HOMEWORK 2 — Report

1 Experimental Work

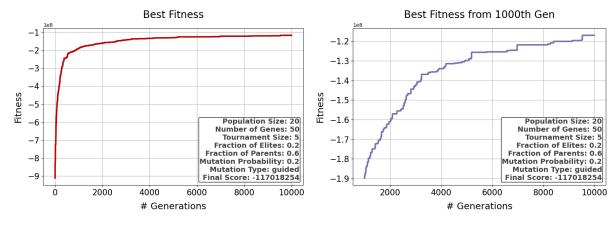
The parameter set provided in Table 1 is used to perform qualitative analysis of different evoluationary algorithm parameters.

Parameter	Values				
<num_inds></num_inds>	5	10	20	40	60
<num_genes></num_genes>	15	30	50	80	120
<tm_size></tm_size>	2	5	8	16	
$<$ frac_elites $>$	0.04	0.2	0.35		
<frac_parents></frac_parents>	0.15	0.3	0.6	0.75	
<mutation_prob></mutation_prob>	0.1	0.2	0.4	0.75	
<mutation type=""></mutation>	unguided	guided			

Table 1: Parameter set used in this experiment.

1.1 Default Parameter Set

Let us first start with the results of default parameter set provided in the homework description to be used as a baseline throughout the experiment. Figure 1 shows the plots associated with the fitness value of the best individual. The default parameter set is indicated as bold in Table 1.



- (a) From first generation to 10000th generation.
- (b) From 1000th generation to 10000th generation.

Figure 1: Fitness curves.

The figures tells us that in the first 1000 generations, the fitness value of the best individual increases rapidly. However, after the 1000th generation, the increase in fitness value slows down. However, the fitness value of the best individual is still increasing and we can see this from the figure on the right. That means the algorithm continue to improve even after rapid jump.

The Figure 2 shows the evolution of the best individual quantitatively. The first image is the image in 1000th generation and the last image is the final image. The images in between are the images of the best individual in every 1000 generation. The images are generated by overlaying the circle represented by each gene one by one on the plain white image as instructed.



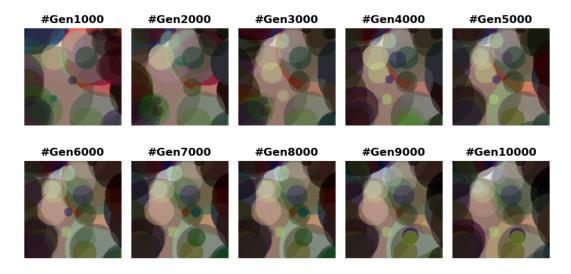


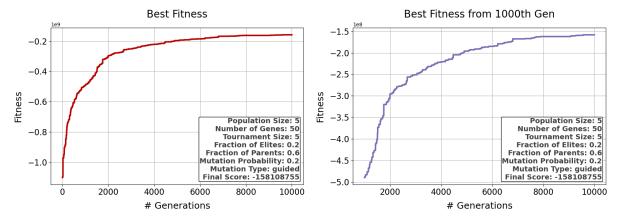
Figure 2: Quantitative evolution of the best individual in the population.

1.2 Number of Individuals

Let us first provide necessary plots for the parameter < num inds>.

1.2.1 5 Individuals

Figure 3 shows the plots associated with the fitness value of the best individual for 5 individuals.



- (a) From first generation to 10000th generation.
- (b) From 1000th generation to 10000th generation.

Figure 3: Fitness curves for 5 individuals.

The Figure 4 shows the evolution of the best individual quantitatively for 5 individuals.

Evolution of Images

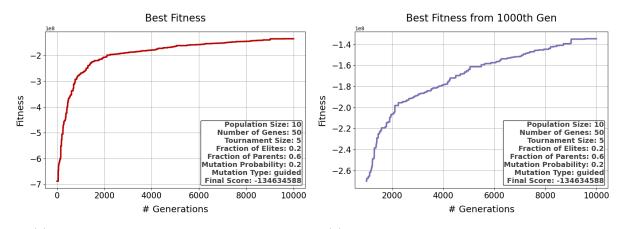
Population Size: 5 Number of Genes: 50 Tournament Size: 5 Fraction of Elites: 0.2 Fraction of Parents: 0.6 Mutation Probability: 0.2 Mutation Type: guided



Figure 4: Quantitative evolution of the best individual in the population for 5 individuals.

1.2.2 10 Individuals

Figure 5 illustrates the plots associated with the fitness value of the best individual for 10 individuals.



(a) From first generation to 10000th generation.

(b) From 1000th generation to 10000th generation.

Figure 5: Fitness curves for 10 individuals.

The Figure 6 shows the evolution of the best individual quantitatively for 10 individuals.

1.2.3 40 Individuals

Figure 7 shows the plots associated with the fitness value of the best individual for 40 individuals.

The Figure 8 indicates the evolution of the best individual quantitatively for 40 individuals.

Population Size: 10 Number of Genes: 50 Tournament Size: 5 Fraction of Elites: 0.2 Fraction of Parents: 0.6 Mutation Probability: 0.2 Mutation Type: guided

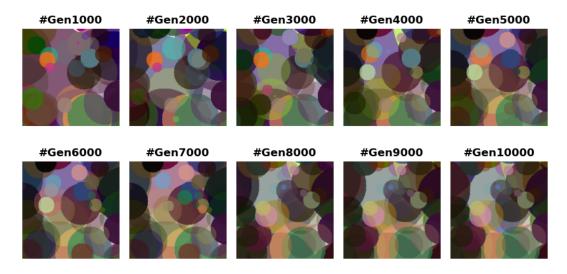
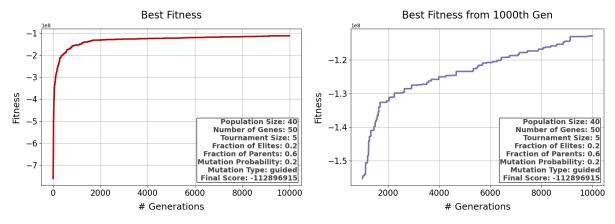


Figure 6: Quantitative evolution of the best individual in the population for 10 individuals.



- (a) From first generation to 10000th generation.
- (b) From 1000th generation to 10000th generation.

Figure 7: Fitness curves for 40 individuals.

1.2.4 60 Individuals

Figure 9 shows the plots associated with the fitness value of the best individual for 60 individuals.

The Figure 10 shows the evolution of the best individual quantitatively for 60 individuals.

Discussion: The results show that the number of individuals in the population has a significant effect on the performance of the algorithm. It can be deduced that as the number of individuals in the population increases, the algorithm converges to a better solution. This is because the diversity in the population increases as the number of individuals increases.

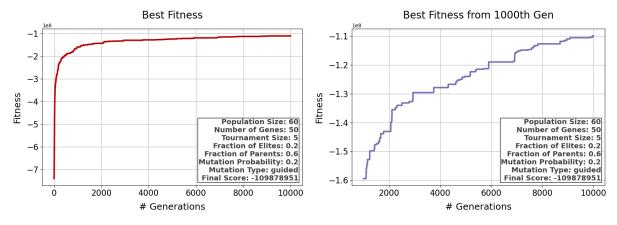
1.3 Number of Genes

Let us first provide necessary plots for the parameter < num genes>.

Population Size: 40 Number of Genes: 50 Tournament Size: 5 Fraction of Elites: 0.2 Fraction of Parents: 0.6 Mutation Probability: 0.2 Mutation Type: guided



Figure 8: Quantitative evolution of the best individual in the population for 40 individuals.



- (a) From first generation to 10000th generation.
- (b) From 1000th generation to 10000th generation.

Figure 9: Fitness curves for 60 individuals.

1.3.1 15 Genes

Figure 11 shows the plots related to the fitness value of the best individual for 15 genes.

The Figure 12 illustrates the evolution of the best individual figuratively for 15 genes.

1.3.2 30 Genes

Figure 13 shows the plots related to the fitness value of the best individual for 30 genes. The Figure 14 shows the evolution of the best individual figuratively for 30 genes.

1.3.3 80 Genes

Figure 15 shows the plots related to the fitness value of the best individual for 80 genes.

Population Size: 60 Number of Genes: 50 Tournament Size: 5 Fraction of Elites: 0.2 Fraction of Parents: 0.6 Mutation Probability: 0.2 Mutation Type: guided

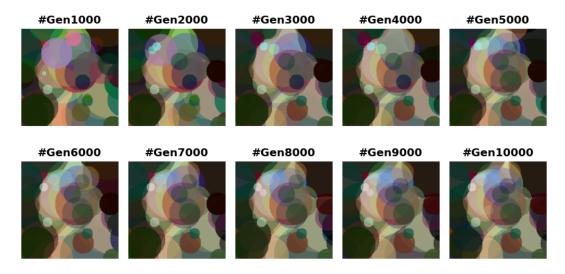
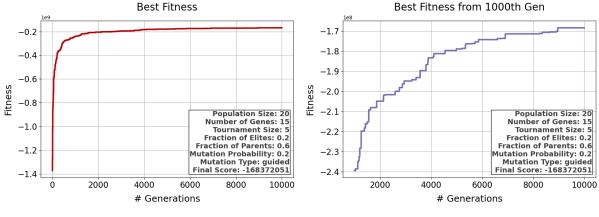


Figure 10: Quantitative evolution of the best individual in the population for 60 individuals.



- (a) From first generation to 10000th generation.
- (b) From 1000th generation to 10000th generation.

Figure 11: Fitness curves for 15 genes.

The Figure 16 shows the evolution of the best individual figuratively for 80 genes.

1.3.4 120 Genes

Figure 17 shows the plots related to the fitness value of the best individual for 120 genes.

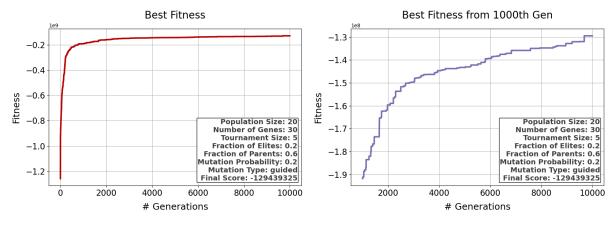
The Figure 18 is given for the evolution of the best individual for 120 genes.

Discussion: We see that, as the number of genes in the individual increases, the algorithm converges to a better solution. This is because the number of genes in the individual increases, the search space increases and the algorithm can explore more possibilities. This was somewhat expected since the number of genes in the individual is directly related to "paint brushes" we have in this problem. However, from 80 genes to 120 genes the improvement is not obserable 80 genes converged to a better solution. We can

Population Size: 20 Number of Genes: 15 Tournament Size: 5 Fraction of Elites: 0.2 Fraction of Parents: 0.6 Mutation Probability: 0.2 Mutation Type: guided



Figure 12: Quantitative evolution of the best individual in the population for 15 genes.



- (a) From first generation to 10000th generation.
- (b) From 1000th generation to 10000th generation.

Figure 13: Fitness curves for 30 genes.

interpret that for 120 genes there need to be more generations to converge to a better solution. So, as the number of genes decreases, the number of generations needed to converge is also decreased. Given we fixed the number of generations, the results are natural in this sense.

1.4 Tournament Size

Let us first provide necessary plots for the parameter <tm size>.

1.4.1 2 Tournament Size

Figure 19 shows the plots related to the fitness value of the best individual for 2 tournament size.

The Figure 20 shows the evolution of the best individual figuratively for 2 tournament size.

Population Size: 20 Number of Genes: 30 Tournament Size: 5 Fraction of Elites: 0.2 Fraction of Parents: 0.6 Mutation Probability: 0.2 Mutation Type: guided

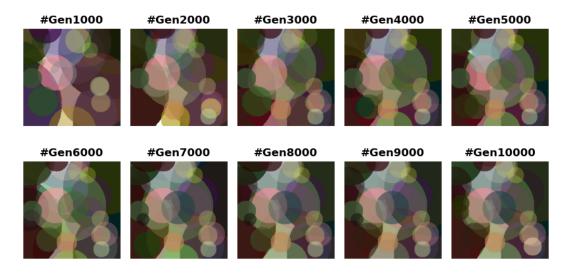
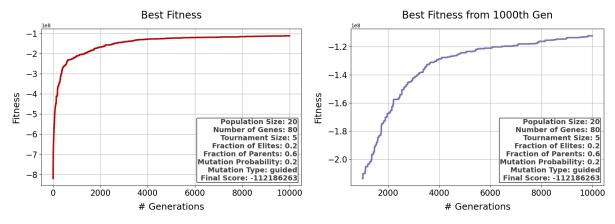


Figure 14: Quantitative evolution of the best individual in the population for 30 genes.



- (a) From first generation to 10000th generation.
- (b) From 1000th generation to 10000th generation.

Figure 15: Fitness curves for 80 genes.

1.4.2 8 Tournament Size

Figure 21 shows the plots related to the fitness value of the best individual for 8 tournament size. The Figure 22 shows the evolution of the best individual figuratively for 8 tournament size.

1.4.3 16 Tournament Size

Figure 23 shows the plots related to the fitness value of the best individual for 16 tournament size. The Figure 24 shows the evolution of the best individual figuratively for 16 tournament size.

Discussion:

Population Size: 20 Number of Genes: 80 Tournament Size: 5 Fraction of Elites: 0.2 Fraction of Parents: 0.6 Mutation Probability: 0.2 Mutation Type: guided

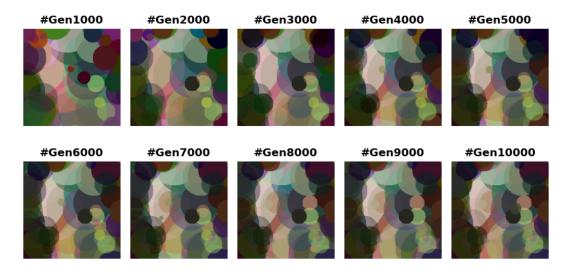
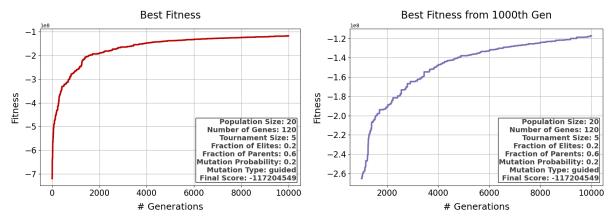


Figure 16: Quantitative evolution of the best individual in the population for 80 genes.



- (a) From first generation to 10000th generation.
- (b) From 1000th generation to 10000th generation.

Figure 17: Fitness curves for 120 genes.

Population Size: 20 Number of Genes: 120 Tournament Size: 5 Fraction of Elites: 0.2 Fraction of Parents: 0.6 Mutation Probability: 0.2 Mutation Type: guided

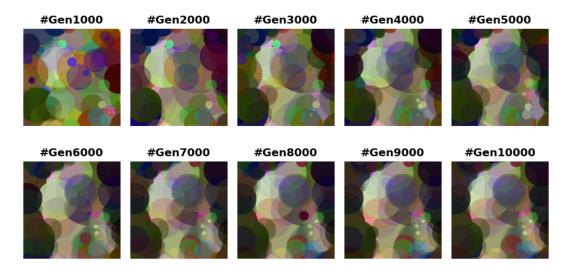
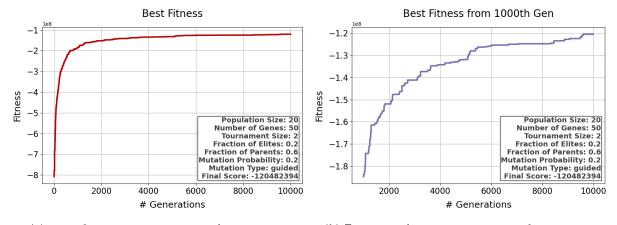


Figure 18: Quantitative evolution of the best individual in the population for 120 genes.



- (a) From first generation to $10000 \mathrm{th}$ generation.
- (b) From $1000 \mathrm{th}$ generation to $10000 \mathrm{th}$ generation.

Figure 19: Fitness curves for 2 tournament size.

- 1.5 Number of Individuals Advancing to Next Generation
- 1.6 Number of Parents Used In Crossover
- 1.7 Mutation Probability
- 1.8 Mutation Type

2 Discussion

Appendix

The code set used throughout this homework is provided as follows.

Population Size: 20 Number of Genes: 50 Tournament Size: 2 Fraction of Elites: 0.2 Fraction of Parents: 0.6 Mutation Probability: 0.2 Mutation Type: guided

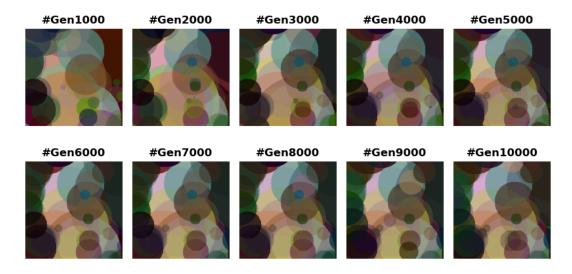
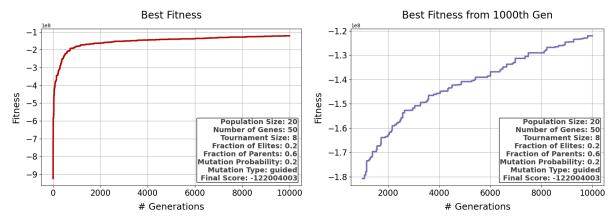


Figure 20: Quantitative evolution of the best individual in the population for 2 tournament size.



- (a) From first generation to 10000th generation.
- (b) From 1000th generation to 10000th generation.

Figure 21: Fitness curves for 8 tournament size.

Submitted by Ahmet Akman 2442366 on April 28, 2024.

Population Size: 20 Number of Genes: 50 Tournament Size: 8 Fraction of Elites: 0.2 Fraction of Parents: 0.6 Mutation Probability: 0.2 Mutation Type: guided



Figure 22: Quantitative evolution of the best individual in the population for 8 tournament size.

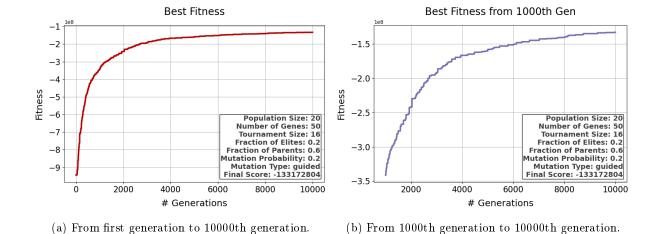


Figure 23: Fitness curves for 16 tournament size.

Population Size: 20 Number of Genes: 50 Tournament Size: 16 Fraction of Elites: 0.2 Fraction of Parents: 0.6 Mutation Probability: 0.2 Mutation Type: guided

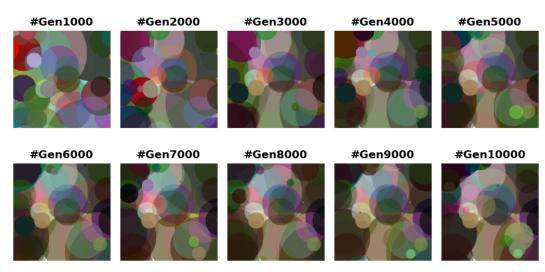


Figure 24: Quantitative evolution of the best individual in the population for 16 tournament size.