

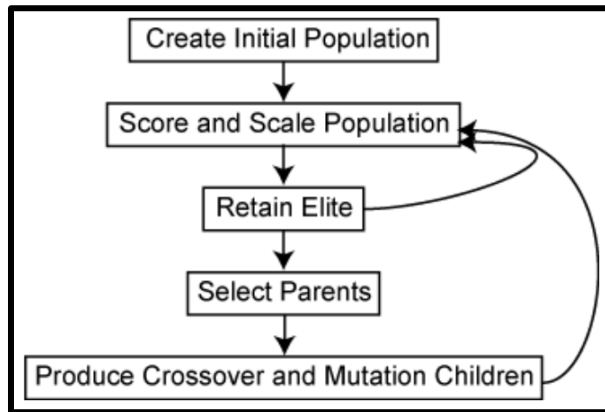
1. Problem Statement

Use a binary-coded GA to minimize the function $f(x_1, x_2) = x_1 + x_2 - 2x_1^2 - x_2^2 + x_1x_2$, in the range of $0.0 \leq x_1, x_2 \leq 0.5$. Using a random population of size $N=6$ and assume 5 bits for each variable.

2. Parameters used

Number of Variables	2
Number of Bits for each variable	5
Total Number of Bits	10
x_{min}, x_{max}	0, 0.5
Minimization Problem converted to maximization using	$F(x) = -f(x)$
Population Size (N)	6
Maximum Number of Generation	1,00,000
Selection technique	Tournament and Elitism
Elitism Selection size	0 (no elite is retained)
Tournament Selection size	3
Crossover Probability (Pc)	0.9
Mutation Probability (Pm)	0.1

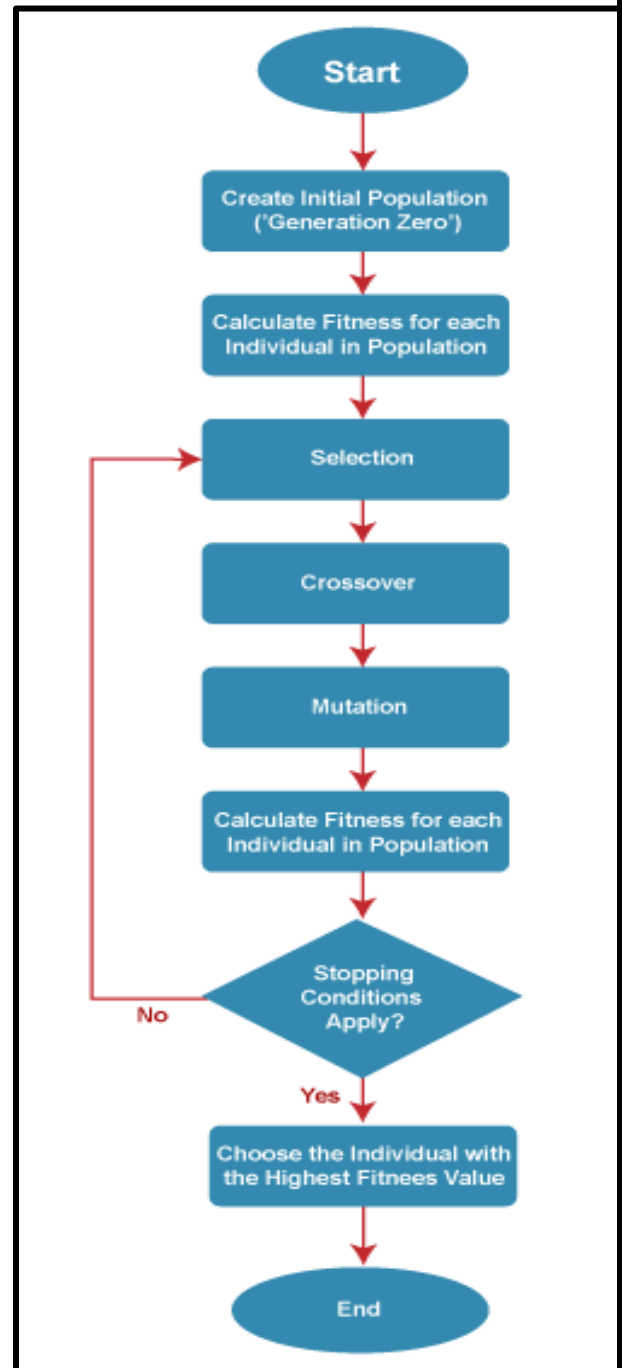
3. Procedure for Binary Coded Genetic Algorithm



- Initialize a population of size $N=6$ with 6 strings of length 10 bits at random (5 bits for each variable).
- Then decoded values all strings with first five bit for 1st variable (x_1) and next five bit for 2nd variable(x_2).
- The real values of x_1 and x_2 are then computed using the formula below.

$$x_i = x_i^{(L)} + \frac{x_i^{(U)} - x_i^{(L)}}{2^{\ell_i} - 1} DV(s^i),$$

- The fitness value is then determined. The fitness function $F(x) = -f$ is used to turn the minimization problem into a maximization problem.
- Then, by tournament selection of size 3, reproduction is carried out, and winners are obtained.
- Then, using the Elitism approach, the best answers can be maintained the same (for our problem zero elite is retained).
- The Single Point Crossover technique is then used with a probability of $p_c = 0.9$ for the remaining four solutions.
- The population is then subjected to mutations with a mutation probability of $p_m = 0.1$.
- After then, the procedure is repeated for 100000 generations.



4. Results

Initial population :

1. 1010011101
2. 1001100000
3. 0101000011
4. 1010100000
5. 0001100100
6. 0010100011

Final Population after 100000 generation :

1. 0000000000
2. 0000000000
3. 0000000000
4. 0000000000
5. 0000000000
6. 0000000000

The final value of all population x_1 , x_2 , $f(x_1, x_2)$ and fitness values are

Population #	x_1	x_2	$f(x_1, x_2)$	Fitness value
1	0	0	0	0
2	0	0	0	0
3	0	0	0	0
4	0	0	0	0
5	0	0	0	0
6	0	0	0	0

The minimum of given problem lies at $x_1 = 0$, $x_2 = 0$, and minimum function value is $f(x_1, x_2) = 0$.

There is another minimum solution of this problem which lies at $x_1 = 0.5$, $x_2 = 0$, and $f(x_1, x_2) = 0$.

CODE is attached below.

```

1: #include<bits/stdc++.h>
2:
3: using namespace std;
4:
5:
6: double f(double , double );
7:
8:
9: int main(){
10:     int n = 6, l1 = 5, l = 10, no_of_variable = 2;
11:     double xmin = 0, xmax = 0.5;
12:     int e = 0; // no of element for elitism (keep even number if n is
even, and odd if n is odd)
13:     int toursize = 3; // no. of element select for tournament
14:     double pc = 0.9, pm = 0.1; // crossover and mutation probability
15:     int max_generation = 100000;
16:
17:
18:     vector <vector <int> > POPULATION( n+1, vector<int> (l+1));
19:     vector <vector <int> > DECODE_VALUE( n+1, vector<int>
(no_of_variable+1));
20:     vector <vector <double> > X( n+1, vector<double>
(no_of_variable+1));
21:     vector <vector <double> > FUNCTION_VALUE( n+1, vector<double>
(2));
22:     vector <vector <double> > FITNESS( n+1, vector<double> (2));
23:     vector <vector <int> > POOL( n+1, vector<int> (l+1));
24:     vector <vector <int> > CROSSOVER( n+1, vector<int> (l+1));
25:     vector <vector <int> > MUTATION( n+1, vector<int> (l+1));
26:     vector <vector <int> > NEW_GENERATION( n+1, vector<int> (l+1));
27:
28:
29:     int i, j, k, generation;
30:     double x1,x2, maxfit, prob;
31:     vector <double> temp( n+1 );
32:     vector <double> best_ele( e+1 );
33:     vector <int> best_position( e+1 );
34:     int sum=0, p , r, maxpos;
35:     double avgfit;
36:
37:
38:     ofstream OUT;
39:     OUT.open("OUTPUT.txt");
40:
41:     // generating random population
42:     cout<<"Initial population : \n";

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43:     OUT<<"Initial population : \n";
44:     srand(time(0));
45:     for(i=1;i<=n;i++){
46:         cout<<i<<"  ";
47:         OUT<<i<<"  ";
48:         for(j=1;j<=l;j++){
49:             POPULATION[i][j] = (rand()%(2));
50:             cout<<POPULATION[i][j];
51:             OUT<<POPULATION[i][j];
52:         }
53:         cout<<endl;
54:         OUT<<endl;
55:     }
56:     cout<<endl<<endl;
57:     OUT<<endl<<endl;
58:
59:     for(i=1;i<=n;i++){
60:         for(j=1;j<=l;j++){
61:             NEW_GENERATION[i][j] = POPULATION[i][j];
62:         }
63:     }
64:
65:
66:     for(generation=1;generation<=max_generation;generation++){
67:
68:         for(i=1;i<=n;i++){
69:             for(j=1;j<=l;j++){
70:                 POPULATION[i][j] = NEW_GENERATION[i][j];
71:             }
72:         }
73:
74:         // finding decord value and x value
75:         for(i=1;i<=n;i++){
76:             sum=0;
77:             for(j=1;j<=l1;j++){
78:                 sum=sum+pow(2,5-j)*POPULATION[i][j];
79:             }
80:             DECODE_VALUE[i][1] = sum;
81:             sum=0;
82:             for(j=l1+1;j<=l;j++){
83:                 sum=sum+pow(2,5+l1-j)*POPULATION[i][j];
84:             }
85:             DECODE_VALUE[i][2] = sum;
86:         }
87:         for(i=1;i<=n;i++){
88:

```

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89:         X[i][1] = xmin+(xmax-xmin)*DECODE_VALUE[i][1]/(pow(2,l1)-
1);
90:         X[i][2] = xmin+(xmax-xmin)*DECODE_VALUE[i][2]/(pow(2,l1)-
1);
91:     }
92:
93:     // calculating function value to find fitness value
94:     for(i=1;i<=n;i++){
95:
96:         x1 = X[i][1];
97:         x2 = X[i][2];
98:         FUNCTION_VALUE[i][1] = f(x1,x2);
99:     }
100:    for(i=1;i<=n;i++){
101:
102:        FITNESS[i][1] = -1*FUNCTION_VALUE[i][1]; // to conver
minimization problem into maximization
103:    }
104:
105:    // Selection of best e solution by Elitism and other by
tournament selection for mating pool
106:    for(i=1;i<=n;i++){
107:        temp[i] = FITNESS[i][1];
108:    }
109:    sort(temp.begin()+1,temp.end());
110:
111:    // for elitism of e no. of element
112:    for(i=1;i<=e;i++){
113:        best_ele[i] = temp[n-i+1];
114:    }
115:    for(j=1;j<=e;j++){
116:        for(i=1;i<=n;i++){
117:            if(FITNESS[i][1]==best_ele[j]){
118:                best_position[j] = i;
119:            }
120:        }
121:    }
122:
123:    // tranfering elitism solution to new generation
124:    for(i=1;i<=e;i++){
125:        p=best_position[i];
126:        for(j=1;j<=l;j++){
127:
128:            NEW_GENERATION[e][j] = POPULATION[p][j];
129:        }
130:    }

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131:
132: //tournament selection of n-e no. of element
133: for(i=e+1;i<=n;i++){
134:     r = 1+(rand()%(n)); // random number between 1 to n
135:     maxfit=FITNESS[r][1];
136:     maxpos=r;
137:     for(j=2;j<=toursize;j++){
138:         r = 1+(rand()%(n));
139:         if(FITNESS[r][1]>maxfit){
140:             maxfit=FITNESS[r][1];
141:             maxpos=r;
142:         }
143:     }
144:
145:     for(j=1;j<=l;j++){
146:         POOL[i][j] = POPULATION[i][maxpos];
147:     }
148: }
149:
150: // CROSS-OVER
151: for(i=e+1;i<=n;i=i+2){
152:     prob=fabs(sin(rand())); // random number between 0 and 1
153:     if(prob<=pc){
154:         r = 1+(rand()%(l-1));
155:         for(j=1;j<=r;j++){
156:             CROSSOVER[i][j] = POOL[i][j];
157:             CROSSOVER[i+1][j] = POOL[i+1][j];
158:         }
159:         for(j=r+1;j<=l;j++){
160:             CROSSOVER[i][j] = POOL[i+1][j];
161:             CROSSOVER[i+1][j] = POOL[i][j];
162:         }
163:     }
164:     else{
165:         for(j=1;j<=l;j++){
166:             CROSSOVER[i][j] = POOL[i][j];
167:             CROSSOVER[i+1][j] = POOL[i+1][j];
168:         }
169:     }
170: }
171:
172: vector <vector <double> > X( n+1, vector<double> (no_of_variable+1));
173: for(i=e+1;i<=n;i=i+2){
174:     for(j=1;j<=l;j++){
175:         MUTATION[i][j] = CROSSOVER[i][j];
176:     }

```

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177:     }
178:     for(i=e+1;i<=n;i=i+2){
179:         prob=fabs(sin(rand())); // random number between 0 and 1
180:         if(prob<=pm){
181:             r = 1+(rand()%(1-1));
182:             if(CROSSOVER[i][r]==0){
183:                 MUTATION[i][r]=1;
184:             }
185:             else{
186:                 MUTATION[i][r]=0;
187:             }
188:         }
189:     }
190:
191:     for(i=e+1;i<=n;i++){
192:         for(j=1;j<=l;j++){
193:             NEW_GENERATION[i][j] = MUTATION[i][j];
194:         }
195:     }
196: }
197:
198: cout<<"Final Population after "<<max_generation<<" generation :
199: \n";
200: OUT<<"Final Population after "<<max_generation<<" generation :
201: \n";
202: for(i=1;i<=n;i++){
203:     cout<<i<<" ";
204:     OUT<<i<<" ";
205:     for(j=1;j<=l;j++){
206:         cout<<POPULATION[i][j];
207:         OUT<<POPULATION[i][j];
208:     }
209:     cout<<endl;
210:     OUT<<endl;
211: }
212:
213: cout<<"Position and Function value of all population : \n";
214: cout<<"Population#\tx1\tx2\tf(x1,x2)\n";
215: cout<<setprecision(3);
216: OUT<<"Position and Function value of all population : \n";
217: OUT<<"Population#\tx1\tx2\tf(x1,x2)\n";
218: OUT<<setprecision(3);
219: for(i=1;i<=n;i++){
220:     x1 = X[i][1];

```



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221:         x2 = X[i][2];
222:         FUNCTION_VALUE[i][1] = f(x1,x2);
223:
224:         cout<<i<<"\t\t"<<x1<<"\t"<<x2<<"\t"<<FUNCTION_VALUE[i][1]<<endl;
225:         OUT<<i<<"\t\t"<<x1<<"\t"<<x2<<"\t"<<FUNCTION_VALUE[i][1]<<endl;
226:     }
227:     cout<<endl<<endl;
228:     OUT<<endl<<endl;
229:     p = best_position[1];
230:     cout<<"The minimum of given problem lies at x1 = "<<X[p][1]<<",
231:     x2 = "<<X[p][1]<<
232:     "<<FUNCTION_VALUE[p][1]<<". ";
233:     OUT<<"The minimum of given problem lies at x1 = "<<X[p][1]<<", x2
234:     = "<<X[p][1]<<
235:     "<<FUNCTION_VALUE[p][1]<<". ";
236:
237:     return 0;
238: }
239:
240: double f(double x1, double x2){
241:     return x1+x2-2*x1*x1-x2*x2+x1*x2;
242: }

```