

Ques Use the method of Lagrange multipliers to solve the problem.

$$\text{Min}_{x_1, x_2} x_1^2 + x_2^2, \text{ s.t. } \begin{cases} x_1 \geq 1 \\ x_2 \geq 2 \end{cases}$$

Ans. we rewrite the problem as follows.

$$L(x_1, x_2, \alpha_1, \alpha_2) = x_1^2 + x_2^2 - \alpha_1(x_1 - 1) - \alpha_2(x_2 - 2), \quad \alpha_1, \alpha_2$$

$$\max_{\alpha \geq 0} \min_x L(x, \alpha) \quad \nabla L(x, \alpha) = \begin{bmatrix} 2x_1 - \alpha_1 \\ 2x_2 - \alpha_2 \end{bmatrix} = 0$$

$$\begin{aligned} 2x_1 - \alpha_1 &= 0 & \frac{\alpha_1}{2} &= x_1 & \text{sign} \\ 2x_2 - \alpha_2 &= 0 & \frac{\alpha_2}{2} &= x_2 \end{aligned}$$

$$L(x_1, x_2, \alpha_1, \alpha_2) = \frac{\alpha_1^2}{4} + \frac{\alpha_2^2}{4} - \alpha_1\left(\frac{\alpha_1}{2} - 1\right) - \alpha_2\left(\frac{\alpha_2}{2} - 2\right)$$

$$\max_{\alpha_1, \alpha_2 \geq 0} -\frac{\alpha_1^2}{4} + \alpha_1 - \frac{\alpha_2^2}{4} + 2\alpha_2 \Rightarrow \alpha_1 = 4, \alpha_2 = 4$$

$$\Rightarrow x_1 = 1 \quad x_2 = 2$$

$$\text{So } x = 5 \quad \checkmark$$

2'. Min $x_1^2 + x_2^2$ subject to $x_1 + x_2 \geq 5$.

a) Express the problem in the matrix form as in 11.20 $C = [1 1]$

b) $\begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}, \quad x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \text{ AS } \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \quad b = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$
 $A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$

$$[1 1] \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = 5$$

c) $\begin{bmatrix} A & C^T \\ C & 0 \end{bmatrix} = \begin{bmatrix} 2 & 0 & 1 \\ 0 & 2 & 1 \\ 1 & 1 & 0 \end{bmatrix}, \quad \begin{bmatrix} b \\ c \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 5 \end{bmatrix}$ So constraints
 Soln is

d) $\begin{bmatrix} 2 & 0 & 1 \\ 0 & 2 & 1 \\ 1 & 1 & 0 \end{bmatrix}^{-1} \begin{bmatrix} 0 \\ 0 \\ 5 \end{bmatrix} = \begin{bmatrix} x_1 \\ x_2 \\ x \end{bmatrix} = \begin{bmatrix} 2.5 \\ 2.5 \\ -5 \end{bmatrix}$ $x_1 \neq x_2 = 2.5$
 $\lambda^* = 5$

↓

$$2.5 + 2.5 = 5 \checkmark$$

$$\begin{bmatrix} \frac{1}{2} & -\frac{1}{4} & \frac{1}{2} \\ -\frac{1}{4} & \frac{1}{4} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} & -1 \end{bmatrix}$$

11.3: So for the constraint line, it is the only tangent line for $c=12.5$ with $x_1 = x_2 = 2.5$ thus an optimal solution to the problem. ✓

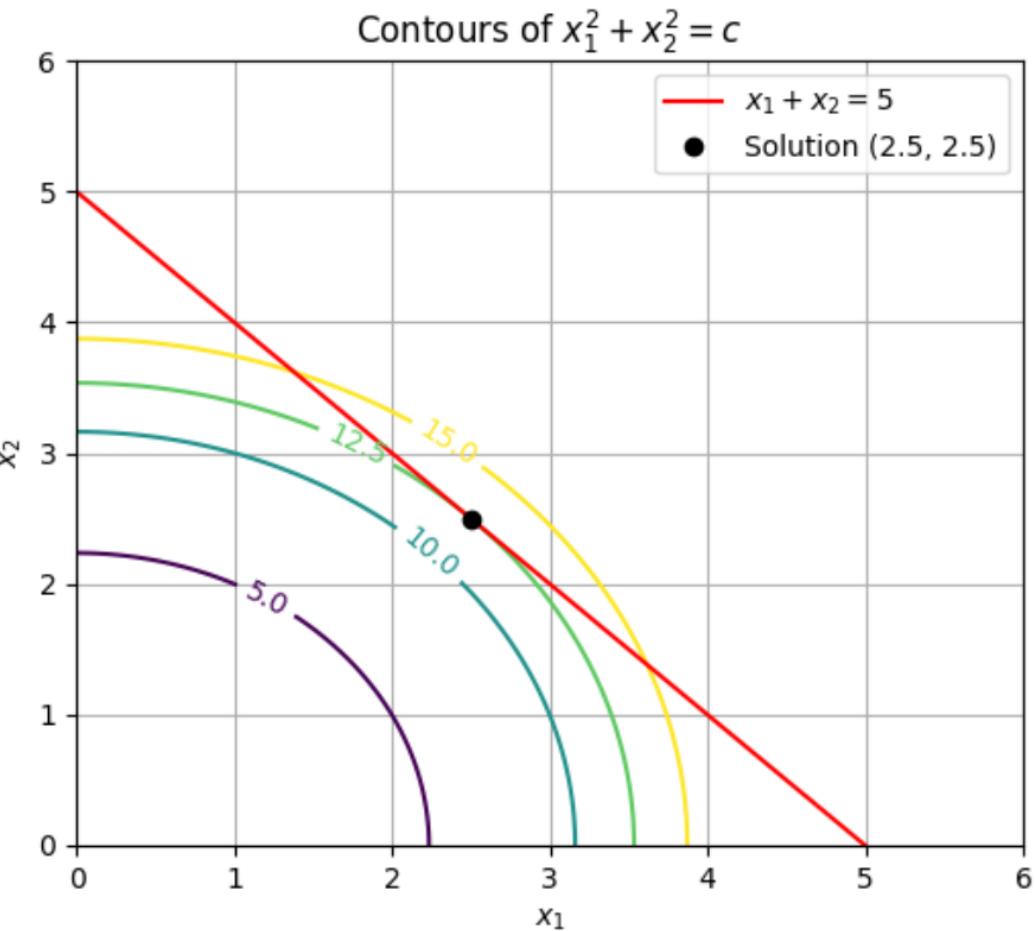
11.4: The given solution you gave is incorrect as it does not satisfy the constraints.

a) diag $\begin{bmatrix} 4 & 0 & 0 \\ 0 & 5 & 0 \\ 0 & 0 & 2 \end{bmatrix} \rightarrow A = \begin{bmatrix} 8 & -3 & 2 \\ -3 & 10 & 3 \\ 2 & 3 & 4 \end{bmatrix} \quad C = \begin{bmatrix} 2 & -1 & -3 \\ 1 & -3 & 3 \end{bmatrix}$ Rest is done in code.

b) $\begin{bmatrix} x_1 & x_2 & x_3 \end{bmatrix} \begin{bmatrix} 4 & -1.5 & 1 \\ -1.5 & 5 & 1.5 \\ 1 & 1.5 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} + \boxed{\begin{bmatrix} 1 & 1 & 1 \end{bmatrix}} \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}, \quad C = \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}$, $\begin{bmatrix} 2 & -1 & -3 \\ 1 & -3 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$

Numerical Analysis HW 11

11.3)



```
import numpy as np
import matplotlib.pyplot as plt

# (a) Define a grid for x1, x2
x1_vals = np.linspace(0, 6, 200)
x2_vals = np.linspace(0, 6, 200)
X1, X2 = np.meshgrid(x1_vals, x2_vals)
Z = X1**2 + X2**2 # This is x1^2 + x2^2

# We'll plot contours for c in [5, 10, 12.5, 15]
contour_levels = [5, 10, 12.5, 15]

plt.figure(figsize=(6,5))
CS = plt.contour(X1, X2, Z, levels=contour_levels, cmap='viridis')
plt.clabel(CS, inline=True, fontsize=10)
plt.title(r"Contours of $x_1^2 + x_2^2 = c$")
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# (b) Constraint line: x1 + x2 = 5 => x2 = 5 - x1
x1_line = np.linspace(0, 5, 200)
x2_line = 5 - x1_line
plt.plot(x1_line, x2_line, 'r-', label=r"x\u2081 + x\u2082 = 5")

# (c) The solution to minimize x1^2 + x2^2 subject to x1 + x2 = 5
# occurs at x1 = x2 = 2.5 (by symmetry or by Lagrange multipliers).
x_star, y_star = 2.5, 2.5
plt.plot(x_star, y_star, 'ko', label=f"Solution ({x_star}, {y_star})")

plt.xlabel(r"x\u2081")
plt.ylabel(r"x\u2082")
plt.legend()
plt.grid(True)
plt.show()

```

D). So the minimum is $(x_1, x_2) = (2.5, 2.5)$ for a $c=12.5$. Geometrically, the smallest cirecle centered at the origin that touches the line $x_1 + x_2 = 5$ is tangent at $(2.5, 2.5)$ which is the solution.

11.4:

```
A:  
[[ 4. -1.5  1. ]  
[-1.5  5.  1.5]  
[ 1.  1.5  2. ]]  
b:  
[-2  0  0]  
c:  
[[ 2 -1 -3]  
[ 1 -3  3]]  
c:  
[ 1 -1]  
**(b) Lagrangian Formulation On Paper  
KKT Matrix =  
[[ 8. -3.  2.  2.  1.]  
[-3. 10.  3. -1. -3.]  
[ 2.  3.  4. -3.  3.]  
[ 2. -1. -3.  0.  0.]  
[ 1. -3.  3.  0.  0.]]  
Right-hand side (RHS) =  
[ 2  0  0  1 -1]  
  
CONSTRAINT CHECK:  1.0  
CONSTRAINT CHECK:  -1.0  
  
Optimal x: [ 0.2972973   0.22297297 -0.20945946]  
Lagrange multipliers λ: [0.28378378  0.14189189]  
Objective value f(x): 0.2263513513513514  
  
import numpy as np  
# Step 1: Define A, c  
Q = np.array([
```

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[ 4.0, -1.5,  1.0],
[-1.5,  5.0,  1.5],
[ 1.0,  1.5,  2.0]
])
A = 2 * Q # so that 1/2 x^T A x = x^T Q x
c = np.array([-2,0,0])

# Step 2: Define C, d
C = np.array([
    [ 2, -1, -3],
    [ 1, -3,  3]
])
d = np.array([1, -1])

# Dimensions
n = 3 # number of variables
m = 2 # number of constraints

# Step 3: Build the KKT matrix
# K = [ A   C^T ]
#      [ C   0   ]
top_block = np.hstack([A, C.T])
bottom_block = np.hstack([C, np.zeros((m, m))])
K = np.vstack([top_block, bottom_block])

# Step 4: Build the RHS = [ -c, d ]^T = [0, d]^T since c=0
rhs = np.concatenate([-c, d])

# Step 5: Solve for (x, lambda)
solution = np.linalg.solve(K, rhs)
x_opt = solution[:n]
lambda_opt = solution[n:]

# Step 6: Compute the objective f(x) = x^T Q x = 1/2 x^T A x
f_opt = x_opt @ Q @ x_opt # or 0.5 * x_opt @ A @ x_opt

print("\nA:\n", Q)
print("b:\n", c)
print("C:\n", C)
print("c:\n", d)

print("** (b) Lagrangian Formulation On Paper")

print("KKT Matrix =\n", K)
print("Right-hand side (RHS) =\n", rhs)

# E: Check if your solution satisfies the equality constraints
print("\nCONSTRAINT CHECK: ", np.dot(C[0], x_opt))
print("CONSTRAINT CHECK: ", np.dot(C[1], x_opt))

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```
# Print results
print("\nOptimal x:", x_opt)
print("Lagrange multipliers λ:", lambda_opt)
print("Objective value f(x):", f_opt)
```