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HW #1

For solving 8 queens problem, i used Genetic algorithm, with below specification:

representation: permutation of 8 numbers

data structure: list

each list element show queens position.

element value: queens row position - element index: queens column position

search space: 8!

fitness: inverse sum of each queens of phenotype penalty

one queen penalty: number of queens she can check.

```
def fitness(pheno):
    fitnessInvers = 0
    for i in range(len(pheno)):
        for j in range(len(pheno)):
            if (abs(pheno[i] - pheno[j]) == abs(i - j) and i != j):
                fitnessInvers += 1
    return fitnessInvers
```

population: 100 fixed phenotype

initialization: 100 random permutation of 1-8

```
def initializePopulation(numOfqueens, numPopultion):
    population = []
    for i in range(numPopultion):
        population.append(random.sample(range(1, numOfqueens+1), numOfqueens))
    return population
```

mutation: swapping values of two randomly chosen positions

mutation probability 80%

```
def mutation(child):
    i = random.randint(0, 7)
    j = random.randint(0, 7)
    child[i], child[j] = child[j], child[i]
```

cross over: for each 2 parent, after permutation we'll have 2 child

select random crossover point, copy first parts of each parent to children, and for second parts of each child insert values from opposite parent.

```
def crossOver(parents):
    crossoverPoint = random.randint(1, 7)
    child = []
    for i in range(0,2):
        child.append( parents[i][:crossoverPoint] )
        oppositParentIndex = 1 - i
        for j in range(crossoverPoint, 8):
            if (parents[oppositParentIndex][j] not in child[i]):
                child[i].append(parents[oppositParentIndex][j])

        child[i] + parents[oppositParentIndex][:crossoverPoint]
        for k in range(crossoverPoint):
            el = parents[oppositParentIndex][:crossoverPoint][k]
            if (el not in child[i]):
                child[i].append(el)
    return (child)
```

parent selection: select 5 parents randomly, choose the two fittest to do cross over

```
def parentSelection(population):
    random5parents = []
    indexList = list(range(len(population)))
    for i in range(5):
        selectedIndex = random.choice(indexList)
        random5parents.append(population[selectedIndex])
        indexList.remove(selectedIndex)

    parentFitenss = []
    for i in range(5):
        parentFitenss.append((random5parents[i], fitness(random5parents[i])))
    parentFitenss.sort(key=lambda tup: tup[1])
    return (parentFitenss[0][0], parentFitenss[1][0])
```

survivor selection: sort whole population in order by increasing inverse fitness, select the top 100 for new generation

```
def survivorSelection(newPopulation, numPopultion):
    phenoFitness = []
    for i in range(len(newPopulation)):
        phenoFitness.append((newPopulation[i], fitness(newPopulation[i])))
    phenoFitness.sort(key=lambda tup: tup[1])
    populationSurvivor = []
    for item in phenoFitness[:numPopultion]:
        populationSurvivor.append(item[0])
    return populationSurvivor
```

termination condition: if reverse fitness of a phenotype is equal to 0(which is solution phenotype) or if we pass 1000 iteration(which we couldn't find any solution)

```
if (fitness(survivorPopulation[0]) == 0):  
    print ("the solution is", survivorPopulation[0], "with", iterations, "iterations")  
    return population
```

```
if (iterations >= 1000):  
    print ("no solution found with 1000 iterations")  
    return population
```