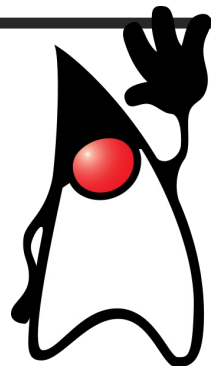


Artificial Immune System Algorithms & Implementation

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Bandung, 22 April 2022

Biodata



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Alumni:

S1 Informatika ITB
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Rekayasa Perangkat Lunak,
Algoritma Optimasi, Soft
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Programming



Pekerjaan:

Studi Teknik Informatika, 2004
Ketua Program Studi &
Reviewer Internal Universitas
Pasundan



Narasumber di:

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Daya dan Perangkat Pos dan
Informatika (Ditjen SDPPI), 2008
-



Asesor BNSP, 2018 – sekrang
(di LSP P1 Unpas)



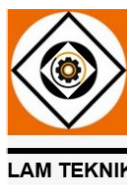
Sekretaris Umum Asosiasi
Pendidikan Tinggi Informatika
dan Komputer (Aptikom)
Propinsi Jawa Barat



Anggota Badan Otonom
Aptikom Pusat – Pendidikan
Jarak Jauh (PJJ)

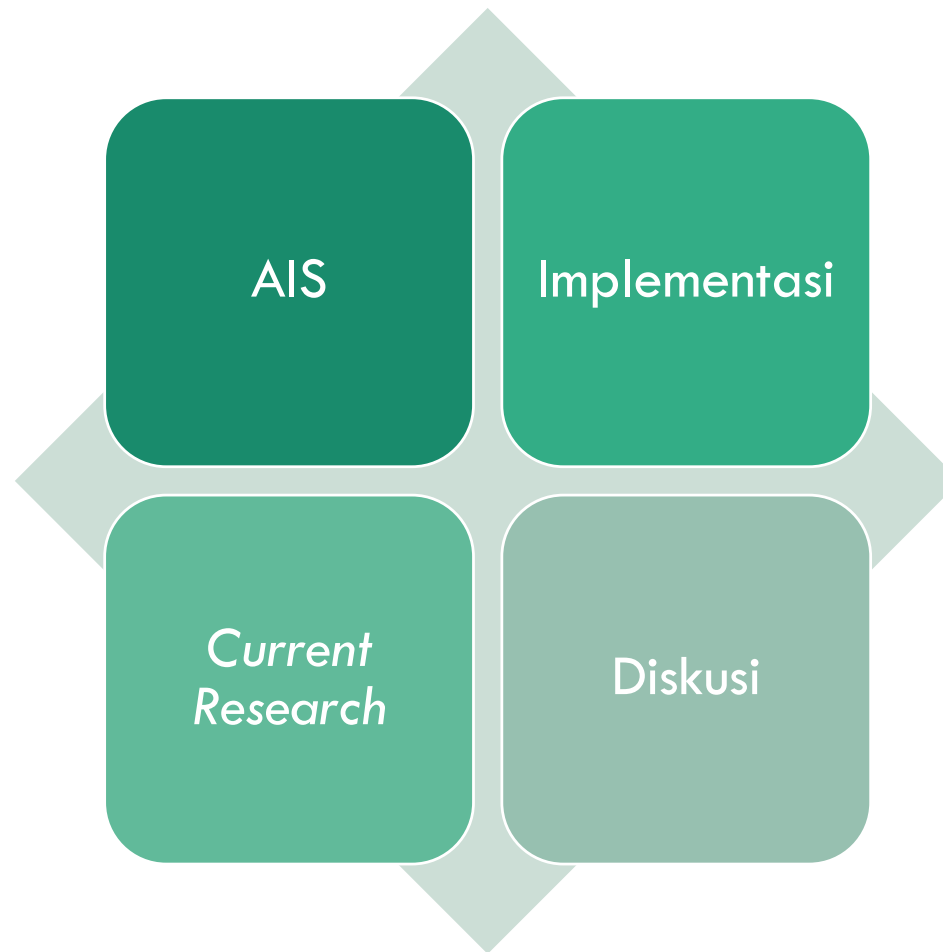


IABEE; Indonesian Accreditation
Board for Engineering Education



Lembaga Akreditasi Mandiri
(LAM) Teknik

Outline



AIS, Definition



01 Artificial Intelligence

- ✓ In artificial intelligence, artificial immune systems (AIS) are a class of computationally intelligent, rule-based machine learning systems inspired by the principles and processes of the vertebrate immune system.

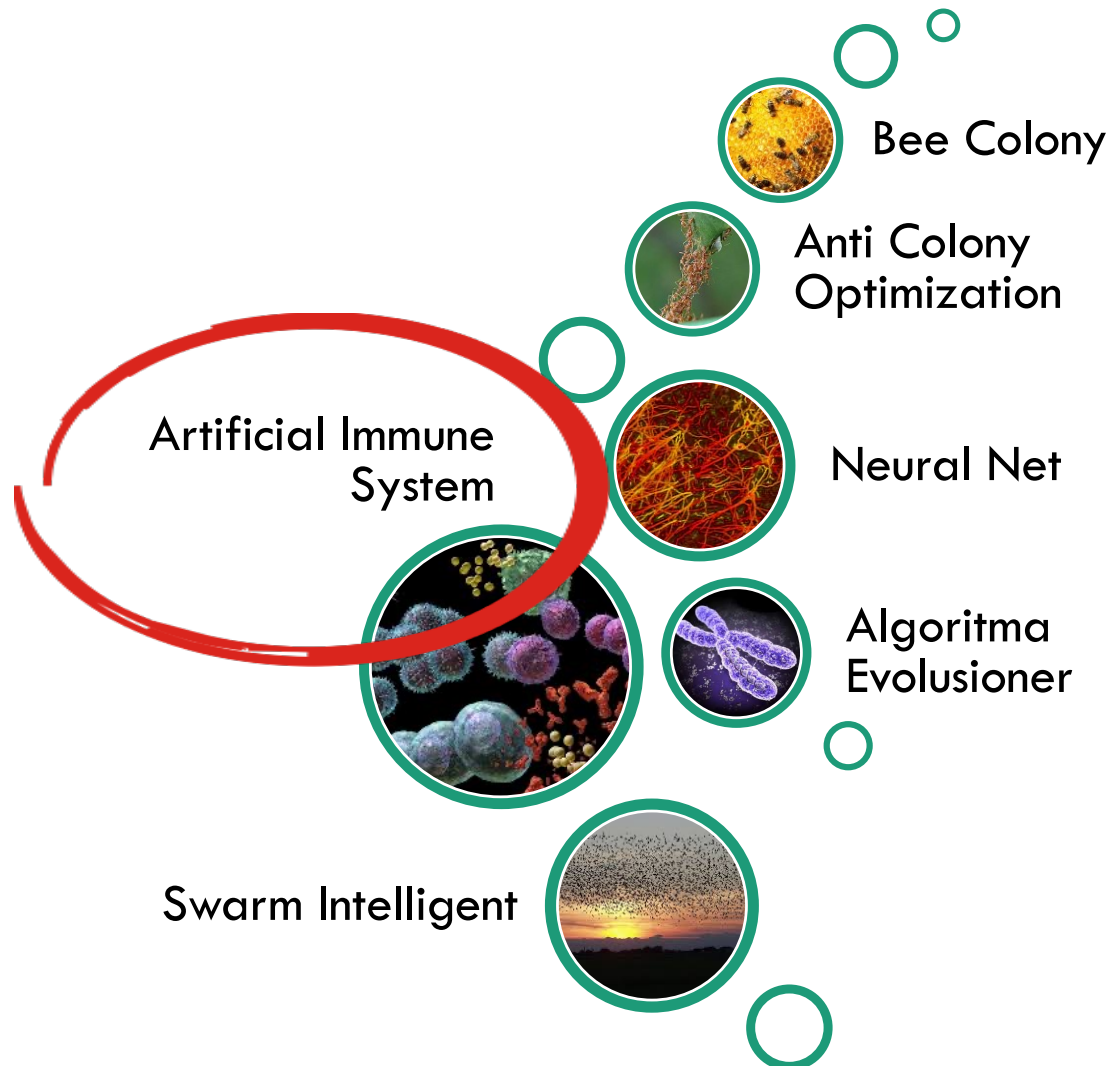
02 Sub-field

- ✓ AIS is a sub-field of **biologically-inspired computing**, and natural computation, with interests in machine learning and belonging to the broader field of artificial intelligence.

03 Applied to problem solving

- ✓ Artificial immune systems (AIS) are adaptive systems, inspired by theoretical immunology and observed immune functions, principles and models, which are applied to problem solving.

Bio-Inspired Computing

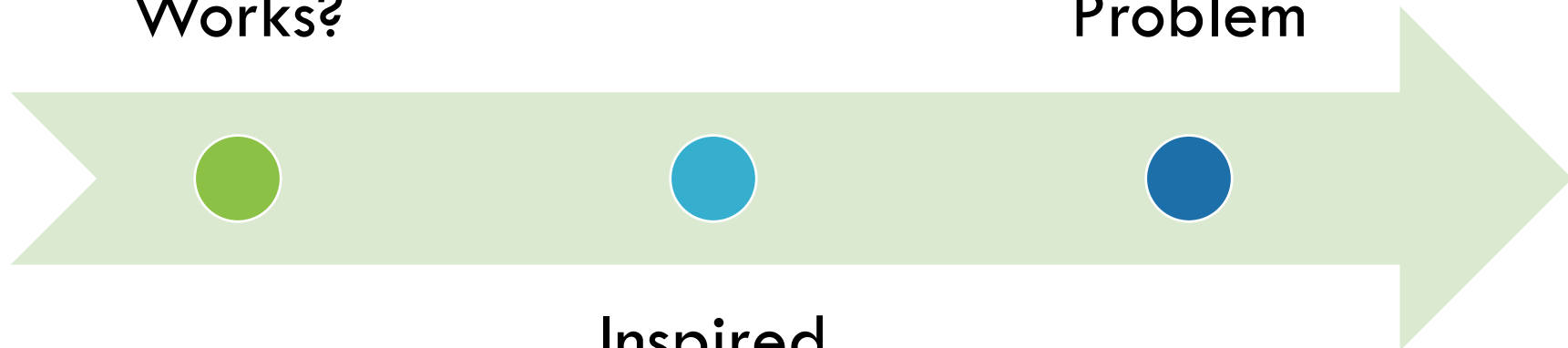


Artificial Immune System



How Immune
System
Works?

To Solve
Engineering
Problem



Inspired
Algorithm

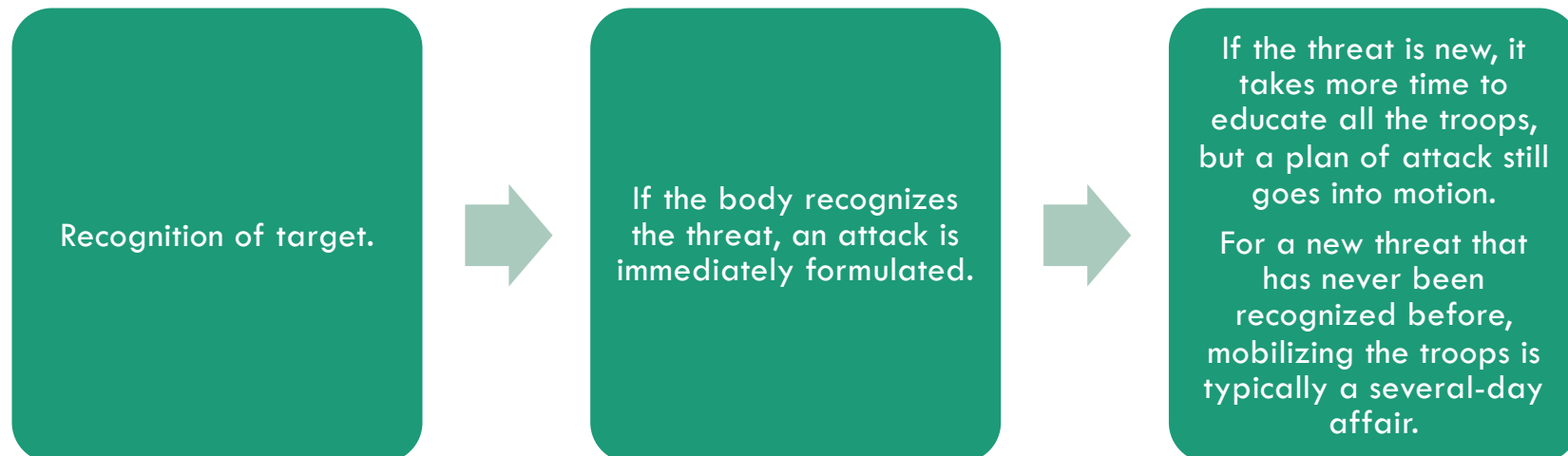
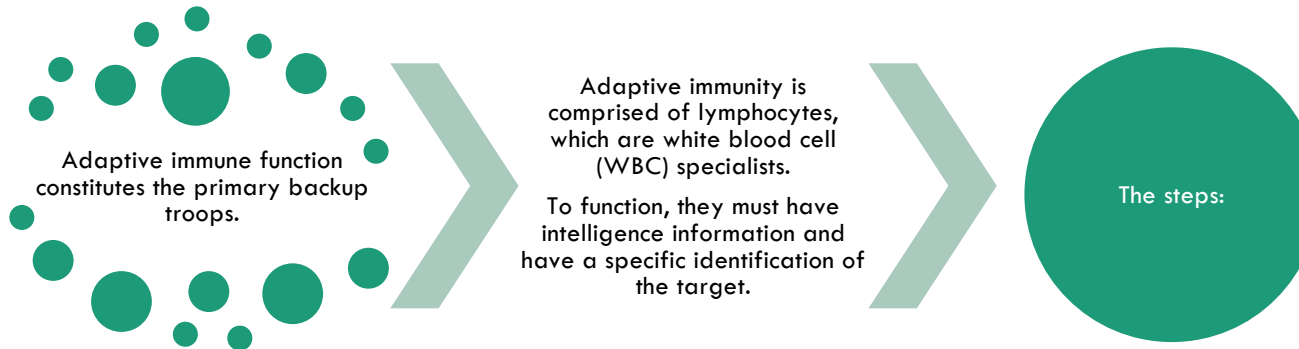
How Immune System Works

Immune function can be divided into two major components:

The innate immune system; it targets everything perceived to be a foreign threat.

The adaptive immune system; It targets specific threats that it recalls from past memory.

The Adaptive Immune System



How Immune System Works



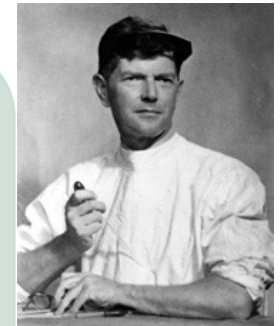
Artificial Immune System



Kampus
Merdeka
INDONESIA JAYA

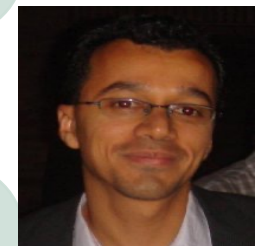
Immune System

- Immune Network Theory (Jerne)
- Clonal Selection Theory (Burnet)
- Negative Selection
- Danger Theory (Matzinger)

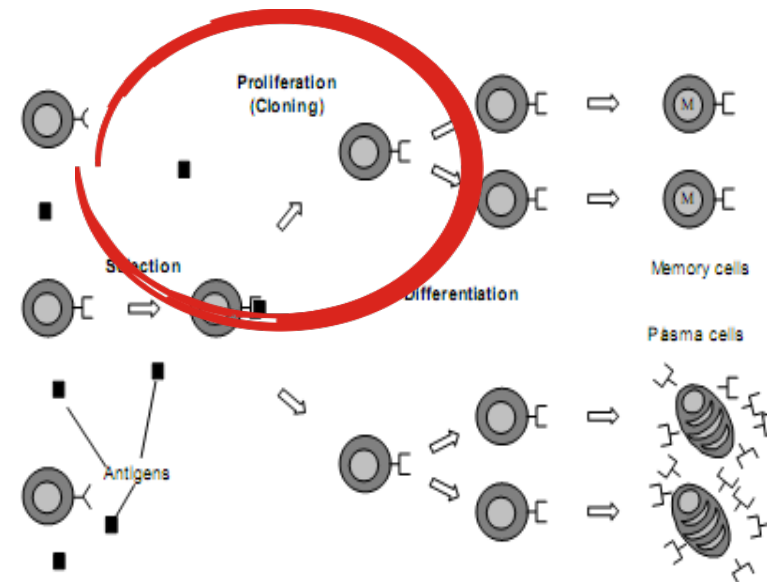
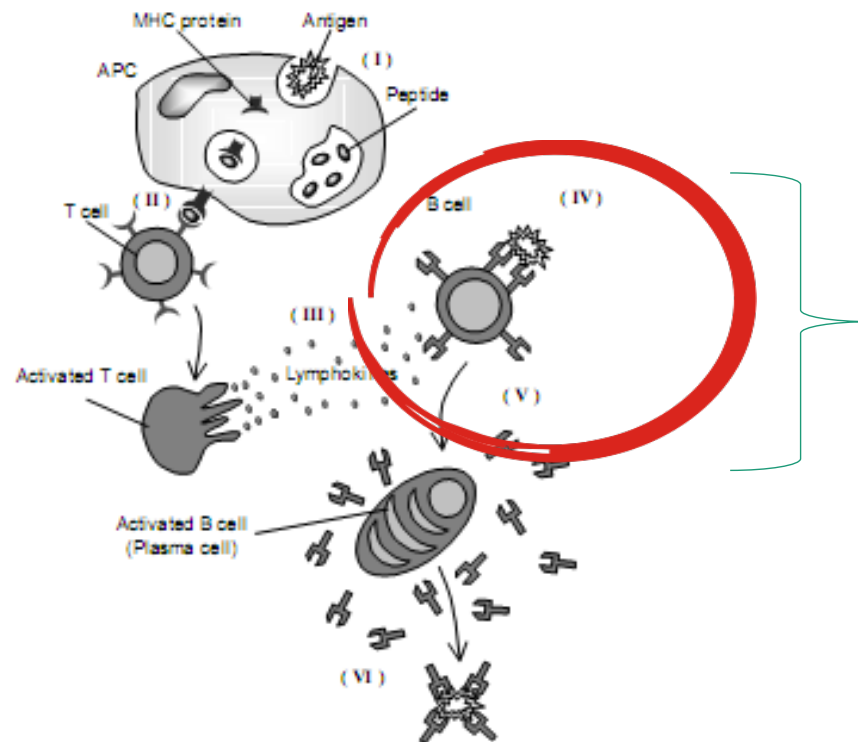


Artificial Immune System

- Immune Network Algorithm
- Clonal Selection Algorithm (de Castro & Von Zuben)
- Negative Selection Algorithm
- Dendrit Cell Algorithm



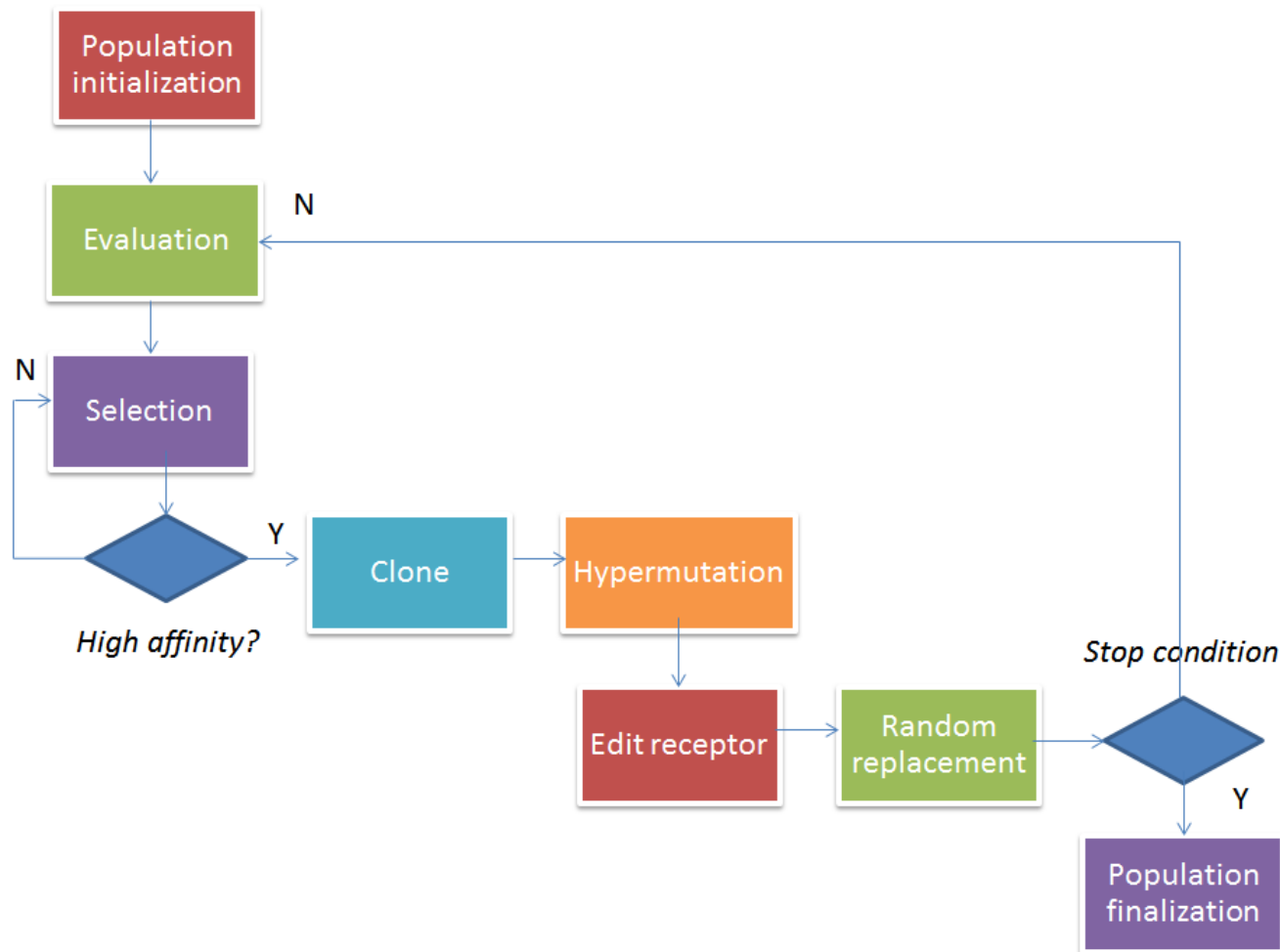
Clonal Selection: Theory (Burnet, 1978)



Clonal Selection Theory

Immune Response

Clonal Selection Algorithm (de Castro & von Zuben, 2000)



Clonal Selection Algorithm (de Castro & von Zuben, 2000)



In artificial immune systems, clonal selection algorithms are a class of algorithms inspired by the clonal selection theory of acquired immunity that explains how B and T lymphocytes improve their response to antigens over time called affinity maturation.

These algorithms focus on the **Darwinian attributes of the theory where selection is inspired by the affinity of antigen-antibody interactions**, reproduction is inspired by cell division, and variation is inspired by somatic hypermutation.

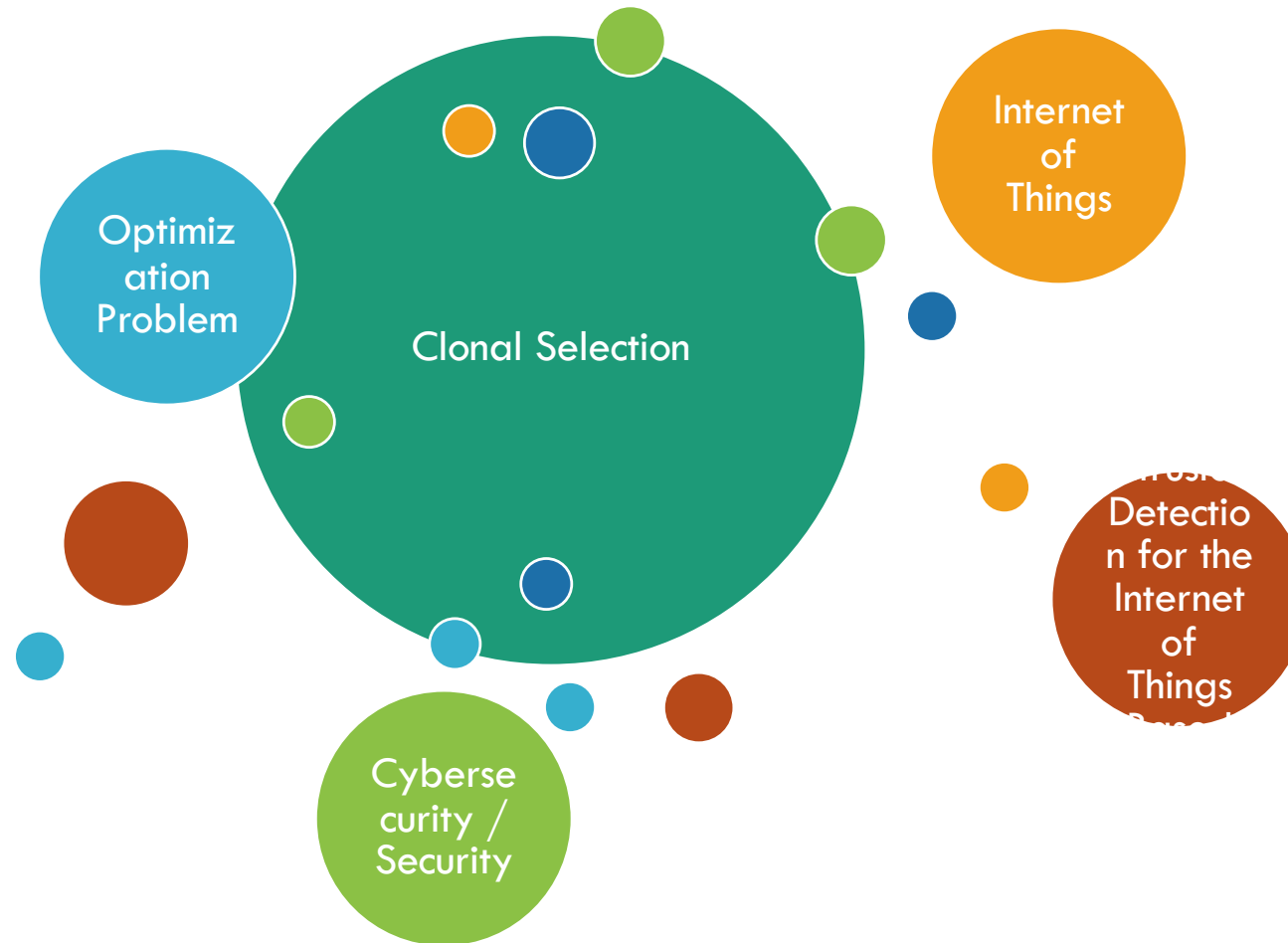
Clonal selection algorithms are most commonly applied to **optimization and pattern recognition domains**, some of which resemble parallel hill climbing and the genetic algorithm without the recombination operator.

Clonal Selection: in Engineering Problem



- De Castro & Von Zuben:
 - a binary character recognition task will be used to test its learning and memory acquisition capabilities;
 - a multi-modal optimization task; and
 - a 30 cities instance of the Travelling Salesman Problem (TSP).
- Nowadays:

Clonal Selection: in Engineering Problem



Implementasi AIS

Clonal Selection Algorithm untuk persoalan Optimasi

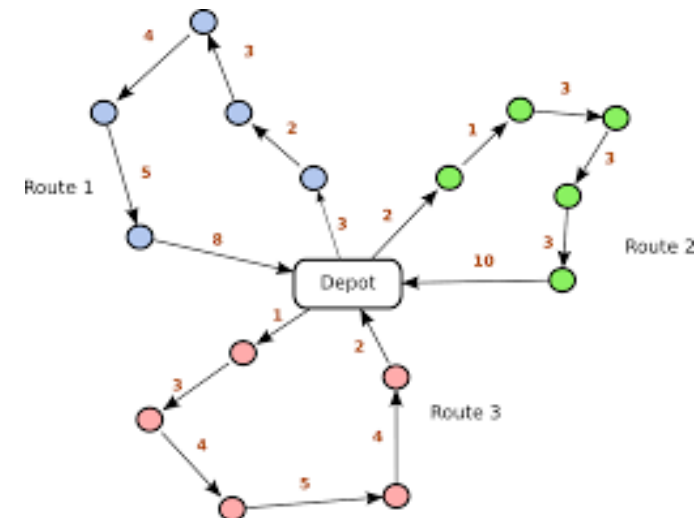
AIS for VRP



Vehicle Routing Problem

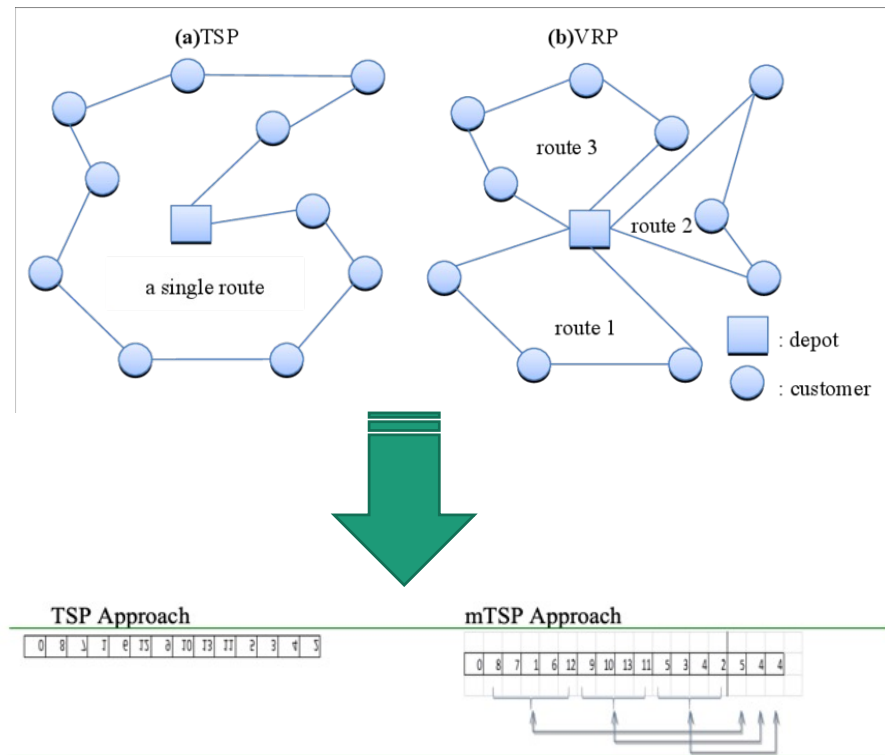


- Vehicle Routing Problem (VRP) is an optimization problem in determining routes with limited vehicle capacity.
- There is an initial depot and a number of n locations with various demands.
- A vehicle departs from the initial depot, visits all locations to fulfill requests from each location.
- The purpose of this problem is to minimize the total distance traveled by the vehicle by arranging the order of places to visit along with when the vehicle will return to fill its capacity again.



Vehicle Routing Problem

- Basically, VRP can be solved using multi-Traveling Salesperson Problem (mTSP). Where a salesperson must visit the entire city exactly once, and return to his hometown.
- At mTSP, there are m salesperson who each visit a particular city exactly once, and return to the original city.
- MTSP is a complex problem, namely a difficult problem where there is no solution with a polynomial time to solve it and requires a non-polynomial-time algorithm (NP Problem).
- In resolving NP problems, the exact approach cannot handle the problem large.
- The heuristic approach is the choice of solutions, namely a quick approach to get a feasible solution.



Solving VRP with CSA



Stage	Name	Description
1	Initialization	Initially individual population (N) randomly.
2	Evaluation	a) Select Antigen – A single antigen is selected at random without replacement (for the current generation) from the pool of antigens. b) Exposure – The system is exposed to the selected antigen. Affinity values are calculated for all antibodies against the antigen. Affinity is a measure of similarity, and is problem dependent. It is common to use Hamming distance.
3	Selection	Selection – A set of n antibodies are selected from the entire antibody pool that have the highest affinity with the antigen.
4	Cloning	Cloning – The set of selected antibodies are then cloned in proportion to their affinity (rank based).
5	Hypermutation	The clones (set of duplicate antigens) are then subjected to an affinity maturation process to better match the antigen in question. Here, the degree of maturation is inversely proportional to their parent's affinity (rank based), meaning that the greater the affinity, the lower the mutation.
6	Editing Receptor	The d individuals in the remaining r antigen pool with the lowest affinity are replaced with new random antibodies.
		Loop, Stage 1-5 until stop condition e.g. number of generation or convergent or optimal.
7	Population Finalization	Population Finalization

Mapping VRP with CSA



Clonal Selection Algorithm Stage	Vehicle Routing Problem
Population initialization	Set of randomly generated combination solutions. There are $(n-1)!$ possibilities that the solutions may be raised. This population is part of the whole combination solutions. The number of solutions is generated by the specified population size.
Affinity evaluation	Evaluation of affinity checks each combination solution that has raised, find the cost required to form the combination solution.
Selection: affinity maturation	Affinity is how close the cost of a combination solution with the optimal/best cost. The combination solutions with higher affinity will be selected.
Cloning	Cloning is process to copy selected combination solutions, number of copies are depends on clone factor:
Hypermutation	Cloned/copied combination solutions will be mutated according to hypermutation probability mutate factor:
Edit receptor/elicitation	After mutate, we will have the best solutions-that will be replaced the worst combination solutions in the initial population. The number worst combination solutions replaced will be depends on some random size replacement d .
Stop condition	Clonal selection process will be repeated until a stop condition obtained. Stopping criteria could be the number of generations, or numbers of populations are evaluated, or best cost found.

Case Study: Waste Management In Bandung City

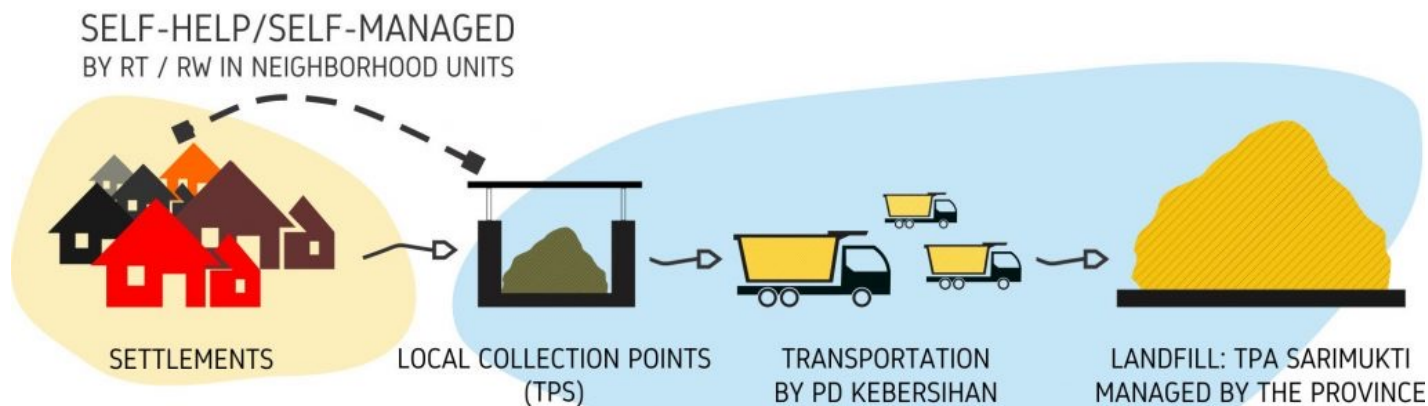


- Bandung City



Waste Management

HOUSING WASTE MANAGEMENT**



KAMPUNG AREA (46.12%)*
IS THE BIGGEST LAND USE IN BANDUNG CITY
COMPARED TO INDUSTRY (21.37%)
FORMAL HOUSING COMPLEX (10.61%)
SERVICES (7.42%)
OTHERS (UP TO 3.45%)

*Bappeda (2010) in Kampung-Kota Bandung (Widjaja, 2013)

** PD Kebersihan Kota Bandung 2016 Profile Slides



Experiment



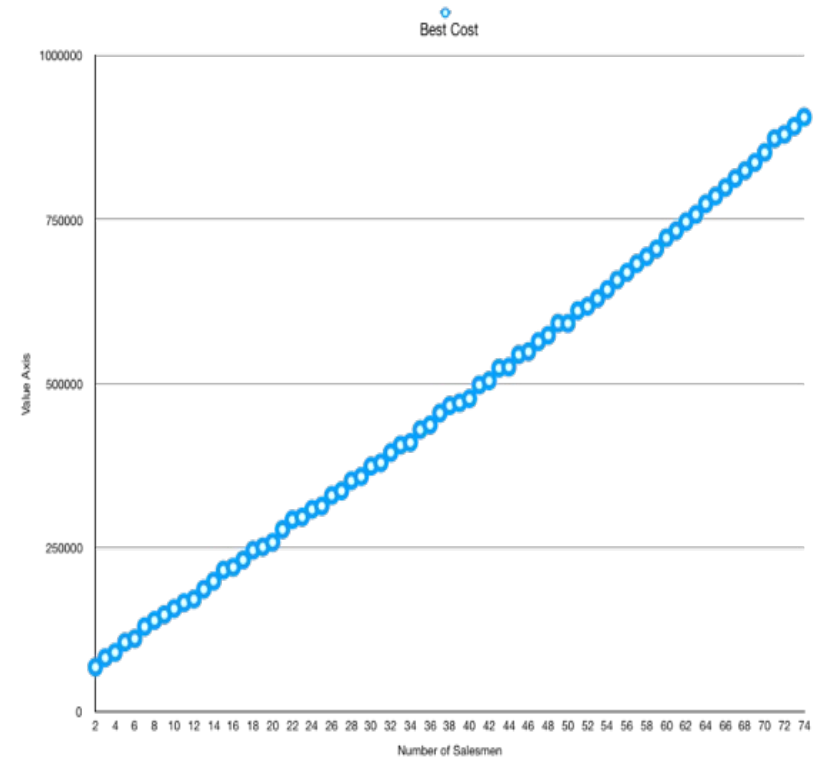
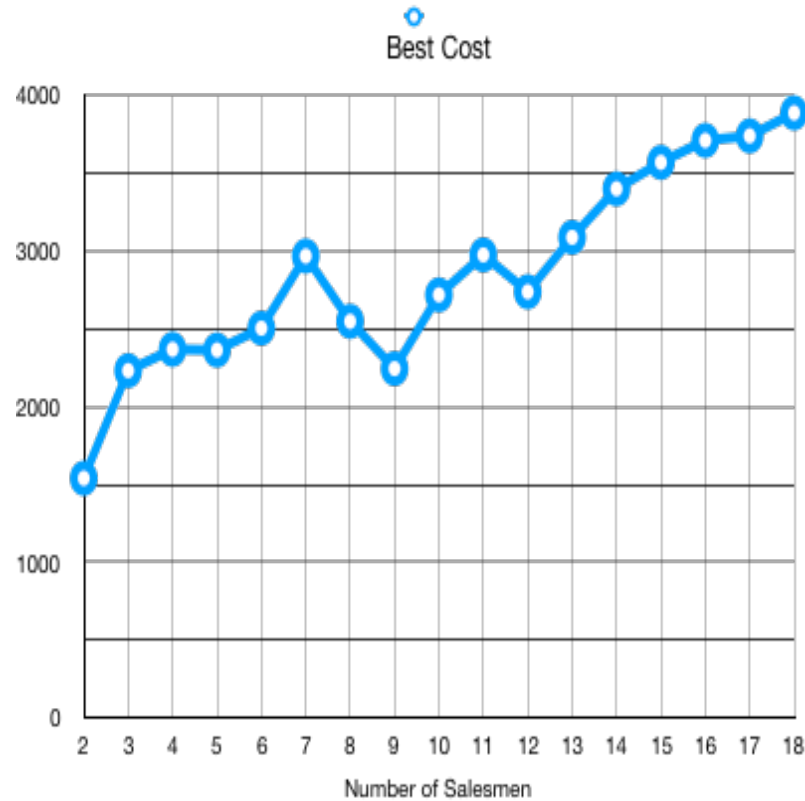
Dataset	Waste transportation In Northern Bandung, West Java, Indonesia: Bandung36.vrp and Bandung148.vrp
VRP parameter:	Number of vertices: 36 or 148
	Number of salesmen: between 2- 17 and between 2 - 74
CSA parameters	Number of population (N) = 50; Selection size (n) = 10; The value for the cloning factor (β) = 0.1; Value for mutation factor (rho) 2.5; Size for random replacement (d) = 5
Determination of criteria stops	Number of generations (g) = 1000, 10,000, 100,000

Experiment Result



Number of salesmen	Sub-tour Cost																Total Cost		
2	1106 434																1540		
3	66	1102	1064														2232		
4	885	108	113	1263													2369		
5	994	254	222	643	251												2364		
6	250	377	42	1260	168	408											2505		
7	693	42	106	930	222	40	936										2969		
8	72	40	780	206	42	1031	156	222								2549			
9	860	66	108	222	168	42	72	40	667							2245			
10	42	1043	222	293	108	206	156	543	40	66						2719			
11	959	72	40	42	206	147	222	156	66	108	958						2976		
12	1007	206	156	42	168	66	222	106	40	146	108	472					2739		
13	974	42	423	108	72	40	198	206	156	66	106	476	222				3089		
14	206	66	474	156	198	106	685	146	40	222	42	72	108	880				3401	
15	601	42	250	803	168	222	106	206	40	198	146	72	156	108	453				3571
16	494	108	168	606	40	66	156	106	222	250	42	146	250	206	72	780			3712
17	222	206	278	40	338	1057	66	106	250	108	198	146	288	168	42	72	156	3741	
18	892	206	42	146	108	198	338	66	72	156	250	40	106	290	311	278	222	168	3889

Experiment Result (2)



Conlusion AIS for VRP



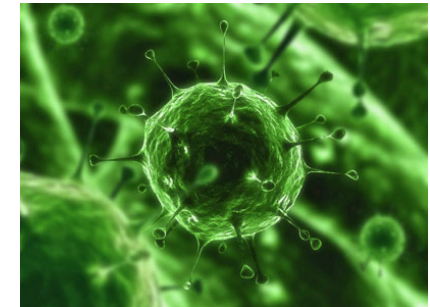
- A clonal selection based algorithm was proposed for solving the VRP with mTSP approach.
- This algorithm is can show solution for mTSP for Bandung waste transportation dataset.
- Two dataset in experiments show that the best cost (minimal cost) result from 2 salesmen / driver in VRP.
- This is because, the cost represent distance between nodes.
- Actually we need other representation of cost, like capacity of the truck and the time limit as VRP constraint.
- This is will be future research.
- VRP can be solved by mTSP approach using bio-inspired computation like Clonal Selection Algorithm.
- This algorithm has been shown the good performance, but has several limitations since the need to define constraint for the VRP constraints.

Current Research

Immune-inspired Optimization Model untuk Winner Determination Problem pada Spectrum Frequency Combinatorial Auction



Immune Inspired Computation



*Spectrum Frequency Combinatorial
Auction*



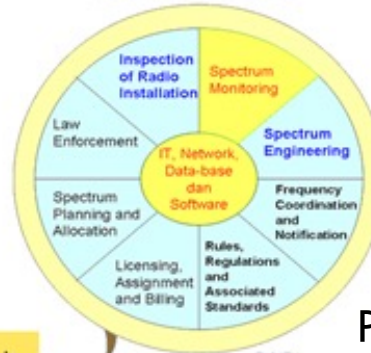
International
Telecommunication
Union



KOMINFO
Administrator Indonesia
Cq Ditjen SDPPI



Manajemen Spektrum
Frekuensi Radio



Ref: ITU

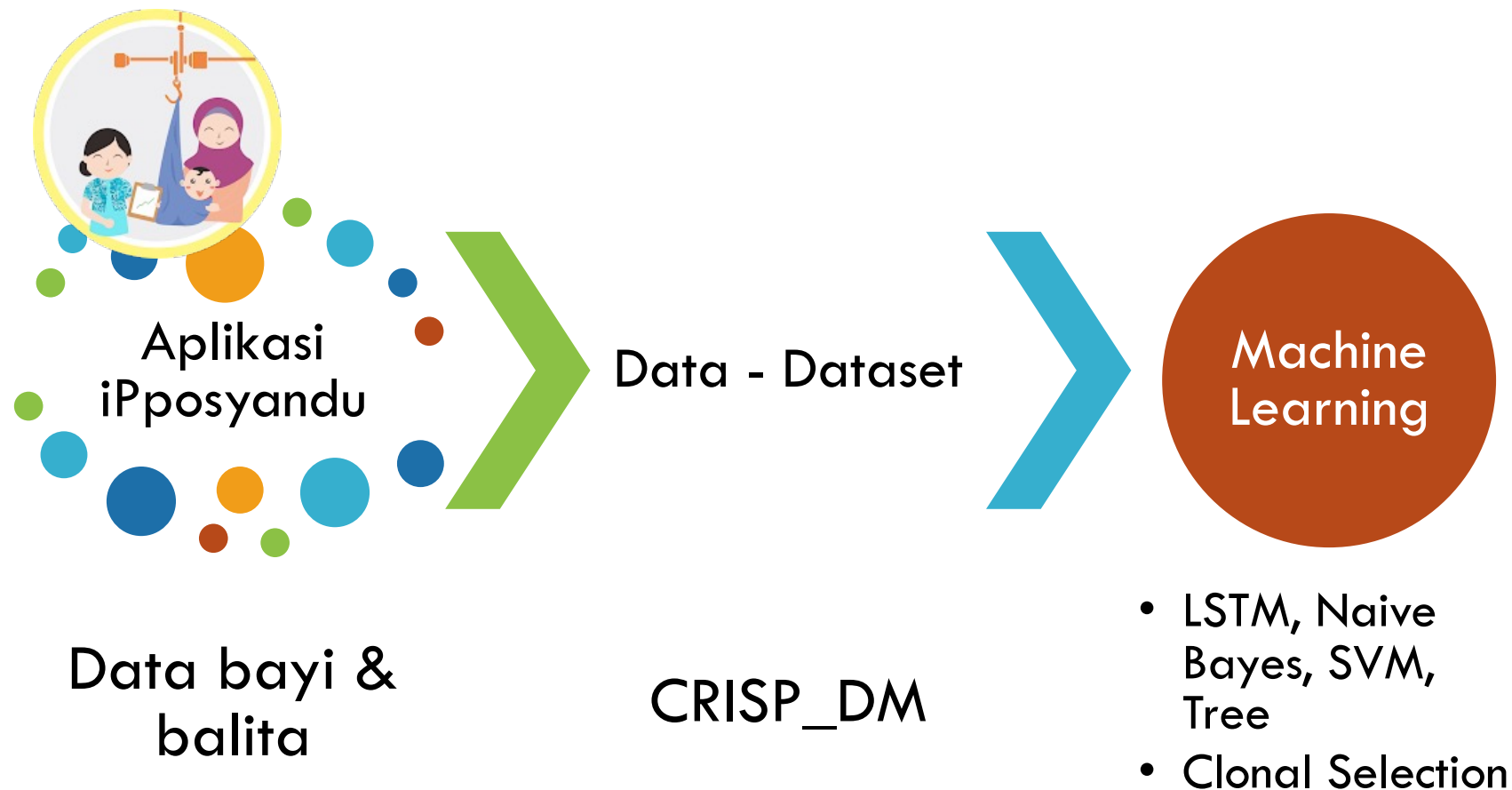
Perizinan Frekuensi Radio merupakan salah satu aspek penting dalam Manajemen Spektrum Frekuensi Radio

02/07/22

Pengelolaan Spektrum Frekuensi Radio
[<http://postel.go.id>]

'urbasari

Machine Learning for Stunting Prediction



<https://play.google.com/store/apps/details?id=id.iposyandu.kader&hl=en&gl=US>

Clonal Selection: in Engineering Problem

Clonal Selection: in Engineering Problem



- Optimization Problem. Eg:
 - The use of Clonal Selection Algorithm for the Vehicle Routing Problem with Time Windows
 - A Clone Selection Algorithm for the Open Vehicle Routing Problem - IEEE Conference Publication
 - An improved immune clonal selection algorithm and its applications for VRP - IEEE Conference Publication
 - Clonal selection algorithm for vehicle routing - IEEE Conference Publication

Clonal Selection: in Engineering Problem



- Optimization Problem:
 - Application of Clonal Selection Algorithm in Construction Site Utilization Planning Optimization (Xi Wang, 2016)
 - CSUP is the decision making process for determining the locations of temporary facilities within the boundary of a construction site by identifying spatial, functional relationships between the temporary facilities. Its objective is to identify an optimal layout from a large number of alternative solutions so that a set of predetermined facilities are appropriately located while satisfying site specific constraints.

Clonal Selection: in Engineering Problem



- IoT. Eg:

- Research on Intrusion Detection for the Internet of Things Based on Clone Selection Principle (Liu, 2012)
- An Immunity-Based IOT Environment Security Situation Awareness Model (Yunquan, 2017)

Research on Intrusion Detection for the Internet of Things Based on Clone Selection Principle



- The fast development of the Internet of Things (IoT) makes its security problems appear gradually.
- It is urgent to study the intrusion detection technology for IoT security threats, based on the clone selection principle.
- The key elements in the clone selection theory are simulated.
- The clone selection algorithm is realized to be applied in IoT.
- Detection elements for security threats evolve to adapt the real IoT environment.
- The proposed method is expected to improve the detection efficiency of IoT security threats.

Clonal Selection Algorithm and Industry



- Industry 4.0 - Artificial intelligence
- Computational intelligence
 - Soft Computing
 - Subset of AI, Bezdek (1994)
- CI:
 - Fuzzy logic
 - Neural networks
 - Evolutionary computation
 - Artificial Immune System
 - Clonal Selection Algorithms

Clonal Selection: in Engineering Problem



- Cybersecurity

- A botnet detection approach based on the clonal selection algorithm (Lysenco, 2018)
 - A new technique for the botnets' detection in the corporate area networks. It is based on the usage of the algorithms of the artificial immune systems. Proposed approach is able to distinguish benign network traffic from malicious one using the clonal selection algorithm taking into account the features of the botnet's presence in the network. An approach present the main improvements of the BotGRABBER system. It is able to detect the IRC, HTTP, DNS and P2P botnets.

Clonal Selection: in Engineering Problem



- Security

- The Methods of Artificial Intelligence for Malicious Applications Detection in Android OS (Bezobrazov, 2016)
- Detection of Mobile Malware: An Artificial Immunity Approach (Brown, 2016)
- Adaptive artificial immune networks for mitigating DoS flooding attacks (Vidal et al, 2017)
- Applications of artificial immune systems to computer security: A survey (Fernandes, 2017)

Conclusion



- Optimization Algorithms are used in engineering fields
- Clonal Selection Algorithm as an algorithm for optimization problem is used in engineering fields
- Beside optimization problem, CSA is used for Security, esp. in IoT Security
- As part of Computational Intelligence, CSA has potential contribution in industry 4.0 technology

Thank you