



INF 421 PROJECT

Evolutionary Algorithms
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INTRODUCTION

This report presents a solution to the programming project **Evolutionary Algorithms** for the course INF421: Design and Analysis of Algorithms at École Polytechnique. Each task is developed in a section of the report which also contains the code implemented using the Nim programming language.

The source code can be accessed on the project's Github repository.

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TASK 1: INDIVIDUALS AND BENCHMARK FUNCTIONS

Write code that allows to use individuals as well as the three functions `OneMax`, `LeadingOnes`, and `Jump_k`. For individuals, do not use libraries but implement a data type that fully utilizes the memory. That is, do not store each bit value of an individual in a byte but in an actual bit.

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TASK 2: RUNTIME ANALYSIS

Prove mathematically (preferably rather tight) upper bounds on the expected run time of the $(1 + 1)$ EA on **OneMax** and on **LeadingOnes**.

We use the fitness levels approach (1).

3.1 THEORETICAL BOUND FOR **OneMax**

3.2 THEORETICAL BOUND FOR **LeadingOnes**

Complement your theoretical bounds with empirical results and compare them.

Furthermore, run empirical tests for the $(\mu + 1)$ EA on **OneMax** with various, self- chosen values of μ . Visualize the expected run time. What do you see? What μ would you recommend?

In your report, do not forget to add a brief discussion about the parameter choices you made yourself, especially the number of tries for each value of μ you chose.

4 TASK 3

5 TASK 4

6 TASK 5

REFERENCES

- [1] Doerr, B., Kötzing, T. Lower Bounds from Fitness Levels Made Easy. *Algorithmica* (2022).
<https://doi.org/10.1007/s00453-022-00952-w>