

ORF307_HW6

March 23, 2022

ORF307 Homework 6

Due: Friday, April 1, 2022 9:00 pm ET

- Please export your code with output as pdf.
- If there is any additional answers, please combine them as **ONE** pdf file before submitting to the Gradescope.

Question 1

Let A be a given matrix. Show that exactly one of the following alternatives must hold.

- (a) There exists some $x \neq 0$ such that $Ax = 0, x \geq 0$.
- (b) There exists some y such that $A^T y > 0$.

Question 2

An alternative to the phase-I/phase-II method for solving the LP

$$\begin{aligned} &\text{minimize} && c^T x \\ &\text{subject to} && Ax = b, \\ &&& x \geq 0 \end{aligned} \tag{1}$$

is the “big-M”-method, in which we solve the auxiliary problem

$$\begin{aligned} &\text{minimize} && c^T x + M\mathbf{1}^T z \\ &\text{subject to} && Ax + z = b \\ &&& x \geq 0, z \geq 0, \end{aligned} \tag{2}$$

where $M > 0$ is a parameter and $z \in \mathbf{R}^m$ is an auxiliary variable. Here, $A \in \mathbf{R}^{m \times n}$, $c \in \mathbf{R}^n$, and $b \in \mathbf{R}^m$. Note that this auxiliary problem has an initial basic feasible solution $(x, z) = (0, b) \geq 0$.

(a) Derive the dual LP of (2).

(b) Prove the following property:

If $M > -y_i^*$ for $i = 1, \dots, m$, where y^* is an optimal solution of the dual of (1), then the optimal z in (2) is zero, and therefore the optimal x in (2) is also an optimal solution of (1).

Hint: Use complementary slackness.

Question 3

Consider the following LP:

$$\begin{aligned} & \text{minimize} && 13x_1 + 10x_2 + 6x_3 \\ & \text{subject to} && 5x_1 + x_2 + 3x_3 = 8 \\ & && 3x_1 + x_2 = 3 \\ & && x_1, x_2, x_3 \geq 0. \end{aligned} \tag{3}$$

- (a) Solve it using the big-M formulation as in Q3 obtaining optimal primal and dual variables (use the provided function).
- (b) Derive the dual LP for (3)
- (c) Solve the dual using CVXPY and compare the optimal primal-dual variables with the ones from (a).

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[2]: '''
This code is provided to help with question 4.
This code returns optimal primal variables x
and dual variables y.
'''

import cvxpy as cp
import numpy as np
import numpy.linalg as la

def simplex_iteration(x, B, problem):
    """Perform one simplex iteration given
    - basic feasible solution x
    - basis B

    It returns new x, new basis, new dual variable,
    and termination flag (true/false)
    """
    A, b, c = problem['A'], problem['b'], problem['c']
    m, n = A.shape
    A_B = A[:, B]

    # Compute reduced cost vector
    p = la.solve(A_B.T, c[B])
    c_bar = c - A.T @ p

    # Check optimality
    if np.all(c_bar >= 0):
        print("Optimal solution found!")
        return x, B, -p, True

    # Choose j such that c_bar < 0 (first one)
    j = np.where(c_bar < 0)[0][0]
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# Compute search direction d
d = np.zeros(n)
d[j] = 1
d[B] = la.solve(A_B, -A[:, j])

# Check for unboundedness
if np.all(d >= 0):
    print("Unbounded problem!")
    return None, None, True

# Compute step length theta
d_i = np.where(d[B] < 0)[0]
theta = np.min(- x[B[d_i]] / d[B[d_i]])
i = B[d_i[np.argmin(- x[B[d_i]] / d[B[d_i]])]]

# Compute next point
x_next = x + theta * d

# Compute next basis
B_next = B
B_next[np.where(B == i)[0]] = j

return x_next, B_next, -p, False

def simplex_algorithm(x, B, problem, max_iter=1000):
    """Run simplex algorithm"""

    for k in range(max_iter):

        x, B, y, end = simplex_iteration(x, B, problem)

        if end:
            break
    return x, B, y

```