

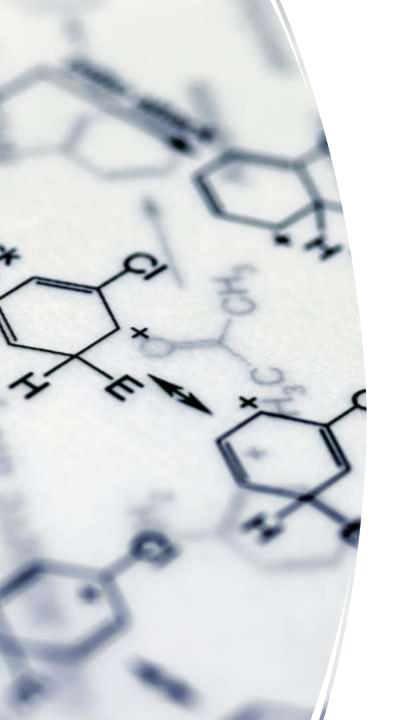
Artificial Immune Systems (AIS)

A Bio-Inspired Computational Approach

Course: Nature inspired Soft

Computing

Course Code: 23CS3202



Introduction to AIS

- Artificial Immune Systems (AIS) are computational models inspired by the human immune system.
- Mimic biological immune responses for problem-solving.
- Used in optimization, anomaly detection, and pattern recognition.
- Derived from immunological principles such as self/nonself discrimination, clonal selection, and immune memory.

Biological Inspiration

The human immune system protects against harmful pathogens.

Key immune mechanisms include:

Antibody recognition and response.

Clonal selection (adaptive immunity).

Negative selection (self-tolerance).

Immune memory (fast response to known threats).



Components of AIS



1. ANTIGENS (PROBLEM INSTANCES): INPUT DATA OR OPTIMIZATION PROBLEMS.



4. IMMUNE
MEMORY: BEST
SOLUTIONS ARE
STORED FOR
FUTURE USE.



2. ANTIBODIES (SOLUTIONS):

CANDIDATE SOLUTIONS EVOLVED OVER ITERATIONS.



5. **NEGATIVE SELECTION:** SELF-

REACTIVE SOLUTIONS ARE ELIMINATED.



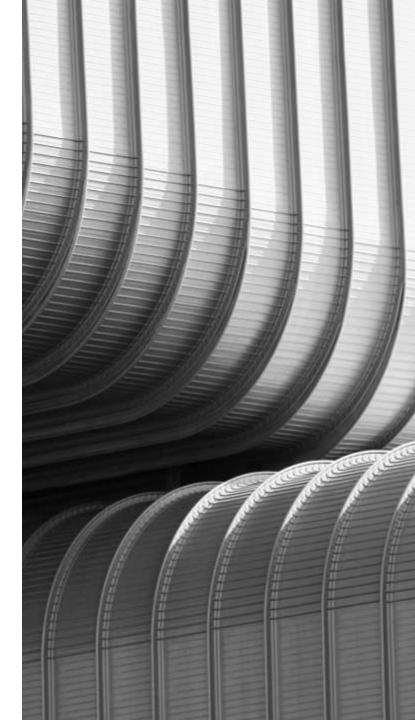
3. CLONAL SELECTION: HIGH-

AFFINITY
ANTIBODIES ARE
CLONED AND
IMPROVED.



6. AFFINITY MATURATION:

SOLUTIONS
IMPROVE THROUGH
MUTATION AND
LEARNING.



Types of AIS Algorithms



Clonal Selection Algorithm (CSA):

Mimics immune response by generating clones of high-affinity antibodies.



Negative Selection Algorithm (NSA): Used in anomaly detection by distinguishing self from non-self.



Immune Network Algorithm (INA):

Models the interaction of antibodies in a dynamic network.



Dendritic Cell Algorithm (DCA):

Inspired by dendritic cells, used for security applications.

AIS Algorithm Workflow

1. Initialize a random population of antibodies (solutions).

4. Affinity
Maturation:
Clones undergo
small mutations
to improve.

2. Evaluate

Affinity:

Determine how well antibodies recognize antigens (fitness

evaluation)

5. Selection:

Best antibodies are retained, others are replaced.

3. Clonal Expansion: High-affinity antibodies are cloned.

6. Memory
Update: Best
solutions are
stored for
future use.

7. Repeat until convergence is achieved.

Example: AIS for Anomaly Detection



CONSIDER NETWORK SECURITY MONITORING.



DEFINE 'NORMAL'
TRAFFIC AS 'SELF'
AND ABNORMAL
TRAFFIC AS 'NONSELF'.



TRAIN THE SYSTEM USING NORMAL DATA (NEGATIVE SELECTION).



DETECT ANOMALIES
BY IDENTIFYING
MISMATCHES IN
INCOMING DATA.



AIS-BASED
INTRUSION
DETECTION SYSTEMS
(IDS) CAN DETECT
NOVEL ATTACKS.

Example: AIS for Optimization



Consider function optimization:



Objective: Minimize $f(x) = x^2 - 4x + 4$.



1. Generate Initial Antibodies (solutions randomly).



2. Evaluate Affinity (compute function values).



3. Clonal Selection & Mutation (mutate better solutions).



4. Retain Best Solutions in memory.



5. Repeat until convergence to an optimal solution.

Advantages and Limitations

Advantages

- Self-learning capability.
- Adaptability to dynamic environments.
- Robustness to noise and uncertainty.
- Can handle high-dimensional problems.

Limitations:

- Computational complexity in large-scale problems.
- Requires fine-tuning of mutation and selection parameters.
- Slower convergence in some optimization tasks.

Applications of AIS



Cybersecurity:
Intrusion
detection,
malware
detection.



Optimization: Engineering design, scheduling problems.



Medical
Diagnosis:
Disease
classification,
bioinformatics.



Pattern
Recognition:
Image
processing,
fraud detection.



Autonomous Systems: Robotics, swarm intelligence.

Conclusion

AIS is a powerful bioinspired technique for solving complex problems.

Mimics immune responses for learning and adaptation.

Successfully applied in anomaly detection, optimization, and pattern recognition.

Future improvements include hybridization with other AI techniques.