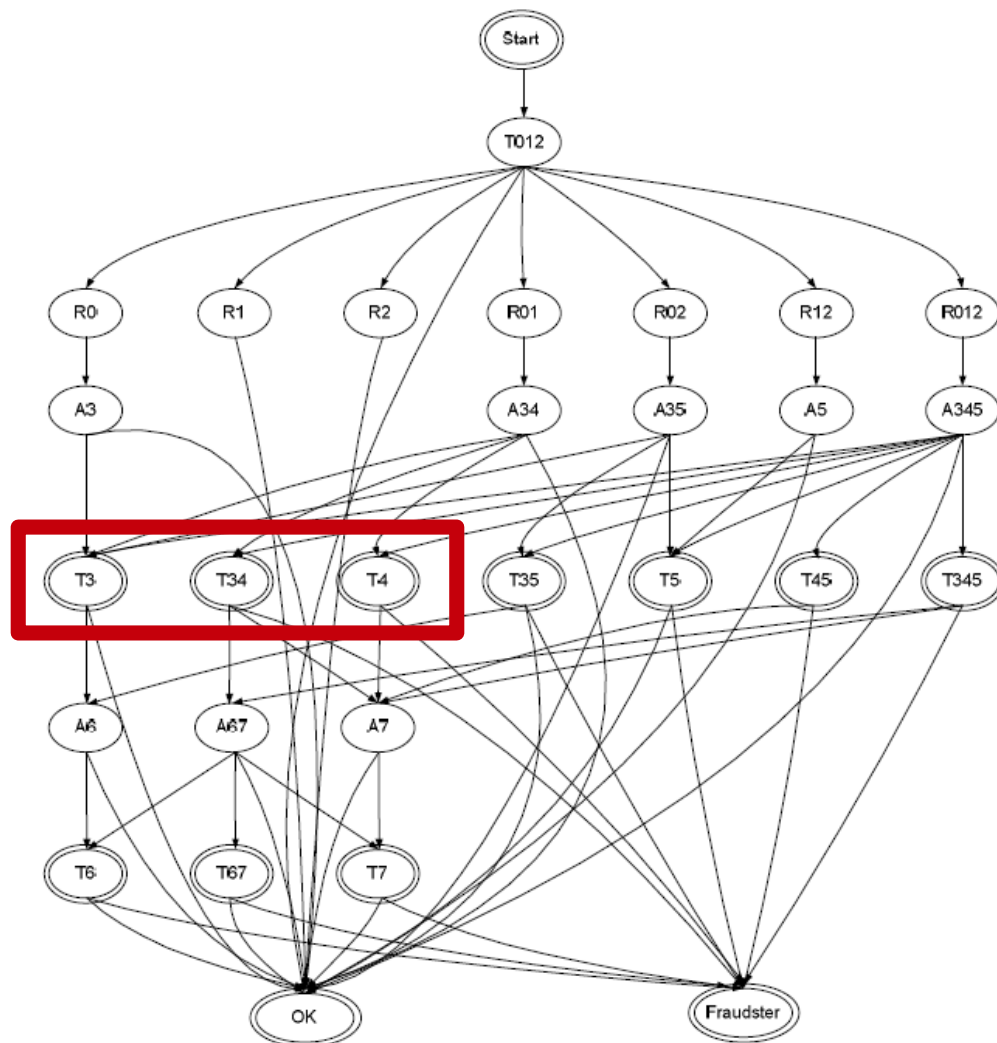
사이버보안연구본부 2014

# Machine Learning Association Rule Mining

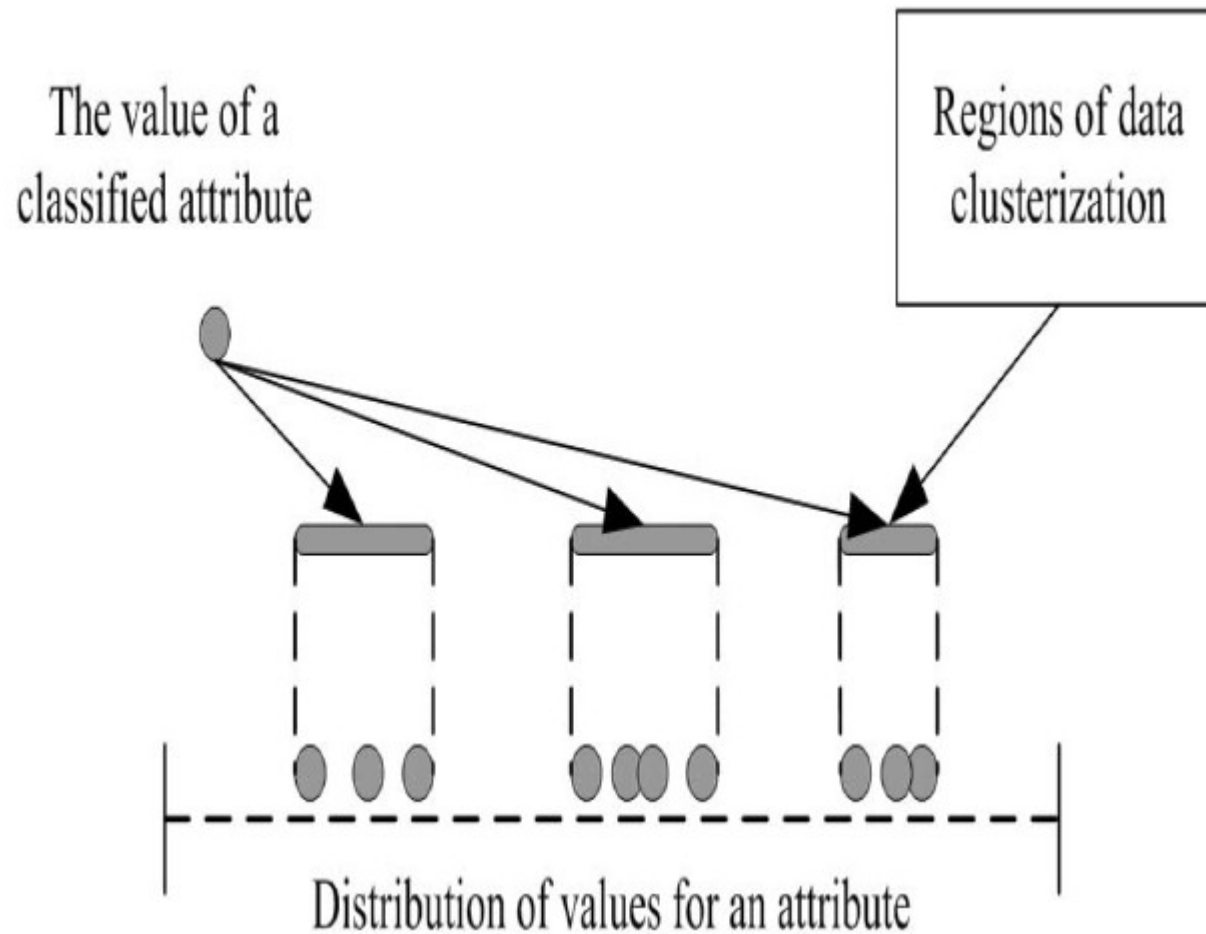


# • Machine Learning Finite State Automata (FSA)

Since the tests in can be **grouped**, the states can represent the several tests being performed at **the same time**. For example, T34 means that T3 and T4 can be done simultaneously



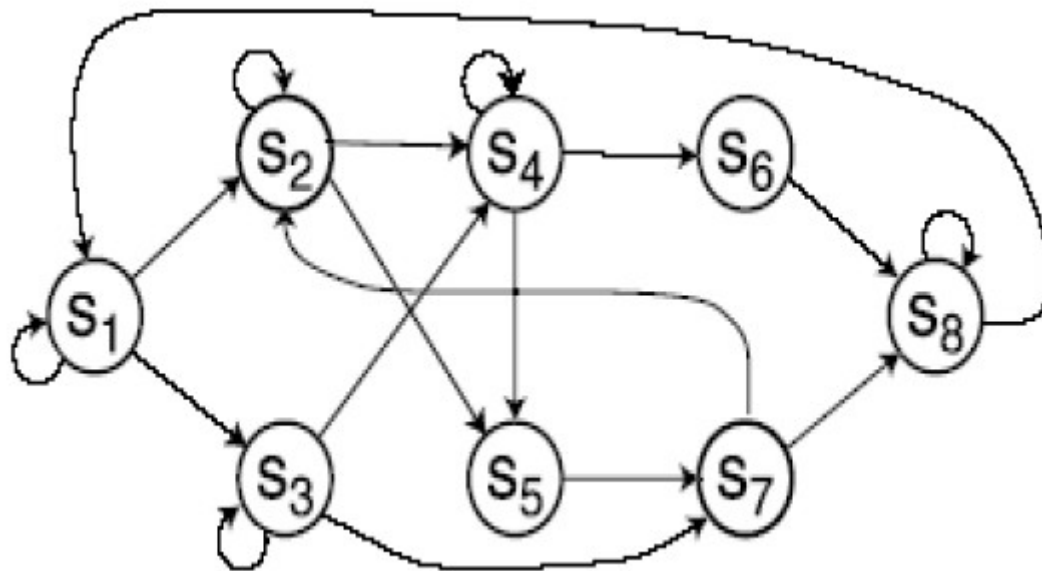
# • Machine Learning Clustering



# • Machine Learning Hidden Markov

## Sequence Based Algorithm

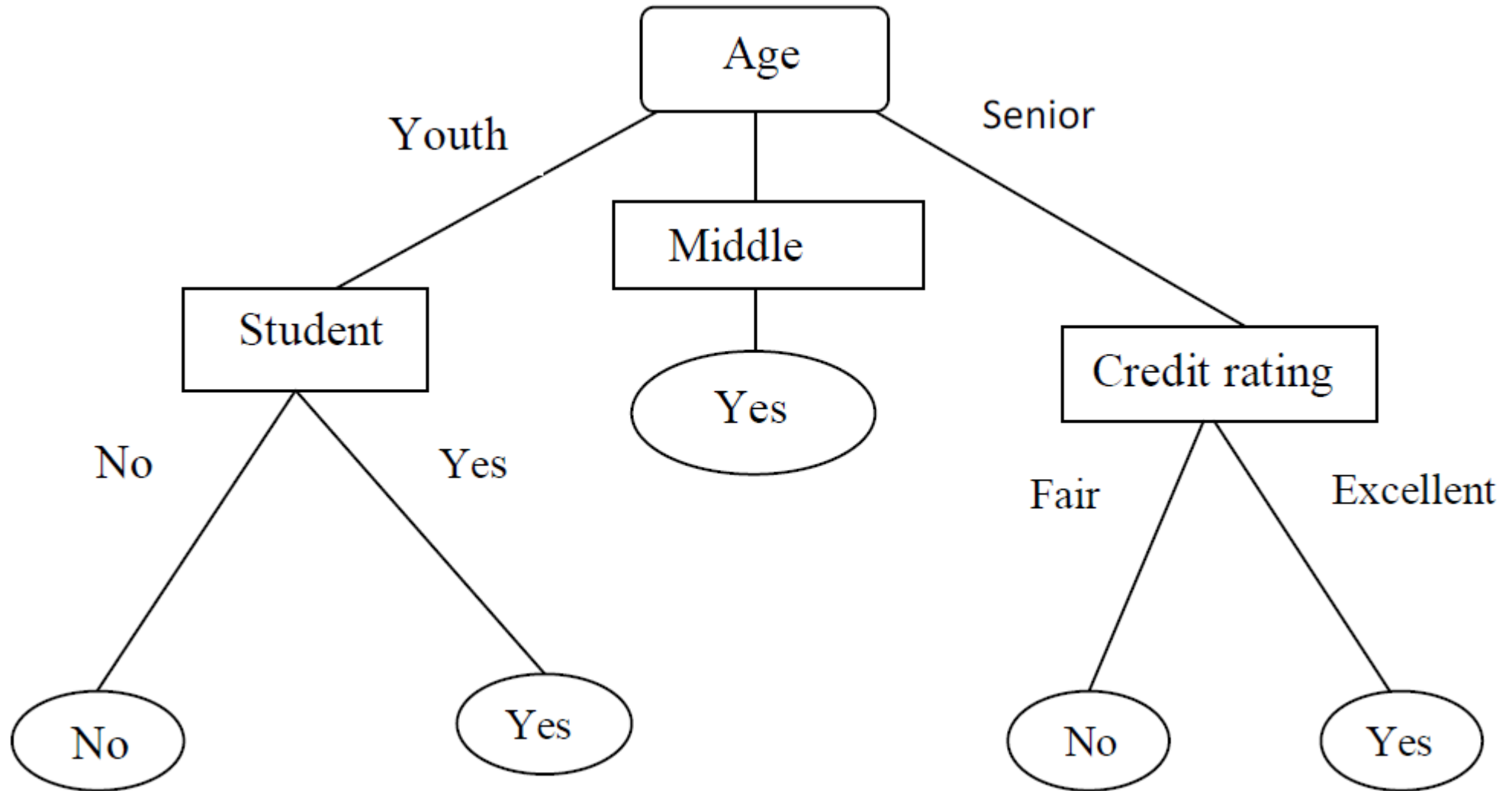
- Certain fraudulent activities may **not be detectable with instance** based algorithms
- **small amount** of money, **instance based** algorithms will **fail** to detect the fraud



# • Machine Learning

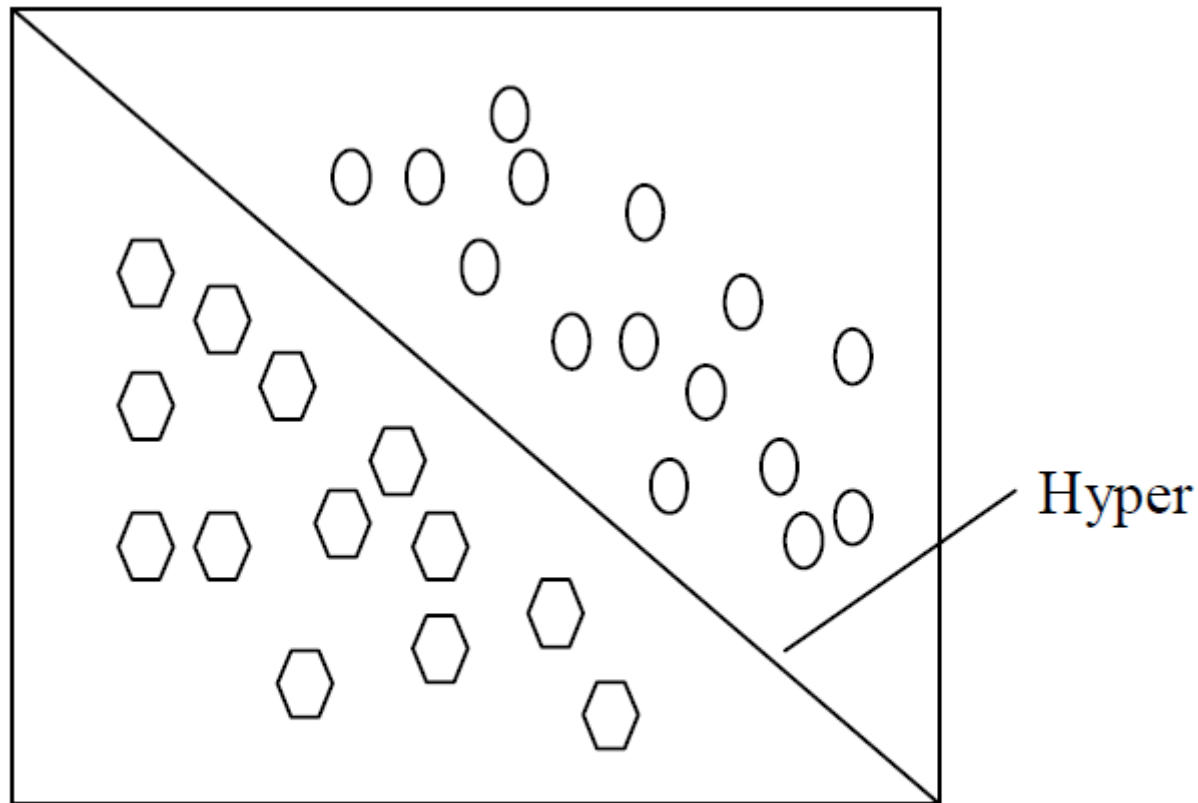
## Decision Tree

Profiling?





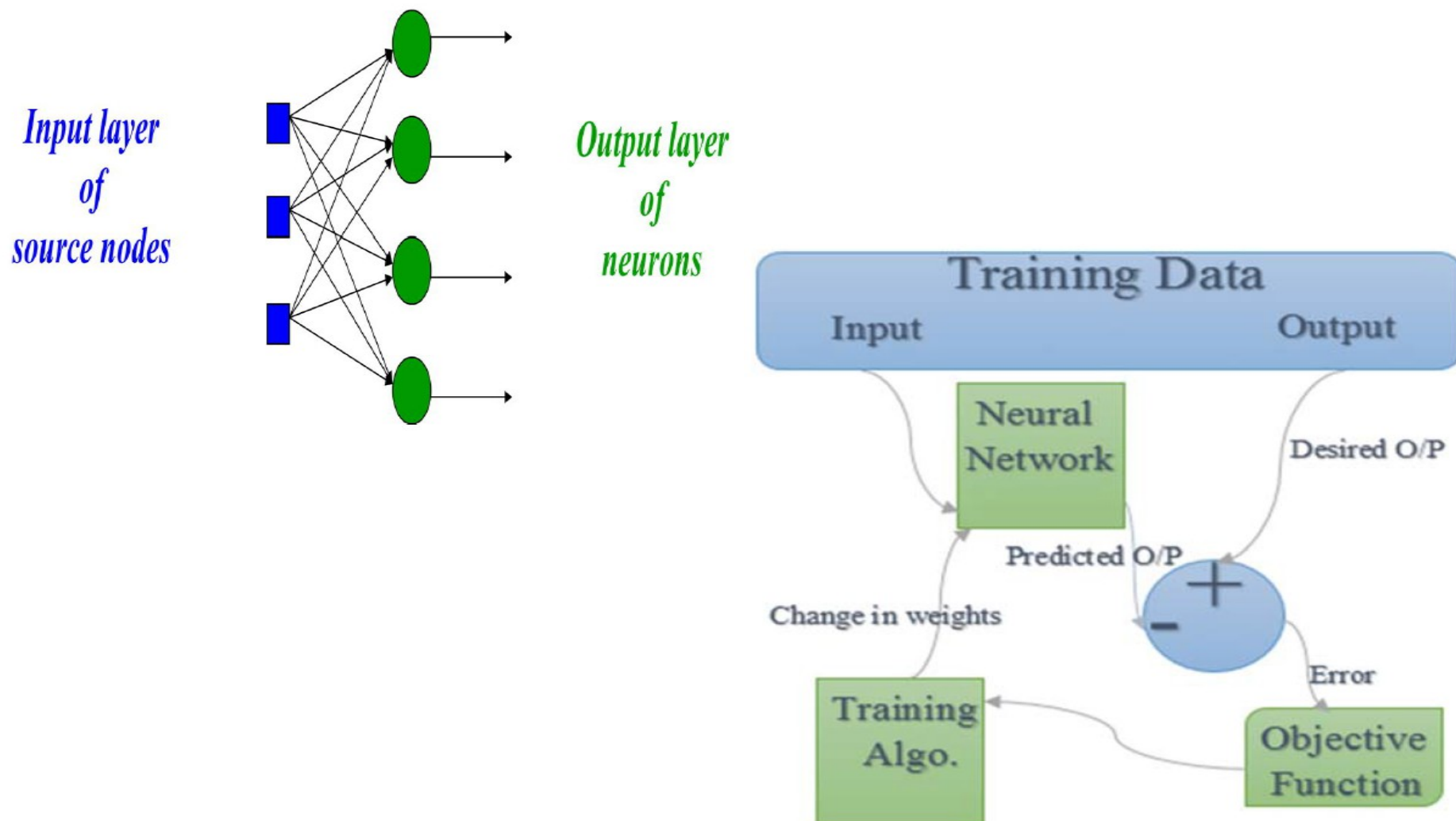
# • Machine Learning Support Vector Machine



This shows the hyperplane which classify the data from one class to another class

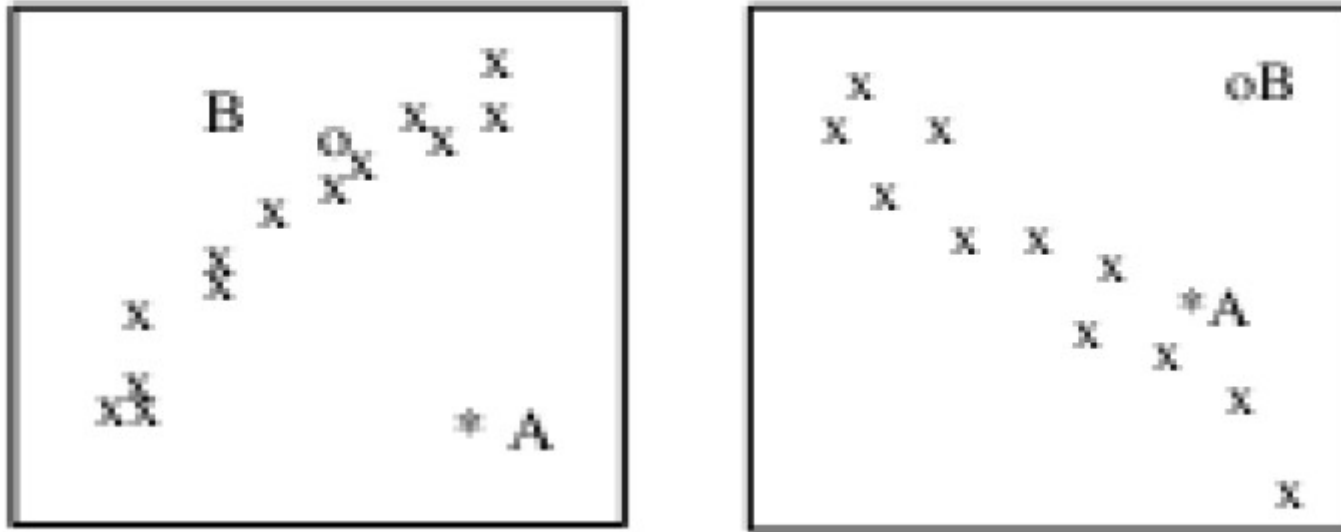
# • Machine Learning

## Neural Network



Single Layer Feed Forward Model

# • Machine Learning anti-k nearest neighbor



**Outlier Detection**

# • Machine Learning

## Comparison of Three Algorithms

	GA (Optimisation)	NN (Classification)	AIS
Components	Chromosome Strings	Artificial Neurons	Attribute Strings
Location of Components	Dynamic	Pre-Defined	Dynamic
Structure	Discrete Components	Networked Components	Discrete components / Networked Components
Knowledge Storage	Chromosome Strings	Connection Strengths	Component Concentration / Network Connections
Dynamics	Evolution	Learning	Evolution / Learning
Meta-Dynamics	Recruitment / Elimination of Components	Construction / Pruning of Connections	Recruitment / Elimination of Components
Interaction between Components	Crossover	Network Connections	Recognition / Network Connections
Interaction with Environment	Fitness Function	External Stimuli	Recognition / Objective Function

# • Solutions

## Classical rule-based approach

- Always “too late”:
  - New fraud pattern is “invented” by criminals
  - Cardholders lose money and complain
  - Banks investigate complains and try to understand the new pattern
    - A new rule is implemented a few weeks later
- Expensive to build (knowledge intensive)
- Difficult to maintain:
  - Many rules
  - The situation is dynamically changing, so frequently
  - rules have to be added, modified, or removed ...



# • Solutions

## Neural Stream

### ● Storage

- hadoop
  - HDFS: Distributed File System(DFS)
  - MapReduce : parallel processing

### ● Algorithms

- on—line learning (Immune System and Genetic Algorithms)
- batch model
- direct data

### ● Stream

- Neural stream
  - Decentralize decision process
  - Cell base detection
  - Network for Artificial Immune Systems
- Storm, Samja can ' t use on—line learning



# • Solutions

## A system based on profiles

- Every bank user gets a vector of parameters that describe his/her behavior: an “average-behavior” profile
- The system constantly compares this “long-term” profile with the recent behavior of cardholder
- Transactions that do not fit into bank user ’ s profile are flagged as suspicious (or are blocked)
- Profiles are updated with every single transaction, so the system constantly adopts to (slow and small) changes in bank user ’ behavior

# Q&A

# Thanks

