

דו"ח מעבדה 2 – אלגוריתמים גנטיים:

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חלק א' – חיפוש לוקאלי:

1. חישוב ממוצע ה-*fitness* של האוכלוסייה בכל דור וסטיית התקן ממנו:

```
314 static void calc_ga_stat(ga_vector &population)
315 {
316     unsigned int sumFitness = 0;
317     for (int i = 0; i < GA_POPSIZE; i++) // sum all
318     {
319         sumFitness += population[i].fitness;
320     }
321     double mean = sumFitness / GA_POPSIZE; // uniform mean
322     double std = 0;
323     for (int i = 0; i < GA_POPSIZE; i++)
324     {
325         double fitness = population[i].fitness;
326         std += pow(fitness - mean, 2);
327     }
328     std = sqrt(std); // simple std as we recognize with
329     for (int i = 0; i < GA_POPSIZE; i++)
330     {
331         population[i].mean = mean;
332         population[i].std = std;
333     }
334 }
```

דיווח ממוצע ה-*fitness* של האוכלוסייה בכל דור וסטיית התקן ממנו:

```
Best: Hello Wprld! (1) -> [mean=11,std=1035.21] -> CPU Clock Ticks=592,Elapsed Run Time=0 -> Generation=34
Best: Hello World! (0) -> [mean=10,std=984.898] -> CPU Clock Ticks=607,Elapsed Run Time=1 -> Generation=35
End-up after: 623.00 ticks and 1.00 seconds
```

2. חישוב זמן ריצה עד להתכנסות לפי *clock ticks and elapsed time* בכל דור:

```
482 static inline void print_best(ga_vector &gav,int generation)
483 {
484     string temp;
485     for (int i = 0; i < GA_TARGET.size(); i++)
486     {
487         temp += gav[0].str[i];
488     }
489     cout << " Best: " << temp << " (" << gav[0].fitness << ") " << " -> [mean=" << gav[0].mean << ",std=" << gav[0].std << "]"
490     << " -> " << "CPU Clock Ticks=" << gav[0].cpuTicks << ",Elapsed Run Time=" << gav[0].elapsedTime << " -> " << "Generation=" << generation << endl;
491 }
492 static inline void swap(ga_vector *&population, ga_vector *&buffer)
493 {
494     ga_vector *temp = population;
495     population = buffer;
496     buffer = temp;
497 }
```

דיווח של זמן הריצה עד להתכנסות לפי *clock ticks and elapsed time* בכל דור:

```
Best: Hello Wprld! (1) -> [mean=14,std=1014.18] -> CPU Clock Ticks=532,Elapsed Run Time=0 -> Generation=31
Best: Hello World! (0) -> [mean=12,std=956.333] -> CPU Clock Ticks=548,Elapsed Run Time=0 -> Generation=32
End-up after: 563.00 ticks and 0.00 seconds
```

3. היוריסטיקת "בול פגיעה" כחלק מהפונקציה *calc_fitness*:

```
346     else if (fitnessEstimation.compare("BACS") == 0)    // Bull And Cows
347     {
348         for (int i = 0; i < GA_POPSIZE; i++)
349         {
350             int fitness = 0;
351             for (int j = 0; j < tsize; j++)
352             {
353                 if (population[i].str[j] != target[j])
354                 {
355                     fitness++; //caught different places to count the same as compliment
356                 }
357             }
358             int localFitness = 0;
359             for (int j = 0; j < tsize; j++)
360             {
361                 bool SEEN = false;
362                 for (int k = 0; k < tsize; k++)
363                 {
364                     if (population[i].str[k] != target[k] && (j != k) && population[i].str[j] != target[k] && !SEEN)
365                     {
366                         localFitness++; // pay more for the difference places!
367                         SEEN = true;
368                     }
369                 }
370             }
371             // update and assign values:
372             population[i].caughtPlaces = tsize - fitness;
373             population[i].fitness = fitness + localFitness;
374             population[i].cpuTicks = ((double)(clock() - startClock));
375             population[i].elapsedTime = time(&population[i].elapsedTime) - startTime;
376         }
377     }
378     Population::calc_ga_stat(population);
379 }
```

השוואה בין היוריסטיקת "בול פגיעה" להיוריסטיקה המקורית:

מקורית	בול פגיעה
פונקציית מרחק מנהטן משקלים מפולגים אחד במחרוזת	פונקציה שמערבת בתוכה חוכמה בנוגע לתוכן הנתונים. מכיוון שאלו מחרוזות והיא רוצה למצוא התאמה היא ממשקלת טעויות והצלחות בהתאם ולא משקלים שמפולגים באופן אחד כמו בפונקציה המקורית.
לינארית באורך הקלט	ריבועית באורך הקלט
דורשת הרבה דורות	דורשת מעט דורות

שימוש בהיוריסטיקת "בול פגיעה":

```
Best: Hello World! (0) -> [mean=15,std=52.6688] -> CPU Clock Ticks=403,Elapsed Run Time=0 -> Generation=15
End-up after: 424.00 ticks and 0.00 seconds
```

שימוש בהיוריסטיקה המקורית:

```
Best: Hello World! (0) -> [mean=11,std=1015.48] -> CPU Clock Ticks=639,Elapsed Run Time=0 -> Generation=36
End-up after: 654.00 ticks and 0.00 seconds
```

- ההיוריסטיקה של "בול פגיעה" קשורה למוטציות במהלך האלגוריתם באופן שבו כאשר יתבצע שינוי מוטנטי חלק מאותם גנים מאוד יתקרבו ויקטינו את "קירבתם" לפתרון משמעותית (כי לאחר שינוי זה "בול פגיעה" תחושב שוב) ולכן זה יצור התכנסות מהירה יותר לפתרון.
 - ההיוריסטיקה של "בול פגיעה" אכן משפרת את ההיוריסטיקה המקורית כפי שנאמר בטבלה הנ"ל מסיבת הבינה שמוכנסת אליה בהתאם לנתונים המעורבים.
4. תמיכה ב-SUS:

```

66 static int SUS(ga_vector &population, ga_vector &buffer)
67 {
68     ga_vector leftOver(GA_POPSIZE);
69     double m = 0;
70     int sum = population[0].fitness;
71     for (int i = 0; i < population.size(); i++)
72     {
73         m += population[i].fitness;
74     }
75     m /= population.size(); // finding avg of population
76     double r = (double)rand() / (RAND_MAX + 1); // pick random number between 0 to 1
77     double delta = r * m;
78     int j = 0, i = 0, k = 0;
79     do {
80         if (delta < sum) // for some delta we divide the population by sum - that is the point of this method [part of roulette]
81         {
82             buffer[i] = population[j];
83             buffer[i].str = population[j].str;
84             i++;
85             delta += sum;
86         }
87         else
88         {
89             // keep into leftOver all individuals about sum less of equals than delta
90             sum += population[j].fitness;
91             leftOver[k] = population[j];
92             leftOver[k].str = population[j].str;
93             k++;
94             j++;
95         }
96     } while (j < population.size());
97     int startRest = i;
98     for (k = i, j = 0; k < GA_POPSIZE - i; k++, j++) // coping leftOvers to the buffer's back
99     {
100         buffer[k] = leftOver[j];
101         buffer[k].str = leftOver[j].str;
102     }
103     return startRest;

```

```
116 static int Tournament(ga_vector &population, ga_vector &buffer)
117 {
118     int K = 2; //rand() % GA_POPSIZE;
119     vector<int> rest;
120     ga_struct best;
121     bool isBestNull = true;
122     for (int i = 0; i < K; i++) // taking K best random indecies
123     {
124         int ind = rand() % population.size();
125         if (isBestNull || (best.fitness > population[ind].fitness))
126         {
127             best = population[ind];
128             isBestNull = false;
129             rest.push_back(ind);
130         }
131     }
132     // copy relevant object to buffer
133     int j = 0, k = 0;
134     for (int i = 0; i < GA_POPSIZE; i++)
135     {
136         if ((k < rest.size()) && (i == rest[k]))
137         {
138             buffer[j] = population[i];
139             j++;
140             k++;
141         }
142     }
143     int restStart = j; // esize index for next mutation the rest population before next round
144     k = 0;
145     // copy the rest to buffer - the order between two groups is very important
146     for (int i = 0; (k < rest.size()) && (j < GA_POPSIZE) && (i < GA_POPSIZE); i++)
147     {
148         if (i != rest[k])
149         {
150             buffer[j] = population[i];
151             j++;
152             k++;
153         }
154     }
155     return restStart;
156 }
```

5. אסטרטגיית בחירה – "טורניר":

```
146 static int Turnir(ga_vector &population, ga_vector &buffer)
147 {
148     int K = 2; // rand() % GA_POPSIZE;
149     vector<int> rest;
150     int j = 0;
151     for (int i = 0; i < K; i++)
152     {
153         int ipos = rand() % GA_POPSIZE;
154         int jpos = rand() % GA_POPSIZE;
155         bool SEEN = false;
156         for (int h = 0; h < i; h++)
157         {
158             if (rest[h] == ipos || rest[h] == jpos)
159             {
160                 // pick k random individuals and check there is no duplications
161                 h = i;
162                 SEEN = true;
163             }
164         }
165         if (SEEN == true)
166         {
167             continue;
168         }
169         if (population[ipos].fitness < population[jpos].fitness)
170         {
171             // if one individual is bigger than the other he will win and pass to the next generation
172             buffer[j] = population[ipos];
173             rest.push_back(jpos);
174         }
175         else
176         {
177             buffer[j] = population[jpos];
178             rest.push_back(ipos);
179         }
180         j++;
181     }
182     sort(rest.begin(), rest.end());
183     int restStart = j;
184     int k = 0;
185     for (int i = 0, k = 0; i < GA_POPSIZE; i++) // put in the buffer's back
186     {
187         if (i == rest[k])
188         {
189             buffer[j] = population[i]; // rest[j]
190             j++;
191             k++;
192         }
193     }
194     return restStart;
195 }
```

6. אסטרטגיית שחלוף נוספת - "twoPoints":

```

48 static void twoPoints(ga_struct &population1, ga_struct &population2, ga_struct &bufferi, int spos1, int spos2)
49 {
50     bufferi.str = population1.str.substr(0, spos1) +
51     population2.str.substr(spos1, spos2 - spos1) +
52     population1.str.substr(spos2, GA_TARGET.size() - spos2);
53 }

```

אסטרטגיית שחלוף נוספת:

```

54 static void uniform(ga_struct &population1, ga_struct &population2, ga_struct &bufferi)
55 {
56     for(int i = 0; i < GA_TARGET.size(); i++){
57         int gen_to_select = rand() % 2;
58         if(gen_to_select == 0){
59             bufferi.str[i] = population1.str[i];
60         } else {
61             bufferi.str[i] = population2.str[i];
62         }
63     }
64 }

```

אסטרטגיית מוטציה נוספת:

```

240 static void swap(ga_struct &member)
241 {
242     // swapping between two indecies - very simple method:
243     int tsize = GA_TARGET.size();
244     int ipos = rand() % tsize;
245     int jpos = rand() % tsize;
246
247     char temp = member.str[ipos];
248     member.str[ipos] = member.str[jpos];
249     member.str[jpos] = temp;
250 }

```

7. בדיקת רגישות:

- נסמן – (1) = בעיית המחזורות עם ההיוריסטיקה המקורית ו-(2) = בעיית המחזורות עם היוריסטיקת "בול פגיעה".

א.

גודל אוכלוסייה	(1)	(2)
2048	29 דורות, 0 שניות, 638 טיקים	14 דורות, 1 שניות, 449 טיקים
5500	28 דורות, 2 שניות, 1679 טיקים	14 דורות, 1 שניות, 1191 טיקים
10000	25 דורות, 3 שניות, 2763 טיקים	12 דורות, 2 שניות, 1859 טיקים
20000	22 דורות, 5 שניות, 5056 טיקים	11 דורות, 4 שניות, 3577 טיקים

- ניתן להבחין שגודל האוכלוסייה משפיע באופן דרמטי על כמות הטיקים של שעון המחשב ועל זמן החישוב. יתרה מזאת ניתן לראות באופן מובהק שמספר הדורות באופן ממוצע נשאר זהה – רגישות לגודל האוכלוסייה בפרמטר הזמן.

ב.

הסתברות למוטציות	(1)	(2)
10%	43 דורות, 1 שניות, 968 טיקים	14 דורות, 1 שניות, 421 טיקים
25%	36 דורות, 1 שניות, 807 טיקים	14 דורות, 1 שניות, 422 טיקים
55%	38 דורות, 1 שניות, 854 טיקים	16 דורות, 1 שניות, 503 טיקים
90%	98 דורות, 2 שניות, 2309 טיקים	17 דורות, 0 שניות, 533 טיקים

- ניתן להבחין בשינוי מספר הדורות של אלגוריתם (1) בפונקציה של הסתברות למוטציות לעומת אלגוריתם (2) שהוא יותר יציב. בפרמטרי הזמן בממוצע אין שינוי – רגישות להסתברות למוטציות בפרמטר של מספר דורות עבור אלגוריתם (1).

ג.

פרופורציית האוכלוסייה האליטיסטית	(1)	(2)
10%	28 דורות, 1 שניות, 611 טיקים	14 דורות, 1 שניות, 445 טיקים
30%	34 דורות, 1 שניות, 725 טיקים	16 דורות, 1 שניות, 491 טיקים
60%	46 דורות, 1 שניות, 973 טיקים	22 דורות, 0 שניות, 667 טיקים
90%	278 דורות, 6 שניות, 5699 טיקים	56 דורות, 1 שניות, 1644 טיקים

- ניתן להבחין שהפרמטרים ששני האלגוריתמים גדלים כפונקציה של פרופורציית האוכלוסייה האליטיסטית פרט לפרמטר הזמן שהוא באופן ממוצע יציב אצל אלגוריתם (2) – רגישות לפרופורציית האוכלוסייה האליטיסטית פרט לזמן באלגוריתם (2).

ד.

אסטרטגיית הבחירה	(1)	(2)
<i>elitism</i>	35 דורות, 1 שניות, 765 טיקים	15 דורות, 1 שניות, 441 טיקים
<i>Turnir</i>	37 דורות, 0 שניות, 860 טיקים	14 דורות, 1 שניות, 438 טיקים
<i>SUS</i>	67 דורות, 2 שניות, 1570 טיקים	13 דורות, 0 שניות, 414 טיקים
<i>Tournament</i>	49 דורות, 1 שניות, 1085 טיקים	15 דורות, 1 שניות, 486 טיקים

- ניתן להבחין שאלגוריתם 1 יותר מושפע משינוי שיטת הבחירה מאשר אלגוריתם (2) שיציב עם שונות נמוכה – רגישות לאסטרטגיית הבחירה באלגוריתם (1).

ה.

אסטרטגיית השחלוף	(1)	(2)
<i>onePoint</i>	53 דורות, 1 שניות, 906 טיקים	15 דורות, 0 שניות, 402 טיקים
<i>twoPoints</i>	26 דורות, 1 שניות, 568 טיקים	15 דורות, 0 שניות, 457 טיקים
<i>uniform</i>	96 דורות, 2 שניות, 1562 טיקים	14 דורות, 0 שניות, 379 טיקים

- כמו באסטרטגיית הבחירה הנ"ל גם כאן ניתן להבחין ברגישות אלגוריתם (1) לשינוי שיטת השחלוף בניגוד ליציבות של אלגוריתם (2).
- איכות הפתרון לפי כל פרמטר הוא מצוין, כלומר ברוב המוחלט של המקרים האלגוריתם התכנס.

חלק ב' – חיפוש עם אילוצים:

8. ייצוג מתאים לגן באורך N :

```
23 struct stateBoard {  
24     vector<int> _board;  
25     unsigned int _fitness;  
26 };
```

- וקטור באורך N כאשר העמודה ה- i מכיל את השורה שבה המלכה ממוקמת.

:simple reverse

```

222 static void simpleReverse(state &member)
223 {
224     // simple reverse random chunk
225     int tsize = GA_NQUE_SIZE;
226     int ipos = rand() % tsize;
227     int jpos = rand() % tsize;
228
229     int low = 0, high = 0;
230     if (ipos < jpos)
231     {
232         low = ipos;
233         high = jpos;
234     }
235     else
236     {
237         low = jpos;
238         high = ipos;
239     }
240     vector<int> s(GA_NQUE_SIZE);
241     for (int i = 0; i < low; i++)
242     {
243         s.push_back(member._board[i]);
244     }
245     vector<int> toRev;
246     for (int i = low; i <= high; i++)
247     {
248         toRev.push_back(member._board[i]);
249     }
250     reverse(toRev.begin(), toRev.end());
251     for (int i = 0; i < toRev.size(); i++)
252     {
253         s.push_back(toRev[i]);
254     }
255     for (int i = high + 1; i < GA_NQUE_SIZE; i++)
256     {
257         s.push_back(member._board[i]);
258     }
259
260     for(int i = 0;i < GA_NQUE_SIZE;i++){
261         member._board[i] = s[i];
262     }

```

:insertion

```
274 static void insertion(state &member)
275 {
276     // insert new random letter to random place
277     int tsize = GA_NQUE_SIZE;
278     int ipos = rand() % tsize;
279     int nextPlace = rand() % tsize;
280
281     vector<int> temp(GA_NQUE_SIZE);
282     for(int i = 0; i < GA_NQUE_SIZE; i++){
283         if(i != ipos){
284             temp.push_back(member._board[i]);
285         } else if(i == nextPlace){
286             temp.push_back(member._board[i]);
287             temp.push_back(member._board[ipos]);
288         }
289     }
290     for(int i = 0; i < GA_NQUE_SIZE; i++){
291         member._board[i] = temp[i];
292     }
293 }
```

אופרטורים נוספים לשחלוף:

:PMX operator

```
76 static void PMX(state &populationi1, state &populationi2, state &bufferi)
77 {
78     //initial buffer for result usage
79     for(int i = 0; i < GA_NQUE_SIZE; i++){
80         bufferi._board[i] = -1;
81     }
82     //picking randomly block chunk
83     int ipos = rand() % GA_NQUE_SIZE;
84     int jpos = rand() % GA_NQUE_SIZE;
85     int low = 0, high = 0;
86     if(ipos <= jpos){
87         low = ipos;
88         high = jpos;
89     } else {
90         low = jpos;
91         high = ipos;
92     }
93     //directly transform the p1 segment to the children keeping the alels order
94     for(int i = low; i <= high; i++){
95         bufferi._board[i] = populationi1._board[i];
96     }
97     vector<int> p2_src; // values from p1
98     vector<int> p2_dest; // values from p2
99     for(int i = low; i <= high; i++){
100         bool IS_THERE = false;
101         for(int j = low; j <= high; j++){
102             if(populationi2._board[j] == populationi1._board[i]){
103                 IS_THERE = true;
104                 break;
105             }
106         }
107         //building the mapping between values in p2 to values same location in p1 where these values are no in child already
108         if(!IS_THERE){
109             p2_src.push_back(populationi2._board[j]);
110             p2_dest.push_back(populationi1._board[i]);
111         }
112     }
113     for(int i = 0; i < p2_dest.size(); i++){
114         for(int j = 0; j < GA_NQUE_SIZE; j++){
115             // find mapped value in the p2 array
116             if(populationi2._board[j] == p2_dest[i]){
117                 //check if this place is empty
118                 if(bufferi._board[j] == -1){
119                     bufferi._board[j] = p2_src[i];
120                 }
121                 //else {
122                 // because of duplication is allowed better is to live this case because of Inf loop and bad performance we could achieve.
123                 // this is working well without it!
124                 //}
125             }
126         }
127     }
128     for(int i = 0; i < GA_NQUE_SIZE; i++){
129         if(bufferi._board[i] == -1){
130             bufferi._board[i] = populationi2._board[i];
131         }
132     }
133 }
```

:OX operator

```
134 static void OX(state &population1, state &population2, state &buffer1)
135 {
136     //initial buffer for result usage
137     for(int i = 0; i < GA_NQUE_SIZE; i++){
138         buffer1._board[i] = -1;
139     }
140     // pick chunk of aels from population1
141     int ipos = rand() % GA_NQUE_SIZE;
142     int jpos = rand() % GA_NQUE_SIZE;
143     int low = 0, high = 0;
144     if(ipos <= jpos){
145         low = ipos;
146         high = jpos;
147     } else {
148         low = jpos;
149         high = ipos;
150     }
151     //directly transform the p1 segment to the children keeping the aels order
152     for(int i = low; i <= high; i++){
153         buffer1._board[i] = population1._board[i];
154     }
155     //picking each value in index i to be in buffer index i -> this operation is handling duplications otherwise there is no any coverage.
156     for(int i = 0; i < low; i++){
157         buffer1._board[i] = population2._board[i];
158     }
159     for(int i = (high + 1); i < GA_NQUE_SIZE; i++){
160         buffer1._board[i] = population2._board[i];
161     }
162 }
```

הרצת האלגוריתם לגדלי לוח שונים:

:n = 8

```
8 Queens solution using Genetic algorithms:
Best Position: <1,5,7,0,6,3,2,7> (fitness=2,generation=1)
Best Position: <5,1,4,7,0,3,1,6> (fitness=1,generation=2)
Best Position: <0,6,1,7,5,3,2,4> (fitness=1,generation=3)
Best Position: <5,1,4,7,0,3,1,6> (fitness=1,generation=4)
Best Position: <7,4,1,5,2,6,0,3> (fitness=1,generation=5)
Best Position: <4,4,7,0,3,6,2,5> (fitness=1,generation=6)
Best Position: <2,4,7,3,0,6,1,5> (fitness=0,generation=7)

Solution visualization:
-----
0 0 0 0 X 0 0 0
0 0 0 0 0 0 X 0
X 0 0 0 0 0 0 0
0 0 0 X 0 0 0 0
0 X 0 0 0 0 0 0
0 0 0 0 0 0 0 X
0 0 0 0 0 X 0 0
0 0 X 0 0 0 0 0

End-up after: 179.00 ticks and 0.00 seconds
```

:n = 15

```
15 Queens solution using Genetic algorithms:
Best Position: <4,2,10,6,1,3,12,3,0,11,7,2,7,13,8> (fitness=5,generation=1)
Best Position: <4,2,10,6,1,3,12,3,0,11,7,2,7,13,8> (fitness=5,generation=2)
Best Position: <9,2,0,13,11,8,12,3,3,14,4,10,5,1,7> (fitness=4,generation=3)
Best Position: <9,2,0,13,11,8,12,3,3,14,4,10,5,1,7> (fitness=4,generation=4)
Best Position: <4,14,11,3,10,6,1,9,2,0,7,1,7,11,13> (fitness=4,generation=5)
Best Position: <12,5,11,8,10,4,14,0,11,6,1,7,10,13,3> (fitness=3,generation=6)
Best Position: <12,5,11,8,10,4,14,0,11,6,1,7,10,13,3> (fitness=3,generation=7)
Best Position: <3,1,11,14,9,6,12,0,0,13,5,8,2,4,10> (fitness=3,generation=8)
Best Position: <12,5,11,8,10,4,14,0,11,6,1,7,10,13,3> (fitness=3,generation=9)
Best Position: <3,1,11,14,9,6,12,0,0,13,5,8,2,4,10> (fitness=3,generation=10)
Best Position: <12,5,11,8,10,4,14,0,11,6,1,7,10,13,3> (fitness=3,generation=11)
Best Position: <3,1,11,14,9,6,12,0,0,13,5,8,2,4,10> (fitness=3,generation=12)
Best Position: <4,1,14,0,6,7,12,10,13,2,9,3,8,11,1> (fitness=2,generation=13)
Best Position: <13,4,5,8,0,12,4,2,14,11,9,1,3,10,7> (fitness=2,generation=14)
Best Position: <4,1,14,0,6,7,12,10,13,2,9,3,8,11,1> (fitness=2,generation=15)
Best Position: <13,4,5,8,0,12,4,2,14,11,9,1,3,10,7> (fitness=2,generation=16)
Best Position: <4,1,14,0,6,7,12,10,13,2,9,3,8,11,1> (fitness=2,generation=17)
Best Position: <13,4,5,8,0,12,4,2,14,11,9,1,3,10,7> (fitness=2,generation=18)
Best Position: <4,1,14,0,6,7,12,10,13,2,9,3,8,11,1> (fitness=2,generation=19)
Best Position: <13,4,5,8,0,12,4,2,14,11,9,1,3,10,7> (fitness=2,generation=20)
Best Position: <4,1,14,0,6,7,12,10,13,2,9,3,8,11,1> (fitness=2,generation=21)
Best Position: <13,4,5,8,0,12,4,2,14,11,9,1,3,10,7> (fitness=2,generation=22)
Best Position: <3,5,10,14,11,6,12,2,0,8,12,9,1,13,7> (fitness=2,generation=23)
Best Position: <4,1,14,0,6,7,12,10,13,2,9,3,8,11,1> (fitness=2,generation=24)
Best Position: <13,4,5,8,0,12,4,2,14,11,9,1,3,10,7> (fitness=2,generation=25)
Best Position: <3,5,10,14,11,6,12,2,0,8,12,9,1,13,7> (fitness=2,generation=26)
Best Position: <4,1,14,0,6,7,12,10,13,2,9,3,8,11,1> (fitness=2,generation=27)
Best Position: <13,4,5,8,0,12,4,2,14,11,9,1,3,10,7> (fitness=2,generation=28)
Best Position: <13,4,5,11,6,14,12,2,0,8,10,1,3,9,7> (fitness=1,generation=29)
Best Position: <13,4,5,11,6,14,12,2,0,8,10,1,3,9,7> (fitness=1,generation=30)
Best Position: <13,4,5,11,6,14,12,2,0,8,10,1,3,9,7> (fitness=1,generation=31)
Best Position: <13,4,5,11,6,14,12,2,0,8,10,1,3,9,7> (fitness=1,generation=32)
Best Position: <13,4,5,11,6,14,12,2,0,8,10,1,3,9,7> (fitness=1,generation=33)
Best Position: <1,11,4,12,7,0,8,3,13,9,14,2,5,10,6> (fitness=1,generation=34)
Best Position: <13,4,5,11,6,14,12,2,0,8,10,1,3,9,7> (fitness=1,generation=35)
Best Position: <1,11,4,12,7,0,8,3,13,9,14,2,5,10,6> (fitness=1,generation=36)
Best Position: <1,11,4,12,7,0,8,3,13,9,14,2,5,10,6> (fitness=1,generation=37)
Best Position: <4,1,12,5,9,13,3,8,14,7,2,0,11,6,10> (fitness=0,generation=38)
```

Solution visualization:

```
-----
0 0 0 0 0 0 0 0 0 0 0 0 X 0 0 0
0 X 0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 X 0 0 0
0 0 0 0 0 0 0 X 0 0 0 0 0 0 0 0
X 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 X 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0 X 0
0 0 0 0 0 0 0 0 0 0 X 0 0 0 0 0
0 0 0 0 0 0 0 0 X 0 0 0 0 0 0 0
0 0 0 0 X 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 X
0 0 0 0 0 0 0 0 0 0 0 0 0 0 X 0 0
0 0 X 0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 X 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 X 0 0 0 0 0 0
```

End-up after: 1203.00 ticks and 1.00 seconds

:n = 25

```
25 Queens solution using Genetic algorithms:
Best Position: <17,12,14,11,3,23,6,16,3,10,21,16,9,19,6,22,20,8,22,24,1,12,2,0,7> (fitness=12,generation=1)
Best Position: <13,23,13,0,8,22,17,7,21,19,15,9,10,20,4,12,18,11,24,0,16,22,13,3,16> (fitness=11,generation=2)
Best Position: <0,12,22,11,3,23,6,16,3,10,21,16,14,2,6,22,20,13,9,15,18,4,19,17,7> (fitness=10,generation=3)
Best Position: <0,12,22,11,3,23,6,16,3,10,21,16,14,2,6,22,20,13,9,15,18,4,19,17,7> (fitness=10,generation=4)
Best Position: <17,8,1,4,11,7,16,22,19,2,14,20,6,5,0,13,7,18,7,19,23,15,10,3,23> (fitness=9,generation=5)
Best Position: <17,8,1,4,11,7,16,22,19,2,14,20,6,5,0,13,7,18,7,19,23,15,10,3,23> (fitness=9,generation=6)
Best Position: <17,8,1,4,11,7,16,22,19,2,14,20,6,5,0,13,7,18,7,19,23,15,10,3,23> (fitness=9,generation=7)
Best Position: <10,5,9,19,7,12,6,20,3,21,15,5,14,24,15,23,22,4,6,1,16,2,13,18,8> (fitness=8,generation=8)
Best Position: <14,22,3,9,16,10,5,20,0,21,7,20,23,8,17,23,6,2,18,21,4,1,10,10,13> (fitness=8,generation=9)
Best Position: <10,5,9,19,7,12,6,20,3,21,15,5,14,24,15,23,22,4,6,1,16,2,13,18,8> (fitness=8,generation=10)
Best Position: <14,22,3,9,16,10,5,20,0,21,7,20,23,8,17,23,6,2,18,21,4,1,10,10,13> (fitness=8,generation=11)
Best Position: <10,5,9,19,7,12,6,20,3,21,15,5,14,24,15,23,22,4,6,1,16,2,13,18,8> (fitness=8,generation=12)
Best Position: <6,15,0,10,19,5,11,17,3,22,16,14,24,8,4,23,9,21,2,20,7,1,18,20,20> (fitness=7,generation=13)
Best Position: <15,9,21,14,7,22,0,6,13,16,2,24,12,3,22,4,23,0,20,10,21,1,8,10,19> (fitness=7,generation=14)
Best Position: <6,15,0,10,19,5,11,17,3,22,16,14,24,8,4,23,9,21,2,20,7,1,18,20,20> (fitness=7,generation=15)
Best Position: <15,9,21,14,7,22,0,6,13,16,2,24,12,3,22,4,23,0,20,10,21,1,8,10,19> (fitness=7,generation=16)
Best Position: <6,15,0,10,19,5,11,17,3,22,16,14,24,8,4,23,9,21,2,20,7,1,18,20,20> (fitness=7,generation=17)
Best Position: <7,18,13,2,22,20,8,10,21,14,9,17,12,23,12,19,5,24,1,3,11,17,15,4,16> (fitness=7,generation=18)
```

...

```
Best Position: <6,13,15,10,7,2,20,23,0,11,21,12,16,9,24,4,17,22,3,18,14,19,5,1,8> (fitness=1,generation=120)
Best Position: <6,13,15,10,7,2,20,23,0,11,21,12,16,9,24,4,17,22,3,18,14,19,5,1,8> (fitness=1,generation=121)
Best Position: <6,13,15,2,7,10,20,23,0,16,14,1,21,9,24,4,17,11,3,12,22,19,5,18,8> (fitness=1,generation=122)
Best Position: <6,13,15,10,7,2,20,23,0,11,21,12,16,9,24,4,17,22,3,18,14,19,5,1,8> (fitness=1,generation=123)
Best Position: <6,13,15,10,7,2,20,23,0,11,21,12,16,9,24,4,17,22,3,18,14,19,5,1,8> (fitness=1,generation=124)
Best Position: <6,13,15,10,7,2,20,23,0,11,21,12,16,9,24,4,17,22,3,18,14,19,5,1,8> (fitness=1,generation=125)
Best Position: <6,13,15,21,7,2,20,23,0,16,10,1,11,9,24,4,17,22,3,12,14,19,5,18,8> (fitness=0,generation=126)
```

Solution visualization:

```
-----
0 0 0 0 0 0 0 0 X 0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 X 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 X 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 X 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0 X 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 X 0 0
X 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 X 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 X
0 0 0 0 0 0 0 0 0 0 0 X 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 X 0 0 0 0 0 0 0 0 0 0 0
0 X 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 X 0 0 0 0 0
0 0 X 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 X 0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 X 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 X 0
0 0 0 0 0 0 0 X 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 X 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 X 0 0 0 0 0 0
0 0 0 0 0 0 0 X 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0 X 0 0 0 0 0 0 0
```

End-up after: 5913.00 ticks and 5.00 seconds

9. השוואה בין ביצועי האלגוריתם הגנטי לבין *Minimal conflicts*:

פרמטר	GA	MC
זמן	סביר	מהיר
מקום	צריכה גבוהה	לינארי באורך הקלט
יכולת לפתור לוחות גדולים ללא מצב התחלה טוב	טובה מאוד	גרועה
יכולת לפתור לוחות גדולים עם מצב התחלה טוב	מצוינת	מצוינת
איטרציות ללא מצב התחלה טוב	סבירות	גבוהות
איטרציות עם מצב התחלה טוב	מעטות	מעטות

- ההכלאה בין האלגוריתמים הינה אפשרית באופן שבו *minimal conflicts* יכול להיות מוטציה לצורך שיפור הגן לקראת הדור הבא וע"י כך לשפר את הדור הבא בצורה חכמה יותר ולשפר את מהירות ההתכנסות בצורה משמעותית.

10. הדגמה ושימוש באלגוריתם הגנטי בכדי לפתור את "בעיית השק":

פריט יראה מהצורה הבאה:

```

33 struct item {
34     vector<pair<pair<unsigned int, unsigned int>, unsigned int>> _products; // one vector usage because of conceptually understanding the problem
35     unsigned int _weights; // constraint
36     unsigned int _fitness; // profit
37 };
38
39 typedef vector<item> ga_couples;

```

- ה-*typedef* נועד לצורך שימוש במערך של פריטים באורך נוח במהלך התוכנית.

באופן דומה ל-*framework* הנ"ל נשתמש:

1. שיטת בחירה – *elitism*
2. שיטת שחלוף – *onePoint*
3. שיטת מוטציה - *flip one random bit*

פתרון בעיה 1:

```
Computing problem number 1 solution...
Best Solution: <1,1,0,1,0,0,1,0,0,0> (Weight = 161,Profit = 284,Generation = 1)
Best Solution: <1,1,1,1,0,1,0,0,0,0> (Weight = 165,Profit = 309,Generation = 2)
Best Solution: <1,1,1,1,0,1,0,0,0,0> (Weight = 165,Profit = 309,Generation = 3)
Best Solution: <1,1,1,1,0,1,0,0,0,0> (Weight = 165,Profit = 309,Generation = 4)
Best Solution: <1,1,1,1,0,1,0,0,0,0> (Weight = 165,Profit = 309,Generation = 5)
Best Solution: <1,1,1,1,0,1,0,0,0,0> (Weight = 165,Profit = 309,Generation = 6)
Best Solution: <1,1,1,1,0,1,0,0,0,0> (Weight = 165,Profit = 309,Generation = 7)
Best Solution: <1,1,1,1,0,1,0,0,0,0> (Weight = 165,Profit = 309,Generation = 8)
Best Solution: <1,1,1,1,0,1,0,0,0,0> (Weight = 165,Profit = 309,Generation = 9)
Best Solution: <1,1,1,1,0,1,0,0,0,0> (Weight = 165,Profit = 309,Generation = 10)
Best Solution: <1,1,1,1,0,1,0,0,0,0> (Weight = 165,Profit = 309,Generation = 11)
Best Solution: <1,1,1,1,0,1,0,0,0,0> (Weight = 165,Profit = 309,Generation = 12)
Best Solution: <1,1,1,1,0,1,0,0,0,0> (Weight = 165,Profit = 309,Generation = 13)
Best Solution: <1,1,1,1,0,1,0,0,0,0> (Weight = 165,Profit = 309,Generation = 14)
Best Solution: <1,1,1,1,0,1,0,0,0,0> (Weight = 165,Profit = 309,Generation = 15)
Best Solution: <1,1,1,1,0,1,0,0,0,0> (Weight = 165,Profit = 309,Generation = 16)
Best Solution: <1,1,1,1,0,1,0,0,0,0> (Weight = 165,Profit = 309,Generation = 17)
Best Solution: <1,1,1,1,0,1,0,0,0,0> (Weight = 165,Profit = 309,Generation = 18)
Best Solution: <1,1,1,1,0,1,0,0,0,0> (Weight = 165,Profit = 309,Generation = 19)
Best Solution: <1,1,1,1,0,1,0,0,0,0> (Weight = 165,Profit = 309,Generation = 20)
Best Solution: <1,1,1,1,0,1,0,0,0,0> (Weight = 165,Profit = 309,Generation = 21)
Best Solution: <1,1,1,1,0,1,0,0,0,0> (Weight = 165,Profit = 309,Generation = 22)

An Optimal Packing: <1,1,1,1,0,1,0,0,0,0> (Weight = 165,Profit = 309,Generation = 22)
End-up after: 123.00 ticks and 0.00 seconds
```

פתרון בעיה 2:

```
Computing problem number 2 solution...
Best Solution: <0,1,1,1,0> (Weight = 26,Profit = 51,Generation = 1)
Best Solution: <0,1,1,1,0> (Weight = 26,Profit = 51,Generation = 2)
Best Solution: <0,1,1,1,0> (Weight = 26,Profit = 51,Generation = 3)
Best Solution: <0,1,1,1,0> (Weight = 26,Profit = 51,Generation = 4)
Best Solution: <0,1,1,1,0> (Weight = 26,Profit = 51,Generation = 5)
Best Solution: <0,1,1,1,0> (Weight = 26,Profit = 51,Generation = 6)
Best Solution: <0,1,1,1,0> (Weight = 26,Profit = 51,Generation = 7)
Best Solution: <0,1,1,1,0> (Weight = 26,Profit = 51,Generation = 8)
Best Solution: <0,1,1,1,0> (Weight = 26,Profit = 51,Generation = 9)
Best Solution: <0,1,1,1,0> (Weight = 26,Profit = 51,Generation = 10)
Best Solution: <0,1,1,1,0> (Weight = 26,Profit = 51,Generation = 11)
Best Solution: <0,1,1,1,0> (Weight = 26,Profit = 51,Generation = 12)
Best Solution: <0,1,1,1,0> (Weight = 26,Profit = 51,Generation = 13)
Best Solution: <0,1,1,1,0> (Weight = 26,Profit = 51,Generation = 14)
Best Solution: <0,1,1,1,0> (Weight = 26,Profit = 51,Generation = 15)
Best Solution: <0,1,1,1,0> (Weight = 26,Profit = 51,Generation = 16)
Best Solution: <0,1,1,1,0> (Weight = 26,Profit = 51,Generation = 17)
Best Solution: <0,1,1,1,0> (Weight = 26,Profit = 51,Generation = 18)
Best Solution: <0,1,1,1,0> (Weight = 26,Profit = 51,Generation = 19)
Best Solution: <0,1,1,1,0> (Weight = 26,Profit = 51,Generation = 20)
Best Solution: <0,1,1,1,0> (Weight = 26,Profit = 51,Generation = 21)

An Optimal Packing: <0,1,1,1,0> (Weight = 26,Profit = 51,Generation = 21)
End-up after: 97.00 ticks and 0.00 seconds
```


פתרון בעיה 3:

```
Computing problem number 3 solution...
Best Solution: <1,1,0,0,1,0> (Weight = 190,Profit = 150,Generation = 1)
Best Solution: <1,1,0,0,1,0> (Weight = 190,Profit = 150,Generation = 2)
Best Solution: <1,1,0,0,1,0> (Weight = 190,Profit = 150,Generation = 3)
Best Solution: <1,1,0,0,1,0> (Weight = 190,Profit = 150,Generation = 4)
Best Solution: <1,1,0,0,1,0> (Weight = 190,Profit = 150,Generation = 5)
Best Solution: <1,1,0,0,1,0> (Weight = 190,Profit = 150,Generation = 6)
Best Solution: <1,1,0,0,1,0> (Weight = 190,Profit = 150,Generation = 7)
Best Solution: <1,1,0,0,1,0> (Weight = 190,Profit = 150,Generation = 8)
Best Solution: <1,1,0,0,1,0> (Weight = 190,Profit = 150,Generation = 9)
Best Solution: <1,1,0,0,1,0> (Weight = 190,Profit = 150,Generation = 10)
Best Solution: <1,1,0,0,1,0> (Weight = 190,Profit = 150,Generation = 11)
Best Solution: <1,1,0,0,1,0> (Weight = 190,Profit = 150,Generation = 12)
Best Solution: <1,1,0,0,1,0> (Weight = 190,Profit = 150,Generation = 13)
Best Solution: <1,1,0,0,1,0> (Weight = 190,Profit = 150,Generation = 14)
Best Solution: <1,1,0,0,1,0> (Weight = 190,Profit = 150,Generation = 15)
Best Solution: <1,1,0,0,1,0> (Weight = 190,Profit = 150,Generation = 16)
Best Solution: <1,1,0,0,1,0> (Weight = 190,Profit = 150,Generation = 17)
Best Solution: <1,1,0,0,1,0> (Weight = 190,Profit = 150,Generation = 18)
Best Solution: <1,1,0,0,1,0> (Weight = 190,Profit = 150,Generation = 19)
Best Solution: <1,1,0,0,1,0> (Weight = 190,Profit = 150,Generation = 20)
Best Solution: <1,1,0,0,1,0> (Weight = 190,Profit = 150,Generation = 21)

An Optimal Packing: <1,1,0,0,1,0> (Weight = 190,Profit = 150,Generation = 21)
End-up after: 98.00 ticks and 0.00 seconds
```

פתרון בעיה 4:

```
Computing problem number 4 solution...
Best Solution: <1,0,0,1,0,0,0> (Weight = 50,Profit = 107,Generation = 1)
Best Solution: <1,0,0,1,0,0,0> (Weight = 50,Profit = 107,Generation = 2)
Best Solution: <1,0,0,1,0,0,0> (Weight = 50,Profit = 107,Generation = 3)
Best Solution: <1,0,0,1,0,0,0> (Weight = 50,Profit = 107,Generation = 4)
Best Solution: <1,0,0,1,0,0,0> (Weight = 50,Profit = 107,Generation = 5)
Best Solution: <1,0,0,1,0,0,0> (Weight = 50,Profit = 107,Generation = 6)
Best Solution: <1,0,0,1,0,0,0> (Weight = 50,Profit = 107,Generation = 7)
Best Solution: <1,0,0,1,0,0,0> (Weight = 50,Profit = 107,Generation = 8)
Best Solution: <1,0,0,1,0,0,0> (Weight = 50,Profit = 107,Generation = 9)
Best Solution: <1,0,0,1,0,0,0> (Weight = 50,Profit = 107,Generation = 10)
Best Solution: <1,0,0,1,0,0,0> (Weight = 50,Profit = 107,Generation = 11)
Best Solution: <1,0,0,1,0,0,0> (Weight = 50,Profit = 107,Generation = 12)
Best Solution: <1,0,0,1,0,0,0> (Weight = 50,Profit = 107,Generation = 13)
Best Solution: <1,0,0,1,0,0,0> (Weight = 50,Profit = 107,Generation = 14)
Best Solution: <1,0,0,1,0,0,0> (Weight = 50,Profit = 107,Generation = 15)
Best Solution: <1,0,0,1,0,0,0> (Weight = 50,Profit = 107,Generation = 16)
Best Solution: <1,0,0,1,0,0,0> (Weight = 50,Profit = 107,Generation = 17)
Best Solution: <1,0,0,1,0,0,0> (Weight = 50,Profit = 107,Generation = 18)
Best Solution: <1,0,0,1,0,0,0> (Weight = 50,Profit = 107,Generation = 19)
Best Solution: <1,0,0,1,0,0,0> (Weight = 50,Profit = 107,Generation = 20)
Best Solution: <1,0,0,1,0,0,0> (Weight = 50,Profit = 107,Generation = 21)

An Optimal Packing: <1,0,0,1,0,0,0> (Weight = 50,Profit = 107,Generation = 21)
End-up after: 116.00 ticks and 0.00 seconds
```

פתרון בעיה 5:

```
Computing problem number 5 solution...
Best Solution: <1,0,1,1,1,0,1,1> (Weight = 104,Profit = 900,Generation = 1)
Best Solution: <1,0,1,1,1,0,1,1> (Weight = 104,Profit = 900,Generation = 2)
Best Solution: <1,0,1,1,1,0,1,1> (Weight = 104,Profit = 900,Generation = 3)
Best Solution: <1,0,1,1,1,0,1,1> (Weight = 104,Profit = 900,Generation = 4)
Best Solution: <1,0,1,1,1,0,1,1> (Weight = 104,Profit = 900,Generation = 5)
Best Solution: <1,0,1,1,1,0,1,1> (Weight = 104,Profit = 900,Generation = 6)
Best Solution: <1,0,1,1,1,0,1,1> (Weight = 104,Profit = 900,Generation = 7)
Best Solution: <1,0,1,1,1,0,1,1> (Weight = 104,Profit = 900,Generation = 8)
Best Solution: <1,0,1,1,1,0,1,1> (Weight = 104,Profit = 900,Generation = 9)
Best Solution: <1,0,1,1,1,0,1,1> (Weight = 104,Profit = 900,Generation = 10)
Best Solution: <1,0,1,1,1,0,1,1> (Weight = 104,Profit = 900,Generation = 11)
Best Solution: <1,0,1,1,1,0,1,1> (Weight = 104,Profit = 900,Generation = 12)
Best Solution: <1,0,1,1,1,0,1,1> (Weight = 104,Profit = 900,Generation = 13)
Best Solution: <1,0,1,1,1,0,1,1> (Weight = 104,Profit = 900,Generation = 14)
Best Solution: <1,0,1,1,1,0,1,1> (Weight = 104,Profit = 900,Generation = 15)
Best Solution: <1,0,1,1,1,0,1,1> (Weight = 104,Profit = 900,Generation = 16)
Best Solution: <1,0,1,1,1,0,1,1> (Weight = 104,Profit = 900,Generation = 17)
Best Solution: <1,0,1,1,1,0,1,1> (Weight = 104,Profit = 900,Generation = 18)
Best Solution: <1,0,1,1,1,0,1,1> (Weight = 104,Profit = 900,Generation = 19)
Best Solution: <1,0,1,1,1,0,1,1> (Weight = 104,Profit = 900,Generation = 20)
Best Solution: <1,0,1,1,1,0,1,1> (Weight = 104,Profit = 900,Generation = 21)

An Optimal Packing: <1,0,1,1,1,0,1,1> (Weight = 104,Profit = 900,Generation = 21)
End-up after: 109.00 ticks and 1.00 seconds
```

פתרון בעיה 6:

```
Computing problem number 6 solution...
Best Solution: <0,1,0,1,0,0,1> (Weight = 169,Profit = 1735,Generation = 1)
Best Solution: <0,1,0,1,0,0,1> (Weight = 169,Profit = 1735,Generation = 2)
Best Solution: <0,1,0,1,0,0,1> (Weight = 169,Profit = 1735,Generation = 3)
Best Solution: <0,1,0,1,0,0,1> (Weight = 169,Profit = 1735,Generation = 4)
Best Solution: <0,1,0,1,0,0,1> (Weight = 169,Profit = 1735,Generation = 5)
Best Solution: <0,1,0,1,0,0,1> (Weight = 169,Profit = 1735,Generation = 6)
Best Solution: <0,1,0,1,0,0,1> (Weight = 169,Profit = 1735,Generation = 7)
Best Solution: <0,1,0,1,0,0,1> (Weight = 169,Profit = 1735,Generation = 8)
Best Solution: <0,1,0,1,0,0,1> (Weight = 169,Profit = 1735,Generation = 9)
Best Solution: <0,1,0,1,0,0,1> (Weight = 169,Profit = 1735,Generation = 10)
Best Solution: <0,1,0,1,0,0,1> (Weight = 169,Profit = 1735,Generation = 11)
Best Solution: <0,1,0,1,0,0,1> (Weight = 169,Profit = 1735,Generation = 12)
Best Solution: <0,1,0,1,0,0,1> (Weight = 169,Profit = 1735,Generation = 13)
Best Solution: <0,1,0,1,0,0,1> (Weight = 169,Profit = 1735,Generation = 14)
Best Solution: <0,1,0,1,0,0,1> (Weight = 169,Profit = 1735,Generation = 15)
Best Solution: <0,1,0,1,0,0,1> (Weight = 169,Profit = 1735,Generation = 16)
Best Solution: <0,1,0,1,0,0,1> (Weight = 169,Profit = 1735,Generation = 17)
Best Solution: <0,1,0,1,0,0,1> (Weight = 169,Profit = 1735,Generation = 18)
Best Solution: <0,1,0,1,0,0,1> (Weight = 169,Profit = 1735,Generation = 19)
Best Solution: <0,1,0,1,0,0,1> (Weight = 169,Profit = 1735,Generation = 20)
Best Solution: <0,1,0,1,0,0,1> (Weight = 169,Profit = 1735,Generation = 21)

An Optimal Packing: <0,1,0,1,0,0,1> (Weight = 169,Profit = 1735,Generation = 21)
End-up after: 109.00 ticks and 1.00 seconds
```

פתרון בעיה 7:

```

Computing problem number 7 solution...
Best Solution: <1,1,0,0,0,1,1,1,0,0,0,0,1,1> (Weight = 750,Profit = 1455,Generation = 1)
Best Solution: <1,1,0,0,0,1,1,1,1,0,0,0,0,1,1> (Weight = 750,Profit = 1455,Generation = 2)
Best Solution: <1,1,0,0,0,1,1,1,1,0,0,0,0,1,1> (Weight = 750,Profit = 1455,Generation = 3)
Best Solution: <0,1,1,1,0,0,1,1,1,0,0,0,0,1,1> (Weight = 750,Profit = 1456,Generation = 4)
Best Solution: <0,1,1,1,0,0,1,1,1,0,0,0,0,1,1> (Weight = 750,Profit = 1456,Generation = 5)
Best Solution: <0,1,1,1,0,0,1,1,1,0,0,0,0,1,1> (Weight = 750,Profit = 1456,Generation = 6)
Best Solution: <1,0,1,0,1,0,1,1,1,0,0,0,0,1,1> (Weight = 749,Profit = 1458,Generation = 7)
Best Solution: <1,0,1,0,1,0,1,1,1,0,0,0,0,1,1> (Weight = 749,Profit = 1458,Generation = 8)
Best Solution: <1,0,1,0,1,0,1,1,1,0,0,0,0,1,1> (Weight = 749,Profit = 1458,Generation = 9)
Best Solution: <1,0,1,0,1,0,1,1,1,0,0,0,0,1,1> (Weight = 749,Profit = 1458,Generation = 10)
Best Solution: <1,0,1,0,1,0,1,1,1,0,0,0,0,1,1> (Weight = 749,Profit = 1458,Generation = 11)
Best Solution: <1,0,1,0,1,0,1,1,1,0,0,0,0,1,1> (Weight = 749,Profit = 1458,Generation = 12)
Best Solution: <1,0,1,0,1,0,1,1,1,0,0,0,0,1,1> (Weight = 749,Profit = 1458,Generation = 13)
Best Solution: <1,0,1,0,1,0,1,1,1,0,0,0,0,1,1> (Weight = 749,Profit = 1458,Generation = 14)
Best Solution: <1,0,1,0,1,0,1,1,1,0,0,0,0,1,1> (Weight = 749,Profit = 1458,Generation = 15)
Best Solution: <1,0,1,0,1,0,1,1,1,0,0,0,0,1,1> (Weight = 749,Profit = 1458,Generation = 16)
Best Solution: <1,0,1,0,1,0,1,1,1,0,0,0,0,1,1> (Weight = 749,Profit = 1458,Generation = 17)
Best Solution: <1,0,1,0,1,0,1,1,1,0,0,0,0,1,1> (Weight = 749,Profit = 1458,Generation = 18)
Best Solution: <1,0,1,0,1,0,1,1,1,0,0,0,0,1,1> (Weight = 749,Profit = 1458,Generation = 19)
Best Solution: <1,0,1,0,1,0,1,1,1,0,0,0,0,1,1> (Weight = 749,Profit = 1458,Generation = 20)
Best Solution: <1,0,1,0,1,0,1,1,1,0,0,0,0,1,1> (Weight = 749,Profit = 1458,Generation = 21)
Best Solution: <1,0,1,0,1,0,1,1,1,0,0,0,0,1,1> (Weight = 749,Profit = 1458,Generation = 22)
Best Solution: <1,0,1,0,1,0,1,1,1,0,0,0,0,1,1> (Weight = 749,Profit = 1458,Generation = 23)
Best Solution: <1,0,1,0,1,0,1,1,1,0,0,0,0,1,1> (Weight = 749,Profit = 1458,Generation = 24)
Best Solution: <1,0,1,0,1,0,1,1,1,0,0,0,0,1,1> (Weight = 749,Profit = 1458,Generation = 25)
Best Solution: <1,0,1,0,1,0,1,1,1,0,0,0,0,1,1> (Weight = 749,Profit = 1458,Generation = 26)
Best Solution: <1,0,1,0,1,0,1,1,1,0,0,0,0,1,1> (Weight = 749,Profit = 1458,Generation = 27)

An Optimal Packing: <1,0,1,0,1,0,1,1,1,0,0,0,0,1,1> (Weight = 749,Profit = 1458,Generation = 27)
End-up after: 202.00 ticks and 0.00 seconds

```

פתרון בעיה 8:

[illegible]

An Optimal Packing: <1,1,0,1,1,1,0,0,0,1,1,0,1,0,0,1,0,0,0,0,0,0,1,1,1> (Weight = 6402560, Profit = 13549094, Generation = 39)
End-up after: 477.00 ticks and 0.00 seconds

- כפי שניתן לראות הנ"ל הבעיות נפתרו מאוד מהר הן כפרמטר של מספר דורות עד להתכנסות והן כפרמטר של זמן.

תודה רבה על זמן הקריאה