# **Assignment 1**

## **Team Information**

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## Link to the Product

• The product is available at: GitHub

• We prepared Jupyter Notebook tests.ipynb that contains 8 test cases for the Simplex method.

## **Programming Language**

• Programming Language: Python

## **Linear Programming Problem**

#### Problem 1

• Maximization or Minimization? Maximization

• Objective Function: Maximize

$$Z = 40x_1 + 30x_2$$

• Constraint Functions:

$$x_1 + x_2 \le 12 \ 2x_1 + x_2 \le 16$$

#### Input

•  $\mathbf{C} = [40, 30]$ 

•  $\mathbf{A} = [[1, 1], [2, 1]]$ 

• **b** = [12, 16]

• Accuracy = 0.1

#### Output

· Maximum value of

at 
$$x_1=4, x_2=5$$

## Problem 2

• Maximization or Minimization? Maximization

• Objective Function: Maximize

$$Z = 2x_1 + 5x_2$$

Z = 33

• Constraint Functions:

$$x_1 + 4x_2 \le 24 \ 3x_1 + 1x_2 \le 21 \ x_1 + x_2 \le 9$$

#### Input

- **C** = [2, 5]
- $\mathbf{A} = [[1, 4], [3, 1], [1, 1]]$
- $\mathbf{b} = [24, 21, 9]$
- **Accuracy** = 0.5

#### Output

• Maximum value of

at

$$Z = 33$$

 $x_1 = 4, x_2 = 5$ 

#### Problem 3

- Maximization or Minimization? Maximization
- Objective Function: Maximize

$$Z = x_1 + 2x_2 + 3x_3$$

• Constraint Functions:

$$x_1 + x_2 + x_3 \le 12 \ 2x_1 + x_2 + 3x_3 \le 18$$

#### Input

- $\mathbf{C} = [1, 2, 3]$
- $\mathbf{A} = [[1, 1, 1], [2, 1, 3]]$
- **b** = [12, 18]
- **Accuracy** = 0.7

#### Output

· Maximum value of

at

$$Z=27$$

 $x_1 = 0, x_2 = 9, x_3 = 3$ 

### Problem 4

- Maximization or Minimization? Maximization
- Objective Function: Maximize

$$Z = 9x_1 + 10x_2 + 16x_3$$

• Constraint Functions:

$$18x_1 + 15x_2 + 12x_3 \le 360$$
  
 $6x_1 + 4x_2 + 8x_3 \le 192$   
 $5x_1 + 3x_2 + 3x_3 \le 180$ 

#### Input

- $\mathbf{C} = [9, 10, 16]$
- $\mathbf{A} = [[18, 15, 12], [6, 4, 8], [5, 3, 3]]$
- **b** = [360, 192, 180]
- Accuracy = 0.00001

#### Output

· Maximum value of

$$Z = 400$$

at

$$x_1 = 0, x_2 = 8, x_3 = 20$$

#### Problem 5

- Maximization or Minimization? Maximization
- Objective Function: Maximize

$$Z = 6x_1 + 2x_2 + 2.5x_3 + 4x_4$$

• Constraint Functions:

$$5x_1+x_2+2x_4 \leq 1000 \ 4x_1+2x_2+2x_3+x_4 \leq 600 \ x_1+2x_3+x_4 \leq 150$$

#### Input

- $\mathbf{C} = [6, 2, 2.5, 4]$
- $\mathbf{A} = [[5, 1, 0, 2], [4, 2, 2, 1], [1, 0, 2, 1]]$
- **b** = [1000, 600, 150]
- Accuracy = 0.00001

#### Output

• Maximum value of

$$Z=1050$$
 
$$x_1=0, x_2=225, x_3=0, x_4=150$$

at

## Problem 6

- Maximization or Minimization? Not Applicable
- Objective Function: Maximize

$$Z = 4x_1 + 5x_2 + 4x_3$$

• Constraint Functions:

$$2x_1 + 3x_2 - 6x_3 \le 240$$
  
 $4x_1 + 2x_2 - 4x_3 \le 200$   
 $4x_1 + 6x_2 - 8x_3 \le 160$ 

#### Input

- $\mathbf{C} = [4, 5, 4]$
- $\mathbf{A} = [[2, 3, -6], [4, 2, -4], [4, 6, -8]]$
- **b** = [240, 200, 160]
- Accuracy = 0.001

#### Output

• The problem is unsolvable!

#### Problem 7

- Maximization or Minimization? Minimization
- Objective Function: Minimize

$$Z = -x_1 - x_2$$

• Constraint Functions:

$$x_1 + x_2 \leq 1 \ -x_1 - x_2 \leq -3$$

#### Input

- $\mathbf{C} = [-1, -1]$
- A = [[1, 1], [-1, -1]]
- **b** = [1, -3]
- Accuracy = 0.001

#### Output

• The method is not applicable!

#### Problem 8

- Maximization or Minimization? Maximization
- Objective Function: Maximize

$$Z = 2x_1 + x_2$$

- Constraint Functions:

$$-x_1+x_2\geq 1$$

#### Input

- **C** = [2, 1]
- **A** = [[-1, 1]]
- b = [1]
- Accuracy = 0.001

#### Output

• The problem is unsolvable!

## Setup and Run

```
cd assignment_1 python main.py
```

## Code

simplex.py

```
from typing import List, Tuple
import numpy as np

class Simplex:

The Simplex class implements the Simplex method for solving linear programm

def __init__(
    self, C: List[float], A: List[float], b: List[float], accuracy: float
) -> None:

Initializes the Simplex method with the following inputs:
    C: Coefficients of the objective function.
```

```
A: Coefficients of the inequality constraints.
  b: Right-hand side values of the inequality constraints.
  accuracy: Precision for detecting optimality (helps handle floating-point erro
  self.C_coef = np.array(C) # Objective function coefficients
  self.A_coef = np.array(A) # Coefficients of the constraints
  self.b_coef = np.array(b) # Right-hand side values
  self.accuracy = accuracy # Desired accuracy
  self.table = None # Simplex table
  self.optimised = False # Indicates whether the solution is optimized
  self.solvable = True # Indicates whether the problem is solvable
def check_infeasibility(self) -> bool:
  If any value in the right-hand side vector b is negative
  and the corresponding row in the matrix A has no positive coefficients,
  the problem is infeasible.
  for i in range(len(self.b_coef)):
    if self.b_coef[i] < 0 and all(</pre>
       self.A_coef[i][j] <= 0 for j in range(len(self.A_coef[i]))
    ):
       return True
  return False
def check_unboudedness(self, ratios: np.ndarray) -> bool:
  If the objective function can grow indefinitely in the direction
  of the feasible region, then the problem is unbounded.
  if np.all(np.isinf(ratios)):
    print("The problem is unsolvable!")
    self.solvable = False
    return True
  return False
def fill_initial_table(self) -> None:
```

```
Initializes the Simplex table by combining the constraint matrix A,
  the identity matrix (for slack variables), and the right-hand side vector b.
  Also appends the objective function row with negative coefficients of C.
  self.table = np.hstack(
       self.A_coef, # Coefficients of the constraints
       np.eye(self.A_coef.shape[0]), # Identity matrix for slack variables
       np.reshape(self.b_coef, (-1, 1)), # Right-hand side vector b
    )
  )
  # Objective function row (negative coefficients of C)
  func = np.hstack((-self.C_coef, np.zeros(self.A_coef.shape[0] + 1)))
  # Add the objective function row at the bottom
  self.table = np.vstack((self.table, func))
def make_iteration(self) -> None:
  .....
  Performs one iteration of the Simplex algorithm:
  1. Finds the pivot column.
  2. Checks for unboundedness.
  3. Performs the pivot operation to transform the table.
  if self.table is None:
    print("Table was not initialized!")
    return
  # Find the most negative value in the objective row
  pivot_column = np.argmin(self.table[-1, :-1])
  # Check if the solution is already optimal
  if self.table[-1, :-1][pivot_column] >= -self.accuracy:
    self.optimised = True
    return
  # Compute the ratios for the ratio test
  ratios = np.divide(
```

```
self.table[:-1, -1], # Right-hand side values (b)
    self.table[:-1, pivot_column], # Pivot column values
    out=np.full_like(
       self.table[:-1, -1], np.inf
    ), # Fill with inf where division is not valid
    where=self.table[:-1, pivot_column]
    > 0, # Only consider positive entries in the pivot column
  )
  if self.check_unboudedness(ratios):
    return
  # Select the pivot row
  pivot_row = np.argmin(ratios)
  # Normalize the pivot row
  self.table[pivot_row] = (
    self.table[pivot_row] / self.table[pivot_row][pivot_column]
  # Make all other elements in the pivot column zero
  for row in range(self.table.shape[0]):
    if row != pivot_row:
       self.table[row] = (
         self.table[row]
         - self.table[row][pivot_column] * self.table[pivot_row]
       )
def get_solution(self) -> Tuple[List[float], float]:
  Returns the decision variables and the optimized objective function value if t
  # Check if the problem is infeasible
  if self.check_infeasibility():
    print("The method is not applicable!")
    self.solvable = False
  # Perform iterations while the solution is not optimized
  while (not self.optimised) and self.solvable:
```

```
self.make_iteration()
# If the problem is unsolvable, return empty results
if not self.solvable:
  return ∏, None
# Initialize solution array (size of decision variables + slack variables)
solution = np.zeros(self.C_coef.shape[0] + self.A_coef.shape[0])
for row in range(self.A_coef.shape[0]):
  # Find the column index in this row where the value is 1
  for col in range(self.C_coef.shape[0] + self.A_coef.shape[0]):
    # Check if this column is a basic variable
    if self.table[row, col] == 1 and np.sum(self.table[:, col]) == 1:
       # This is a basic variable column
       solution[col] = self.table[row, -1]
       break # Move to the next row
# Extract decision variables from the solution
decision_vars = solution[: self.C_coef.shape[0]]
# Round to 10 decimal places
decision_vars = [round(var, 10) for var in decision_vars]
max_value = round(self.table[-1, -1], 10)
return decision_vars, max_value
```

main.py

```
from src.simplex import Simplex

command = ""

while command.lower() != "end":
    print("What a nice day to solve optimization with simplex! (enter end to finish)")
    print("Enter function coeficients: ")
    command = input()
    if command.lower() == "end":
```

```
break
try:
  function_row = list(map(float, command.split(" ")))
except Exception:
  print("Invalid function coefficients. Please, try again.")
  break
print("Enter number of constraints and then coeficients of constraints: ")
try:
  n = int(input())
  constraint_coef = □
  for _ in range(n):
    constraint_coef.append(list(map(float, input().split(" "))))
except Exception:
  print("Invalid constraints. Please, try again.")
print("Enter right hand side: ")
try:
  rhs = list(map(float, input().split()))
except Exception:
  print("Invalid right-hand side coefficients. Please, try again.")
  break
print("Enter accuracy: ")
  acc = float(input())
except Exception:
  print("Invalid accuracy value. Please, try again.")
  break
try:
  simplex = Simplex(function_row, constraint_coef, rhs, acc)
  simplex.fill_initial_table()
  answer, max_value = simplex.get_solution()
except Exception:
  print("You entered invalid problem. Please, try again.")
  break
print("Solution: ")
for i in range(len(answer)):
  print(f"x{i + 1} = {answer[i]}")
```

print("Max value: ")
print(max\_value)