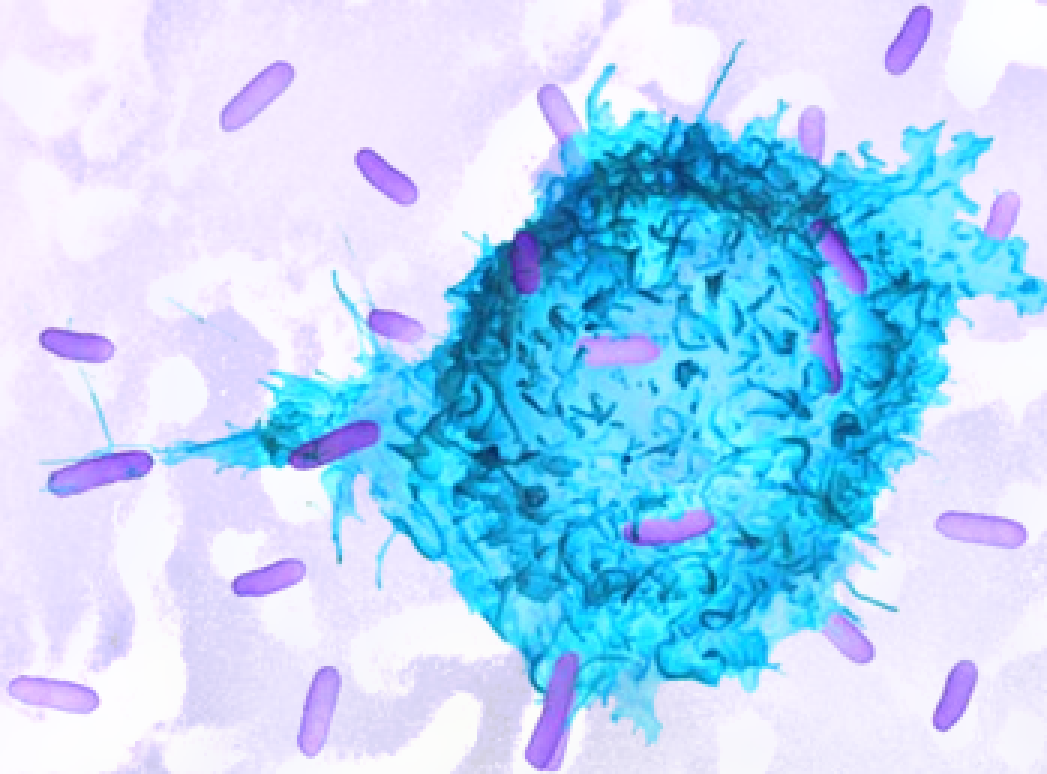
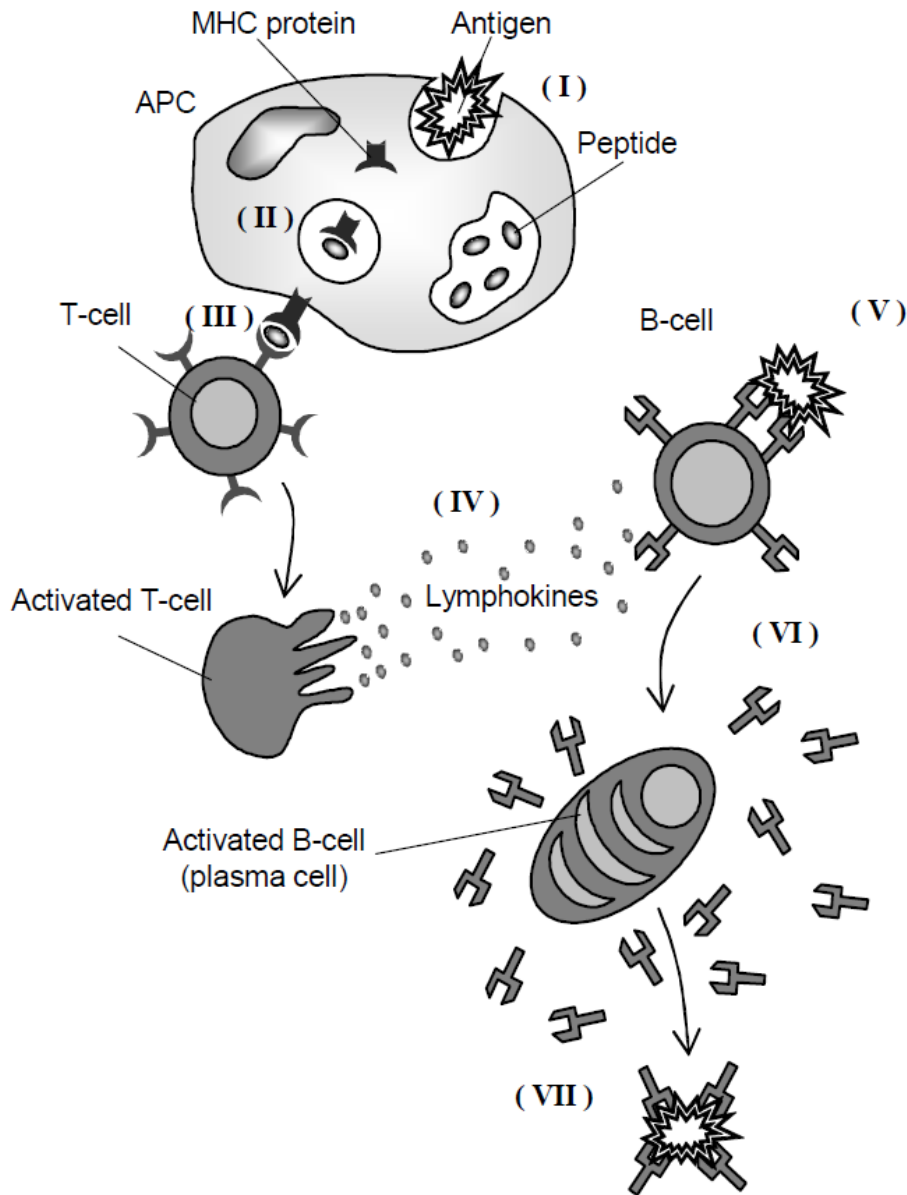


VALIS, an immune-inspired classification algorithm



Peter Karpov
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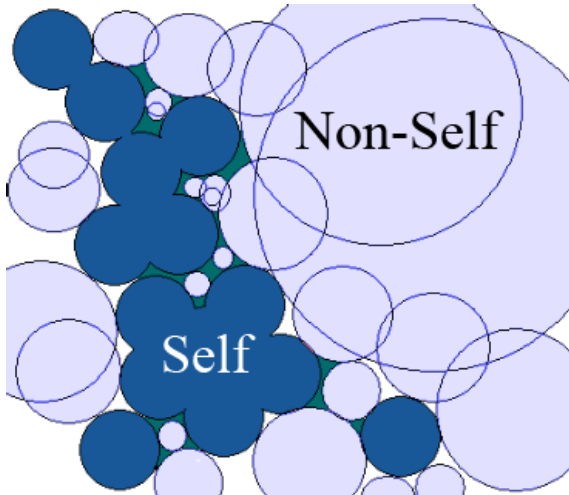
Natural immune system



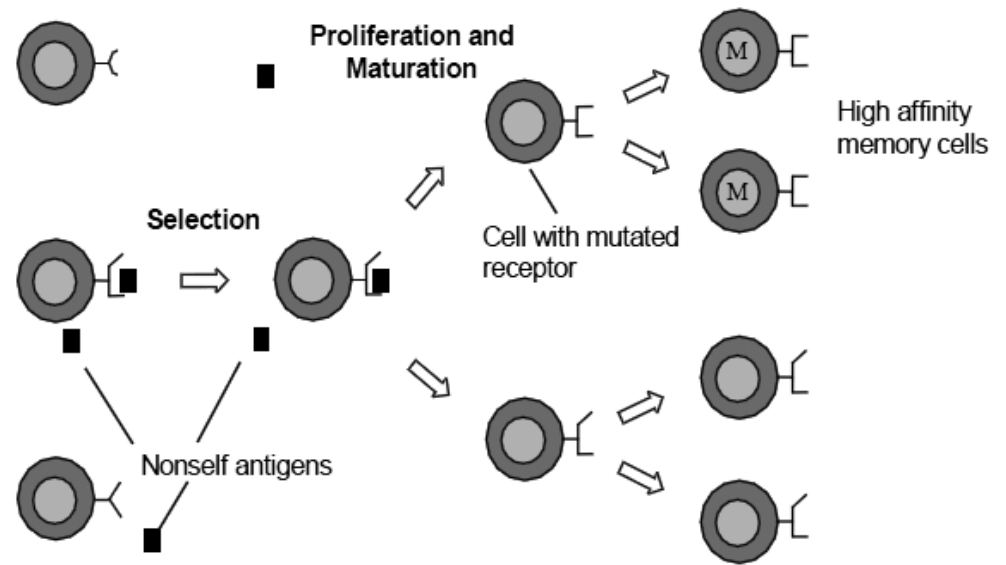
- Antigen presentation
- T-cell activation
- B-cell activation
- Antigen matching
- Antibody production
- Pathogen destruction

Immune-inspired algorithms

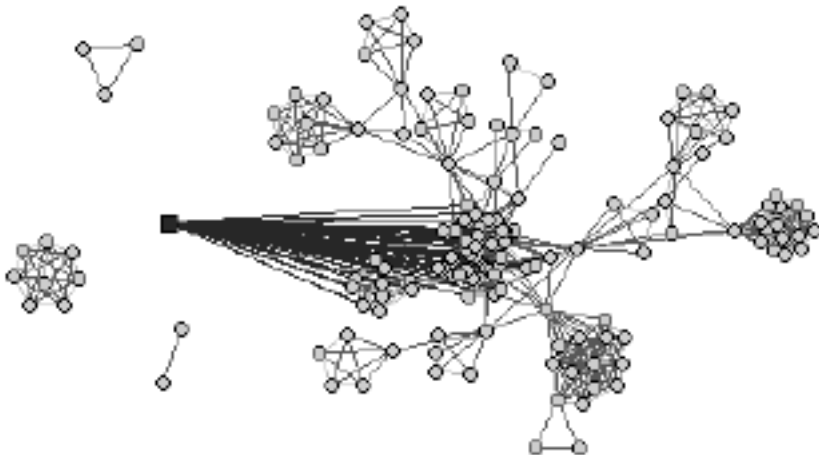
Negative selection



Clonal selection



Immune networks

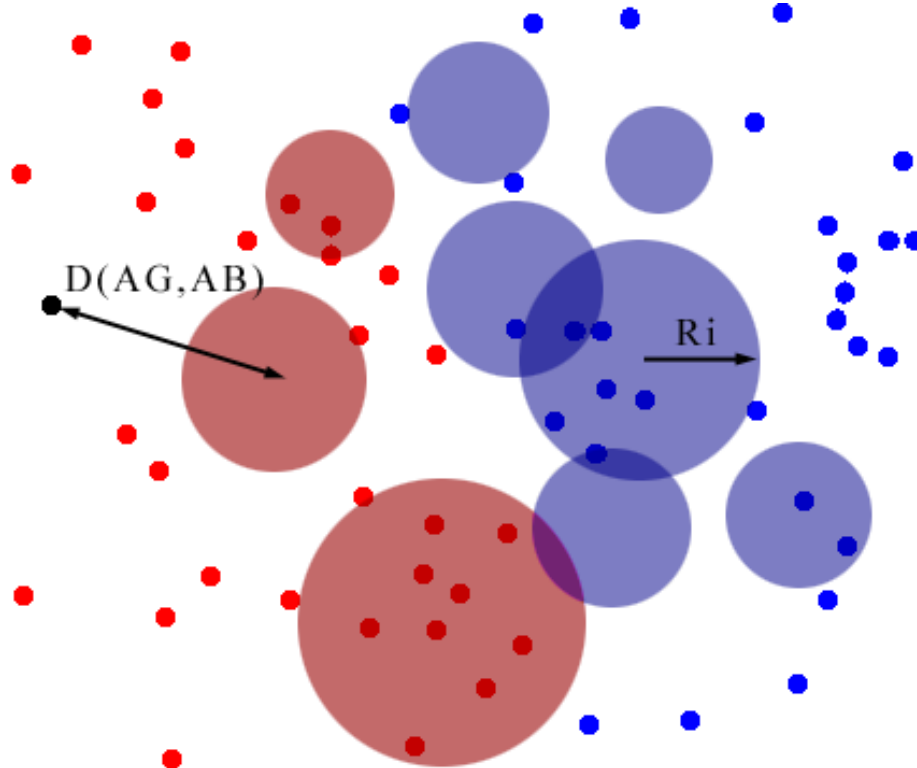


Dendritic cells,
danger theory,

... ..

Immune system — classification analogy

Antigen	←---→	Data sample AG
Antibody	←---→	Data structure AB
Immune response	←---→	Classification result



Binding:

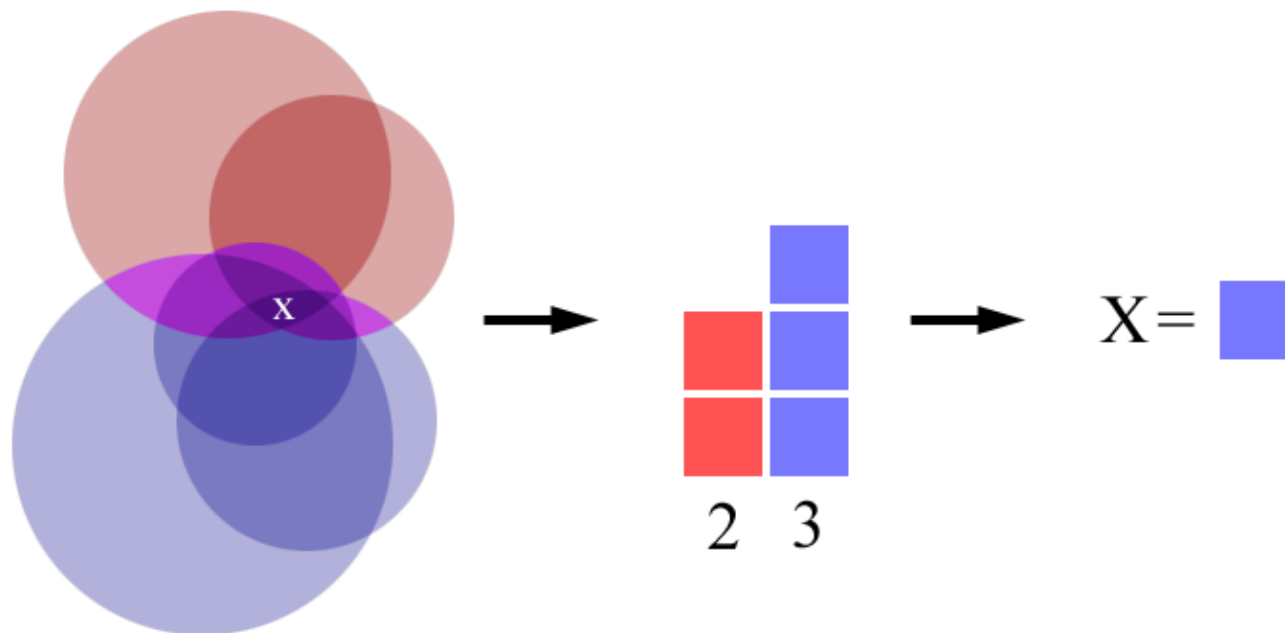
$$D(AB_i, AG) < R_i,$$

D — distance function,

R_i — binding threshold

Classification by voting

VALIS — Vote Allocating Immune System



Bound antibodies vote for their class. Majority of votes determines the classification result.

Training algorithm

1. Initialization

Create $PopSize$ random antibodies

2. Dataset presentation

Present dataset to the system, for each antibody calculate:

- Number of bound and correctly classified antigens
- Probability of correct classification $PCorrect$
- Overlapping coefficient $KShare$
- Fitness $f = PCorrect / KShare$

3. Class assignment

Set antibody classes according to the majority of bound antigens

4. Reproduction

Create $NChildren = LearnRate \times PopSize$ children by crossover and mutation

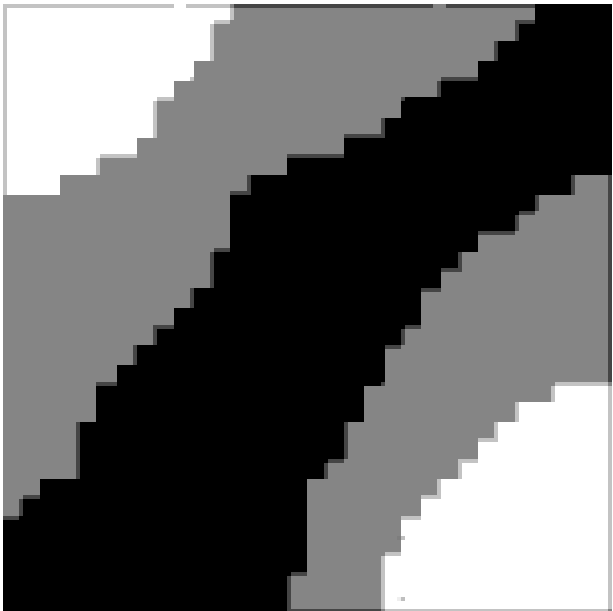
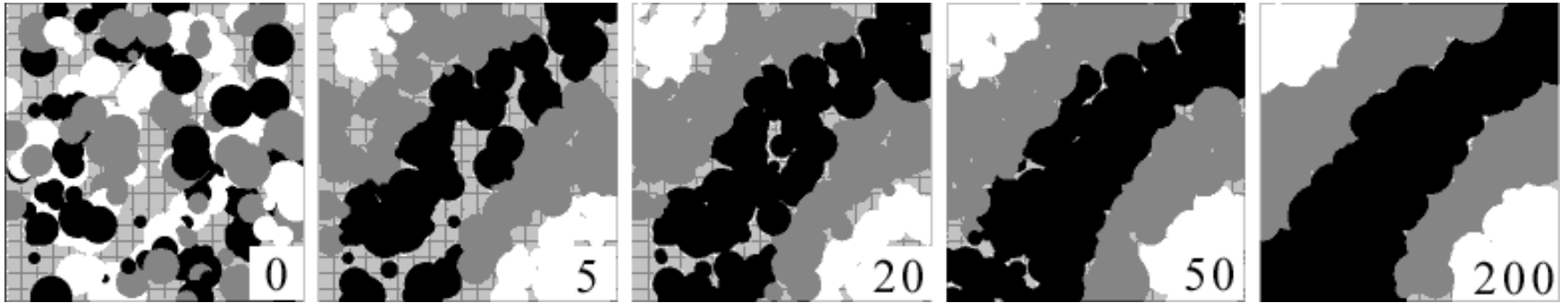
5. Replacement

Replace $NChildren$ lowest fit antibodies with children

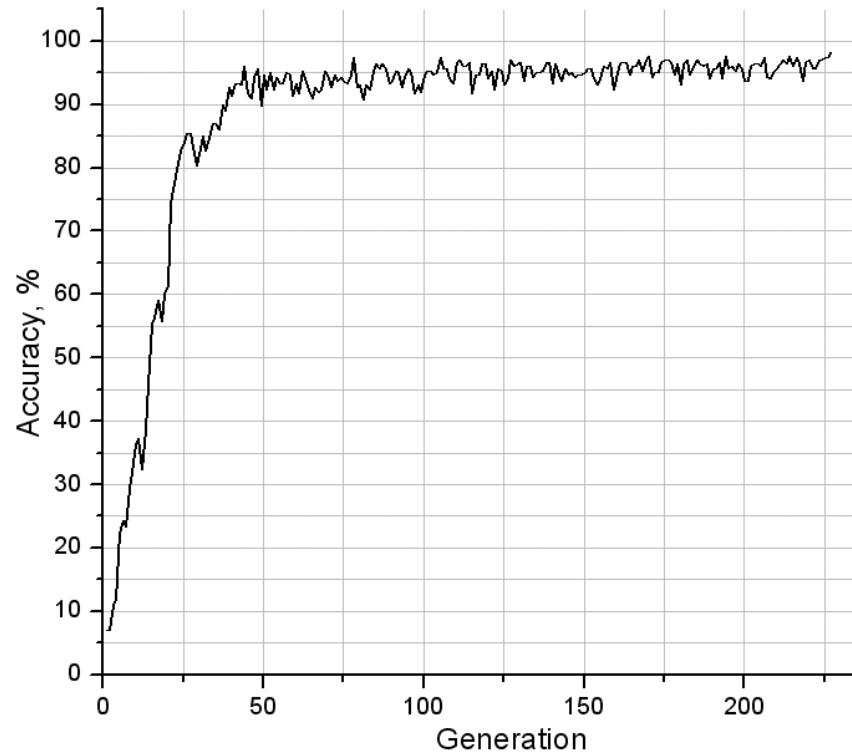
6. End of one generation

Go to step 2 until a termination criterion is satisfied

Test problem



True class distribution



Training dynamics

Character recognition

Antigens

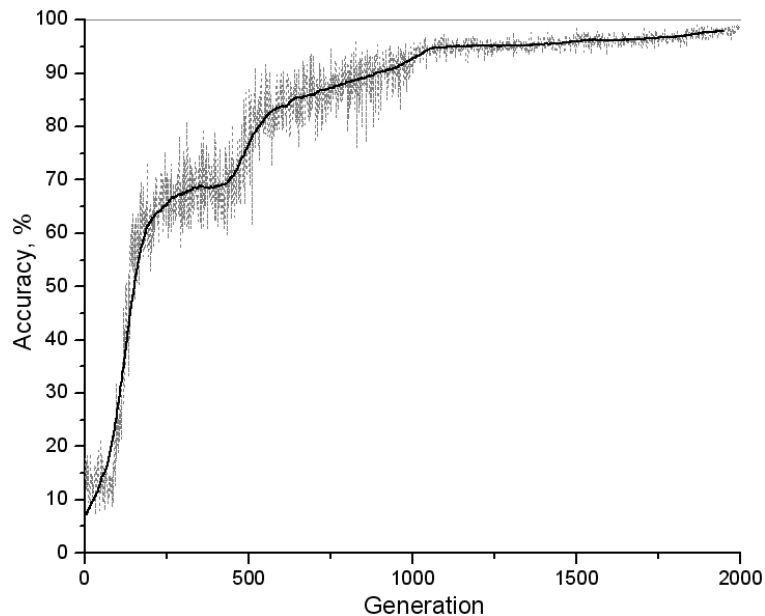
Numerals from 0 to 9, randomly scaled and rotated

Antibodies

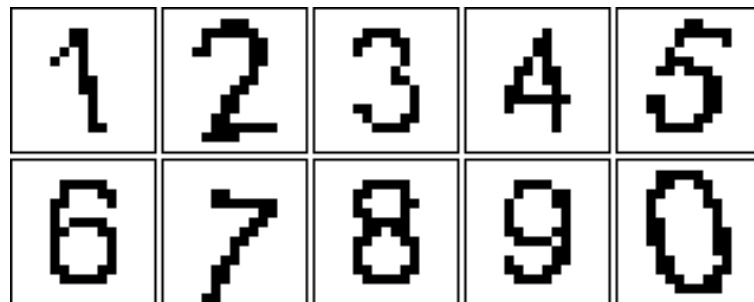
Binary arrays

Classification accuracy

99%



Generalization ability: recognition invariance to scale and rotation



Text classification

Antigens

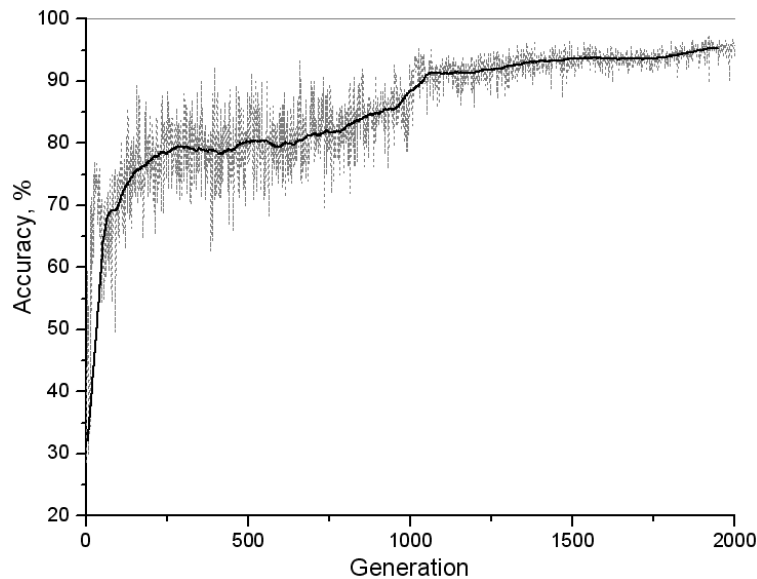
Pieces of source code in one of 6 programming languages

Antibodies

Variable length strings

Classification accuracy

96%



Typical antibody elements:

Assembler @, mov, lf

Basic \$

C fp, {#13, }, <, \n

Fortran #10c

Lisp ;;,),))#13

Pascal en, in#13, :=

Conclusion

Features of the proposed system:

- Effective training
- High generalization ability
- Decentralized memory
- Collective classification
- Robust to antibody deletion

Future work:

- Testing on various problems
- Variable learning rate
- Soft binding
- Antibody population postprocessing

Project's web page is available at:

inversed.ru/AIS.htm