AI Programming [Week 12] Practice

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- Random Restart 실습
- Simulated Annealing 실습
- main.py 실습
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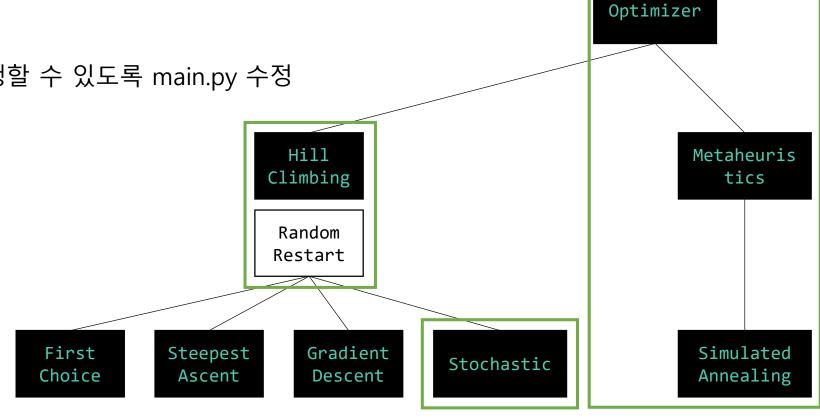
HW07 본인 제출 파일 준비

- main.py
- optimizer.py
- problem.py
- setup.py

실습 목표



- 1. Stochastic hill climbing 추가
- 2. Random Restart 추가
- 3. Simulated Annealing 추가
- 4. exp.txt를 사용해서 실험을 수행할 수 있도록 main.py 수정





optimizer.py 실습

Inference 구성하기

```
class HillClimbing(Optimizer):
    ...

def __init__(self):
    Optimizer.__init__(self)
    self._pType = 0
    self._limitStuck = 100
```

```
class Optimizer(Setup):
   def __init__(self):
       Setup.__init__(self)
class MetaHeuristic(Optimizer):
   def __init__(self):
        Optimizer.__init__(self)
class SimulatedAnnealing(MetaHeurstic):
   def __init__(self):
        MetaHeurstic.__init__(self)
class Stochastic(HillClimbing):
   def __init__(self):
        pass
```



Stochastic Hill climbing 실습 (10분)

Stochastic Class 구현하기 *some useful codes 사용

```
class Stochastic(HillClimbing):
   def displaySetting(self):
       print()
       print("Search Algorithm: Stochastic Hill Climbing")
       print()
       HillClimbing.displaySetting(self)
   def run(self, p):
       # hint; Stochastic 알고리즘은 Steepest Ascent 알고리즘과 흐름이 유사함
       p.storeResult(current, valueC)
   def stochasticBest(self, neighbors, p):
       # some useful codes 사용하기
       return neighbors[i], valuesForMin[i]
```



```
def selectAlgorithm(pType):
    print()
    print("Select the search algorithm:")
    print(" 1. Steepest-Ascent")
    print(" 2. First-Choice")
    print(" 3. Gradient Descent")
    print(" 4. Stocahstic")
    while True:
        aType = int(input("Enter the number: "))
        if not invalid(pType, aType):
            break
    optimizers = { 1: 'SteepestAscent()',
                   2: 'FirstChoice()',
                   3: 'GradientDescent()',
                   4: 'Stochastic()' }
    alg = eval(optimizers[aType])
    alg.setVariables(pType)
    return alg
```

Main문에 추가하여 Stochastic 실행해보기

*일부 문제는 **시간이** 오래 걸릴 수 있음

Ackley.txt or Griewank.txt or tsp 문제 풀어보기

(5분)



Random Restart 실습 (10분)

Random Restart는 기존의 HillClimbing methods들을 주어진 실험 수 만큼 반복 적으로 수행하여 더 좋은 결과를 찾는 방법으로,

HillClimbing methods들이 공통적으로 수행할 수 있도록 HillClimbing Class에 randomRestart Method를 추가

재대로 실행될 수 있도록 화살표된 부분 구현해주기

```
class HillClimbing(Optimizer):
    def __init__(self):
        self. numRestart = 1
    def randomRestart(self, p):
        i = 1
        self.run(p)
        bestSolution = p.getSolution()
        bestMinimum = p.getValue()
        numEval = p.getNumEval()
        while i < self._numRestart:</pre>
            self.run(p)
            newSolution = p.getSolution()
            newMinimum = p.getValue()
            numEval += p.getNumEval()
            if newMinimum < bestMinimum:</pre>
                bestSolution = newSolution
                bestMinimum = newMinimum
            i += 1
        p.storeResult(bestSolution, bestMinimum)
```



Random Restart 실습 (5분)

```
def main():
    p, pType = selectProblem()
    alg = selectAlgorithm(pType)
    alg randomRestart(p)
    p.describe()
    alg.displaySetting()
    p.report()
```

run을 randomRestart로 변경해서

Numeric – Ackley.txt – Gradient Descent 방식으로 문제 풀기

Hillclimbing – self._numRestart = 1 and 10 으로 각각 실행해서 결과(min value) 비교해보기



Simulated Annealing 실습

```
function SIMULATED-ANNEALING(problem, schedule) returns a solution state
inputs: problem, a problem
    schedule, a mapping from time to "temperature"
```

 $current \leftarrow Make-Node(problem.Initial-State)$

```
for t \leftarrow 1 to \infty do
T \leftarrow schedule[t]
```

While 문으로 대체하시면 됩니다.

if T = 0 then return current

 $next \leftarrow$ a randomly selected successor of *current*

 $\Delta E \leftarrow next. VALUE - current. VALUE$

if $\Delta E < 0$ then current \leftarrow next

else *current* \leftarrow *next* only with probability $e^{-\Delta E/T}$



Simulated Annealing 실습 (10분)

```
class SimulatedAnnealing:
   def init (self):
        self. numSample = 100
       self._limitEval = 100000
        self. when Best Found = 0
   def displaySetting(self):
        print("Search Algorithm: Simulated Annealing")
        print("Number of evaluations until termination: {0:,}"
              .format(self. limitEval))
   def run(self, p):
       current = p.randomInit()
       valueC = p.evaluate(current)
       best, valueBest = current, valueC
       whenBestFound = i = 1
       t = self.initTemp(p)
       while True:
            ... # Implement
        self. whenBestFound = whenBestFound
        p.storeResult(best, valueBest)
```

Some useful code 활용하여 SimulatedAnnealing 구현

Some useful code 안의 함수들은 SimulatedAnnealing Class 안으로 옮겨서 사용

Simulated Annealing 수행 (5분)

Simulated Annealing 코드 수행해보기

- 1. main.py-selectAlgorithm에 Simulated Annealing 추가
- 2. main.py-main에 if 분기문 추가
- 3. 임시적으로 구현한 Simulated Annealing이 잘 수행될 수 있도록 setVriables method 추가

Tsp100.txt 문제 풀어보고 hill climbing 알고리즘과 결과 비교해보기

```
class SimulatedAnnealing(MetaHeuristic):
    def setVariables(self, _):
        pass
```

```
def selectAlgorithm(pType):
    print(" 4. Stocahstic")
    print(" 5. SimulatedAnnealing")
    while True:
        aType = int(input("Enter the number: "))
        if not invalid(pType, aType):
            break
    optimizers = { 1: 'SteepestAscent()',
                   2: 'FirstChoice()',
                   3: 'GradientDescent()',
                   4: 'Stochastic()',
                   5: 'SimulatedAnnealing()'}
    alg = eval(optimizers[aType])
    def main():
        if issubclass(type(alg), HillClimbing):
            alg.randomRestart(p)
        else:
            alg.run(p)
```



main.py 실습

Text 파일(exp.txt)로부터 상수를 읽 어와서 사용하도록 변경

알고리즘을 여러 번 실험할 수 있도록 Main 함수 변경

Main 함수에서 여러 번 실험을 통해 결과를 report할 수 있도록 여러가지 기능 확장 예정

```
# If you are solving a function optimization problem,
  what should be the step size for axis-parallel mutation?
  Mutation step size (delta ): 0.01
# If your algorithm choice is 2 or 3,
   what should be the number of consecutive iterations without improvement?
  Give the number of iterations (limitStuck): 1000
# If your algorithm choice is 4 (gradient descent),
# what should be the update step size and increment for derivative?
  Update rate for gradient descent (alpha): 0.01
  Increment for calculating derivative (dx): 10 ** (-4)
# If you want a random-restart hill climbing,
   enter the number of restart.
# Enter 1 if you do not want a random-restart.
  Number of restarts (numRestart): 10
# If you are running a metaheuristic algorithm,
  what should be the total number of evaluations until temination?
 Enter the number (limitEval): 100000
# Enter the total number of experiments
  Enter the number (numExp): 10
```



main.py 실습

main – skeleton.py 완성하고 코드 이해하기

```
def createProblem(parameters):
    # debugger로 parameters 인자 확인하기
    pType = parameters['pType']
    if pType == 1:
        p = Numeric()
    elif pType == 2:
        p = Tsp()
    # p.setVariables 함수 수정하기
    p.setVariables(parameters)
    return p
```

main.py 실습

```
class Problem(Setup):
   def __init__(self):
       Setup.__init__(self)
        self._solution = []
        self. value = 0
        self._numEval = 0
        self. pFileName = ''
        self._bestSolution = []
        self._bestMinimum = 0
        self. avgMinimum = 0
        self._avgNumEval = 0
        self. sumOfNumEval = 0
        self._avgWhen = 0
   def setVariables(self, parameters):
        Setup.setVariables(self, parameters)
        self._pFileName = parameters['pFileName']
```

```
class Optimizer(Setup):
    def __init__(self):
        Setup.__init__(self)
        self._numExp = 0
    def setVariables(self, parameters):
        Setup.setVariables(self, parameters)
        self._numExp = parameters['numExp']
class Setup:
    def __init__(self):
        self. pType = 0
        self._aType = 0
        self._delta = 0
        self._alpha = 0
        self. dx = 0
    def setVariables(self, parameters):
        self._pType = parameters['pType']
        self._aType = parameters['aType']
        self._delta = parameters['delta']
        self._alpha = parameters['alpha']
        self._dx = parameters['dx']
    def getAType(self):
        return self._aType
```



main.py 실습

exp.txt 설정: tsp - tsp50.txt - first choice - limit stuck = 1000, numExp = 10 위 설정으로 구현했던 main.py debugging하며 필요한 위치에 method 추가 및 변경하기

*출력 결과도 예시와 같이 나올 수 있도록 변경

Number of cities: 50								
City locations	::							
(96, 22)	(56, 12)	(19, 24)	(83, 58)	(62, 5)				
(79, 31)	(1, 0)	(29, 71)	(17, 89)	(43, 66)				
(82, 74)	(52, 35)	(84, 92)	(93 , 45)	(41, 24)				
(36, 83)	(82, 35)	(89, 71)	(93 , 89)	(67, 10)				
(71, 82)	(68, 50)	(84, 81)	(74, 94)	(53 , 13)				
(81, 31)	(17, 92)	(99, 82)	(25, 63)	(0, 2)				
(21, 83)	(70, 64)	(79, 6)	(31, 53)	(90, 50)				
(48, 14)	(41, 26)	(80, 56)	(49, 51)	(19, 38)				
(2, 0)	(29, 63)	(18, 59)	(10, 44)	(49, 7)				
(37, 9)	(19, 14)	(90, 85)	(100, 5)	(34, 55)				

Number of experiments: 10

Search Algorithm: First-Choice Hill Climbing
Max evaluations with no improvement: 1,000 iterations

Best order of visits:

37	34	13	16	5	25	0	48	32	19
4	44	1	24	35	11	36	14	45	40
6	29	46	2	39	43	33	49	42	28
41	7	30	8	26	15	9	38	21	31
20	23	12	47	18	27	17	22	10	3

Minimum tour cost: 588

Total number of evaluations: 57,792

Number of cities: 50 City locations:

Number of experiments: 10

Search Algorithm: First-Choice Hill Climbing

Max evaluations with no improvement: 1,000 iterations

Average tour cost: 649

Average number of evaluations: 5,455

Best tour found:

1	28	41	10	37	26	49	2	44	8
27	47	12	22	43	7	25	14	39	13
24	34	40	18	29	30	46	15	36	9
2 3	48	42	31	11	19	0	16	17	35
38	5	20	4	33	32	45	3	21	6

Best tour cost: 624

Total number of evaluations: 54,546

과제 안내

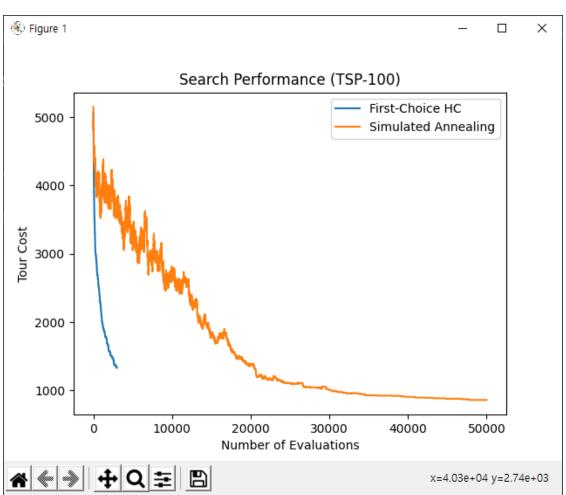


제공된 main – skeleton 코드를 사용하여 아래 알고리즘들 모두 정상적으로 수행될 수 있도록 구현 (output으로 실험 결과가 정상적으로 출력될 수 있어야 함)

* 반드시 구현했던 Class들을 활용하여 실행되어야 함

Hill climbing (random start)
Stochastic
first-choice
steepest ascent
gradient descent
Meta Heuristics
Simulated Annealing

+ First Choice, Simulated Annealing에서 NumEval에 따른 current Value를 시각화해서 제출 (tsp100.txt) 사용



과제 안내



제출물:

```
파이썬 파일 총 5개를 HW08_NAME 폴더로 묶어서 압축하여 제출 (.zip) main.py optimizer.py problem.py setup.py plot.py
```

리포트 제출 (.pdf)

Stochastic, simulated Anneaing으로 Numeric, TSP 문제 각각 실행한 terminal 스크린샷 총 4개 Plot.py로 만든 그래프 스크린샷