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/NL/NL2.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 a chemist in charge of a process that requires two liquids to reach the end product. Your goal is to maximize t
!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

```
# Print response
print(response_p1)
🚁 Let's formulate the mathematical optimization model for the given problem. We'll define the parameters, decision variabl
     - L: Maximum capacity of the mixture tank (80 l)
    Decision Variables:
     - x1: Litres of the first liquid component used
     - X2: Litres of the second liquid component used
    Maximize the quality of the product, which is measured by the difference between the Napierian logarithm of the product
    Maximize: ln(x_1 \times x_2) - \sqrt[3]{(x_1 + 2x_2)}
    Constraints:
    1. Stability constraint: The squared difference between the litres employed of each component and the mean litres utilis
        (x_1 - (x_1 + x_2) / 2)^2 + (x_2 - (x_1 + x_2) / 2)^2 \le 10
    2. Capacity constraint: The total litres of both components should not exceed the mixture tank's capacity (80 l).
    3. Non-negativity constraints: The litres used for each component should be non-negative.
        X_2 \ge 0
    The complete mathematical optimization model:
    Maximize:
     ln(x_1 \times x_2) - \sqrt[3]{(x_1 + 2x_2)}
    Subject to:
     (x_1 - (x_1 + x_2) / 2)^2 + (x_2 - (x_1 + x_2) / 2)^2 \le 10
    X_1 + X_2 \leq 80
    X_1 \ge 0
    X_2 \geq 0
    This model aims to maximize the quality of the product while satisfying the stability and capacity constraints, ensuring
```

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)
Here's a Python code using Pyomo to solve the optimization problem:
    ```python
 from pyomo.environ import *
 from math import log, sqrt
 # Create a Pyomo model
 model = ConcreteModel()
 # Define decision variables
 model.x1 = Var(domain=NonNegativeReals)
 model.x2 = Var(domain=NonNegativeReals)
 # Define the objective function
 def objective_rule(model):
 return log(model.x1 * model.x2) - (model.x1 + 2 * model.x2)**(1/3)
 model.objective = Objective(rule=objective_rule, sense=maximize)
 # Define constraints
 def stability_constraint_rule(model):
 mean_litres = (model.x1 + model.x2) / 2
 return (model.x1 - mean_litres)**2 + (model.x2 - mean_litres)**2 <= 10
 model.stability_constraint = Constraint(rule=stability_constraint_rule)
```

```
def capacity_constraint_rule(model):
 return model.x1 + model.x2 <= 80
model.capacity_constraint = Constraint(rule=capacity_constraint_rule)
Solve the optimization problem
solver = SolverFactory('ipopt') # Use the IPOPT solver
solver.solve(model)
Print the results
print("Optimal solution:")
print(f"x1 = {value(model.x1):.2f} litres")
print(f"x2 = {value(model.x2):.2f} litres")
print(f"Objective value: {value(model.objective):.2f}")
In this code, we:
1. Import the necessary Pyomo components and mathematical functions.
Create a Pyomo `ConcreteModel`.

 Define the decision variables `x1` and `x2` as non-negative real numbers.
 Define the objective function using the `objective_rule` and add it to the model with `sense=maximize`.

5. Define the stability constraint using the `stability_constraint_rule` and add it to the model.6. Define the capacity constraint using the `capacity_constraint_rule` and add it to the model.
7. Solve the optimization problem using the IPOPT solver (you may need to install it separately).
8. Print the optimal solution values for x1, x2, and the objective value.
Note: In this code, we don't use any sample data as the problem is self-contained and doesn't require external data. The
```

To run this code, make sure you have Pyomo and the IPOPT solver installed in your Python environment. You can install Py

## 4. Input Problem Data and Test Model Code

```
Installing pyomo nd solver
!pip install -q pyomo
!wget -N -q "https://matematica.unipv.it/gualandi/solvers/ipopt-linux64.zip"
!unzip -o -q ipopt-linux64
 - 12.8/12.8 MB 31.1 MB/s eta 0:00:00
\overline{2}
 ---- 49.6/49.6 kB 5.3 MB/s eta 0:00:00
from pyomo.environ import *
import math # MODIFIED IMPORT SO CODE EXECUTES
Create a Pvomo model
model = ConcreteModel()
Define decision variables
model.x1 = Var(domain=NonNegativeReals)
model.x2 = Var(domain=NonNegativeReals)
Define the objective function
def objective rule(model):
 return log(model.x1 * model.x2) - (model.x1 + 2 * model.x2)**(1/3)
model.objective = Objective(rule=objective_rule, sense=maximize)
Define constraints
def stability_constraint_rule(model):
 mean_litres = (model.x1 + model.x2) / 2
 return (model.x1 - mean_litres)**2 + (model.x2 - mean_litres)**2 <= 10
model.stability_constraint = Constraint(rule=stability_constraint_rule)
def capacity_constraint_rule(model):
 return model.x1 + model.x2 <= 80
model.capacity_constraint = Constraint(rule=capacity_constraint_rule)
Solve the optimization problem
solver = SolverFactory('ipopt') # Use the IPOPT solver
solver.solve(model)
Print the results
print("Optimal solution:")
print(f"x1 = {value(model.x1):.2f} litres")
print(f"x2 = {value(model.x2):.2f} litres")
print(f"Objective value: {value(model.objective):.2f}")
```

Optimal solution: x1 = 42.24 litres x2 = 37.76 litres Objective value: 2.47

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