

**Question 1** There is a factory located at each of the two places  $P$  and  $Q$ . From these locations, a certain commodity is delivered to each of the three depots situated at  $A$ ,  $B$  and  $C$ . The daily requirements of the depot are  $a$ ,  $b$  and  $c$  units of the commodity, respectively while the production capacity of the factories at  $P$  and  $Q$  are  $p$  and  $q$  units, respectively. Further, the cost of transportation from any factory to any depot is given below:

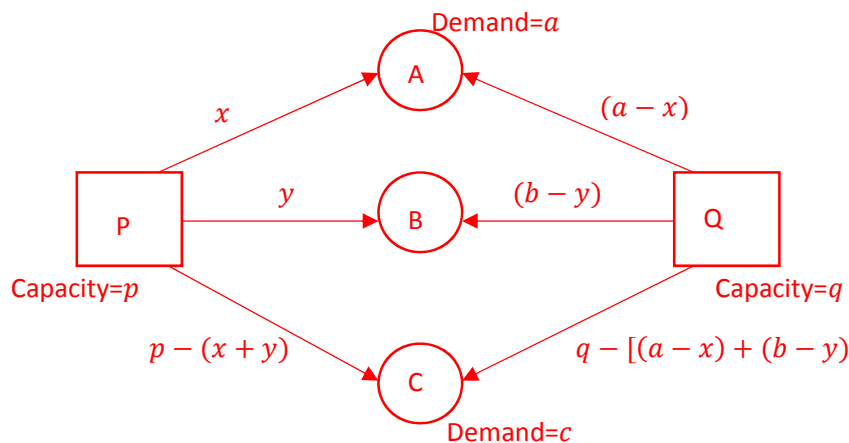
	$A$	$B$	$C$
$P$	$c_{pa}$	$c_{pb}$	$c_{pc}$
$Q$	$c_{qa}$	$c_{qb}$	$c_{qc}$

(a) Formulate the above problem as an optimization problem.

[10]

**Answer**

The given problem can be exhibited diagrammatically as follows:



Here, let the factory at  $P$  transport  $x$  and  $y$  units to  $A$  and  $B$  respectively. Since the requirements are always non-negative quantities.

We have,

$$x \geq 0, y \geq 0 \text{ and } p - (x + y) \geq 0$$

$$\text{Similarly, } (a - x) \geq 0, (b - y) \geq 0 \text{ and } q - [(a - x) + (b - y)] \geq 0$$

The transportation cost from the factory at  $P$  to  $A$ ,  $B$  and  $C$  are  $c_{pa}$ ,  $c_{pb}$  and  $c_{pc}$ , respectively. Similarly, the transportation cost for  $Q$  to  $A$ ,  $B$  and  $C$  are  $c_{qa}$ ,  $c_{qb}$  and  $c_{qc}$ .

Hence, the cost of transportation is,

$$Z = c_{pa}x + c_{pb}y + c_{pc}(p - x - y) + c_{qa}(a - x) + c_{qb}(b - y) + c_{qc}(q - (a - x) - (b - y))$$

Clearly this is to minimize the cost of transportation, hence the optimization problem can be formally stated as:

**OBJECTIVE:**

$$\text{Minimize } Z = (c_{pa} - c_{pc} + c_{qc})x + (c_{pb} - c_{pc} - c_{qb} + c_{qc})y + p \cdot c_{pc} + a \cdot c_{qa} + b \cdot c_{qb} + q \cdot c_{qc} - (a \cdot c_{qc} + b \cdot c_{qc})$$

Subject to

$$(x + y) \leq p$$

$$x - y \geq q(a + b) \text{ and } 0 \leq x \leq a, 0 \leq y \leq b$$

Here,  $x$  and  $y$  are the design parameters only.

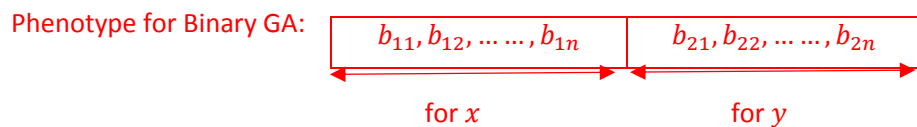
- (b) It is proposed to solve the above optimization problem using Binary coded GA. Decide the genotype for the chromosome structure to do this. **[5]**

**Answer**

As there are only two design parameters, the genotype of the chromosome would look like



For binary encoded GA, let each parameter be represented by  $n$  bits. Since,  $x$  and  $y$  are with the same domain of values, the, phenotype for binary-coded GA will look as follows.

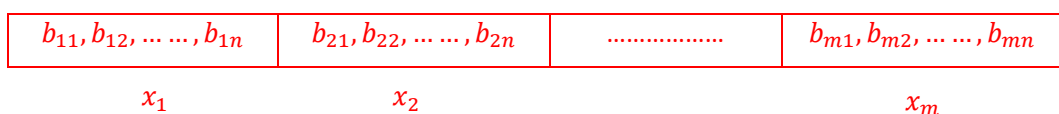


- (c) Suppose the problem needs to be adapted with  $m$  factories and  $n$  depots. What change in the chromosome structure you should devise? **[5]**

**Answer**

We see that for two factory problems, we need two design parameters to be considered. Thus, generalizing the above, for  $m+1$  number of factories, we need to accommodate  $m$  design parameters. Hence, the chromosome structure would be

Genotype for  $m + 1$  factory Binary coded GA



**Question 2** Answer the following:

(a) What is the use of selection operation in Genetic algorithm? [2]

**Answer**

The selection operation in Genetic algorithm is used to select individuals from the current population. The selected individuals will form a mating pool.

(b) Mention **four** criteria, which you should consider to judge the efficiency of a selection strategy? [4]

**Answer**

The four criteria which are usually considered as the measure of efficiency of any selection strategy are:

- (i) Selection pressure: How good individuals are selected.
- (ii) Population diversity: How varieties of individuals are selected.
- (iii) Convergence rate: Selection should lead to a faster search towards optimal solution.
- (iv) Solution quality: Selection should lead to global optimum or near global optimum solution.

(c) Precisely state the **two** major steps in *Tournament selection* strategy. [4]

**Answer**

*Tournament selection strategy* can be stated as follows:

Suppose, the size of the population is  $N$  and our task is to select  $N_p$  ( $N_p \leq N$ ) individuals from the current population. Further, let us consider  $N_u$  ( $N_u \ll N$ ) be the size of each team. The two steps in the Tournament selection is as follows.

Step 1:

Select  $N_u$  members randomly from the current population of size  $N$ .

Step 2:

Select a winner from the  $N_u$  individual from the currently chosen. Put the winner in the mating pool.

As we have to select  $N_p$  individuals, we have to repeat Step 1-2 for  $N_p$  times.

- (d) It is planned to apply *Roulette wheel selection strategy* into *Tournament selection*.  
Give your suggestion, how the same can be realized. [2]

**Answer**

In the second step of *Tournament selection strategy*, there is a task to select a winner from a team. This selection of winner can be carried out applying *Roulette wheel scheme*. (Here only one round to select an individual from the team.)

- (e) How *Tournament selection strategy* is comparable to *Roulette wheel selection strategy*, if they are individually applied in Genetic algorithm? You may give your comparison in the form of a table with reference to **four** efficiency measurement criteria you have mentioned as your answer to *Problem 2(b)*. [8]

**Answer**

A chart showing a summary of comparison is shown in the table given below.

Scheme \ Criteria	Population diversity	Selection pressure	Convergence rate	Solution quality
Roulette Wheel selection	LOW: It favors the individual with higher fitness values.	HIGH: The chance of getting selected individual with higher fitness values.	FAST: As the population diversity is low and selection pressure is high, GA terminates at a faster rate.	Usually terminated with a local optimum solution.
Tournament selection	HIGH-MODERATE: A random selection is made to form team and hence a fair chance is given to all individuals in the population.	LOW-MODERATE: If a random selection is performed to select a winner, then selection pressure is low/moderate.	MODERATE: As population diversity and selection pressure is low to moderate its termination rate is high to moderate.	Near optimal solution is expected.

**Question 3** Answer the following:

- (a) Obtain the offspring chromosome(s), it will produce from the reproduction of three parent chromosomes  $P_1$ ,  $P_2$  and  $P_3$  following the “Three parent crossover” technique? [4]

$P_1$ : 

1	1	0	1	0	0	0	1
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$P_2$ : 

0	1	1	0	1	0	0	1
---	---	---	---	---	---	---	---

$P_3$ : 

0	0	1	1	0	1	1	0
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**Answer**

Three parent crossover technique: Each bit of  $P_1$  is compared with the bit in the same position of  $P_2$ . If the two bits are same, then it is copied to offspring else the bit (at the same position) of  $P_3$  is copied to offspring. Thus, offspring produced is

0	1	1	1	0	0	0	1
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- (b) Two parent chromosomes in Order GA are given as under:

$P_1$ : 

1	2	3	4	5	6	7	8	9
---	---	---	---	---	---	---	---	---

$P_2$ : 

1	2	3	4	5	6	7	8	9
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Obtain the two offspring that can be obtained following the “Partially mapped crossover” technique. [4]

**Answer**

Let  $c_1$  and  $c_2$  are the two offspring

- (i) Select any two points as crossover points on the parent chromosomes.
- (ii) A schema in  $P_1$  (or  $P_2$ ) is directly copied in the same location and order in  $c_1$  (or  $c_2$ ).
- (iii) The rest of the gene values in  $c_1$  (or  $c_2$ ) are obtained from  $P_2$  (or  $P_1$ ) in the same order as they appear.

Thus, two offspring chromosomes  $c_1$  and  $c_2$  will be obtained as

1	2	3	4	5	6	7	8	9
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1	2	3	4	5	6	7	8	9
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(c) “Linear crossover” technique in Real-coded GA takes the following form:

$$c_i = \alpha_i \cdot p_1 + \beta_i \cdot p_2$$

Where  $p_1$  and  $p_2$  are any two values and  $\alpha_i, \beta_i$  are any two values chosen by the user.

Explain for the following two parent chromosomes, how two offspring chromosomes can be produced. [4]

$P_1$ :

5	10
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$P_2$ :

6	9
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### Answer

Let us arbitrarily decide three set of values for  $\langle \alpha_i, \beta_i \rangle$  for three offspring chromosomes as follows:

$$c_{11} = 0.5(p_{11} + p_{21}); c_{12} = 0.5(p_{12} + p_{22});$$

$$c_{21} = 1.5p_{11} - 0.5p_{21}; c_{22} = 1.5p_{21} - 0.5p_{22}$$

(d) Explain the working of “Flipping” as mutation operator in Binary-coded GA.

[4]

### Answer

An offspring can be mutated to a “mutated offspring” using the help of a “mutation chromosome.” The mutation chromosome has a bit pattern with fewer 1’s at random location. The number of 1’s in “mutation chromosome” is decided by mutation

operator  $\mu$ , which denotes the probability that a bit is 1 at a random location in “mutation chromosome”.

A bit in the offspring chromosome is flipped 0 to 1 and vice versa if a bit in the same position in mutation chromosome is 1 only.

For example,

<b>Offspring:</b>	0	1	1	0	0	1	1	1
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<b>Mutation chromosome:</b>	0	1	0	0	0	1	0	0
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<b>Mutated chromosome:</b>	0	0	1	0	0	0	1	1
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(e) Consider the following is a chromosome in Order-GA encoding scheme.

<b>Offspring:</b>	B	H	F	G	C	E	A	D
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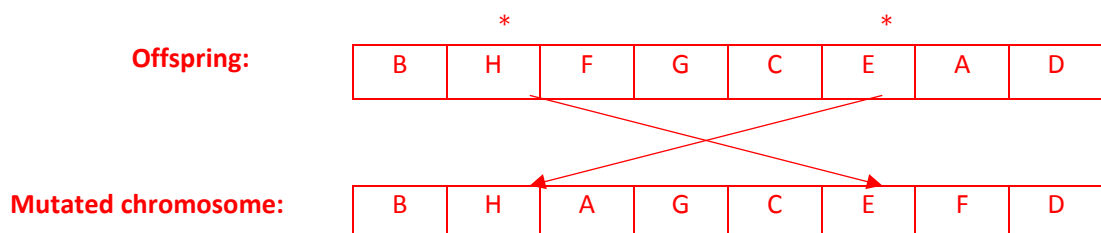
Explain how such a chromosome can be mutated. Consider the least change in chromosomes between the offspring and the mutated offspring. [4]

### Answer

Let us consider the change in gene values between two positions as the least change possible.

- (i) First select two random locations in the offspring chromosome.
- (ii) Swap the gene values between the selected locations.

For example, two random locations are shown as \*



**Question 4** Each of the following question includes a statement and possibly the statements are **wrong**. You have to **rewrite** the sentence in their correct form. **[10 X 2=20]**

- (a) Vector Evaluated Genetic Algorithm (VEGA) is a multi-objective evolutionary algorithm (MOEA), which is an a priori technique and Pareto based.

**Answer**

Vector Evaluated Genetic Algorithm (VEGA) is a multi-objective evolutionary algorithm (MOEA), which is an a posteriori technique and non-Pareto-based.

- (b) A Pareto front is also a Pareto optimal front but the reverse is not true.

**Answer**

A Pareto optimal front is a Pareto front but the reverse is not true.

- (c) Both NSGA and NSGA-II follow the “Crowding Tournament” selection strategy to create mating pool.

**Answer**

NSGA follows “Stochastic remainder selection” whereas NSGA-II follows “Crowding Tournament selection” to create mating pool.

- (d) The length of chromosomes in encoding scheme while solving a multi-objective optimization problem (MOOP) is proportional to the number of objective functions in the MOOP.

**Answer**

The chromosome length in MOOP is independent of the number of chromosomes.

- (e) If  $c_1$  and  $c_2$  are two offspring chromosomes, then according to NPGA,  $c_1$  will be preferable to be selected for mating pool if  $c_1$ 's niche count is higher than that of  $c_2$ .

**Answer**

If  $c_1$  and  $c_2$  are two offspring chromosomes, then according to NPGA,  $c_1$  will be preferable to be selected for mating pool if  $c_1$ 's niche count is lower than that of  $c_2$ .



- (f) According to MOGA, the rank of a solution is defined as the number of solutions by which it is dominated.

**Answer**

According to MOGA, the rank of a solution  $x_i$  is defined as

$$\text{rank}(x_i) = 1 + p_i$$

Where  $p_i$  is the number of solutions which dominates  $x_i$ .

- (g) The a priori high level information that is required in “Lexicographic ordering” is the scalar weights of each objective function.

**Answer**

The a priori high level information that is required in “Lexicographic ordering” is the descending ordering of the rank of the importance of objective functions.

- (h) Crowded comparison operator ( $<_c$ ) (as defined in NSGA-II) to select between  $x$  and  $y$  is defined as  $\text{rank}(x) < \text{rank}(y)$  or  $\text{rank}(x) > \text{rank}(y)$  and  $x_d > y_d$ .

**Answer**

Crowded comparison operator ( $<_c$ ) (as defined in NSGA-II) to select between  $x$  and  $y$  is defined as  $\text{rank}(x) < \text{rank}(y)$  or  $\text{rank}(x) = \text{rank}(y)$  and  $x_d < y_d$ .

- (i) A solution  $x_i$  is said to dominate another solution  $x_j$  if
1.  $x_j$  is worse than  $x_i$  and
  2.  $x_i$  is strictly better than  $x_j$ .

**Answer**

A solution  $x_i$  is said to dominate another solution  $x_j$  if both conditions (1) and (2) are true.

1.  $x_i$  is no worse than  $x_j$ .
  2.  $x_i$  is strictly better than  $x_j$  in at least one objective.
- (j) MOGA and NSGA follow their own steps to assign fitness values to all solutions in the current population whereas NPGA and NSGA-II do not require any fitness value calculation.

**Answer**

The statement is correct.

**Question 5** Answer the following:

- (a) What is niche count? What it does signify? Give an idea how niche count of a solution in a population can be calculated. [1+1+4=6]

**Answer**

Niche count: Niche count of a solution is a measure of how crowded the solution is by its neighboring solutions.

Physical meaning of Niche count: A solution with a higher niche count implies that the solution is a good representative of many other solutions with almost similar result. On the other hand, a solution with a lower niche count, if it is selected for mating pool then it ensures population diversity.

Measuring Niche count of a solution in a population: Let  $x_i$  be a solution whose Niche count is to be measured. Further, let  $x_j$  be any solution ( $x_i \neq x_j$ ) in the given population of size  $N$  with solutions  $x_1, x_2, \dots, \dots, x_N$ .

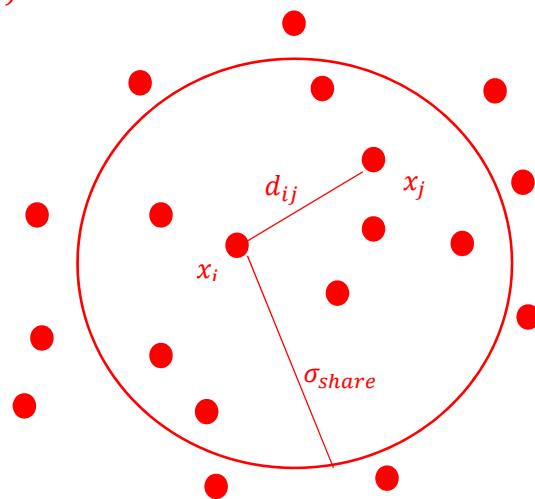
Let  $d_{ij}$  denotes the phenotype distance between  $x_i$  and  $x_j$  and  $\sigma_{share}$  is the maximum phenotype distance allowed between any two individuals to become members of a niche.

The, we calculate a sharing value  $sh(d_{ij})$ , which is as follows:

$$sh(d_{ij}) = \begin{cases} 1 - \left(\frac{d_{ij}}{\sigma_{share}}\right)^2, & \text{if } d_{ij} < \sigma_{share} \\ 0 & \text{otherwise} \end{cases}$$

Finally, Niche count of the solution  $x_i$  will be obtained as

$$Niche\ Count(x_i) = \sum_{\substack{j=1 \\ j \neq i}}^N sh(d_{ij})$$



- (b) Explain the **concept** of non-dominated sorting procedure as proposed in NSGA.  
What is the use of non-dominated sorting output in NSGA? [3+5=8]

**Answer**

Concept of non-dominated sorting:

Suppose,  $X$  denotes a set of solutions of a multi-objective optimization problem and,  $|X| = N$ .

Given  $X$ , we can find all non-dominated solution's  $X_1 \subseteq X$ . All solutions in  $X_1$  lies in a front is called first non-dominated front (see Figure).

After removing  $X_1$  from  $X$ , and repeating the same procedure we get  $X_2$  such  $X_2 \subseteq (X - X_1)$ .

The above procedure can be continued until all solutions are sorted into front. Thus, given  $X$ , we obtain  $X_1, X_2, \dots, X_N$  which are called non-dominated sorting fronts in  $X$ .

Fitness assignments with non-dominated sorting fronts

Let  $X_i$  be the  $i$ -th non-dominated sorting fronts ( $i = 1, 2, \dots, n$ ). The steps to assign fitness values to all solutions in  $X_i$  are as follows.

1. Assign a dummy fitness value say  $f_{X_i}$  to each solution belongs to  $X_i$ . The dummy fitness value is a very large number and typically is proportional to the number of solutions in all fronts up to the current front. That is,  $f_{X_i} = |X_1 \cup X_2 \cup \dots \cup X_i|$ , a proportional constant is  $\geq 1$ .
2. For each solution  $x_j \in X_i$ , it then calculates Niche count,  $\gamma(x_j)$ .
3. The fitness value of  $x_j \in X_i$ , is then calculated as

$$f_j = \frac{f_{X_i}}{\gamma(x_j)}$$

The above procedure is repeated to all solutions belongs to all fronts in  $X$ .

- (c) The non-dominated sorting GA (NSGA-II) procedure for finding multiple Pareto optimal solutions in a multi-objective optimization problem has the following three features:
- (i) It uses an elitist principle
  - (ii) It uses an explicit diversity preserving mechanism, and
  - (iii) It emphasizes the non-dominated solutions.

Briefly explain how NSGA-II accomplishes the above mentioned features. You should mention the concept only instead of procedure for each. **[2+2+2=6]**

**Answer**

- (i) NSGA-II adopts elitist principle

A set of solutions which are very close to the optimal solutions are called elite. NSGA-II computes elite solutions with two steps:

- A. Reproduce  $N$  offspring using  $N$  parents in the current population.
- B. Obtain improved  $N$  solutions from the set of  $2N$  solutions, which are non-dominated front by front.

- (ii) NSGA-II follows explicit diversity preserving mechanism

NSGA-II proposes Crowding Tournament selection to ensure population diversity. To do this, it calculates "Crowding distance" for each solution. The crowding distance measure implies how a solution is crowded by its nearest neighbor. NSGA-II prefers a solution having less rank or crowding distance to be selected for mating.

- (iii) NSGA-II emphasizes non-dominated sorting solutions

NSGA-II procedure in its each iteration obtains non-dominated sorting fronts. It then prefers the solutions in *first, second, ... .., i - th* fronts until it obtains a mating pool of *size*  $< N$ . To make the size of the mating pool exactly  $N$ , it selects the remaining solutions from the  $(i + 1)th$  front using Crowding Tournament selection procedure.