## miniProject

## December 8, 2020

```
[1]: import numpy
     import random
     random.seed()
[2]: Nd = 9
     class Population(object):
         """ A set of candidate solutions to the Sudoku puzzle. These candidates are \Box
      \hookrightarrowalso known as the chromosomes in the population. """
         def __init__(self):
             self.candidates = []
             return
         def seed(self, Nc, given):
             self.candidates = []
             # Determine the legal values that each square can take.
             helper = Candidate()
             helper.values = [[[] for j in range(0, Nd)] for i in range(0, Nd)]
             for row in range(0, Nd):
                 for column in range(0, Nd):
                     for value in range(1, 10):
                         if((given.values[row][column] == 0) and not (given.
      →is_column_duplicate(column, value) or given.is_block_duplicate(row, column, __
      →value) or given.is_row_duplicate(row, value))):
                              # Value is available.
                              helper.values[row][column].append(value)
                         elif(given.values[row][column] != 0):
                              # Given/known value from file.
                              helper.values[row][column].append(given.
      →values[row][column])
                              break
             # Seed a new population.
             for p in range(0, Nc):
                 g = Candidate()
```

```
for i in range(0, Nd): # New row in candidate.
               row = numpy.zeros(Nd)
               # Fill in the givens.
               for j in range(0, Nd): # New column j value in row i.
                    # If value is already given, don't change it.
                   if(given.values[i][j] != 0):
                       row[j] = given.values[i][j]
                    # Fill in the gaps using the helper board.
                   elif(given.values[i][j] == 0):
                       row[j] = helper.values[i][j][random.randint(0,__
\rightarrowlen(helper.values[i][j])-1)]
               # If we don't have a valid board, then try again. There must be
\rightarrowno duplicates in the row.
               while(len(list(set(row))) != Nd):
                   for j in range(0, Nd):
                       if(given.values[i][j] == 0):
                            row[j] = helper.values[i][j][random.randint(0, __
→len(helper.values[i][j])-1)]
               g.values[i] = row
           self.candidates.append(g)
       # Compute the fitness of all candidates in the population.
       self.update_fitness()
       print("Seeding complete.")
       return
   def update_fitness(self):
       """ Update fitness of every candidate/chromosome. """
       for candidate in self.candidates:
           candidate.update_fitness()
       return
   def sort(self):
       """ Sort the population based on fitness. """
       self.candidates.sort(reverse = True,key = lambda x: x.fitness)
       return
   def sort_fitness(self, x, y):
       """ The sorting function. """
       if(x.fitness < y.fitness):</pre>
           return 1
       elif(x.fitness == y.fitness):
```

```
return 0
else:
return -1
```

```
[3]: class Candidate(object):
         """ A candidate solutions to the Sudoku puzzle. """
         def __init__(self):
             self.values = numpy.zeros((Nd, Nd), dtype=int)
             self.fitness = None
             return
         def update_fitness(self):
              """ The fitness of a candidate solution is determined by how close it_\sqcup

ightharpoonup is to being the actual solution to the puzzle. The actual solution (i.e. the \sqcup
      _{\hookrightarrow} 'fittest') is defined as a 9x9 grid of numbers in the range [1, 9] where _{\sqcup}
      \hookrightarroweach row, column and 3x3 block contains the numbers [1, 9] without any \sqcup
      →duplicates (see e.g. http://www.sudoku.com/); if there are any duplicates
      \hookrightarrow then the fitness will be lower. """
             row_count = numpy.zeros(Nd)
             column_count = numpy.zeros(Nd)
             block_count = numpy.zeros(Nd)
             row_sum = 0
             column_sum = 0
             block_sum = 0
             for i in range(0, Nd): # For each row...
                  for j in range(0, Nd): # For each number within it...
                      row_count[self.values[i][j]-1] += 1 # ... Update list with_
      →occurrence of a particular number.
                  row_sum += (1.0/len(set(row_count)))/Nd
                  row_count = numpy.zeros(Nd)
             for i in range(0, Nd): # For each column...
                  for j in range(0, Nd): # For each number within it...
                      column_count[self.values[j][i]-1] += 1 # ... Update list with
      →occurrence of a particular number.
                  column_sum += (1.0 / len(set(column_count)))/Nd
                  column_count = numpy.zeros(Nd)
              # For each block...
             for i in range(0, Nd, 3):
                  for j in range(0, Nd, 3):
                      block_count[self.values[i][j]-1] += 1
```

```
block_count[self.values[i][j+1]-1] += 1
               block_count[self.values[i][j+2]-1] += 1
               block_count[self.values[i+1][j]-1] += 1
               block_count[self.values[i+1][j+1]-1] += 1
               block_count[self.values[i+1][j+2]-1] += 1
               block_count[self.values[i+2][j]-1] += 1
               block count[self.values[i+2][j+1]-1] += 1
               block_count[self.values[i+2][j+2]-1] += 1
               block_sum += (1.0/len(set(block_count)))/Nd
               block_count = numpy.zeros(Nd)
       # Calculate overall fitness.
       if (int(row sum) == 1 and int(column sum) == 1 and int(block sum) == 1):
           fitness = 1.0
       else:
           fitness = column_sum * block_sum
       self.fitness = fitness
       return
  def mutate(self, mutation_rate, given):
       """ Mutate a candidate by picking a row, and then picking two values_{\sqcup}
⇒within that row to swap. """
      r = random.uniform(0, 1.1)
      while(r > 1): # Outside [0, 1] boundary - choose another
           r = random.uniform(0, 1.1)
      success = False
       if (r < mutation_rate): # Mutate.</pre>
           while(not success):
               row1 = random.randint(0, 8)
               row2 = random.randint(0, 8)
               row2 = row1
               from_column = random.randint(0, 8)
               to_column = random.randint(0, 8)
               while(from_column == to_column):
                   from_column = random.randint(0, 8)
                   to_column = random.randint(0, 8)
               # Check if the two places are free...
               if(given.values[row1][from_column] == 0 and given.
→values[row1][to_column] == 0):
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# ...and that we are not causing a duplicate in the rows'
→ columns.
                   if(not given.is_column_duplicate(to_column, self.
→values[row1][from_column])
                      and not given.is_column_duplicate(from_column, self.
→values[row2][to_column])
                      and not given.is_block_duplicate(row2, to_column, self.
→values[row1][from_column])
                      and not given.is_block_duplicate(row1, from_column, self.
→values[row2][to_column])):
                       # Swap values.
                       temp = self.values[row2][to_column]
                       self.values[row2][to_column] = self.
→values[row1][from_column]
                       self.values[row1] [from_column] = temp
                       success = True
       return success
```

```
[4]: class Given(Candidate):
         def init (self, values):
             self.values = values
             return
         def is_row_duplicate(self, row, value):
             for column in range(0, Nd):
                 if(self.values[row][column] == value):
                    return True
             return False
         def is_column_duplicate(self, column, value):
             for row in range(0, Nd):
                 if(self.values[row][column] == value):
                    return True
             return False
         def is_block_duplicate(self, row, column, value):
             i = 3*(int(row/3))
             j = 3*(int(column/3))
             if((self.values[i][j] == value)
                or (self.values[i][j+1] == value)
                or (self.values[i][j+2] == value)
                or (self.values[i+1][j] == value)
                or (self.values[i+1][j+1] == value)
```

```
or (self.values[i+1][j+2] == value)
or (self.values[i+2][j] == value)
or (self.values[i+2][j+1] == value)
or (self.values[i+2][j+2] == value)):
    return True
else:
    return False
```

```
[5]: class Tournament(object):
          """ The crossover function requires two parents to be selected from the \sqcup
      →population pool. The Tournament class is used to do this.
          Two individuals are selected from the population pool and a random number \Box
      _{\hookrightarrow} in [0, 1] is chosen. If this number is less than the 'selection rate' (e.g. _{\sqcup}
      \rightarrow 0.85), then the fitter individual is selected; otherwise, the weaker one is \Box
      \hookrightarrow selected.
          11 11 11
         def __init__(self):
              return
         def compete(self, candidates):
              c1 = candidates[random.randint(0, len(candidates)-1)]
              c2 = candidates[random.randint(0, len(candidates)-1)]
              f1 = c1.fitness
              f2 = c2.fitness
              if(f1 > f2):
                  fittest = c1
                  weakest = c2
              else:
                  fittest = c2
                  weakest = c1
              selection_rate = 0.85
              r = random.uniform(0, 1.1)
              while(r > 1):
                  r = random.uniform(0, 1.1)
              if(r < selection rate):</pre>
                  return fittest
              else:
                  return weakest
```

## [6]: class CycleCrossover(object): """ Crossover relates to the analogy of genes within each parent candidate → mixing together in the hopes of creating a fitter child candidate. Cycle → crossover is used here """

```
def __init__(self):
       return
   def crossover(self, parent1, parent2, crossover_rate):
       child1 = Candidate()
       child2 = Candidate()
       child1.values = numpy.copy(parent1.values)
       child2.values = numpy.copy(parent2.values)
       r = random.uniform(0, 1.1)
       while(r > 1):
           r = random.uniform(0, 1.1)
       if (r < crossover_rate):</pre>
           crossover_point1 = random.randint(0, 8)
           crossover_point2 = random.randint(1, 9)
           while(crossover_point1 == crossover_point2):
               crossover_point1 = random.randint(0, 8)
               crossover_point2 = random.randint(1, 9)
           if(crossover_point1 > crossover_point2):
               temp = crossover_point1
               crossover point1 = crossover point2
               crossover_point2 = temp
           for i in range(crossover_point1, crossover_point2):
               child1.values[i], child2.values[i] = self.crossover_rows(child1.
→values[i], child2.values[i])
       return child1, child2
   def crossover_rows(self, row1, row2):
       child_row1 = numpy.zeros(Nd)
       child_row2 = numpy.zeros(Nd)
       remaining = [*range(1, Nd+1)]
       cycle = 0
       while((0 in child_row1) and (0 in child_row2)):
           if(cycle % 2 == 0): # Even cycles.
               # Assign next unused value.
               index = self.find_unused(row1, remaining)
               start = row1[index]
               remaining.remove(row1[index])
               child_row1[index] = row1[index]
               child_row2[index] = row2[index]
```

```
next = row2[index]
                     while(next != start): # While cycle not done...
                         index = self.find_value(row1, next)
                         child_row1[index] = row1[index]
                         remaining.remove(row1[index])
                         child_row2[index] = row2[index]
                         next = row2[index]
                     cycle += 1
                 else: # Odd cycle - flip values.
                     index = self.find_unused(row1, remaining)
                     start = row1[index]
                     remaining.remove(row1[index])
                     child_row1[index] = row2[index]
                     child_row2[index] = row1[index]
                     next = row2[index]
                     while(next != start): # While cycle not done...
                         index = self.find_value(row1, next)
                         child row1[index] = row2[index]
                         remaining.remove(row1[index])
                         child row2[index] = row1[index]
                         next = row2[index]
                     cycle += 1
             return child_row1, child_row2
         def find_unused(self, parent_row, remaining):
     #
               print(parent_row[2])
             for i in range(0, len(parent_row)):
                 if(parent_row[i] in remaining):
                     return i
         def find_value(self, parent_row, value):
             for i in range(0, len(parent_row)):
                 if(parent_row[i] == value):
                     return i
[7]: class Sudoku(object):
         """ Solves a given Sudoku puzzle using a genetic algorithm. """
         def __init__(self):
```

self.given = None

return

```
def load(self, path):
   with open(path, "r") as f:
        values = numpy.loadtxt(f).reshape((Nd, Nd)).astype(int)
        self.given = Given(values)
   return
def save(self, path, solution):
    with open(path, "w") as f:
        numpy.savetxt(f, solution.values.reshape(Nd*Nd), fmt='%d')
    return
def solve(self):
   Nc = 1000 # Number of candidates (i.e. population size).
   Ne = int(0.05*Nc) # Number of elites.
   Ng = 1000 # Number of generations.
   Nm = 0 # Number of mutations.
    # Mutation parameters.
   phi = 0
   sigma = 1
   mutation_rate = 0.06
    # Create an initial population.
    self.population = Population()
    self.population.seed(Nc, self.given)
   stale = 0
    for generation in range(0, Ng):
        print("Generation %d" % generation)
        best_fitness = 0.0
        for c in range(0, Nc):
            fitness = self.population.candidates[c].fitness
            if(fitness == 1):
                print("Solution found at generation %d!" % generation)
                print(self.population.candidates[c].values)
                return self.population.candidates[c]
            if(fitness > best_fitness):
                best_fitness = fitness
        print("Best fitness: %f" % best_fitness)
        next_population = []
```

```
# Select elites (the fittest candidates) and preserve them for the
\rightarrownext generation.
           self.population.sort()
           elites = []
           for e in range(0, Ne):
               elite = Candidate()
               elite.values = numpy.copy(self.population.candidates[e].values)
               elites.append(elite)
           # Create the rest of the candidates.
           for count in range(Ne, Nc, 2):
               t = Tournament()
               parent1 = t.compete(self.population.candidates)
               parent2 = t.compete(self.population.candidates)
               ## Cross-over.
               cc = CycleCrossover()
               child1, child2 = cc.crossover(parent1, parent2,_
\rightarrow crossover_rate=1.0)
               # Mutate child1.
               child1.update_fitness()
               old_fitness = child1.fitness
               success = child1.mutate(mutation_rate, self.given)
               child1.update_fitness()
               if(success):
                   Nm += 1
                   if(child1.fitness > old_fitness):
                       phi = phi + 1
               # Mutate child2.
               child2.update_fitness()
               old_fitness = child2.fitness
               success = child2.mutate(mutation_rate, self.given)
               child2.update_fitness()
               if(success):
                   Nm += 1
                   if(child2.fitness > old_fitness):
                        phi = phi + 1
               # Add children to new population.
               next_population.append(child1)
               next_population.append(child2)
           # Append elites onto the end of the population. These will not have
→been affected by crossover or mutation.
           for e in range(0, Ne):
```

```
next_population.append(elites[e])
           # Select next generation.
           self.population.candidates = next_population
           self.population.update_fitness()
           # Calculate new adaptive mutation rate (based on Rechenberg's 1/5u
\rightarrowsuccess rule). This is to stop too much mutation as the fitness progresses.
\rightarrow towards unity.
           if(Nm == 0):
               phi = 0
           else:
               phi = phi / Nm
           if(phi > 0.2):
               sigma = sigma/0.998
           elif(phi < 0.2):
               sigma = sigma*0.998
           mutation_rate = abs(numpy.random.normal(loc=0.0, scale=sigma,_
→size=None))
           Nm = 0
           phi = 0
           # Check for stale population.
           self.population.sort()
           if(self.population.candidates[0].fitness != self.population.
stale = 0
           else:
               stale += 1
           # Re-seed the population if 100 generations have passed with the
→ fittest two candidates always having the same fitness.
           if(stale >= 100):
               print("The population has gone stale. Re-seeding...")
               self.population.seed(Nc, self.given)
               stale = 0
               sigma = 1
               phi = 0
               Nm = 0
               mutation_rate = 0.06
       print("No solution found.")
       return None
```

```
[10]: s = Sudoku()
      s.load("puzz.txt")
      solution = s.solve()
      if(solution):
          s.save("solution.txt", solution)
     Seeding complete.
     Generation 0
     Best fitness: 0.308642
     Generation 1
     Best fitness: 0.308642
     Generation 2
     Best fitness: 0.349794
     Generation 3
     Best fitness: 0.349794
     Generation 4
     Best fitness: 0.473251
     Generation 5
     Best fitness: 0.473251
     Generation 6
     Best fitness: 0.473251
     Generation 7
     Best fitness: 0.536351
     Generation 8
     Best fitness: 0.536351
     Generation 9
     Best fitness: 0.536351
     Generation 10
     Best fitness: 0.599451
     Generation 11
     Best fitness: 0.599451
     Generation 12
     Best fitness: 0.599451
     Generation 13
     Best fitness: 0.662551
     Generation 14
     Best fitness: 0.777778
     Generation 15
     Best fitness: 0.777778
     Generation 16
     Best fitness: 0.851852
     Generation 17
     Best fitness: 0.851852
     Generation 18
     Best fitness: 0.851852
     Generation 19
     Solution found at generation 19!
```

```
[[8 3 9 2 7 4 6 5 1]
[5 2 4 1 3 6 7 8 9]
[7 6 1 5 8 9 4 2 3]
[1 9 7 8 5 3 2 6 4]
[6 8 3 4 9 2 5 1 7]
[2 4 5 6 1 7 9 3 8]
[3 5 2 7 4 1 8 9 6]
[9 7 8 3 6 5 1 4 2]
[4 1 6 9 2 8 3 7 5]
```

[]: