# Assignment2 Part2 Final

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# 1 Assignment 2

## 1.1 Longest Closed Path Problem

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```
[1]: from pyomo.environ import *
```

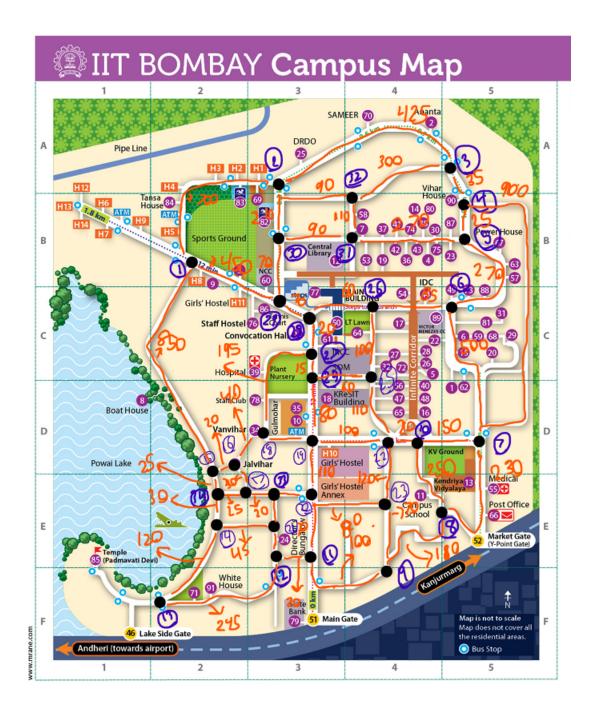
## 1.1.1 Defining the Graph

We have approximated the given map by excluding the very small white roads. The nodes in the map are landmarks, buildings and some juctions are approximated. We have included a picture of the graph for convinience.

```
[2]: V = ['1', '2', '3', '4', '5', '6', '7', '8', '9', '10', '11', '12', '13', '14', ]
      _{\hookrightarrow} '15', '16', '17', '18', '19', '20', '21', '22', '23', '24', '25', '26', '27', _{\sqcup}
      \hookrightarrow '28', '29', '30', '31', '32', '33', '34'] # Vertices in the graph
     E = \Gamma
          ('1', '2'), ('2', '1'),
          ('2', '3'), ('3', '2'),
          ('3', '4'), ('4', '3'),
          ('4', '5'), ('5', '4'),
          ('4', '7'), ('7', '4'),
          ('5', '6'), ('6', '5'),
          ('6', '7'), ('7', '6'),
          ('7', '8'), ('8', '7'),
          ('8', '9'), ('9', '8'),
          ('8', '10'), ('10', '8'),
          ('9', '11'), ('11', '9'),
          ('11', '12'), ('12', '11'),
          ('12', '13'), ('13', '12'),
          ('13', '14'), ('14', '13'),
          ('14', '34'), ('34', '14'),
          ('34', '15'), ('15', '34'),
          ('15', '1'), ('1', '15'),
          ('1', '29'), ('29', '1'),
          ('29', '28'), ('28', '29'),
          ('28', '26'), ('26', '28'),
```

```
('26', '6'), ('6', '26'),
    ('29', '30'), ('30', '29'),
    ('30', '2'), ('2', '30'),
    ('2', '32'), ('32', '2'),
    ('32', '3'), ('3', '32'),
    ('30', '31'), ('31', '30'),
    ('31', '5'), ('5', '31'),
    ('31', '32'), ('32', '31'),
    ('14', '21'), ('21', '14'),
    ('21', '12'), ('12', '21'),
    ('21', '33'), ('33', '21'),
    ('34', '17'), ('17', '34'),
    ('17', '33'), ('33', '17'),
    ('33', '20'), ('20', '33'),
    ('20', '11'), ('11', '20'),
    ('17', '16'), ('16', '17'),
    ('16', '15'), ('15', '16'),
    ('16', '18'), ('18', '16'),
    ('18', '19'), ('19', '18'),
    ('18', '27'), ('27', '18'),
    ('20', '19'), ('19', '20'),
    ('19', '22'), ('22', '19'),
    ('22', '10'), ('10', '22'),
    ('22', '23'), ('23', '22'),
    ('23', '9'), ('9', '23'),
    ('20', '23'), ('23', '20'),
    ('27', '24'), ('24', '27'),
    ('27', '28'), ('28', '27'),
    ('24', '19'), ('19', '24'),
    ('24', '25'), ('25', '24'),
    ('25', '26'), ('26', '25')
]
edge_lengths = {
    ('1', '2'): 700, ('2', '1'): 700,
    ('2', '3'): 425, ('3', '2'): 425,
    ('3', '4'): 35, ('4', '3'): 35,
    ('4', '5'): 25, ('5', '4'): 25,
    ('4', '7'): 900, ('7', '4'): 900,
    ('5', '6'): 270, ('6', '5'): 270,
    ('6', '7'): 500, ('7', '6'): 500,
    ('7', '8'): 230, ('8', '7'): 230,
    ('8', '9'): 180, ('9', '8'): 180,
    ('8', '10'): 250, ('10', '8'): 250,
    ('9', '11'): 100, ('11', '9'): 100,
    ('11', '12'): 20, ('12', '11'): 20,
    ('12', '13'): 245, ('13', '12'): 245,
```

```
('13', '14'): 120, ('14', '13'): 120,
    ('14', '34'): 30, ('34', '14'): 30,
    ('34', '15'): 25, ('15', '34'): 25,
    ('15', '1'): 850, ('1', '15'): 850,
    ('1', '29'): 450, ('29', '1'): 450,
    ('29', '28'): 30, ('28', '29'): 30,
    ('28', '26'): 60, ('26', '28'): 60,
    ('26', '6'): 85, ('6', '26'): 85,
    ('29', '30'): 70, ('30', '29'): 70,
    ('30', '2'): 210, ('2', '30'): 210,
    ('2', '32'): 90, ('32', '2'): 90,
    ('32', '3'): 300, ('3', '32'): 300,
    ('30', '31'): 90, ('31', '30'): 90,
    ('31', '5'): 270, ('5', '31'): 270,
    ('31', '32'): 110, ('32', '31'): 110,
    ('14', '21'): 45, ('21', '14'): 45,
    ('21', '12'): 30, ('12', '21'): 30,
    ('21', '33'): 40, ('33', '21'): 40,
    ('34', '17'): 25, ('17', '34'): 25,
    ('17', '33'): 20, ('33', '17'): 20,
    ('33', '20'): 23, ('20', '33'): 23,
    ('20', '11'): 70, ('11', '20'): 70,
    ('17', '16'): 30, ('16', '17'): 30,
    ('16', '15'): 20, ('15', '16'): 20,
    ('16', '18'): 40, ('18', '16'): 40,
    ('18', '19'): 17, ('19', '18'): 17,
    ('18', '27'): 195, ('27', '18'): 195,
    ('20', '19'): 110, ('19', '20'): 110,
    ('19', '22'): 100, ('22', '19'): 100,
    ('22', '10'): 20, ('10', '22'): 20,
    ('22', '23'): 120, ('23', '22'): 120,
    ('23', '9'): 70, ('9', '23'): 70,
    ('20', '23'): 80, ('23', '20'): 80,
    ('27', '24'): 15, ('24', '27'): 15,
    ('27', '28'): 20, ('28', '27'): 20,
    ('24', '19'): 80, ('19', '24'): 80,
    ('24', '25'): 60, ('25', '24'): 60,
    ('25', '26'): 100, ('26', '25'): 100
}
```



# 1.1.2 Creating a model in Pyomo and defining decision variables

```
[3]: # Create a Pyomo model
model = ConcreteModel()

model.V = Set(initialize=V)
model.E = Set(initialize=E, dimen=2)

# Parameters
```

```
model.1 = Param(model.E, initialize=edge_lengths)

# Decision variables
model.x = Var(model.E, within=Binary) # Binary variable for edge usage
model.y = Var(model.V, within=Binary) # Binary variable for vertex usage
model.u = Var(model.V, bounds=(1, len(V)), within=NonNegativeReals) # Subtour
→elimination using MTZ variables
```

## 1.1.3 Explanation

• Edge Selection Variable (x[i,j]):

This binary variable indicates whether the edge (i,j) belongs to E is part of the solution.

$$x[i,j] = \begin{cases} 1 & \text{if edge } (i,j) \text{ is selected} \\ 0 & \text{otherwise} \end{cases}$$

• Vertex Selection Variable (y[i]):

This binary variable indicates whether vertex i belongs to V is included in the cycle.

$$y[i] = \begin{cases} 1 & \text{if vertex } i \text{ is part of the cycle} \\ 0 & \text{otherwise} \end{cases}$$

• MTZ Variables (u[i]):

These continuous variables are used to prevent subtours. They define an ordering of vertices in the cycle to ensure connectivity and enforce that the selected edges form a single cycle. u[i] is constrained between 1 and |V|, the number of vertices.

## 1.1.4 Objective Function

```
[4]: # Objective function: Maximize the total length of the selected edges
def objective_rule(model):
    return sum(model.1[i, j] * model.x[i, j] for i, j in model.E)
model.obj = Objective(rule=objective_rule, sense=maximize)
```

## 1.1.5 Explanation

• Objective Function (Maximization of Total Edge Length):

The goal is to maximize the total length of the edges selected in the cycle. Each edge (i, j) has a corresponding length l[i,j], and a binary decision variable x[i,j] indicates whether the edge is included in the solution. The objective function sums the lengths of all selected edges:

$$\text{Maximize} \quad \sum_{(i,j) \in E} l[i,j] \cdot x[i,j]$$

Here, l[i,j] represents the length of the edge from vertex i to vertex j, and x[i,j] = 1 if the edge (i, j) is selected, and x[i,j] = 0 otherwise. The function seeks to maximize the total sum of the selected edge lengths, finding the longest possible cycle in the graph.

#### 1.1.6 Constraints

```
[5]: # Constraints:
     # Flow conservation: Inflow must equal outflow for each vertex
     def flow_conservation_rule(model, i):
         return sum(model.x[i, j] for j in model.V if (i, j) in model.E) == sum(model.
      \rightarrow x[j, i] for j in model.V if (j, i) in model.E)
     model.flow_conservation = Constraint(model.V, rule=flow_conservation_rule)
     # Degree constraint: Each vertex in the cycle must have exactly two edges (1 in, _
      \hookrightarrow 1 out)
     def degree_rule(model, i):
         return sum(model.x[i, j] + model.x[j, i] for j in model.V if (i, j) in model.
      \rightarrowE or (j, i) in model.E) == 2 * model.y[i]
     model.degree = Constraint(model.V, rule=degree_rule)
     # MTZ subtour elimination constraints
     def mtz_rule(model, i, j):
         if i != '1' and j != '1' and (i, j) in model.E:
             return model.u[i] - model.u[j] + len(V) * model.x[i, j] <= len(V) - 1
         return Constraint.Skip
     model.mtz = Constraint(model.V, model.V, rule=mtz_rule)
     # Ensure at least one edge is selected
     model.force_edges = Constraint(expr=sum(model.x[i, j] for i, j in model.E) >= 1)
```

## 1.1.7 Explanation

## • Flow Conservation:

This constraint ensures that the inflow equals the outflow for each vertex (i belongs to V). For each vertex i, the sum of flows on the outgoing edges is equal to the sum of flows on the incoming edges:

$$\sum_{j \in V: (i,j) \in E} x[i,j] = \sum_{j \in V: (j,i) \in E} x[j,i] \quad \forall i \in V$$

## • Degree Constraint:

Each vertex in the cycle must have exactly two edges—one incoming and one outgoing. This is ensured by requiring the sum of the binary decision variables x[i,j] (indicating edge usage) for all edges connected to vertex i to equal twice the vertex decision variable y[i]:

$$\sum_{j \in V: (i,j) \in E \text{ or } (j,i) \in E} (x[i,j] + x[j,i]) = 2 \cdot y[i] \quad \forall i \in V$$

#### • MTZ Subtour Elimination:

To prevent disconnected cycles (subtours), the MTZ formulation introduces variables u[i],

which define an order for the vertices in the tour. For each pair of distinct vertices (i, j belongs to V) where there is an edge, the following constraint is imposed:

$$u[i] - u[j] + |V| \cdot x[i, j] \le |V| - 1 \quad \forall i \ne j, (i, j) \in E$$

This ensures that no subtours can form unless the edges between i and j are selected.

## • Force Edge Selection:

This constraint ensures that at least one edge is selected in the final solution, preventing a trivial result of no edges being chosen:

$$\sum_{(i,j)\in E} x[i,j] \ge 1$$

## 1.1.8 Solving and Displaying the Results

```
[6]: # Solve the model
     solver = SolverFactory('glpk')
     result = solver.solve(model, tee=True)
    GLPSOL--GLPK LP/MIP Solver 5.0
    Parameter(s) specified in the command line:
     --write /var/folders/b9/cqzf91qd1cb5kkpcpbgg1pmm0000gn/T/tmptbjve51h.glpk.raw
     --wglp /var/folders/b9/cqzf91qd1cb5kkpcpbgg1pmm0000gn/T/tmpd0r3u8ez.glpk.glp
     --cpxlp /var/folders/b9/cqzf91qd1cb5kkpcpbgg1pmm0000gn/T/tmpw42ee7pa.pyomo.lp
    Reading problem data from
    '/var/folders/b9/cqzf91qd1cb5kkpcpbgg1pmm0000gn/T/tmpw42ee7pa.pyomo.lp'...
    /var/folders/b9/cqzf91qd1cb5kkpcpbgg1pmm0000gn/T/tmpw42ee7pa.pyomo.lp:1608:
    warning: lower bound of variable 'x2' redefined
    /var/folders/b9/cqzf91qd1cb5kkpcpbgg1pmm0000gn/T/tmpw42ee7pa.pyomo.lp:1608:
    warning: upper bound of variable 'x2' redefined
    165 rows, 169 columns, 832 non-zeros
    136 integer variables, all of which are binary
    1744 lines were read
    Writing problem data to
    '/var/folders/b9/cqzf91qd1cb5kkpcpbgg1pmm0000gn/T/tmpd0r3u8ez.glpk.glp'...
    1401 lines were written
    GLPK Integer Optimizer 5.0
    165 rows, 169 columns, 832 non-zeros
    136 integer variables, all of which are binary
    Preprocessing...
    165 rows, 169 columns, 832 non-zeros
    136 integer variables, all of which are binary
    Scaling...
     A: min|aij| = 1.000e+00 max|aij| = 3.400e+01 ratio = 3.400e+01
    GM: min|aij| = 8.900e-01 max|aij| = 1.124e+00 ratio = 1.262e+00
    EQ: min|aij| = 7.929e-01 max|aij| = 1.000e+00 ratio = 1.261e+00
    2N: \min |aij| = 5.000e-01 \max |aij| = 1.062e+00 \text{ ratio} = 2.125e+00
```

```
Constructing initial basis...
    Size of triangular part is 161
    Solving LP relaxation...
    GLPK Simplex Optimizer 5.0
    165 rows, 169 columns, 832 non-zeros
          0: obj = -0.000000000e+00 inf =
                                             1.000e+00 (1)
          2: obj =
                     3.064285714e+02 \text{ inf} = 0.000e+00 (0)
         98: obj =
                     7.672205882e+03 inf =
                                             1.092e-14 (0)
    OPTIMAL LP SOLUTION FOUND
    Integer optimization begins...
    Long-step dual simplex will be used
         98: mip =
                     not found yet <=
                                                                 (1; 0)
                                                     +inf
        535: >>>> 5.417000000e+03 <=
                                          7.631000000e+03 40.9% (31; 1)
    + 9461: >>>> 5.607000000e+03 <=
                                         7.089000000e+03 26.4% (405; 114)
    + 12535: >>>> 5.757000000e+03 <=
                                          7.002000000e+03 21.6% (508; 201)
    + 13929: >>>> 5.982000000e+03 <=
                                          6.954000000e+03 16.2% (493; 328)
    + 20828: >>>>
                   5.987000000e+03 <= 6.752000000e+03 12.8% (528; 714)
                                            tree is empty 0.0% (0; 5569)
    + 65515: mip = 5.987000000e+03 <=
    INTEGER OPTIMAL SOLUTION FOUND
    Time used:
                 1.5 secs
    Memory used: 1.8 Mb (1900297 bytes)
    Writing MIP solution to
    '/var/folders/b9/cqzf91qd1cb5kkpcpbgg1pmm0000gn/T/tmptbjve51h.glpk.raw'...
    343 lines were written
[7]: # Output the results
    print("Status:", result.solver.status)
    print("Termination condition:", result.solver.termination_condition)
     # Print the selected edges
    print("Selected edges in the longest cycle:")
    for i, j in model.E:
        if model.x[i, j].value == 1:
            print(f"Edge ({i}, {j}) with length {model.1[i, j]}")
     # Print the total length of the longest cycle
    print(f"Total length of the longest cycle: {model.obj()}")
    Status: ok
    Termination condition: optimal
    Selected edges in the longest cycle:
    Edge (2, 1) with length 700
    Edge (3, 2) with length 425
    Edge (4, 3) with length 35
    Edge (7, 4) with length 900
    Edge (5, 6) with length 270
    Edge (6, 7) with length 500
    Edge (9, 8) with length 180
```

```
Edge (8, 10) with length 250
Edge (11, 9) with length 100
Edge (12, 11) with length 20
Edge (13, 12) with length 245
Edge (14, 13) with length 120
Edge (1, 15) with length 850
Edge (28, 29) with length 30
Edge (26, 28) with length 60
Edge (29, 30) with length 70
Edge (30, 31) with length 90
Edge (31, 5) with length 270
Edge (21, 14) with length 45
Edge (33, 21) with length 40
Edge (17, 33) with length 20
Edge (16, 17) with length 30
Edge (15, 16) with length 20
Edge (19, 18) with length 17
Edge (18, 27) with length 195
Edge (20, 19) with length 110
Edge (10, 22) with length 20
Edge (22, 23) with length 120
Edge (23, 20) with length 80
Edge (27, 24) with length 15
Edge (24, 25) with length 60
```

Total length of the longest cycle: 5987.0

Edge (25, 26) with length 100

