

Sustainable Manufacturing: Application of Optimization to Textile Manufacturing Plants

Isuru Liyanage¹, Sachira Nuwanga², Ravidu Anjana³, Windhya Rankothge⁴ and Narmada Gamage⁵

¹ Sri Lanka Institute of Information Technology

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Abstract

The main goal of manufacturing industry is to produce the end products on time with good quality and keep the resource wastage low. However, manufacturing industry face several challenges such as bottle necks in the workflow, unsynchronized production, and sudden increase in product demands. In this paper, we are proposing a management platform for textile manufacturing plants with following modules: (1) sewing workflow optimization (2) quality assurance workflow optimization and (3) finishing workflow optimizations. We have used Genetic Programming (GP) approach, to optimize the workflows, considering different factors that affect each workflow. Our results show that, using our proposed platform, the manufacturing workflows can be optimized and reduce the bottle necks in the workflows and resource wastage in the manufacturing plant.

Index terms— manufacturing plant, textile industry, optimization, planning, scheduling, genetic programming (GP).

1 Introduction

The textile industry is an important segment in the world's manufacturing industry, playing a crucial role in economic development. It is an evergrowing market, with key competitors being China, European Union, United States, and India [19]. The textile sector offers huge employment opportunities for people, especially for the people reside in rural area.

In the textile industry, specifically in manufacturing plants, the production process is a combined effort of different sections such as designing, cutting, sewing, packing, and delivering. They ensure that the clothes are produced with good quality and delivered on time with minimal wastage. However, the textile industry faces several challenges such as bottle necks in the workflow, unsynchronized production, sudden increases in product demands and unseen product wastages [20].

Therefore, it is necessary to adopt to a proper planning and scheduling approach, so that the workflows of manufacturing plants can be optimized. The optimal planning and scheduling are traditionally modeled as Integer Linear Programming (ILP) optimization. However, there is the intrinsic constraint that ILP optimization is a NP-complete problem, and even when solutions are obtained for limited classes, it might be too slow [25]. Therefore, ILP calculation is unfeasible for optimization problems with dynamic nature, even though an ILP formalization of the problem gives an exact solution. The best approach would be to use an approximation approach (heuristic-based approach) for planning and scheduling of manufacturing plants activities.

In this paper, we have proposed a platform for planning and scheduling of a manufacturing plant, specifically targeting the textile industry. We have identified three main workflows of a textile manufacturing plants: sewing, quality assurance and finishing. Therefore, the proposed platform includes following modules: (1) sewing workflow optimization (2) quality assurance workflow optimization and (3) finishing workflow optimization. We have used an approximation (heuristic) based approach: Genetic Programming (GP), to optimize the identified

44 work flows. The GP approach is a well-known method in the planning and scheduling context [1][2][3][4]
45 [5][6][7][8][9][10][11][12][13][14][15][16]. Moreover, they have been widely used for research works related to
46 manufacturing planning systems.

47 We have identified several parameters that effect each workflow, and processes required to be optimized in each
48 workflow. We have consulted and visited textile manufacturing plants, their management and workers to identify
49 the factors contributing to optimizing workflows. We have conducted experiments to measure the performances
50 of our algorithms using the Sri Lankan textile manufacturing industry as a case study. With the real data sets,
51 our results show that, our proposed GP approach can decide the optimal planning of each work flow dynamically
52 in the order of seconds.

53 The rest of the paper is organized as follows. Section II presents the related work. Section III introduces the
54 proposed optimization algorithm for planning and scheduling of textile manufacturing plants. In section IV, the
55 result and discussions are presented. Our final remarks can be found on Section V. The references are mentioned
56 in Section VI.

2 II.

3 Related Work

59 The Genetic Programming (GP) approach is a well-known method in the planning and scheduling context
60 [1][2][3][4] [5][6][7][8][9][10][11][12][13][14][15][16], especially for NP-hard problems, In this section, we will discuss
61 the most recent work on scheduling and planning in the textile manufacturing context.

62 The work on [1] shows that resourceconstrained project scheduling problems are NP-hard and therefore,
63 it is recommended to use approximation (heuristic) approaches to solve the problem. They have proposed
64 a multi-objective evolutionary algorithm to solve the problem with multiple activity performance modes and
65 two objectives to minimize project makespan and resource utilization smoothness. They have used GP for their
66 proposed algorithm. The authors of [2] used GP approach to implement an optimal resource allocation algorithm,
67 that minimizes the average execution time of the critical path of workflow under constraints. The authors of [3]
68 show that Master Production Scheduling (MPS) is the key action for success of a textile manufacturing plant.
69 First, they have analyzed the constraints, mainly the installed capacity and the number of workers. Next, with
70 the use of genetic algorithms, the MPS is optimized to carry out production planning, with an improvement of up
71 to 96% of the level of service provided. The research [4] developed a system that uses GP approach to optimize
72 pattern generation in textile production. Their results show that the system can generate enough patterns for
73 fabric designers according to a given set of constraints. The work on [5] focuses on scheduling trousers collection
74 orders, considering dates of starting orders and exporting them, the quantities of each order and the combined
75 importance of customer and style. They have used GP approach to generate the best orders scheduling solution.
76 The study of [6] addresses the production scheduling problem in a textile factory and they have used GP approach
77 to find the minimum total make-span. The authors of [7] presented a methodology to design a control strategy
78 to optimize a complex production process using a neural network combined with GP. The GP approach is used
79 to optimize the architecture and the underlying parameters of the neural network in order to achieve the most
80 effective model of the production process and to obtain set point values and raw material characteristics for an
81 optimal tenacity and elongation of the spinning yarns.

82 Our research work has been inspired by these works, and we are focusing on a proposing a management
83 platform which helps textile manufacturing plants to optimize their workflows and schedule them with minimum
84 bottle necks and wastages.

4 III.

5 Methodology

87 We are proposing to use an approximation approach (heuristic-based approach) for planning and scheduling
88 of textile manufacturing plants workflows, which have a dynamic nature. Specifically, we are proposing to use
89 Genetic Programming (GP) approach. We have identified 3 workflows in the textile manufacturing plant to apply
90 the optimization: sewing, quality assurance and finishing. We have implemented the E-Optimizer platform, which
91 can be used to optimize sewing, quality assurance and finishing workflows. Figure 1 Shows the system diagram
92 of E-Optimizer.

93 It is important to note that, there are several factors to be considered when optimizing selected workflows
94 in the textile manufacturing plant. As the initial step, we have identified parameters that effect each workflow,
95 and processes required to be optimized in each workflow. We have consulted and visited textile manufacturing
96 plants, their management and workers to identify the factors contributing to optimizing workflows.

97 For sewing process, parameters such as number of pieces to be sew, number of pieces that can be handled
98 by one machine, number of sewing machines and time taken for one item to be sewed, effect the effectiveness
99 of workflow. In the quality assurance workflow, parameters such as availability of chemical stain removers, hot
100 water and steam generators, number of washers, spinners, and dryers, effect the effectiveness of workflow. For the
101 finishing workflow, parameters such as availability of hand irons with a vacuum press table, scissors press, carousel
102 machines, steam dollies, availability of plastic bags, boxes, and cartons, effect the effectiveness of workflow.

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Volume XX Issue II Version I GP is a method of processing the initial population (a group of individual solutions) to get adapted to a new better environment, through threemajor genetic operators: (1) selection, (2) crossover and (3) mutation [18]. The group of individual solutions gradually evolved into a better area of the search space by continuing to evolve from generation to generation. They will eventually converge to the most appropriate individual solution, which is the optimal solution for a problem.

GP can be described by the following five main steps [18] The process begins with generating a set of individuals which is called a Population. Each individual is a Solution to the problem. After the initial population is generated, a fitness function is used to measure how good each solution is. Out of all the solutions, the best solutions are selected for next generation population.

Then genetic operations are applied to selected solutions to generate new solutions. We have considered two types of genetic operators to produce offspring: (1) mutation and (2) crossover. Crossover and mutation perform two different activities. Crossover is a convergence operation which is intended to pull the current population towards a local min/max. But the Mutation is a divergence operation, and it infrequently break one or more members of a population out of a local min/max space and possibly discover a better space. Since the end goal is to bring the population to convergence, crossover happen more frequently and mutation, being a divergence operation, happen less frequently.

The newly generated solutions are added to the existing set of solutions, which is known as the current population. This process is continued until y number of generations are explored. In the final generation, the solution that gives the best fitness value is selected as the best solution.

7 b) Optimization of sewing workflow

The optimization goal for sewing workflow is to optimize the sewing capacity arrangement; that is to find the best set of clothes to be sew within a given time limit. Cloth types are prioritized based on their profit generation capabilities, and given a priority, where high profit generating clothes are given a high priority value, to ensure maximum profit generation.

The sewing workflow optimization can be explained as the following constrained optimization: Maximize: $z = \sum_{i=1}^n P_i X_i$

(1) Subject to the constraint: $\sum_{i=1}^n T_i X_i \leq T$ (2)

Where, n

Chromosome size Ci i th gene (Cloth type) Pi Priority of the Cloth type i Ai Time taken to sew the Cloth type i X Time Threshold

The constraint (2) ensures that the selected set of clothes can be sewn within the given time threshold.

When we apply GP approach to the optimization problem, the encoding of a Chromosomes (optimal set of clothes to be sew within a given time limit) is demonstrated in Figure 2. The set of Chromosomes: A1, A2, A3 and A4 build the population. In a Chromosome, each gene supplies two pieces of information: (i) the gene index number represents the cloth type and (ii) the value in the gene signifies the whether the cloth type has been selected to sew or not. A "1" indicates that the respective item is selected and a "0" indicates that it is not selected. For example, the value of the last gene of Chromosome A1, indicates that the cloth type "Denim" has not been selected to sew. (3) Next, the genetic operators are applied to Chromosomes to produce off-springs. Even though the traditional way of producing off-springs is to select the fittest chromosomes and perform genetic operations, it would lead to chromosomes that are closer to one another and results in less diversity. Therefore, we have used Roulette Wheel Selection method [17] and generated better off springs, with more diversity.

In the Roulette Wheel Selection method, a circular wheel is divided into n pies, where n is the number of Chromosomes in the population. Each Chromosome gets a portion of the circle which is proportional to its fitness value.

As given in Table ??, first, the fitness value percentage is calculated for each Chromosome. Then the Roulette Wheel is represented as shown in Figure ??, where each Chromosome gets a portion of the circle which is proportional to its fitness value.

8 Fig. 3: Roulette Wheel Selection representation

To select parents to generate off-springs, first the wheel is rotated, and then the region of the wheel which comes in front of a fixed point is chosen as the parent. For the second parent, the same process is repeated. Genetic operations are performed to the selected two parents. We have performed one-point crossover and mutation to generate off-springs. This process is continued until x number of generations are explored. In the final generation, the solution that gives the best fitness value is selected as the best solution.

9 c) Optimizaiton of quality assurance workflow

The quality assurance process in textile industry ensures that there are no stains on manufactured clothes. Therefore, if a stain in spotted, the cloth is sent for cleaning. The optimization goal for quality assurance workflow is to optimize the cleaning process; that is to find the best set of clothes to be cleaned within a given time limit.

162 The quality assurance workflow optimization can be explained as the following constrained optimization:
 163 Maximize: $f(x) = \sum_{i=1}^n c_i x_i$ (4)

164 Subject to the constraint: The constraint (4) ensures that the selected set of clothes can be cleaned within the
 165 given time threshold. $t = \sum_{i=1}^n t_i x_i \leq T$ (5)

166 The GP approach used to optimize the quality assurance workflow is same as the GP approach used to optimize
 167 the sewing workflow, except for the fitness function, which measures how good a solution is.

168 The fitness function used in optimization of the quality assurance workflow is given in Equation (
 169 $f(x) = \sum_{i=1}^n c_i x_i$) (6)

170 10 d) Optimizing finishing workflow

171 The finishing process in textile manufacturing carries out pressing and ironing of the cloth. The optimization
 172 goal for finishing workflow is to optimize the finishing process; that is to find the best set of clothes to be pressed
 173 and ironed within a given time limit.

174 The finishing workflow optimization can be explained as the following constrained optimization: Maximize:
 175 The constraint (8) ensures that the selected set of clothes can be cleaned within the given time threshold. $f(x) = \sum_{i=1}^n c_i x_i$
 176 $t = \sum_{i=1}^n t_i x_i \leq T$

177 The GP approach used to optimize the quality assurance workflow is same as the GP approach used to optimize
 178 the sewing workflow, except for the fitness function, which measures how good a solution is.

179 The fitness function used in optimization of the quality assurance workflow is given in Equation
 180 (9) $f(x) = \sum_{i=1}^n c_i x_i$ (9)

181 IV.

182 11 Experiments and Results

183 In this section, we are presenting the results and observations of the modules:(1) sewing optimization (2) quality
 184 assurance optimization and (3) finishing optimization, which were implemented using the GP based approaches.
 185 We have consulted and visited textile manufacturing plants, their management and workers to identify the factors
 186 contributing to optimizing workflows and to collect the data sets related to three workflows.

187 We have used GP as our heuristic-based optimization approach because it gives reasonable results fast. We
 188 derived three individual fitness functions for the three work flows, based on the factors effecting each flow. We
 189 have used mutation and one-point cross over as genetic operations and selected parents using Roulette Wheel
 190 approach to generate off-springs.

191 12 a) Optimizing sewing workflow

192 Case 1: As explained in the in previous section, with the GP process the given initial solution is improved by
 193 mutations and crossover operations over the generations. Therefore, we explored how GP process improves the
 194 initial solution when optimizing the sewing workflow. First, we find the initial solutions, and then apply GP
 195 process to improve the solution. As shown in the Figure 4, most of the improvements in the fitness function
 196 happens early on (during first 8 generations) and after that improvements happens very rarely.

197 Case 2: Next we explored the relationship between time and number of generations for sewing optimization
 198 process. As shown in Figure 5, time taken for generations shows a sub-linear growth, and the optimal values are
 199 produced within milliseconds.

200 13 c) Optimizing finishing workflow

201 Case 5: To explore how the initial solution is improved with GP process when optimizing the finishing workflow,
 202 we conducted experiments. First, we find the initial solutions, and then applied GP process to improve the
 203 solution. As shown in the Figure ??, most of the improvements in the fitness function happens early on (during
 204 first 6 generations) and after that improvements happens very rarely.

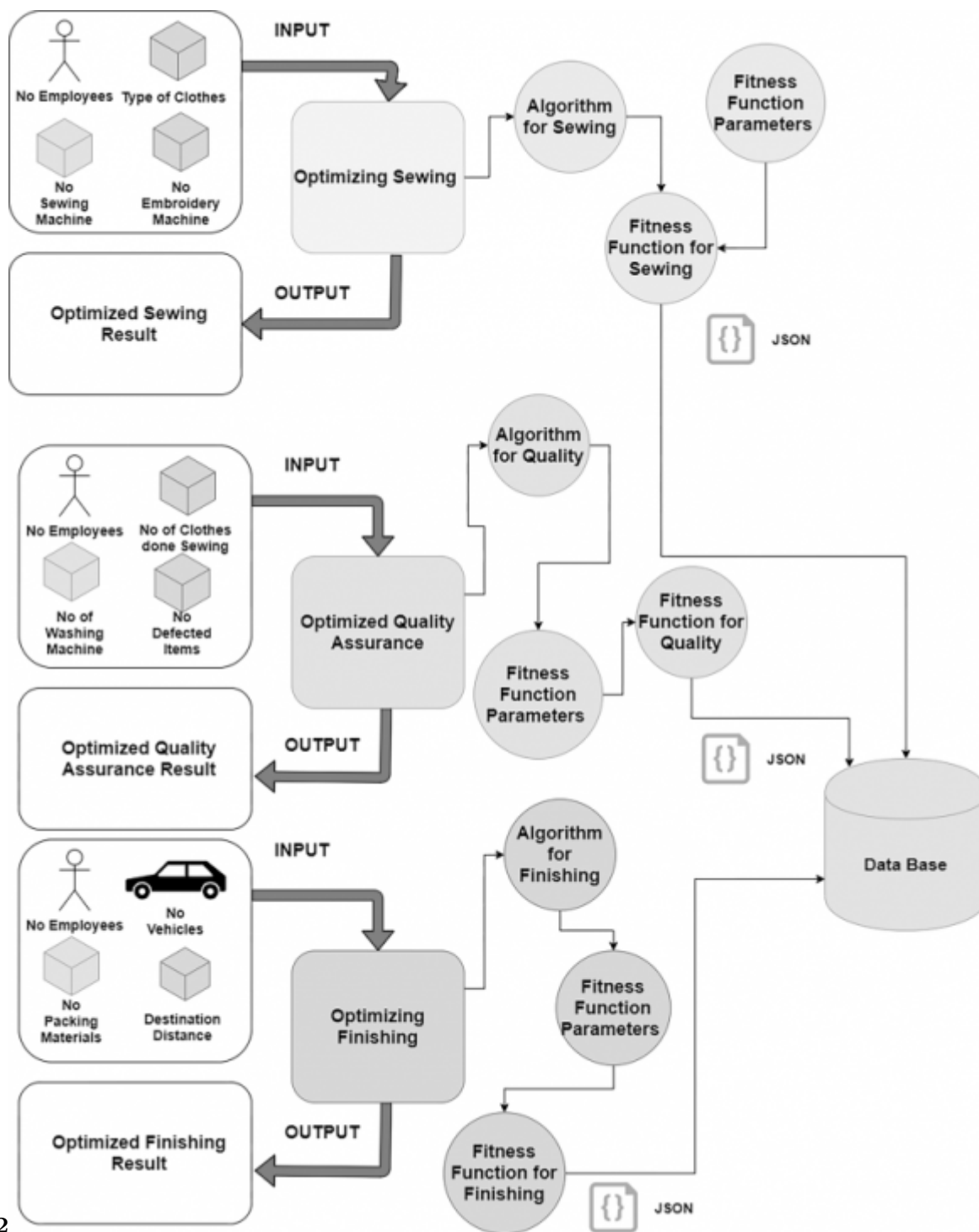
205 14 Final Remarks

206 We have proposed a management platform for optimization and scheduling of textile manufacturing plant,
 207 specifically for following workflows: (1) sewing optimization (2) quality assurance optimization and (3) finishing
 208 optimization. We have used Genetic Algorithms as our heuristic-based optimization approach to get reasonable
 209 results fast. We have consulted and visited textile manufacturing plants, their management and workers to
 210 identify the factors contributing to optimizing workflows. We derived three individual fitness functions for the
 211 three work flows, based on the factors effecting each flow. Our results show that, using our proposed modules,
 212 the manufacturing plant work flows can be optimized and scheduled efficiently and effectively.

213 As the future work, we are planning to extend the optimization research work, specifically to explore
 214 other heuristic approaches such as Iterated Local Search (ILS), and make recommendations on most suitable
 215 optimization approaches.

216 Also, we are planning to work on a prediction module, that predicts the expected time to finish workflow,
 217 specific for different cloth types, instead of using an average time threshold for all cloth types. The prediction

218 module will help us to build a complete platform for optimization and scheduling of textile manufacturing plant.
219 ¹



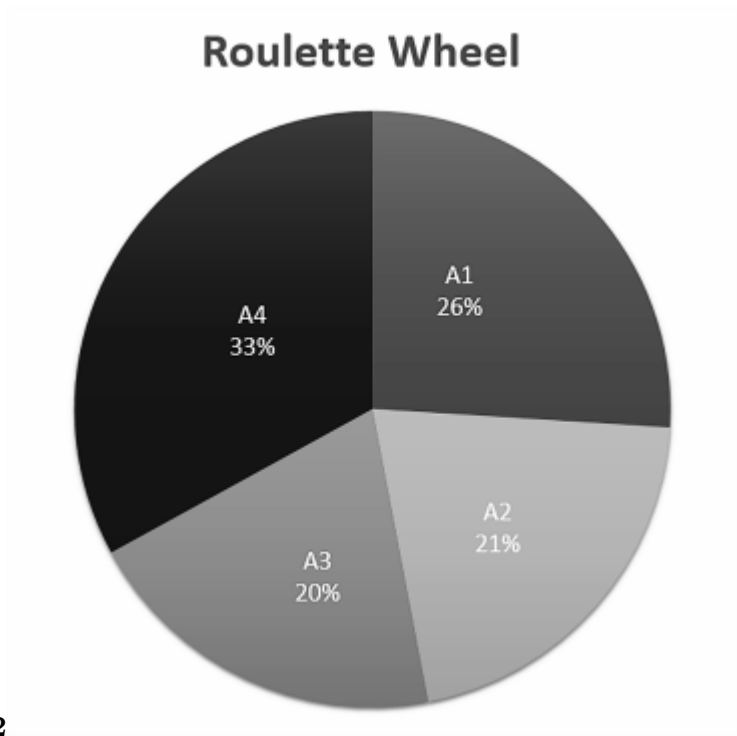
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Figure 1: 12



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Figure 2: Fig. 1 :



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Figure 3: Fig. 2 :

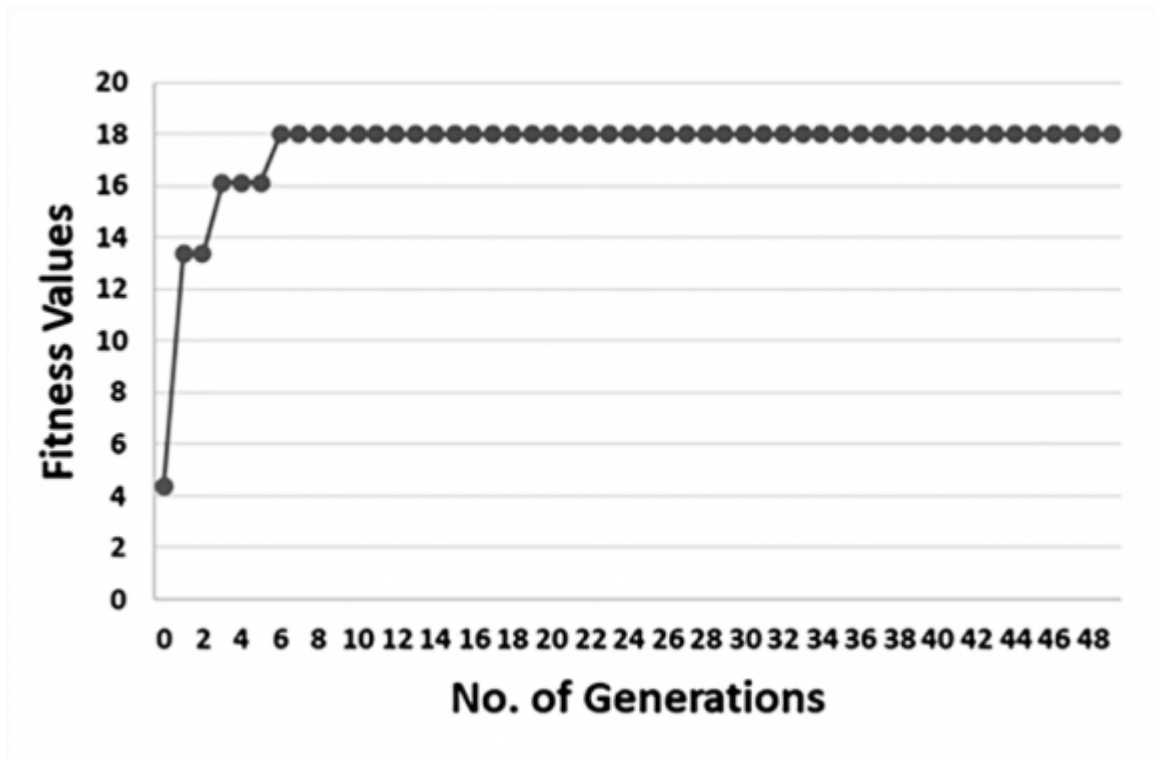


Figure 4:

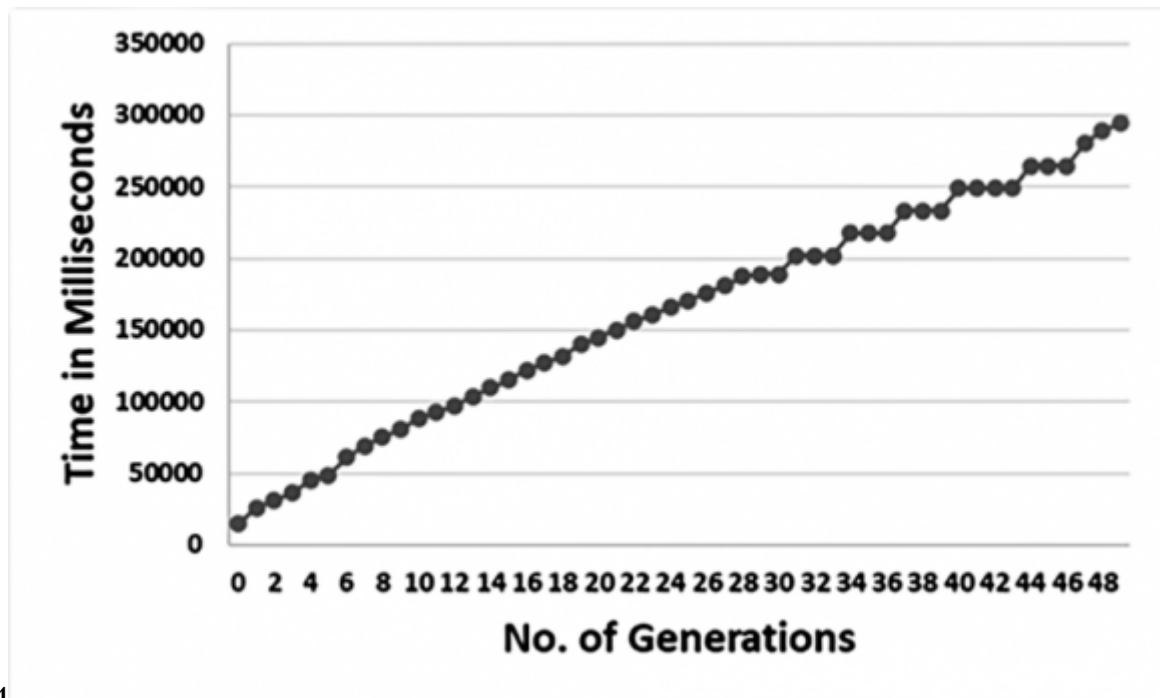
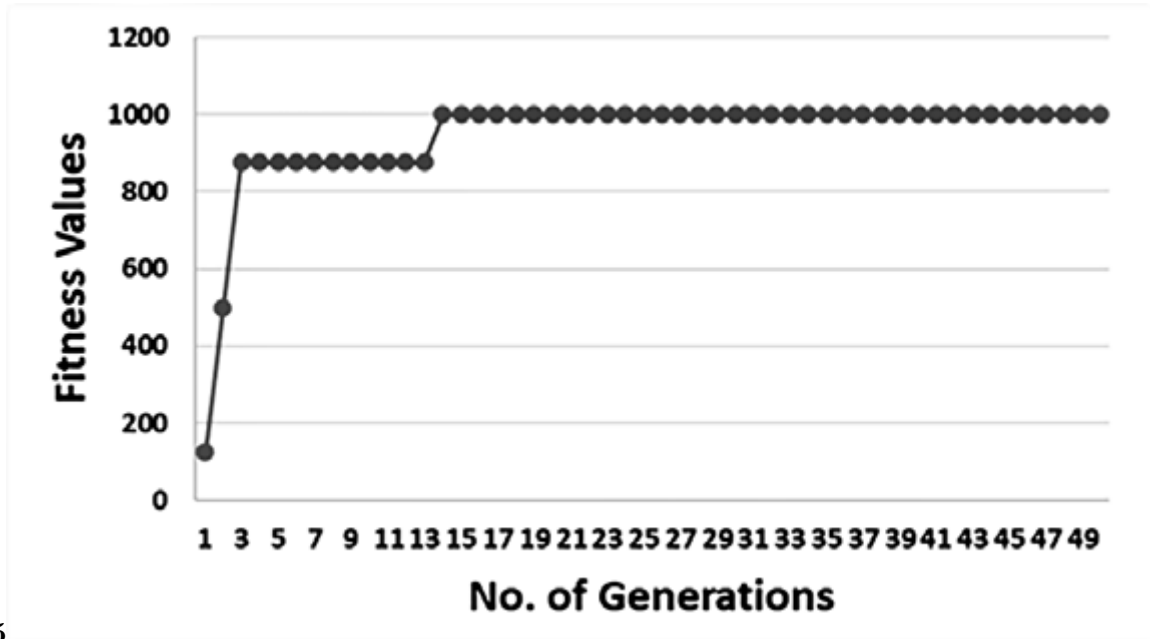


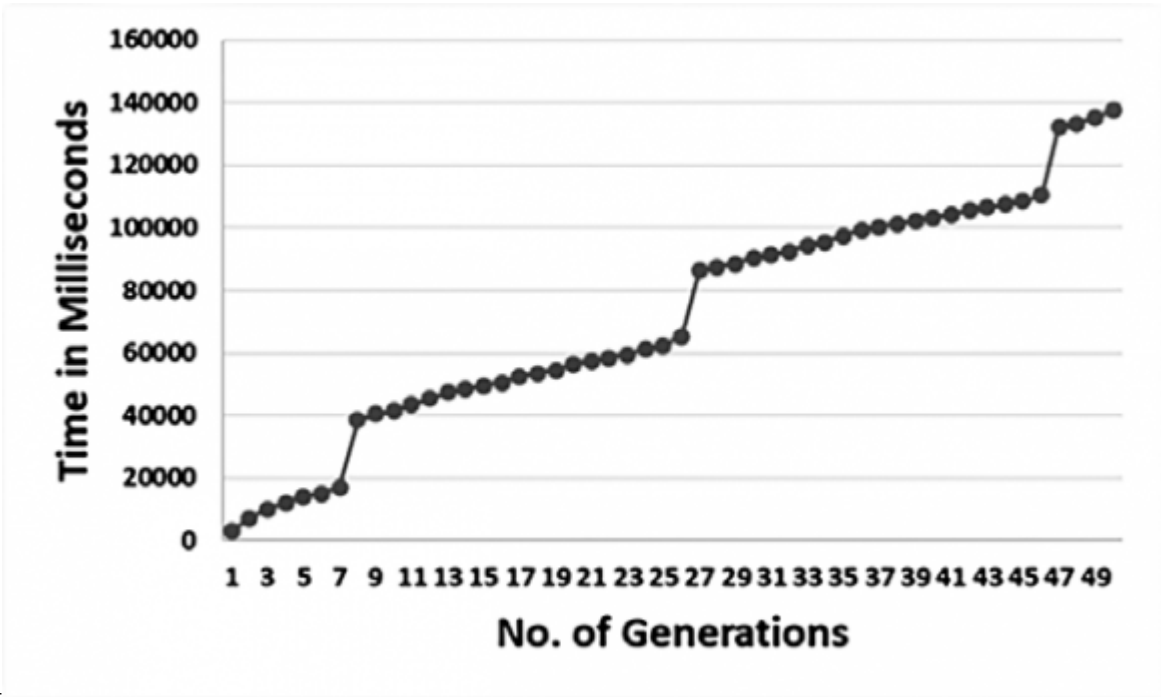
Figure 5: Fig. 4 :

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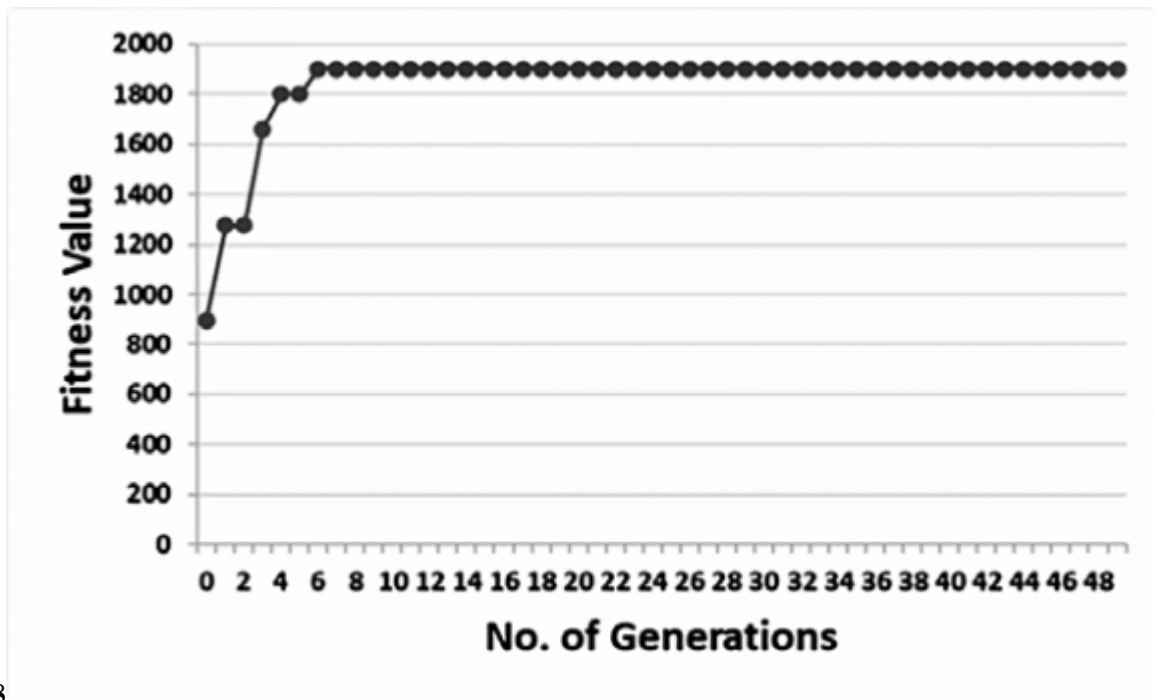
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Figure 6: Fig. 5 :



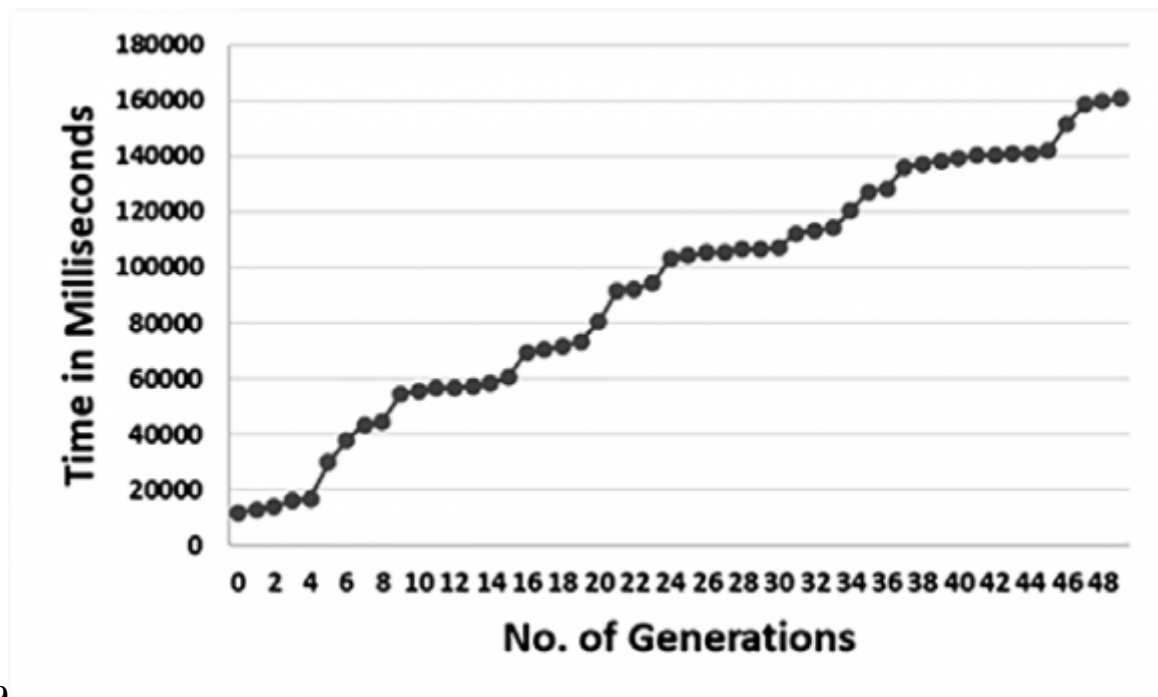
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Figure 7: Fig. 6 : 4 :



78

Figure 8: Fig. 7 :Fig. 8 :



9

Figure 9: Fig. 9 :

1

Chromosome	Fitness	Percentage of Fitness
A1	26	20.8
A2	25	20
A3	42	33.6
A4	32	20.6

Figure 10: Table 1 :

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