## Importing the necessary libraries

In [8]: import itertools
import sympy as sym

# Generalized n variable Template Function for any given mathematical problem with given constraints

**Generalized Optimizer - Lagrangian Solver -** All feasible solutions are given along with local minima/maxima

## Input:

1. Minimize/Maximize Objective Function

$$f(x_1, x_2, \ldots, x_n)$$

2. Equality Constraints -

$$h_i(x_1, x_2, \dots, x_n) = 0 \quad i = 1, 2, \dots, p$$

3. Inequality Constraints

$$g_j(x_1,x_2,\ldots,x_n) \leq 0 \quad j=1,2,\ldots,m$$

## **Solver Variables:**

**1.** Lagrange Function -  $L(f, x, \lambda, h, \mu, g, s)$ 

$$L(f,x,\lambda,h,\mu,g,s) = f(x) + \sum_{i=1}^p \lambda_i h_i(x) + \sum_{j=1}^m \mu_i (g_j(x) + s_j^2)$$

2. Lagrange Multipliers -  $\lambda_i, \mu_i, s_i$ 

$$\mu_j \geq 0, s_j \geq 0 \quad j=1,2,\ldots,m$$

# **Necessary Conditions (Stationarity):**

1. Lagrange Equation

$$rac{\partial L}{\partial x_k} = 0 \quad k = 1, 2, \dots, n$$

2. KKT Conditions

$$rac{\partial L}{\partial \lambda_i} = 0 \quad i = 1, 2, \dots, p$$

$$\frac{\partial L}{\partial \mu_j} = 0 \quad j = 1, 2, \dots, m$$

$$rac{\partial L}{\partial s_j} = 0 \quad j = 1, 2, \dots, m$$

## **Feasibility Conditions:**

$$s_j^2 \geq 0 \quad j=1,2,\ldots,m$$

$$g_j(x) \leq 0 \quad j = 1, 2, \dots, m$$

# **Switching/Orthogonality Conditions:**

$$\mu_j \times g_j(x) = 0$$
  $j = 1, 2, \ldots, m$ 

## **Sufficient Conditions:**

$$\exists \lambda^* = [\lambda_1^*, \lambda_2^*, \dots, \lambda_p^*] \quad \exists \mu^* = [\mu_1^*, \mu_2^*, \dots, \mu_m^*]$$

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```
In [59]: def solve lagrangian(f, h, g, side constraints):
             1.1.1
             This function solves the Constrained Optimization Problem using the Lagr
             f : sympy expression
             h : list of sympy expressions (Equality Constraints)
             g : list of sympy expressions (Inequality Constraints)
             side constraints : list of sympy expressions (Side Constraints)
             # Type check if f, h, g are sympy expressions
             if not isinstance(f, sym.Expr) or not isinstance(h, list) or not isinsta
                 raise Exception("f, h, g must be sympy expressions")
             # Get Free Symbols from f, h, g
             symbols = list(set(f.free symbols).union(*[set(h i.free symbols) for h i
             # Find number of equality constraints
             p = len(h)
             # Creating Lambda matrix
             l = sym.Matrix([sym.symbols(f"l {i}") for i in range(p)])
```

```
# Find number of inequality constraints
m = len(q)
# Creating Mu matrix
u = sym.Matrix([sym.symbols(f"u_{j}") for j in range(m)])
# Creating S matrix
s = sym.Matrix([sym.symbols(f"s_{j}") for j in range(m)])
# Writing Lagrangian
L = f + sum([l[i]*h[i] for i in range(p)]) + sum([u[j]*(g[j] + s[j]**2)f
# Finding the partial derivatives of the Lagrangian
L x = sym.Matrix([sym.diff(L, x) for x in symbols])
# Finding Hessians of f, every h i, and every g j
H f = sym.hessian(f, symbols)
H h = [sym.hessian(h i, symbols) for h i in h]
H g = [sym.hessian(g j, symbols) for g j in g]
# Now we need Switching Conditions
# We will have 2<sup>m</sup> cases
# First we need to find all possible combinations of m binary variables
# We will use 0 to represent m i = 0 and 1 to represent m i > 0
# We will use 0 to represent g i = 0 and 1 to represent g i < 0
# All possible combinations of m binary variables
all combinations = list(itertools.product([0, 1], repeat=m))
# Defining cases and their checks
cases = []
case checks = []
for combination in all combinations:
    # 0 represents m i = 0 i.e. q i < 0 and 1 represents m i > 0 i.e. q
    curr case = []
    curr checks = []
    for i in range(m):
        if combination[i] == 0:
            curr case.append(sym.Eq(u[i], 0))
            curr checks.append(g[i] < 0)</pre>
        else:
            curr case.append(sym.Eq(g[i], 0))
            curr checks.append(u[i] > 0)
    # Adding side constraints
    curr checks += side constraints
    # Append Lagrandian partial derivative equations
    curr case += [sym.Eq(L x[i], 0) for i in range(len(symbols))]
    # Add h i = 0 equations
    curr case += [sym.Eq(h i, 0) for h i in h]
```

```
# Adding the case and its checks
    cases.append(curr case)
    case checks.append(curr checks)
# Let's solve all the cases and check if they are feasible
solutions = []
symbols to solve = symbols.copy()
symbols to solve.extend(l)
symbols_to_solve.extend(u)
for i in range(len(cases)):
    # Solving the current case
    curr solution = sym.solve(cases[i], symbols to solve)
    # Check if curr solution is a list of tuples
    if isinstance(curr solution, list):
        # There can be multiple solutions
        # We will try all the solutions and check if they are feasible
        for solution in curr solution:
            # Flag to check if solution is a real number
            flag = True
            # Converting the solution to a dictionary
            sol dict = {}
            for j in range(len(symbols_to_solve)):
                # Checking if the solution[j] is a Complex Number
                if sym.im(solution[j]) != 0:
                    flag = False
                    break
                sol_dict[symbols_to_solve[j]] = solution[j]
            if not flag:
                solutions.append(None)
                continue
            # Checking if the current solution is feasible
            curr checks res = [check.subs(sol dict) for check in case ch
            solutions.append(solution if all(curr checks res) else None)
    else:
        # Single solution
        # Flag to check if solution is a real number
        flag = True
        # Curr soln is a dictionary
        for sol in curr solution.values():
            if sym.im(sol) != 0:
```

```
flag = False
                break
        if not flag:
            solutions.append(None)
            continue
        curr checks res = [check.subs(curr solution) for check in case of
        solutions.append(curr solution if all(curr checks res) else None
# Now we need to check if the solutions are local minima or local maxima
# We will use the following conditions
# H f + sum(l i*H h i) + sum(u j*H g j) is positive semidefinite
for i in range(len(solutions)):
    if solutions[i] is not None:
        \# Substituting the solution in the Hessian and adding all the He
        R = H f.copy()
        for i in range(p):
            R += l[i]*H h[i]
        for j in range(m):
            R += u[j]*H_g[j]
        R = R.subs(solutions[i])
        # Checking if the Hessian is positive semidefinite
        solutions[i]["type"] = "max" if not R.is positive semidefinite ε
        # Add F-value to the solution
        solutions[i]["F"] = f.subs(solutions[i])
# Return Non-None Solutions
return [solution for solution in solutions if solution is not None]
```

# **Example 1:**

```
Minimize - f(x_1,x_2)=x_1^2+x_2^2-4x_1-6x_2 subject to x_1+x_2-2\leq 0 and 2x_1+3x_2-12\leq 0 and x_1\geq 0 and x_2\geq 0
```

```
In [60]: x1, x2 = sym.symbols("x1 x2")
f = x1**2 + x2**2 - 4*x1 - 6*x2
g = [x1 + x2 - 2, 2*x1 + 3*x2 - 12]
h = []
side_constraints = [x1 >= 0, x2 >= 0]
solve_lagrangian(f, h, g, side_constraints)
```

```
Out[60]: [{x2: 3/2, x1: 1/2, u_0: 3, u_1: 0, 'type': 'min', 'F': -17/2}]
```

# **Example 2:**

```
Minimize - f(x_1,x_2)=(x_1-2.5)^2+(x_2-2.5)^2 subject to 2x_1+2x_2-3\leq 0 and x_1\geq 0 and x_2\geq 0
```

## **Example 3:**

Minimize -  $f(x_1,x_2)=(x_1-2.5)^2+(x_2-2.5)^2$  subject to  $2x_1+2x_2=3$  and  $x_1\geq 0$  and  $x_2\geq 0$ 

```
In [62]: x1, x2 = sym.symbols("x1 x2")
f = (x1 - 2.5)**2 + (x2 - 2.5)**2
g = []
h = [2*x1 + 2*x2 - 3]
side_constraints = [x1 >= 0, x2 >= 0]
solve_lagrangian(f, h, g, side_constraints)
```

## **Tierce-1 Exam Problem 2:**

```
Minimize - f(x_1,x_2,x_3)=x_1^2+2x_1x_2+3x_2^2+4x_2x_3+x_3^2-6x_3
```

```
In [65]: x1, x2, x3 = sym.symbols("x1 x2 x3")
    f = x1**2 + 2*x1*x2 + 3*x2**2 + 4*x2*x3 + x3**2 - 6*x3
    g = []
    h = []
    side_constraints = []

solve_lagrangian(f, h, g, side_constraints)
```

```
Out[65]: [{x3: -3, x2: 3, x1: -3, 'type': 'max', 'F': 9}]
```

### **Tierce-1 Exam Problem 3:**

**Minimize** -  $f(x_1,x_2)=x_1^2+2x_2^2-3x_1-6x_2$  subject to  $x_1+x_2\leq 3$  and  $x_1+3x_2\leq 10$  and  $x_2\geq 0$ 

```
In [66]: x1, x2 = sym.symbols("x1 x2")
    f = x1**2 + 2*x2**2 - 3*x1 - 6*x2
    g = [x1 + x2 - 3, x1 + 3*x2 - 10]
    h = []
    side_constraints = [x2 >= 0]

solve_lagrangian(f, h, g, side_constraints)
```

Out[66]: []

If you wish to test more examples, you can input you functions f, g, h and run it on this code.

# Solving Example 1 Step-by-Step to show the execution of the code

#### **Problem Statement:**

Write a code to minimise  $f(x)=x_1^2+x_2^2-4x_1-6x_2$  and subject to the constraints  $x_1+x_2\leq 2$  and  $2x_1+3x_2\leq 12$  and  $x_1\geq 0$  and  $x_2\geq 0$ .

#### Solution:

We will use the **Lagrange Multiplier Method** to solve this problem.

**Step 1:** We will first define the function f(x) and the constraints g(x) and h(x).

**Step 2:** We will then define the Lagrangian function  $L(x, \lambda, \mu)$ .

**Step 3:** We will then find the partial derivatives of the Lagrangian function with respect to  $x_1$ ,  $x_2$  and put them equal to 0 to get some conditions.

**Step-4:** We will create switching conditions using the equation:  $\mu_i g_i(x) = 0$ .

**Step-5:** Solve all switch cases using Constraints, and another inequality which tells  $\mu_j \geq 0$ .

Step-6: We will then compare solutions of all switch cases and find the optimal solution.

#### Code

**Step-1:** Define the function f(x) and the constraints g(x) and h(x).

In [12]: 
$$x1$$
,  $x2$ ,  $l1$ ,  $l2$ ,  $m1$ ,  $m2$  =  $sym.symbols('x1 x2 l1 l2 m1 m2')
 $f = x1**2 + x2**2 - 4*x1 - 6*x2$   
 $g_1 = x1 + x2 - 2$   
 $g_2 = 2*x1 + 3*x2 - 12$$ 

**Step-2:** Define the Lagrangian function  $L(x, \lambda, \mu)$ .

$$L(x,\lambda,\mu)=f(x)+\sum_{i=1}^p \lambda_i h_i(x)+\sum_{j=1}^m \mu_i(g_j(x)+s_j^2)$$

$$L(x,\lambda,\mu) = x_1^2 + x_2^2 - 4x_1 - 6x_2 + \mu_1(x_1 + x_2 - 2 + s_1^2) + \mu_2(2x_1 + 3x_2 - 12 + s_2^2)$$

(We do not need to consider  $\lambda$  as we do not have equality constraints.)

(Also we do not need to code this function as we will be using the partial derivatives of this function.)

**Step-3:** Find the partial derivatives of the Lagrangian function with respect to  $x_1$ ,  $x_2$  and put them equal to 0 to get some conditions.

$$\frac{\partial L}{\partial x_1} = 2x_1 - 4 + \mu_1 + 2\mu_2 = 0$$

$$\frac{\partial L}{\partial x_2} = 2x_2 - 6 + \mu_1 + 3\mu_2 = 0$$

Out[13]: 
$$(m1 + 2*m2 + 2*x1 - 4, m1 + 3*m2 + 2*x2 - 6)$$

**Step-4:** Create switching conditions using the equation:  $\mu_j g_j(x) = 0$ .

$$\mu_1g_1(x)=0$$
 and  $\mu_2g_2(x)=0$ 

**CASE 1:** 
$$\mu_1 = 0$$
 ( $g_1(x) < 0$ ) and  $\mu_2 = 0$  ( $g_2(x) < 0$ )

CASE 2: 
$$\mu_1 = 0$$
  $(g_1(x) < 0)$  and  $g_2(x) = 0$   $(\mu_2 > 0)$ 

CASE 3: 
$$g_1(x)=0$$
 ( $\mu_1>0$ ) and  $\mu_2=0$  ( $g_2(x)<0$ )

CASE 4: 
$$g_1(x)=0$$
 ( $\mu_1>0$ ) and  $g_2(x)=0$  ( $\mu_2>0$ )

```
In [14]: # Switching Conditions
          case 1 = [sym.Eq(m1, 0), sym.Eq(m2, 0)]
          case 1 checks = [g 1 < 0, g 2 < 0, x1 >= 0, x2 >= 0]
          case_2 = [sym.Eq(m1, 0), sym.Eq(g_2, 0)]
          case_2_checks = [g_1 < 0, m2 > 0, x1 >= 0, x2 >= 0]
          case 3 = [sym.Eq(g 1, 0), sym.Eq(m2, 0)]
          case 3 checks = [m1 > 0, q 2 < 0, x1 >= 0, x2 >= 0]
          case_4 = [sym.Eq(g_1, 0), sym.Eq(g_2, 0)]
          case 4 checks = [m1 > 0, m2 > 0, x1 >= 0, x2 >= 0]
          case 1, case 1 checks, case 2, case 2 checks, case 3, case 3 checks, case 4,
Out[14]: ([Eq(m1, 0), Eq(m2, 0)],
          [x1 + x2 - 2 < 0, 2*x1 + 3*x2 - 12 < 0, x1 >= 0, x2 >= 0],
           [Eq(m1, 0), Eq(2*x1 + 3*x2 - 12, 0)],
           [x1 + x2 - 2 < 0, m2 > 0, x1 >= 0, x2 >= 0],
           [Eq(x1 + x2 - 2, 0), Eq(m2, 0)],
           [m1 > 0, 2*x1 + 3*x2 - 12 < 0, x1 >= 0, x2 >= 0],
           [Eq(x1 + x2 - 2, 0), Eq(2*x1 + 3*x2 - 12, 0)],
           [m1 > 0, m2 > 0, x1 >= 0, x2 >= 0])
          Step-5: Solve all switch cases using Constraints, and another inequality which tells \mu_i \geq 0.
         CASE 1:
         \mu_1 = 0 (g_1(x) < 0) and \mu_2 = 0 (g_2(x) < 0)
                                       2x_1 + 3x_2 - 12 = 0
                                         x_1 + x_2 - 2 = 0
                                             x_1 \geq 0
                                             x_2 > 0
                                             \mu_1 \geq 0
                                             \mu_2 \geq 0
In [15]: # Case 1
          case 1 constraints = [sym.Eq(L x1, 0), sym.Eq(L x2, 0)] + case 1
          case_1_solution = sym.solve(case_1_constraints, [x1, x2, m1, m2])
         case 1 solution
Out[15]: {x1: 2, x2: 3, m1: 0, m2: 0}
In [16]: # Perform checks
          case 1 checks res = [check.subs(case 1 solution) for check in case 1 checks]
          if all(case 1 checks res):
              print('Case 1 is valid')
```

```
else:

print('Case 1 is invalid')

Case 1 is invalid

Case 1 is not feasible

CASE 2:

u_1 = 0 (a_1(x) < 0) and a_2(x) = 0 (u_2 > 0)
```

$$\mu_1=0$$
  $(g_1(x)<0)$  and  $g_2(x)=0$   $(\mu_2>0)$  
$$2x_1+3x_2-12=0$$
 
$$x_1+x_2-2=0$$
 
$$x_1\geq 0$$
 
$$x_2\geq 0$$
 
$$\mu_1\geq 0$$
 
$$\mu_2\geq 0$$

```
In [17]: # Case 2
    case_2_constraints = [sym.Eq(L_x1, 0), sym.Eq(L_x2, 0)] + case_2

# Solve Case 2 and find x1, x2, m1, m2
    case_2_solution = sym.solve(case_2_constraints, [x1, x2, m1, m2])
    case_2_solution
```

```
Out[17]: {x1: 24/13, x2: 36/13, m1: 0, m2: 2/13}
```

```
In [18]: # Perform checks for Case 2
    case_2_checks_res = [check.subs(case_2_solution) for check in case_2_checks]

if all(case_2_checks_res):
    print('Case 2 is valid')
else:
    print('Case 2 is invalid')
```

Case 2 is invalid

#### Case 2 is not feasible

### CASE 3:

$$g_1(x)=0$$
 ( $\mu_1>0$ ) and  $\mu_2=0$  ( $g_2(x)<0$ ) 
$$2x_1+3x_2-12=0$$
 
$$x_1+x_2-2=0$$
 
$$x_1\geq 0$$
  $x_2\geq 0$ 

$$\mu_1 \geq 0$$
 $\mu_2 \geq 0$ 

```
In [19]: # Case 3
         case_3_constraints = [sym.Eq(L_x1, 0), sym.Eq(L_x2, 0)] + case_3
         # Solve Case 3 and find x1, x2, m1, m2
         case 3 solution = sym.solve(case 3 constraints, [x1, x2, m1, m2])
         case 3 solution
```

Out[19]: {x1: 1/2, x2: 3/2, m1: 3, m2: 0}

```
In [20]: # Perform checks for Case 3
         case 3 checks res = [check.subs(case 3 solution) for check in case 3 checks]
         if all(case 3 checks res):
             print('Case 3 is valid')
         else:
             print('Case 3 is invalid')
```

Case 3 is valid

Since Case 3 is feasible, we will solve it.

Finding the function value at  $x_1, x_2 = (\frac{1}{2}, \frac{3}{2})$ 

```
In [21]: f val = f.subs(case 3 solution)
         f val
```

Out[21]:  $-\frac{17}{2}$ 

The function value is -8.5

Checking for local minima/maxima.

To check this, we will use the following condition:

$$egin{aligned} igtriangledown^2 f(x) + \mu_1 igtriangledown^2 g_1(x) + \mu_2 igtriangledown^2 g_2(x) \geq 0 \end{aligned}$$

```
In [22]: # Checking if the solution is a local minimum or a local maximum
         F = sym.hessian(f, [x1, x2])
         G1 = sym.hessian(g 1, [x1, x2])
         G2 = sym.hessian(g 2, [x1, x2])
         R = F + m1*G1 + m2*G2
         R.subs(case 3 solution)
         R
```

```
Out[22]: \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}
In [23]: F, G1, G2, R
Out[23]: (Matrix([
            [2, 0],
            [0, 2]]),
            Matrix([
            [0, 0],
            [0, 0]]),
            Matrix([
            [0, 0],
            [0, 0]]),
            Matrix([
            [2, 0],
            [0, 2]]))
In [24]: if sym.Matrix(R.subs(case 3 solution)).is positive semidefinite:
               print('Case 3 is a local minimum')
           else:
               print('Case 3 is a local maximum')
```

Case 3 is a local minimum

#### The solution is a local minimum.

We cannot say that this is the true minimum value of this function using this method because we have not checked for all the possible solutions till now.

#### CASE 4:

$$g_1(x)=0$$
 ( $\mu_1>0$ ) and  $g_2(x)=0$  ( $\mu_2>0$ )  $2x_1+3x_2-12=0$   $x_1+x_2-2=0$   $x_1\geq 0$   $x_2\geq 0$   $\mu_1\geq 0$ 

```
In [25]: # Case 4
    case_4_constraints = [sym.Eq(L_x1, 0), sym.Eq(L_x2, 0)] + case_4

# Solve Case 4 and find x1, x2, m1, m2
    case_4_solution = sym.solve(case_4_constraints, [x1, x2, m1, m2])
    case_4_solution
```

```
Out[25]: {x1: -6, x2: 8, m1: 68, m2: -26}
```

```
In [26]: # Perform checks for Case 4
    case_4_checks_res = [check.subs(case_4_solution) for check in case_4_checks]

if False in case_4_checks_res:
    print('Case 4 is not feasible')
else:
    print('Case 4 is feasible')
```

Case 4 is not feasible

#### Case 4 is not feasible

**Step-6:** Find the true minimum value returned by this function by checking all the possible solutions.

Since, we have gotten only one feasible situation, and that solution is a local minimum, we can say that this is the true minimum value of this function.

Hence, the true minimum value of this function is -8.5 at  $x_1,x_2=(\frac{1}{2},\frac{3}{2})$