

Norwegian Research Center for Al Innovation

EvolP.jl: Modular Optimisation in Julia using Evolutionary Computation

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Abstract

We present an open-source software framework, EvoLP.jl [2], to support the research in optimisation using Evolutionary Computation (EC) in the Norwegian scientific community. EC is highly relevant in many problems in artificial intelligence, engineering and statistics when non-convex or non-differentiable functions appear. The software is a package in the Julia programming language that provides reusable computing *blocks* for experimenting and analysing several components for single-objective EC algorithms. By stacking these blocks, the user can quickly create **modular solvers** where each of the components can be easily swapped for testing. In addition, it provides a few built-in algorithms ready to use out of the box. A bunch of utilities for analysis are available as well: test functions, result reporting, and statistics logging and overview. EvoLP.jl is an effort of the Norwegian Open Artificial Intelligence Lab.

And here is one **possible output** of the solution above:

```
optimum(result) = 0
optimizer(result) = Any[5, 1, 8, 6, 3, 7, 2, 4]
```

Check the full **step-by-step** example in the **documentation**.

What else can EvoLP do?

Components

A visual example: The 8-queens problem

This example deals with a classical combinatorial problem in AI where the goal is to place 8 queens in a chess board such that no queen checks each other [1]. Fig. 1 shows three configurations where the constraints and possible clashes are highlighted.





Figure 2: Stack the components together to make your own solver.

And what can I use it for?

Combinatorial challenges:

Assignment and packing

Scheduling and search

Constraint satisfaction and optimisation

Jump right into it

• Random **population generators** for vectors and particles

• Parent selectors

• Several **recombinators** and **mutators**

- Test functions for benchmarking your algorithms
- Convenient result reporting and a logbook for **statistics**

• Built-in **algorithms** • Support for **custom operators**

Numerical optimisation and **tuning** for machine learning:

- Hyperparameter tuning
- Neuroevolution
- Feature selection

Figure 1: The 8-queens problem. 1a shows the constraints (in pink) imposed by the placement of a single queen piece (in blue). 1b highlights the conflicts arising from a possible configuration of the board. 1c illustrates one possible solution with no conflicts.

Let's design a solution

Using the **provided blocks** we can set up a solver quickly. We would need:

• A permutations **generator**

• A tournament **selector**

- A permutation **mutator** • An **objective**: **minimise** conflicts
- A permutation **recombinator**

Let's code the solution

What is Julia?

Julia is high-level, high-performance programming language very suitable for scientific computing. It is part of the PetaFLOPS Club (10^{15} floating point operations per second) along with C, C++ and Fortran, and its syntax is similar to Python or MATLAB. This is the Julia code for solving the 8-queen problem using **EvoLP.jl**:

EvoLP.jl is well-tested, provides extensive documentation and is free—available for everyone to use under an open-source license at GitHub. After installing Julia, you can **install EvoLP.jl** by using the Julia **REPL**:

julia> **import** Pkg julia> Pkg.add("EvoLP")

> **Figure 3:** Visit the GitHub repository by scanning this QR code.

Acknowledgements

This should install **EvoLP** in your system.



EvolP.jl is partly funded by Project no. 311284 of The Research Council of Norway. We would like to thank the Norwegian Open Artificial Intelligence Lab for the promotion and hosting of the framework in its GitHub repository, and IDI for access to its computing resources.

using EvolP

- $X = permutation_vector_pop(100, 8, 1:8)$
- S = TournamentSelectionSteady(5)
- C = OrderOneCrossover()
- M = SwapMutation()
- $f = diag_constraints(x)$
- result = mySteadyGA(statsbook, diag_constraints, X, 500, S, C, M, 0.8) \hookrightarrow @show optimum(result) @show optimizer(result)



References

[1] S. Russell and P. Norvig. Artificial Intelligence: A Modern Approach. Pearson Series in Artificial Intelligence. Pearson, 2020.

[2] X. F. C. Sánchez-Díaz and O. J. Mengshoel. EvoLP.jl: a playground for evolutionary computation in Julia. In NAIS'23: Symposium of the Norwegian AI Society, Bergen, Norway, June 2023.

NorwAl Innovate Conference 2023