

# **Project Report**

Basics & Applications of AI/ML for Process Systems Engineering

Submitted by: -

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# Question:-1

$$A \stackrel{k_1}{\to} B$$
$$A \stackrel{k_2}{\to} C$$

Rate equations:

$$\frac{dC_A}{dt} = -(k_1C_A + k_2C_A)$$
$$\frac{dC_B}{dt} = k_1C_A$$
$$\frac{dC_C}{dt} = k_2C_A$$

# **Steps:**

- Importing the numpy and pandas library
  - import numpy as np
  - import pandas as pd
- ❖ Loading the Data excel file and printing the data
  - data = pd.read\_excel("C:\\Users\\ASUD\\Downloads\\data.xlsx")
  - > data.shape
- ❖ Loading the t\_span, concentration of A,B,C from the loaded data and ptinting
  - > t\_span = np.array(data.iloc[:,0])
  - > Ca=np.array(data.iloc[:,1])
  - > Cb=np.array(data.iloc[:,2])
  - Cc=np.array(data.iloc[:,3])

```
❖ Defining the ODEs function
```

```
➤ def odes(t,y,*K):
    dydt=np.zeros(3)
    Ca=y[0]
    Cb=y[1]
    Cc=y[2]
    k1=K[0]
    k2=K[1]
    dydt[0]=-(k1*Ca + k2*Ca)
    dydt[1]=k1*Ca
    dydt[2]=k2*Ca
```

#### **❖** Importing the library

- > from scipy.optimize import minimize
- > import math

return (dydt)

- from scipy.integrate import solve\_ivp
- ➤ import matplotlib.pyplot as plt

# ❖ Defining the objective function for finding the RMSE

```
def object_func(K):
    y0 = [1,0,0]
    t_span = np.array([0,1])
    t_point= np.linspace(0,1,21)

    soln = solve_ivp(odes,t_span,y0,t_eval=t_point, args=(K))
    y= soln.y.T

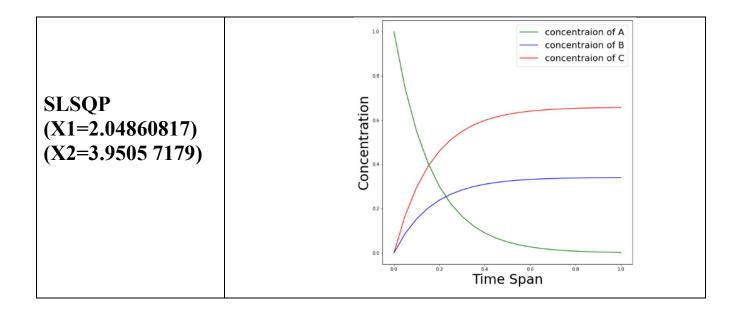
    sum1=0
    sum2=0
    sum3=0
```

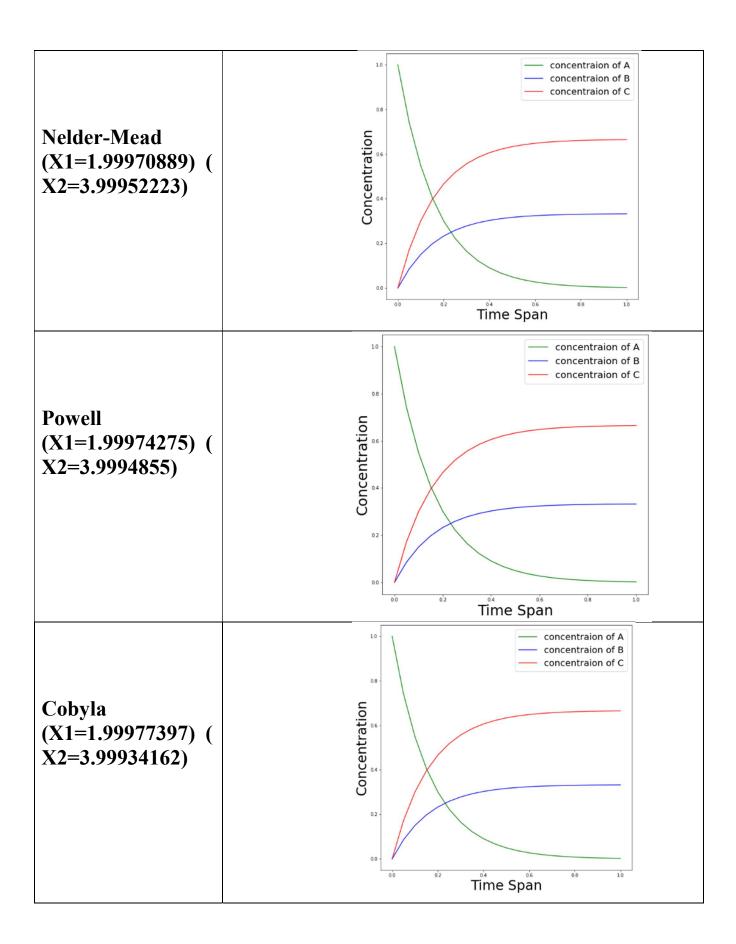
```
for i in np.linspace(0,1,21):
          i=int(i)
          sum1=sum1+((Ca[i])-(y[i,0]))**2
          sum2=sum2+((Cb[i])-(y[i,1]))**2
  >
>
          sum3 = sum3 + ((Cc[i]) - (v[i,2]))**2
       rmse1=((math.sqrt((sum1)))/20)
       rmse2=((math.sqrt((sum2)))/20)
  rmse3 = ((math.sqrt((sum3)))/20)
       rmse=(rmse1+rmse2+rmse3)/3
       return (rmse)
❖ Minimizing the objective function by estimating K1 & K2 by using BFGS
               Initial guess of K1 & K2 K0=[0.1,2]
   method
  > soln1= minimize(object func, K0, method='TNC')
Solving the ODEs function by the solve ivp solver
 \triangleright K=soln1.x
    Initial guess of Y1, Y2, Y3 as v0 = [1,0,0]
    t span = np.array([0,1])
   t point= np.linspace(0,1,21)
   soln = solve ivp(odes,t span,y0,t eval=t point,args=(K))
   y = soln.y.T
❖ Plotting the graph between time and concentration of the A, B,C
 > t=soln.t
    A=soln.y[0]
    B=soln.y[1]
    C=soln.y[2]
   plt.figure(figsize=(10,10))
```

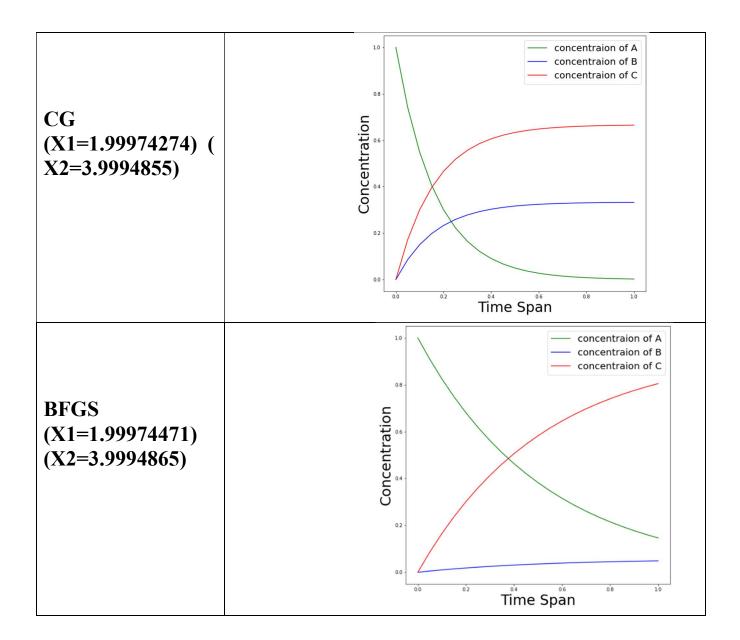
```
plt.plot(t,A,label='concentraion of A',color='green')
plt.plot(t,B,label='concentraion of B',color='blue')
plt.plot(t,C,label='concentraion of C',color='red')
plt.xlabel('Time Span',fontsize=30)
plt.ylabel('Concentration',fontsize=30)
plt.legend(fontsize=20)
plt.show()
```

# **Observation for the Different optimization Algorithms (Methods)**

After applying different method, it is observed that there is not much change in the value of the X1 and X2







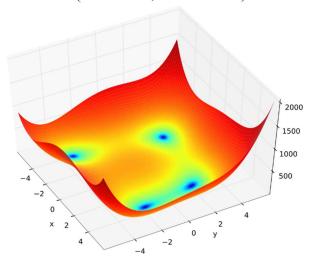
# **Question:-2**

# The given test function is Himmelblau

$$f(x,y) = (x^2 + y - 11)^2 + (x + y^2 - 7)^2$$

The Himmelblau Function has four identical local minimum at:

- f(x\*)=0 at x\*=(3,2)
- f(x\*)=0 at x\*=(-2.805118, 3.283186)
- f(x\*)=0 at x\*=(-3.779310, -3.283186)
- f(x\*)=0 at x\*=(3.584458, -1.848126)



The given Constraint is 
$$x1**2 + x2**2 \ge 25$$
  $-5 \le x1$ ,  $x2 \le 5$ 

Python Implementation of Himmelblau Function

$$((x[0]**2+x[1]-11)**2+(x[0]+x[1]**2-7)**2$$

# SLSQP Optimization Algorithm

#### Steps:-

- ❖ Defining the Himmelblau function
  - $\triangleright$  def fun(x):

```
return((x[0]**2+x[1]-11)**2+(x[0]+x[1]**2-7)**2)
```

- Importing scipy.optimize
  - > import scipy.optimize as optimize
- Giving the constraint of the function
  - $\triangleright$  constraint=({'type':'ineq','fun':lambda x: x[0]\*\*2+x[1]\*\*2-25})
- ❖ Providing the Upper and Lower bounds of the X1 & X2
  - $\rightarrow$  bounds=((-5,10),(-10,5))
- ❖ Minimizing fun with initial conditions & using SLSQP Algorithm
  - res=optimize.minimize(fun,(4,3),method='SLSQP',bounds=bounds,constraints=constraint)
- ❖ Optimize X1 & X2 value
  - > res.x

$$X1=-3.77931004$$
  $X2=-3.28318593$ 

- ❖ Optimize Fun value
  - ➤ Res.fun
  - > fun value= 2.5289416192976698e-12

#### > Genetic Algorithm

# **Steps:-**

- Importing the library
  - > from pymoo.core.problem import Problem
  - > import numpy as np
- Defining the function and constraint and upper ,lower bound of the X1 and X2
- > class G1(Problem):

n 
$$var1 = 2$$
 # Number of variable is 2 i.e. X1 & X2

$$xlo = np.array([-5, -10], dtype=float) # lower bound of x1=-5, X2=-10$$

$$xu = np.array([10, 5], dtype=float)$$
 # Upper bound of X1=10,X2=5

def \_evaluate(self, x, out, \*args, \*\*kwargs):

$$x1 = x[:, 0]$$

$$x2 = x[:, 1]$$

$$f = (x1**2+x2-11)**2+(x1+x2**2-7)**2$$

# Constraints

$$g1 = -x1**2-x2**2+25$$

$$out["F"] = f$$

$$out["G"] = [g1]$$

- Importing the Library
- > from pymoo.algorithms.soo.nonconvex.ga import GA
- > from pymoo.optimize import minimize

```
▶ problem =G1()
algorithm = GA(
pop_size=200,
eliminate_duplicates=True)
res = minimize(problem,
algorithm,
seed=1,
verbose=False)
❖ Printing the value of X1 & X2
▶ res.X
X1=-3.77930249 X2=-3.28315084
❖ Printing the function value
```

function value=5.03080946e-08

res.fun

# Conclusion:- Both the algorithm gives the approximately same value

<u>SLSQP</u>	<u>X1=-</u> 3.77931004	<u>X2=</u> =-3.28318593	<u>Fun=</u> 2.5289416192976698e-12
<u>GA</u>	<u>X1=-</u> 3.77930249	<u>X2=-</u> 3.28315084	<u>Fun=</u> 5.03080946e-08