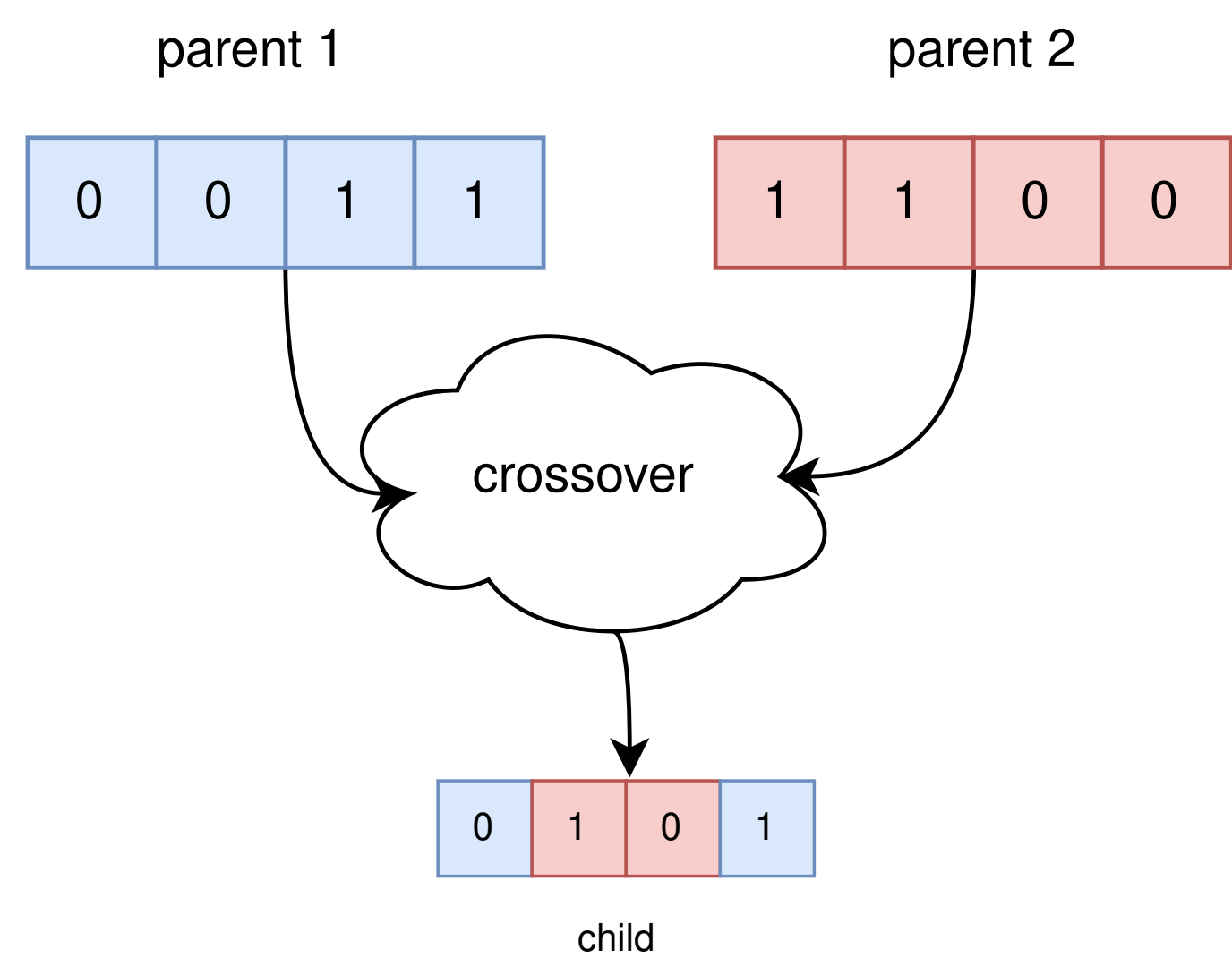


Preserving Inter-gene Relations during Test Case Generation using Intelligent Evolutionary Operators

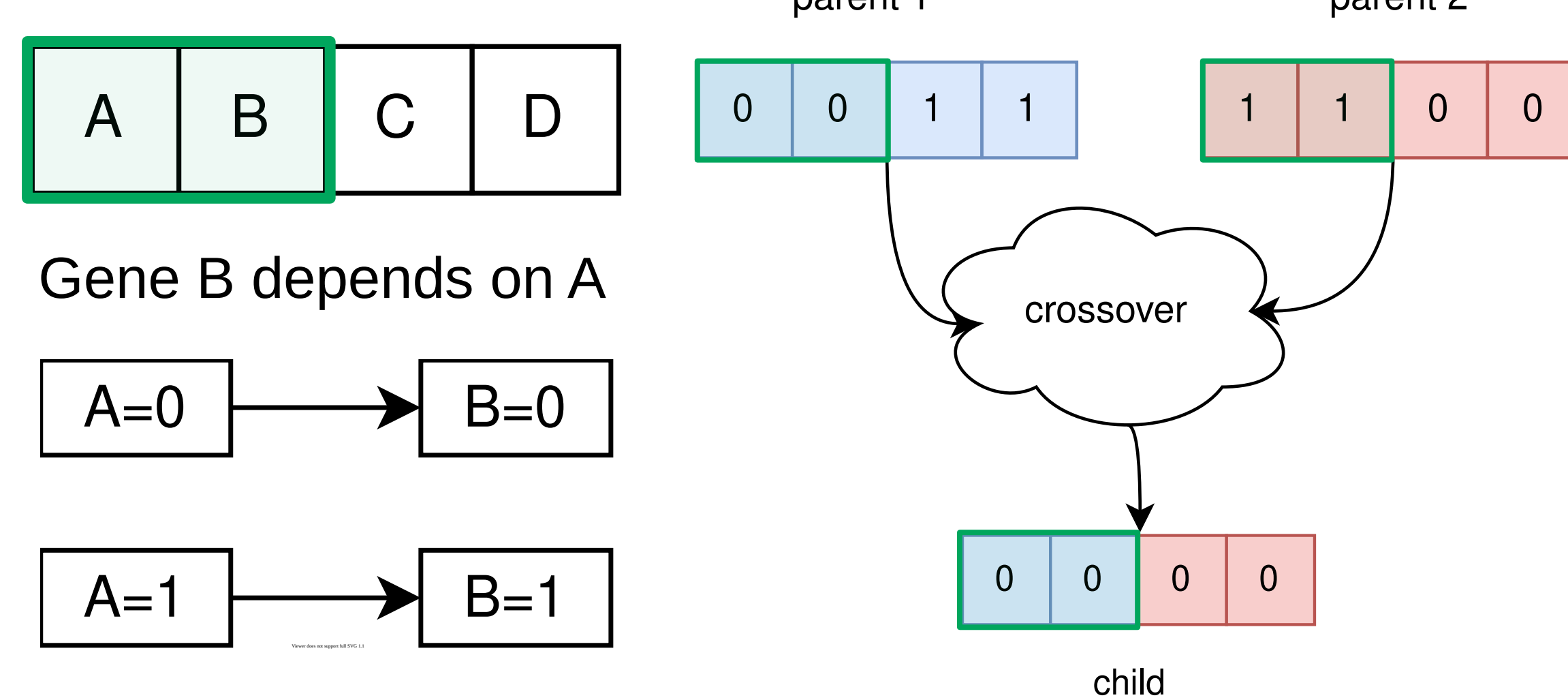
Dimitri Stallenberg

Disruptive Crossover Operators

Random

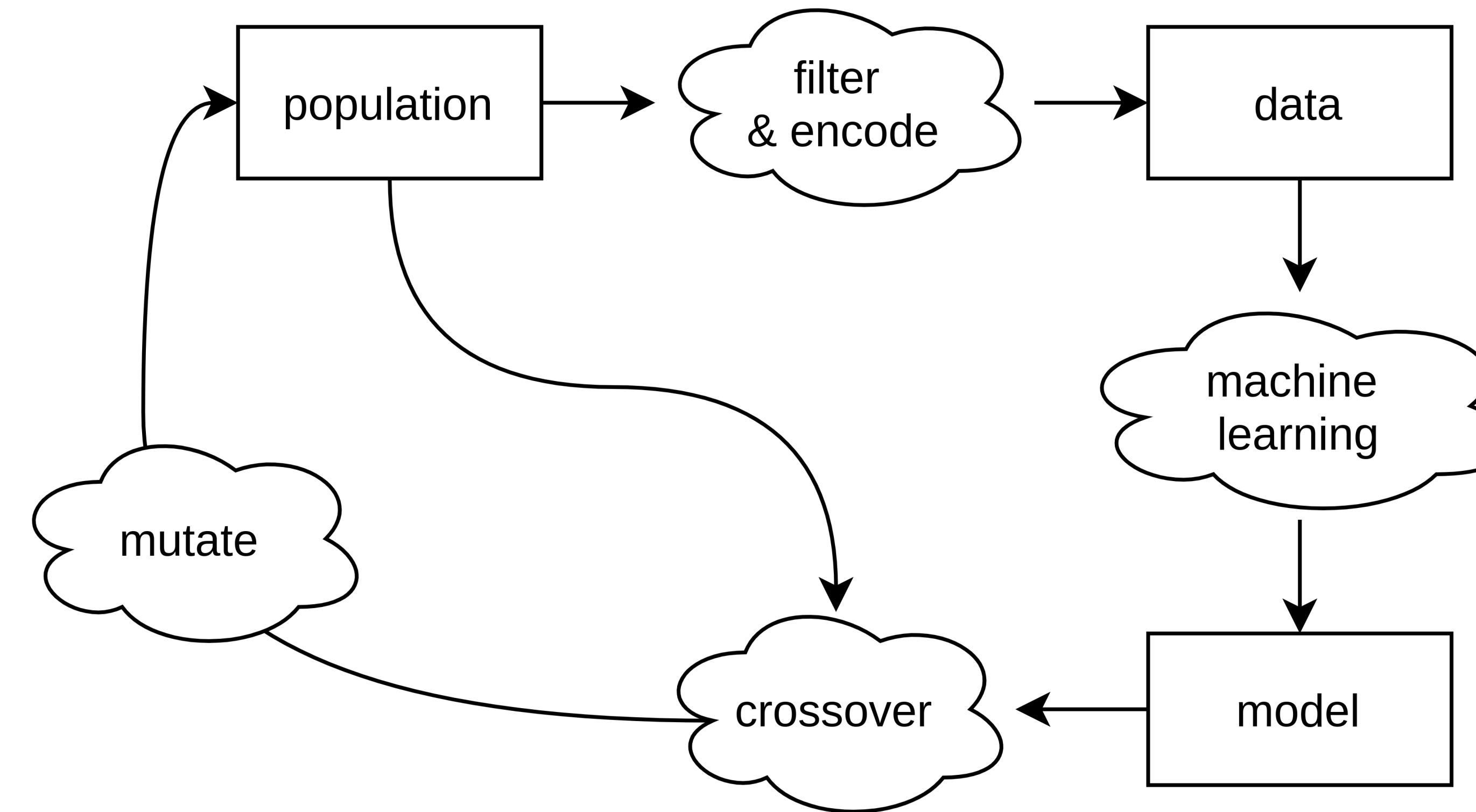


Building Blocks



Process

Process Flow



Encoding

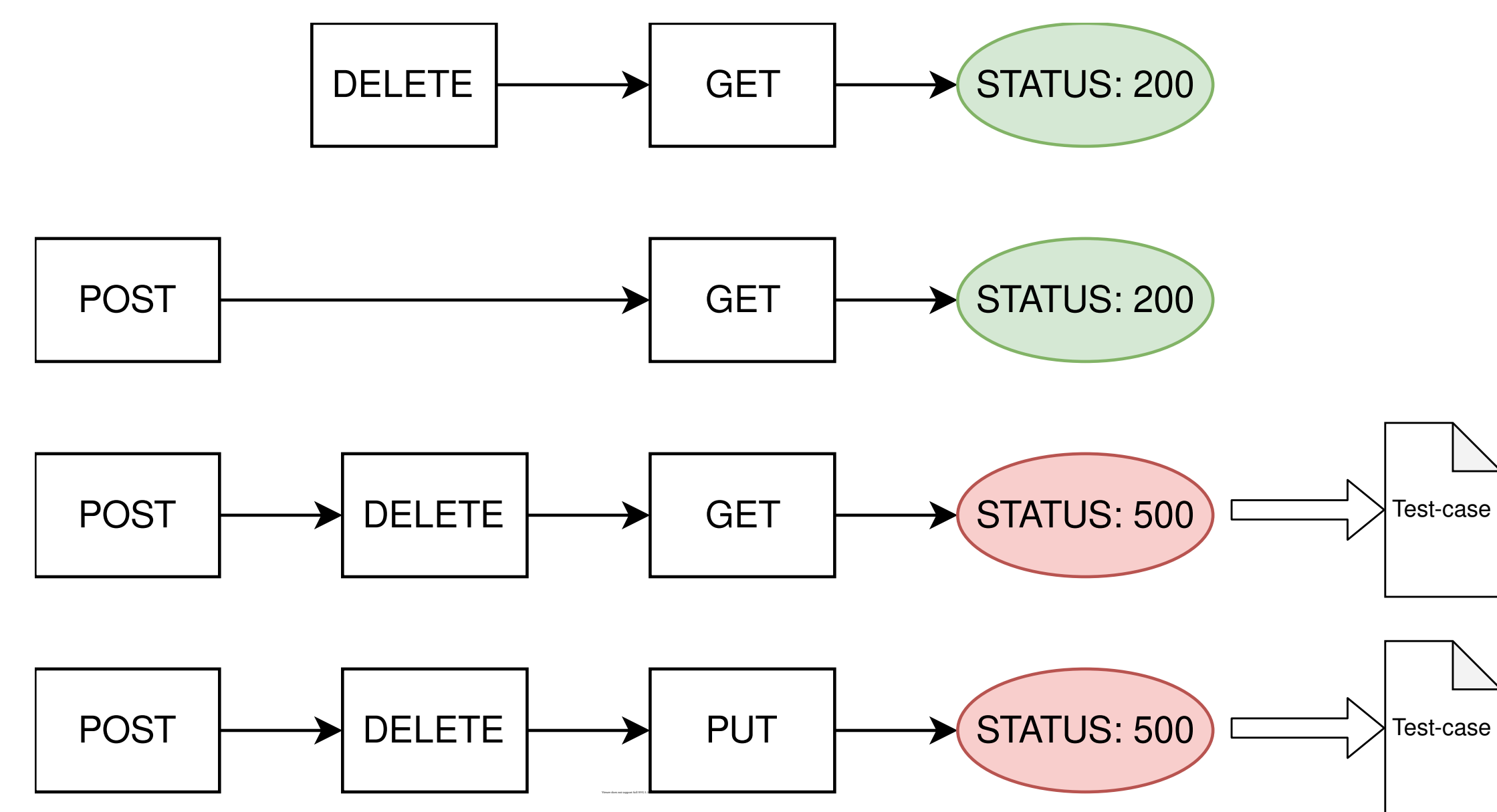
n possible actions
 k solutions in population

a_0	a_1	a_2	...	a_n
a_0	a_1	a_2	...	a_n
a_0	a_1	a_2	...	a_n
...
a_0	a_1	a_2	...	a_n

RESTfull API Testing

Looks like the POST and DELETE request depend on each-other.

Hence, the building block approach can be beneficial to the search process.

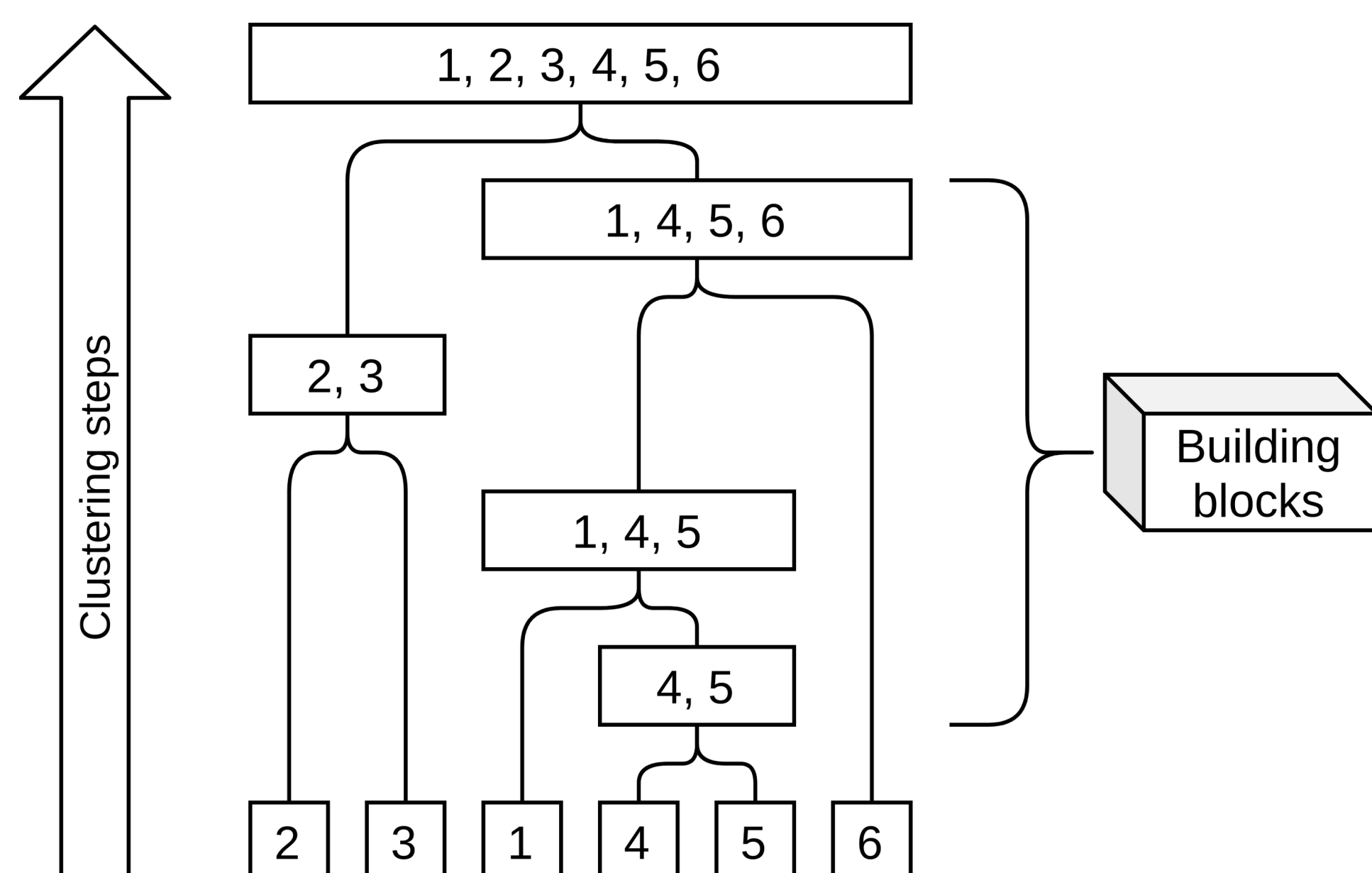


What is the impact of using intelligent evolutionary operators on the performance of test-case generating evolutionary algorithms in terms of covering test targets?

Machine Learning

ACMOSA

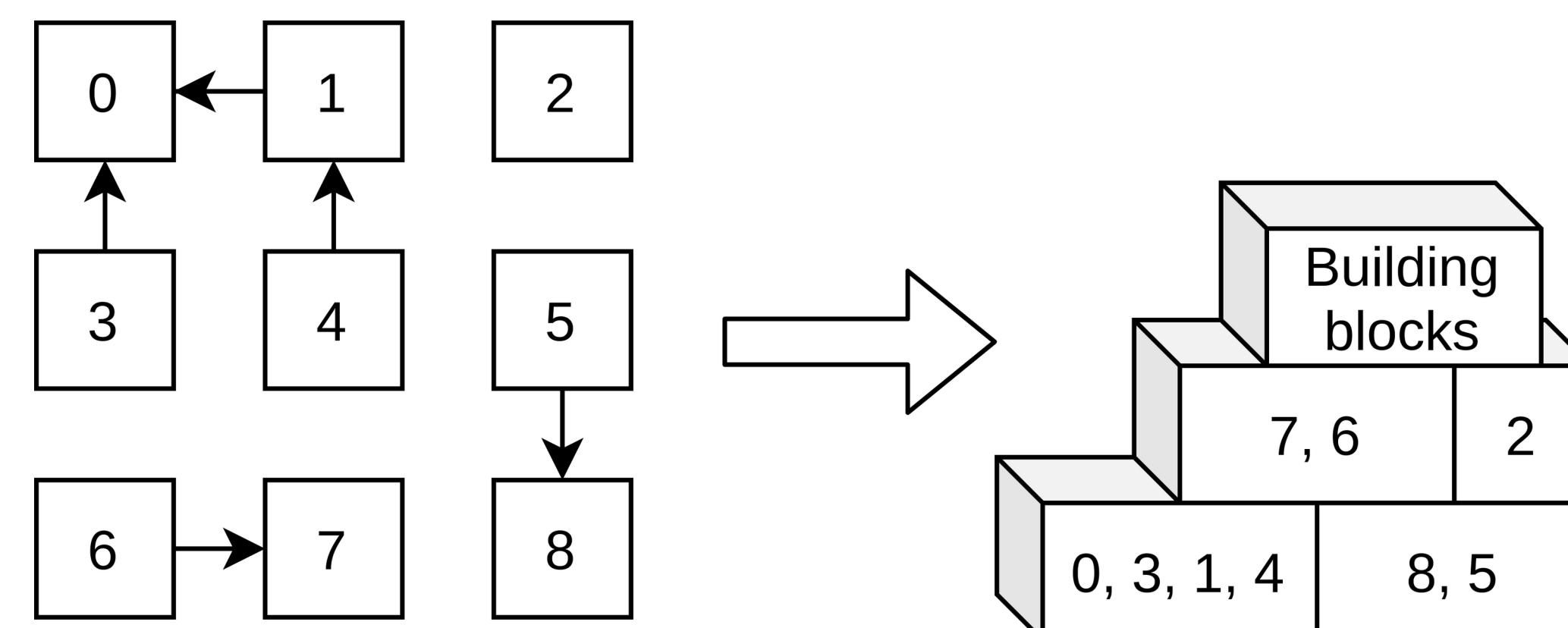
- Uses Agglomerative Hierarchical Clustering
- Finds the most similarly occurring parts of solutions



GOMOSA

- Uses Gene-pool Optimal Mixing
- Finds a Bayesian Network that most accurately fits the distribution of partial solutions

Bayesian Network to Building Blocks



Evaluation

Results

Out of the 5 benchmark APIs ACMOSA and GOMOSA performed:

- better for 2 benchmarks
- equal for 2 benchmarks
- worse for 1 benchmark

Conclusions

- ACMOSA and GOMOSA provide a significant benefit over MOSA for certain APIs
- Performance difference per API is most likely due to a combination of the API size and the level of dependence between API endpoints

Case: NCS benchmark

