

# Optimized Shopping, solving TSP using a parallel Genetic Algorithm

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## ABSTRACT

This work deals with the problem of finding the shortest route during shopping using a shopping list. It is therefore important to find out which route is the best aka shortest route based on the products and their sorting in the store. A problem commonly known as a traveling salesman problem. In this paper, a solution is presented that is based on parallel genetic algorithms. To use the full advantages of the parallel genetic algorithms, a three-part architecture was designed. The architecture consists of a display component, which was implemented as a Unity frontend, and a master/worker architecture. The master component ensures communication with the frontend and manages the master population, which are processed by the workers.

## Author Keywords

traveling salesman; tsp; genetic algorithm; parallel genetic algorithm; distributed systems; th köln; architecture of distributed systems

## INTRODUCTION

This project is part of the Architecture of Distributed Systems module in the Master's program in Computer Science (specializing in software engineering) at the [TH Köln](#). The aim of the module is to work on a problem that can be solved faster or better by using a distributed architecture. Students are free to choose the topic.

For implementation and testing, the students are provided with a PC pool with 12 Macs (TODO: insert exact description). With the help of these computers, the Measurement for this project were carried out.

## PROBLEM DESCRIPTION

The selected problem is a classic traveling salesman problem. The aim is to use a shopping list to find the optimal route through a chosen store to enable the customer to have a stress-free shopping experience.

The solution developed in the project will be tested using the example of the Rewe market in the Gummersbach forum. The range and scale are abstracted to ensure a clear presentation.

## ALGORITHM

When finding the algorithm, the main focus was on the possible parallelization, which made the choice of a genetic algorithm. The advantage of genetic algorithms is that they can be parallelized particularly well in a variety of ways. When calculating genetic algorithms, the most time-consuming operation is determining the fitness of an individual [1]. Since the distribution of the fitness calculation over a wide number of components would generate a large network traffic, the distributed calculation of entire populations was decided.

To ensure an exchange between many populations, different models, such as the island model [2], the pollen model [3] and the nearest neighbor migration model [3, 4] were compared and analyzed.

The nearest neighbor migration was chosen for the algorithm because it offers the advantage of generating a low network load (especially compared to the pollen model) and at the same time ensuring a continuous exchange of individuals among the populations.

Further information on configuration can be found in the Configuration of the Algorithm chapter.

## ARCHITECTURE

The architecture was divided into three components. Firstly, the pure frontend, which was implemented in Unity. Secondly, the implementation of the algorithm in its own master/worker architecture. The master, hereinafter referred to as the scheduler, ensures the exchange between the workers and with the frontend. The workers receive populations and evolve them.

In the design there was a strong focus on the loose coupling of the components so that the failure of individual components does not limit the functionality of the entire system.

REST was used for the exchange between the components. This guarantees a free choice of technology for the individual components.

## COMPONENTS

### Unity-Frontend

The frontend was developed in Unity and, in addition to the visual display, is also responsible for generating the shopping list and a navigation mesh that shows all producty and usable walkways. After creating a shopping list, it is sent to the scheduler. By regularly asking the scheduler, new calculation results are queried and displayed.

### Scheduler

The scheduler represents the master component and therefore offers a REST API, which can be viewed in the [OpenApi 3.0 documentation](#). The scheduler is written in Kotlin and Ktor is used as the server framework.

After receiving a new shopping list, a master population is created. All workers who are already registered or who will register in the future will be assigned a subpopulation for processing. If a worker delivers an updated population, an individual exchange is initiated with the neighboring population.

### Worker

The worker maps the computing component and is implemented as a simple Java application. At the beginning, the worker logs on to the scheduler and waits for a shopping list and thus population to be available. After receiving the population and the navigation mesh, the population is processed.

## CONFIGURATION OF THE ALGORITHM

With genetic algorithms, there are many possible adjustments that can be tuned to influence the result. Different configurations were therefore tested for scheduler and workers.

The most important are named below and briefly explained. More detailed explanations and specific values can be found in the [complete documentation](#) (available only in german).

### Scheduler-Config

- **Worker-Count:** The number of simultaneous workers supported by the scheduler. Based on this, the number of subpopulations is calculated
- **Master-Population-Size:** The size of the master population.
- **Demo-Individual-Size:** Before a final result is passed on to the frontend, a number of individuals is held, which consists of the best individuals from the processed populations. Only when this minimum number is reached is a result passed on.
- **Min-Delta:** The min delta ensures that no final result is passed on to the frontend as long as there are still large discrepancies between the demo individuals collected. So there is still a lot of room for improvement.
- **Individual-Exchange-Count:** This determines how many individuals are exchanged between the subpopulations.

### Worker-Config

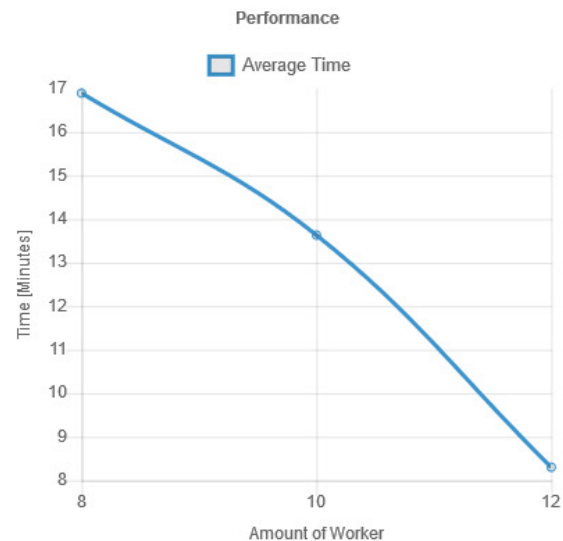
- **Mutation-Rate:** The mutation rate describes the frequency of the mutation of a single gene.
- **Tournament-Size:** The tournament size describes how many individuals are randomly drawn from the population for the selection.

## MEASUREMENT

The measurements were performed in the scheduler using the Char.js library. Before the actual measurements and thus the comparison of the performance was carried out using different numbers of workers, various tests were used out to find the optimal parameters.

This generated attention to important details that had to be adjusted for the different numbers of workers. Using the example of the scheduler setting worker-count, this means, it is not enough just to adjust this setting to the number of workers actually used. A test with six real workers and a worker count of 12 resulted in a computing time of 7.58 minutes. However, when the worker counts were adjusted to six, the computing time was extended to 14 minutes. After further testing, this is related to the population size. Due to the halved number of workers, the population size per worker doubled, which led to considerable increased computing times.

The knowledge gained from the parameter tests was subsequently be used for the tests with differentiating real worker numbers.



**Figure 1.** Comparison of computing time with different numbers of workers

The tests with different numbers of workers quickly showed how much the parallelization helps to shorten the calculation. As in figure 1 to see, the calculated time is greatly reduced. The originally planned tests with fewer workers were avoided due to the longer and longer runtimes.

## CONCLUSION

Even if genetic algorithms do not guarantee optimal results, they offer a good opportunity to approach problems that cannot be solved in a finite amount of time by pure computing power to such an extent that they become qualitative, processable data. Combining these advantages with the strengths of a distributed calculation, complex problems can be solved in

acceptable times. Using the example of the shopping list, it is sufficient to write it before the actual shopping, when you arrive at the store, the route is ready and usable.

If you are interested in the application and procedure, the entire documentation (only available in German) and all sources are available in our [GitHub repository](#).

## ACKNOWLEDGMENTS

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