Homework3

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Problem Description

The Orienteering Problem (OP) is a variant of the Traveling Salesman Problem (TSP). The objective is to determine a subset of nodes to visit and the visiting sequence such that:

- Each node i has a score s_i .
- Traveling from node i to node j incurs a travel time c_{ij} .
- \bullet The total travel time must not exceed a given time budget T.
- The route starts and ends at node 0 (the depot).

The goal is to **maximize the total collected score** while staying within the travel time budget. The input data are randomly generated in the Python script op_random_instance.py.

Mathematical Model

Decision Variables:

$$x_{ij} = \begin{cases} 1 & \text{if the route goes from node } i \text{ to } j \\ 0 & \text{otherwise} \end{cases}$$

$$y_i = \begin{cases} 1 & \text{if node } i \text{ is visited} \\ 0 & \text{otherwise} \end{cases}$$

$$u_i \in \{1, \dots, n-1\}, \text{ position of node } i \text{ in the tour (MTZ constraints)}$$

Objective: Maximize the total collected score:

$$\max \sum_{i=0}^{n-1} s_i y_i$$

Constraints:

- Start and return to the depot: $y_0 = 1$
- Flow conservation:

$$\sum_{i \neq j} x_{ij} = y_j, \quad \sum_{i \neq j} x_{ji} = y_j \quad \forall j$$

• Time budget:

$$\sum_{i \neq j} c_{ij} x_{ij} \le T$$

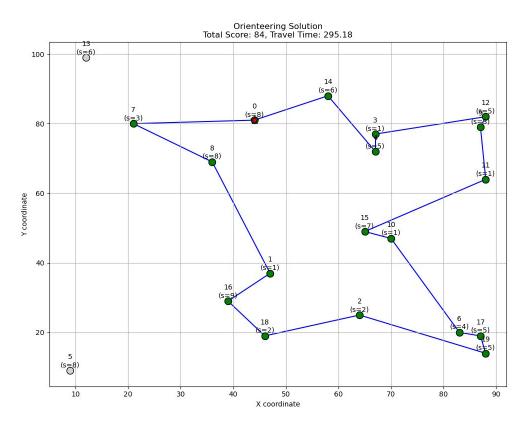
• Subtour elimination (Miller–Tucker–Zemlin constraints):

$$u_i - u_j + nx_{ij} \le n - 1 \quad \forall i \ne j, \ i, j \in \{1, \dots, n - 1\}$$

Solution Output

- Optimal Tour: e.g., [0, 14, 4, 3, 12, 9, 11, 15, 10, 6, 17, 19, 2, 18, 16, 1, 8, 7, 0]
- Total Collected Score: e.g., 84
- Total Travel Time: e.g., 295.18

Optimal Tour Visualization



Source Code

The Gurobi Python code used to solve this problem is:

```
import numpy as np
import matplotlib.pyplot as plt
import gurobipy as gp
from gurobipy import GRB
import networkx as nx

# ---- data ----
# number of customers
n = 20

# generate locations
np.random.seed(0)
loc_x = np.random.randint(0, 100, n)
loc_y = np.random.randint(0, 100, n)

# generate scores
s = np.random.randint(1, 10, n)
```

```
19 # calculate travel time
20 c = {
      (i, j): ((loc_x[i] - loc_x[j]) ** 2 + (loc_y[i] - loc_y[j]) ** 2) ** 0.5
21
      for i in range(n)
      for j in range(n)
23
24 }
25
26 # time budget
_{27} T = 300
28
29 # Create model
30 m = gp.Model("Orienteering")
31
32 # Decision variables
33 \# x[i,j] = 1 if we travel from node i to node j, 0 otherwise
x = m.addVars(c.keys(), vtype=GRB.BINARY, name="x")
36 # u[i] represents the position of node i in the tour (for subtour elimination)
37 u = m.addVars(range(1, n), lb=1, ub=n - 1, vtype=GRB.INTEGER, name="u")
38
39 # y[i] = 1 if node i is visited, 0 otherwise
y = m.addVars(range(n), vtype=GRB.BINARY, name="y")
42 # Objective: maximize the total collected score
43 m.setObjective(gp.quicksum(s[i] * y[i] for i in range(n)), GRB.MAXIMIZE)
45 # Constraints
46 # Node O must be the starting and ending point
47 m.addConstr(y[0] == 1, "visit_depot")
49 # Flow balance constraints
50 for j in range(n):
      m.addConstr(
          gp.quicksum(x[i, j] for i in range(n) if i != j) == y[j], f"in_flow_{j}"
52
53
54
      m.addConstr(
          gp.quicksum(x[j, i] for i in range(n) if i != j) == y[j], f"out_flow_{j}"
55
56
57
58 # Time budget constraint
59 m.addConstr(
      gp.quicksum(c[i, j] * x[i, j] for i, j in c.keys() if i != j) <= T, "time_budget"</pre>
60
61 )
62
63 # Miller-Tucker-Zemlin subtour elimination constraints
64 for i in range(1, n):
      for j in range(1, n):
65
          if i != j:
66
               m.addConstr(u[i] - u[j] + n * x[i, j] \le n - 1, f"mtz_{i}_{j}")
67
69 # Solve the model
70 m.optimize()
72 # Extract solution
73 if m.status == GRB.OPTIMAL:
      tour = []
74
      current = 0  # Start from depot
75
76
      tour.append(current)
77
      while True:
78
          for j in range(n):
79
               if j != current and x[current, j].x > 0.5:
80
                   tour.append(j)
81
                   current = j
82
83
                   break
          if current == 0 and len(tour) > 1:
84
85
               break
86
```

```
# Calculate total collected score and total travel time
 87
                    total_score = sum(s[i] for i in tour)
                    total_time = sum(c[tour[i], tour[i + 1]] for i in range(len(tour) - 1))
 89
 90
                    print(f"Optimal tour: {tour}")
 91
                    print(f"Total collected score: {total_score}")
 92
                     print(f"Total travel time: {total_time:.2f}")
 93
 94
                    # Plot the solution
 95
                    plt.figure(figsize=(10, 8))
 96
 97
                    # Plot all nodes
 98
                    plt.scatter(loc_x, loc_y, s=100, c="lightgray", edgecolors="black", zorder=1)
 99
100
                    # Plot visited nodes
                     visited_nodes = [i for i in range(n) if y[i].x > 0.5]
102
                    plt.scatter(
                                loc_x[visited_nodes],
104
                                loc_y[visited_nodes],
                                s = 100,
106
107
                                c="green",
                                edgecolors="black",
108
                                zorder=2,
                    )
110
                    # Highlight depot
112
                    plt.scatter(
                                loc_x[0], loc_y[0], s=150, c="red", edgecolors="black", marker="*", zorder=3
114
                    # Plot edges
117
                    for i in range(len(tour) - 1):
118
                                plt.plot(
119
120
                                             [loc_x[tour[i]], loc_x[tour[i + 1]]],
                                             [loc_y[tour[i]], loc_y[tour[i + 1]]],
                                             zorder=1,
124
                                )
126
                     # Add node labels
                    for i in range(n):
127
                                plt.annotate(
128
                                            f"{i}\n(s={s[i]})",
129
                                             (loc_x[i], loc_y[i]),
130
                                             textcoords="offset points",
131
                                            xytext=(0, 5),
                                            ha="center",
133
                                )
134
135
                    plt.title(
136
                                f" \texttt{Orienteering Solution} \\ \texttt{NTotal Score} : \\ \texttt{\{total\_score\}}, \\ \texttt{Travel Time} : \\ \texttt{\{total\_time} : .2f\} \\ \texttt{"} \\ \texttt{(total\_time} : .2f\} \\ \texttt{"} \\ \texttt{(total\_time} : .2f\} \\ \texttt{(total\_time} : .2f] \\ \texttt{(total\_time} : .2f]
138
139
                    plt.xlabel("X coordinate")
                    plt.ylabel("Y coordinate")
140
141
                    plt.grid(True)
                    plt.tight_layout()
142
                    plt.savefig(r"d:\Visual Studio Code\Operation Research\orienteering_solution.jpg")
143
144
                    plt.show()
145 else:
print("No solution found.")
```

Listing 1: Gurobi Python Code

Optimization Output

Optimize a model with 384 rows, 439 columns and 2207 nonzeros

Model fingerprint: 0xe2154d74

Variable types: 0 continuous, 439 integer (420 binary)

Coefficient statistics:

Matrix range [1e+00, 1e+02] Objective range [1e+00, 9e+00] Bounds range [1e+00, 2e+01] RHS range [1e+00, 3e+02]

Found heuristic solution: objective 14.0000000

Presolve removed 1 rows and 21 columns

Presolve time: 0.01s

Presolved: 383 rows, 418 columns, 2202 nonzeros

Variable types: 0 continuous, 418 integer (399 binary)

Root relaxation: objective 8.997690e+01, 206 iterations, 0.01 seconds (0.00 work units)

Nodes Currer				rrent	t Node		Objective Bour		s Work		ζ
Ex	pl Une	xpl	Obj	Dept	h Int	Inf	Incumbent	${\tt BestBd}$	Gap	It/Node	Time
	0	0	89.97	690	0	31	14.00000	89.97690	543%	-	0s
Η	0	0					39.0000000	89.97690	131%	-	0s
H	0	0					44.0000000	89.97690	104%	-	0s
	0	0	87.70	828	0	33	44.00000	87.70828	99.3%	-	0s
H	0	0					46.0000000	87.58043	90.4%	-	0s
	0	0	87.58	043	0	39	46.00000	87.58043	90.4%	-	0s
	0	0	87.58	043	0	37	46.00000	87.58043	90.4%	-	0s
	0	0	86.67	193	0	48	46.00000	86.67193	88.4%	-	0s
H	0	0					50.0000000	86.58144	73.2%	-	0s
	0	0	86.58	144	0	51	50.00000	86.58144	73.2%	-	0s
	0	0	86.52	193	0	62	50.00000	86.52193	73.0%	-	0s
	0	0	86.51	788	0	58	50.00000	86.51788	73.0%	-	0s
	0	0	86.51	788	0	58	50.00000	86.51788	73.0%	-	0s
Η	0	0					56.0000000	86.51788	54.5%	-	0s
	0	0	84.13	848	0	67	56.00000	84.13848	50.2%	-	0s
	0	0	84.13	379	0	65	56.00000	84.13379	50.2%	-	0s
H	0	0					60.0000000	84.13379	40.2%	-	0s
	0	0	83.77	327	0	68	60.00000	83.77327	39.6%	-	0s
	0	0	83.15	281	0	65	60.00000	83.15281	38.6%	-	0s
	0	0	83.15	281	0	72	60.00000	83.15281	38.6%	-	0s
	0	0	83.15	281	0	74	60.00000	83.15281	38.6%	-	0s
H	0	0					65.0000000	83.15281	27.9%	-	0s
	0	0	83.15	281	0	74	65.00000	83.15281	27.9%	-	0s
H	0	0					70.000000	83.15281	18.8%	-	0s
	0	0	83.15	281	0	74	70.00000	83.15281	18.8%	-	0s
	0	0	83.15	281	0	60	70.00000	83.15281	18.8%	-	0s
	0	0	83.15	281	0	72	70.00000	83.15281	18.8%	-	0s
	0	0	83.15	281	0	67	70.00000	83.15281	18.8%	-	0s
	0	0	83.15	281	0	70	70.00000	83.15281	18.8%	-	0s
	0	0	83.15	281	0	66	70.00000	83.15281	18.8%	-	0s
	0	0	83.15	281	0	74	70.00000	83.15281	18.8%	-	0s
	0	0	83.15	281	0	74	70.00000	83.15281	18.8%	-	0s
	0	0	83.15	281	0	75	70.00000	83.15281	18.8%	-	0s
	0	0	83.15	281	0	75	70.00000	83.15281	18.8%	-	0s
	0	0	83.15	281	0	74	70.00000	83.15281	18.8%	-	0s
	0	0	83.15	281	0	74	70.00000	83.15281	18.8%	-	0s

```
70.00000
                                               83.15281 18.8%
           0
               83.15281
                           0
                               66
                                                                        0s
                                               83.15281 18.8%
     0
           0
               83.15281
                           0
                               72
                                    70.00000
                                                                        0s
                                    70.00000
                                               83.15281 18.8%
                                                                        0s
     0
           0
               83.15281
                           0
                               75
     0
               83.15281
                                    70.00000
                                               83.15281 18.8%
                                                                        0s
           0
                           0
                               75
     0
           0
               83.15281
                           0
                                    70.00000
                                               83.15281 18.8%
                                                                        0s
     0
           0
                                  72.0000000
                                               83.15281 15.5%
                                                                        0s
Η
     0
           0
               83.15281
                           0
                               75
                                    72.00000
                                               83.15281 15.5%
                                                                        0s
                                               83.11988 13.9%
           0
                                  73.0000000
                                                                        0s
Η
     0
Η
     0
           0
                                  75.0000000
                                               83.11988 10.8%
                                                                        0s
     0
           2
               83.11988
                           0
                                    75.00000
                                               83.11988 10.8%
                                                                        0s
* 2906
         653
                          19
                                  76.0000000
                                               80.58144 6.03% 14.5
                                                                        1s
```

Cutting planes:

Learned: 8
Gomory: 4
Cover: 12

Implied bound: 32

Clique: 6
MIR: 55
StrongCG: 1
Inf proof: 17
Zero half: 7
Mod-K: 2
RLT: 1

Relax-and-lift: 30

BQP: 2 PSD: 9

Explored 5205 nodes (75288 simplex iterations) in 2.07 seconds (0.61 work units) Thread count was 16 (of 16 available processors)

Solution count 10: 76 75 73 ... 46

Optimal solution found (tolerance 1.00e-04)

Best objective 7.600000000000e+01, best bound 7.60000000000e+01, gap 0.0000% Optimal tour: [0, 14, 4, 3, 12, 9, 11, 15, 10, 6, 17, 19, 2, 18, 16, 1, 8, 7, 0]

Total collected score: 84
Total travel time: 295.18