# Project Report

On

# **NURSES' SCHEDULING SYSTEM**

(USING GENETIC ALGORITHM)

**Team Members** 

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#### 1. INTRODUCTION

Nurse scheduling problem (NSP), also called the nurse rostering problem (NRP), is the problem of determining an optimal way to assign nurses to shifts, typically with a set of hard constraints which all valid solutions must follow, and a set of simple constraints which define the relative quality of valid solutions. Answers to the nurse scheduling problem can be applied to constrained scheduling problems in other fields.

The nurse scheduling problem has been analysed since 1969 and is known to have NP-hard complexity!

#### 2. GENERAL DESCRIPTIPON

In the nurse scheduling problem, we have to assign shifts and holidays to nurses, find a schedule which respects nurses' wishes and needs of the hospital, e.g., a nurse doesn't want to work a day shift just after a night shift, the hospital needs to have a certain number of nurses at night.

- Morning shift
- Evening shift
- Night shift

Nurses are categorized, depending upon their skill-set, into either of three classes i.e., class A, class B, class C. Here, class A nurses are most skilled and class C nurses least skilled.

#### 3. ALGORITHM DESIGN

We defined our population size p = 100 schedules.

Total Nurse = 10

Class	Number of Nurses
А	5
В	3
С	2

Initially we randomly initialized all the 100 schedules.

#### > Step 1:

To calculate fitness value of each schedule, we assigned some score to each of the constraint. More influential the constraint, the higher is the penalty associated with it.

Sr. No.	Constraint	Fitness Score
1.	There must be at least one nurse allocated to every shift in	300
	every day	
2.	There must be exactly one holiday in a week for every	250
	nurse	
3.	Certain shift-patterns are not allowed	200
	<ol> <li>Morning shift after night shift</li> </ol>	
	ii. ii. Evening shift after night shift	
4.	There must be at least one nurse of class A in every	150
	morning and evening shift	
5.	There must be exactly one nurse of class A in night shift	100
6.	There must be at least one and at most two-night shifts in	50
	a week for every nurse	

#### **>** Step 2:

If least fitness value in the population is zero, then stop executing the algorithm. The corresponding schedule will be the optimized schedule satisfying all the constraints.

#### **>** Step 3:

#### • Selection Operation:

We sorted the schedules in order of increasing fitness value. Top 10% schedules (in our case: 10 schedules) are directly included in the next generation.

#### • Crossover Operation:

To find remaining 90% schedules of next generation two schedules are chosen at random from top 50% schedules and crossover sites are chosen randomly. Then the shifts at these crossover sites are either exchanged (in 90% cases) and mutated to completely new value (in 10 % cases).

#### Return to Step 1.

#### Note:

There are many possible optimal schedules (schedules having fitness value zero). One of them will be displayed when the algorithm completely executed completely.

### 4. Implementation and Result

Objective function to be minimized:

$$F = C_1 * 300 + C_2 * 250 + C_3 * 200 + C_4 * 150 + C_5 * 100 + C_6 * 50$$

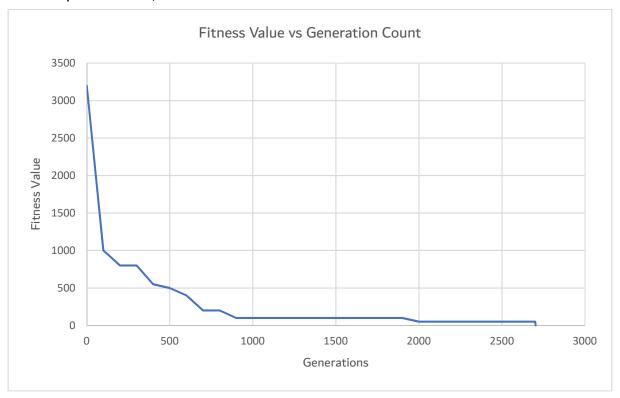
where  $C_1$ ,  $C_2$ ,  $C_3$ ,  $C_4$ ,  $C_5$ , and  $C_6$  denotes number of occurrence of constraints 1, 2, 3, 4, 5 and 6, respectively.

Hence  $C_1$ ,  $C_2$ ,  $C_3$ ,  $C_4$ ,  $C_5$ , and  $C_6 >= 0$ .

After executing the algorithm, one of the possible optimal schedules is displayed as shown in picture.

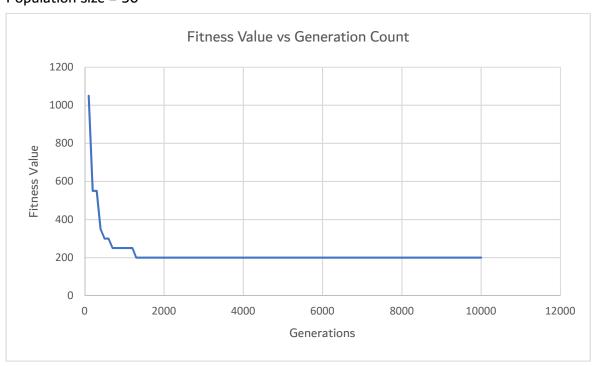
Here M stands for morning shift, E stands for evening Shift, N stands for night shift, H stands for Holiday.

### In our implementation,

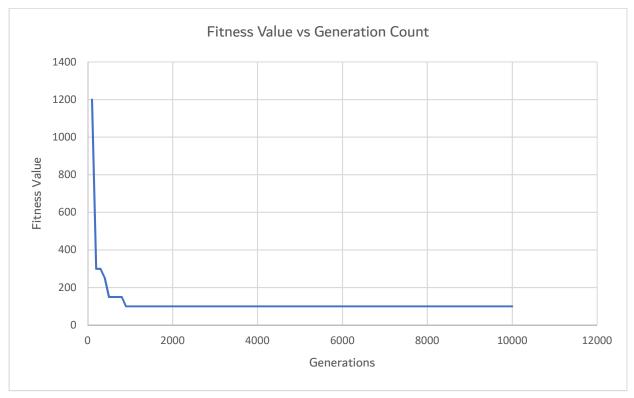


The following are the graphical illustrations of fitness value vs the generation count.

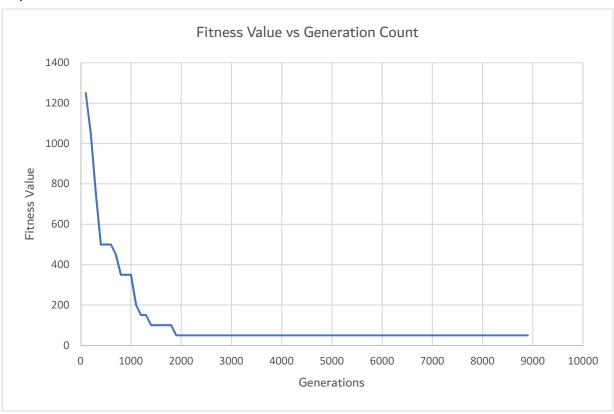
# Population size = 50



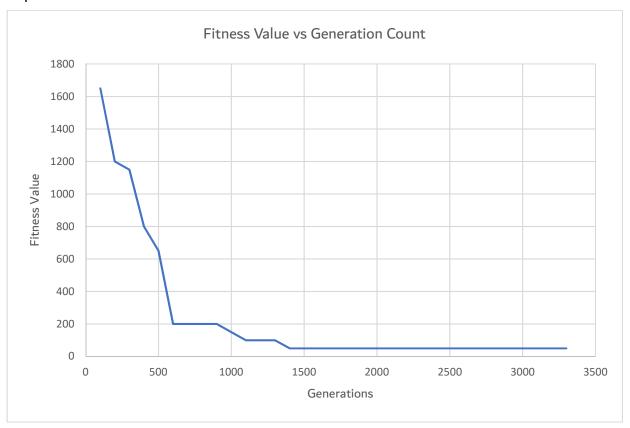
# Population size = 100



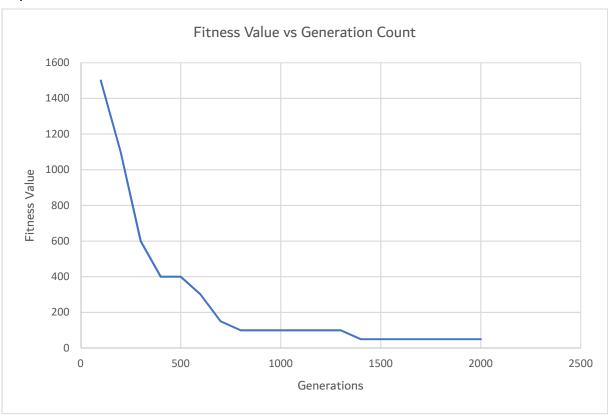
# Population size = 200



# Population size = 500



# Population size = 1000



From the earlier graphs, it can be inferred that as population size grows, the rate of convergence for the algorithm increases. In other words, it takes lesser generations to reach an optimal solution when population size is bigger (Refer the limiting generation for each graph).

#### 5. REFERENCES

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