0. Imports and Setting up Anthropic API Client

```
from google.colab import drive
drive.mount('/content/drive')
→ Mounted at /content/drive
!pip install python-dotenv
import os
import dotenv
dotenv.load_dotenv('/content/drive/MyDrive/.env')

→ Collecting python-dotenv

      Downloading python_dotenv-1.0.1-py3-none-any.whl (19 kB)
    Installing collected packages: python-dotenv
    Successfully installed python-dotenv-1.0.1
# Load Prompts and Problem Description
prompt1_path = '/content/drive/MyDrive/Thesis/Prompts/Prompt1_MathematicalModel.txt'
prompt2_path = '/content/drive/MyDrive/Thesis/Prompts/Prompt2_PyomoCode.txt'
problem_desc_path = '/content/drive/MyDrive/Thesis/ProblemDescriptions/MIP/MIP4.txt'
prompt1_file = open(prompt1_path, "r")
prompt2_file = open(prompt2_path, "r")
problem_desc_file = open(problem_desc_path, "r")
prompt1 = prompt1_file.read()
print("Prompt 1:\n", prompt1)
prompt2 = prompt2_file.read()
print("Prompt 2:\n", prompt2)
problem_desc = problem_desc_file.read()
print("Problem Description:\n", problem_desc)
→ Prompt 1:
     Please write a mathematical optimization model for this problem. Include parameters, decision variables, the objective
    Prompt 2:
     Please write a python pyomo code for this optimization problem.
    Use sample data where needed.
    Indicate where you use sample data.
    Problem Description:
     You are task with maximizing the water flow in a network of pipes over 10 time steps. The water enters the system at po
    These are the pipe capacities:
    AB: 3
    AC: 6
    AF: 1
    BC: 3
    BD: 10
    BE: 4
    CE: 4
    CF: 4
    DE: 5
    EG: 5
    FB: 12
    FG: 7
!pip install anthropic
     Show hidden output
# Importing Anthropic & Setting Headers
import anthropic
client = anthropic.Anthropic(
   # defaults to os.environ.get("ANTHROPIC_API_KEY")
    api_key=os.environ.get("ANTHROPIC_API_KEY"),
```

1. Prompt 1: Create Mathematical Model

```
message = client.messages.create(
        model="claude-3-opus-20240229",
        system=prompt1,
        max_tokens=4096
        messages=[
                {"role": "user", "content": problem_desc}
)
response_p1 = message.content[0].text
# Print response
print(response_p1)
 Based on the given information, we can formulate a mathematical optimization model as follows:
         - T: Total number of time steps (T = 10)
         - N: Total number of points in the network (N = 7, labeled A to G)
         - P: Total number of available pipes (P = 12)
         - K: Maximum number of pipes that can be activated (K = 10)
         - C[i, j]: Capacity of the pipe connecting points i and j (given in the problem statement)
         - f(t): Water inflow at point A at time step t, where f(t) = max(-4/30t^3 + t^2 - 0.234t + 3, 0)
         -x[i, j, t]: Amount of water flowing from point i to point j at time step t (continuous variable) -y[i, j]: Binary variable indicating whether the pipe connecting points i and j is activated (1 if activated, 0 otherwi
         - r[i, t]: Amount of water released out of the system at point i at time step t (continuous variable)
         Objective Function:
         Constraints:
         1. Flow conservation at each point i and time step t (except for point A at t=1 and point G):
               sum(x[j,\ i,\ t]\ for\ j\ in\ range(1,\ N+1)\ if\ (j,\ i)\ is\ a\ valid\ pipe)\ +\ r[i,\ t]\ =\ sum(x[i,\ j,\ t+1]\ for\ j\ in\ range(1,\ N+1)\ 
         2. Inflow at point A at time step t=1:
                sum(x[A, j, 1] \text{ for } j \text{ in range}(1, N+1) \text{ if } (A, j) \text{ is a valid pipe}) = f(1)
         3. Capacity constraints for activated pipes:
               x[i, j, t] \leftarrow C[i, j] * y[i, j] for all valid pipes (i, j) and t in range(1, T+1)
         4. Pipe activation constraint:
               sum(y[i, j] for all valid pipes (i, j)) <= K
         5. Binary constraints for pipe activation variables:
               y[i, j] is binary for all valid pipes (i, j)
         6. Non-negativity constraints:
               x[i, j, t] >= 0 for all valid pipes (i, j) and t in range(1, T+1) r[i, t] >= 0 for all i in range(1, N+1) and t in range(1, T+1)
         The objective function aims to maximize the total amount of water flowing into point G over all time steps. The constrai
```

2. Prompt 2: Write Code based on Mathematical Model

```
message2 = client.messages.create(
    model="claude-3-opus-20240229",
    system=prompt2,
    max_tokens=4069,
    messages=[
        {"role": "user", "content": response_p1}
)
response_p2 = message2.content[0].text
# Print response
print(response_p2)
From Here's a Python code using Pyomo to solve the optimization problem:
     ```python
 import pyomo.environ as pyo
 import numpy as np
 # Define the model
 model = pyo.ConcreteModel()
 # Define the parameters
```

```
T = 10
N = 7
P = 12
K = 10
Sample data for pipe capacities (replace with actual data)
 ('A', 'B'): 5, ('A', 'C'): 7, ('A', 'D'): 3, ('B', 'C'): 4, ('B', 'E'): 6, ('C', 'D'): 2, ('C', 'E'): 5, ('C', 'F'): 8,
 ('D', 'F'): 4,
('E', 'F'): 3, ('E', 'G'): 6,
('F', 'G'): 5
}
Define the water inflow function
def f(t):
 return max(-4/30*t**3 + t**2 - 0.234*t + 3, 0)
Define the sets
model.T = pyo.RangeSet(1, T)
model.N = pyo.RangeSet(1, N)
model.P = pyo.Set(initialize=C.keys())
Define the decision variables
model.x = pyo.Var(model.P, model.T, domain=pyo.NonNegativeReals)
model.y = pyo.Var(model.P, domain=pyo.Binary)
model.r = pyo.Var(model.N, model.T, domain=pyo.NonNegativeReals)
Define the objective function
def obj_rule(model):
 return sum(model.x['F', 'G', t] + model.x['E', 'G', t] for t in model.T)
model.obj = pyo.Objective(rule=obj_rule, sense=pyo.maximize)
Define the constraints
def flow_conservation_rule(model, i, t):
 if i == 1 or i == N:
 return pyo.Constraint.Skip
 inflow = sum(model.x[j, i, t] for j in model.N if (j, i) in model.P) outflow = sum(model.x[i, j, t+1] for j in model.N if (i, j) in model.P)
 return inflow + model.r[i, t] == outflow
model.flow_conservation = pyo.Constraint(model.N, model.T, rule=flow_conservation_rule)
def inflow_rule(model):
 return sum(model.x['A', j, 1] for j in model.N if ('A', j) in model.P) == f(1)
model.inflow = pyo.Constraint(rule=inflow_rule)
```

# 4. Input Problem Data and Test Model Code

```
Installing pyomo nd solver
!pip install -q pyomo
!apt-get install -y -qq glpk-utils
!pip install glpk
!pip install pandas

Show hidden output
```

```
import pyomo.environ as pyo
import numpy as np
Define the model
model = pyo.ConcreteModel()
Define the parameters
T = 10
N = 7
P = 12
K = 10
Sample data for pipe capacities (replace with actual data)
C = \{
 (0, 1): 3, (0, 2): 6, (0, 5): 1, (1, 2): 3,
 (1, 3): 10, (1, 4): 4, (2, 4): 4, (2, 5): 4, (3, 4): 5, (4, 6): 5, (5, 1): 12, (5, 6): 7
Define the water inflow function
def f(t):
 return max(-4/30*t**3 + t**2 - 0.234*t + 3, 0)
Define the sets
model.T = pyo.RangeSet(1, T)
model.N = pyo.RangeSet(1, N)
model.P = pyo.Set(initialize=C.keys())
Define the decision variables
model.x = pyo.Var(model.P, model.T, domain=pyo.NonNegativeReals)
model.y = pyo.Var(model.P, domain=pyo.Binary)
model.r = pyo.Var(model.N, model.T, domain=pyo.NonNegativeReals)
Define the objective function
def obj_rule(model):
 return sum(model.x[5, 6, t] + model.x[4, 6, t] for t in model.T)
model.obj = pyo.Objective(rule=obj_rule, sense=pyo.maximize)
Define the constraints
def flow_conservation_rule(model, i, t):
 if i == 1 or i == N:
 return pyo.Constraint.Skip
 inflow = sum(model.x[j, i, t] for j in model.N if (j, i) in model.P)
 outflow = sum(model.x[i, j, t+1] for j in model.N if (i, j) in model.P)
 return inflow + model.r[i, t] == outflow
model.flow_conservation = pyo.Constraint(model.N, model.T, rule=flow_conservation_rule)
def inflow_rule(model):
 return sum(model.x[1, j, 1] for j in model.N if (1, j) in model.P) == f(1)
model.inflow = pyo.Constraint(rule=inflow_rule)
def capacity_rule(model, i, j, t):
 return model.x[i, j, t] \leftarrow C[i, j] * model.y[i, j]
model.capacity = pyo.Constraint(model.P, model.T, rule=capacity_rule)
def pipe_activation_rule(model):
 return sum(model.y[i, j] for (i, j) in model.P) <= K
model.pipe_activation = pyo.Constraint(rule=pipe_activation_rule)
Solve the model
solver = pyo.SolverFactory('glpk')
results = solver.solve(model)
Print the results
print("Objective value:", pyo.value(model.obj))
for (i, j) in model.P:
 if pyo.value(model.y[i, j]) > 0.5:
 print(f"Pipe {i}-{j} activated")
 for t in model.T:
 print(f" Time {t}: Flow = {pyo.value(model.x[i, j, t])}")
```

```
ERROR:pyomo.core:Rule failed when generating expression for Constraint flow_conservation with index (2, 10): KeyError: "Index '(2, 4, 11)' is not valid for indexed component 'x'"
 ERROR:pyomo.core:Constructing component 'flow_conservation' from data=None failed:
 KeyError: "Index '(2, 4, 11)' is not valid for indexed component 'x'"
 KeyError
 Traceback (most recent call last)
 <ipython-input-12-0909a747c7e8> in <cell line: 46>()
 outflow = sum(model.x[i, j, t+1]) for j in model.N if (i, j) in model.P) return inflow + model.r[i, t] == outflow
 44
 45
 -> 46 model.flow_conservation = pyo.Constraint(model.N, model.T, rule=flow_conservation_rule)
 47
 48 def inflow_rule(model):
 - 3 8 frames -
 /usr/local/lib/python3.10/dist-packages/pyomo/core/base/indexed_component.py in _validate_index(self, idx)
 864
 # Raise an exception
 865
 raise KeyError(
 "Index '%s' is not valid for indexed component '%s'"
 --> 866
 867
 % (normalized_idx, self.name)
 868
 KeyError: "Index '(2, 4, 11)' is not valid for indexed component 'x'"
```

### 5. Correct The Model Code to Test Mathematical Model (if applicable)

```
import pyomo.environ as pyo
import numpy as np
Define the model
model = pyo.ConcreteModel()
Define the parameters
T = 10
N = 7
P = 12
K = 10
Sample data for pipe capacities (replace with actual data)
 (0, 1): 3, (0, 2): 6, (0, 5): 1, (1, 2): 3,
 (1, 3): 10, (1, 4): 4, (2, 4): 4, (2, 5): 4,
 (3, 4): 5, (4, 6): 5, (5, 1): 12, (5, 6): 7
}
Define the water inflow function
def f(t):
 return max(-4/30*t**3 + t**2 - 0.234*t + 3, 0)
Define the sets
model.T = pyo.RangeSet(1, T)
model.N = pyo.RangeSet(1, N)
model.P = pyo.Set(initialize=C.keys())
Define the decision variables
model.x = pyo.Var(model.P, model.T, domain=pyo.NonNegativeReals)
model.y = pyo.Var(model.P, domain=pyo.Binary)
model.r = pyo.Var(model.N, model.T, domain=pyo.NonNegativeReals)
Define the objective function
def obi rule(model):
 return sum(model.x[5, 6, t] + model.x[4, 6, t] for t in model.T)
model.obj = pyo.Objective(rule=obj_rule, sense=pyo.maximize)
Define the constraints
def flow_conservation_rule(model, i, t):
 if i == 1 or i == N or t == T: # MODIFIED TO INCLUDE t == 10 SO CODE IS CONSISTENT WITH MATH MODEL
 return pyo.Constraint.Skip
 inflow = sum(model.x[j, i, t] for j in model.N if (j, i) in model.P)
 outflow = sum(model.x[i, j, t+1]) for j in model.N if (i, j) in model.P)
 return inflow + model.r[i, t] == outflow
model.flow_conservation = pyo.Constraint(model.N, model.T, rule=flow_conservation_rule)
def inflow rule(model):
 return sum(model.x[1, j, 1] for j in model.N if (1, j) in model.P) == f(1)
model.inflow = pyo.Constraint(rule=inflow_rule)
def capacity_rule(model, i, j, t):
 return model.x[i, j, t] \leftarrow C[i, j] * model.y[i, j]
model.capacity = pyo.Constraint(model.P, model.T, rule=capacity_rule)
```

```
def pipe_activation_rule(model):
 return sum(model.y[i, j] for (i, j) in model.P) <= K
model.pipe_activation = pyo.Constraint(rule=pipe_activation_rule)
Solve the model
solver = pyo.SolverFactory('glpk')
results = solver.solve(model)
Print the results
print("Objective value:", pyo.value(model.obj))
for (i, j) in model.P:
 if pyo.value(model.y[i, j]) > 0.5:
 print(f"Pipe {i}-{j} activated")
 for t in model.T:
 print(f" Time {t}: Flow = {pyo.value(model.x[i, j, t])}")
ERROR:pyomo.common.numeric_types:evaluating object as numeric value: x[5,6,1] (object: <class 'pyomo.core.base.var.VarData'>)
 No value for uninitialized NumericValue object x[5,6,1]
 ERROR:pyomo.common.numeric_types:evaluating object as numeric value: obj
 (object: <class 'pyomo.core.base.objective.ScalarObjective'>)
 No value for uninitialized NumericValue object x[5,6,1]
 Traceback (most recent call last)
 <ipython-input-18-18d7bd1e2f82> in <cell line: 65>()
```