scipy-fed-batch

May 21, 2024

```
[151]: # %pip install openpyxl
      import pandas as pd
      import numpy as np
      from matplotlib import pyplot as plt
      import scipy
      import numpy as np
      from scipy.integrate import solve_ivp
      from scipy.optimize import minimize
      from sklearn.metrics import r2_score
      from scipy.optimize import differential_evolution, minimize
[152]: df = pd.read_excel("ANN_amox/Semibat_convertido_clean.xlsx")
      df.columns = ___
        →['Ester_t', 'Ester_gl', 'Ester_mM', 'Amox_t', 'Amox_gl', 'Amox_mM', 'APA_t', 'APA_gl', 'APA_mM', 'PO
      df
[152]:
             Ester_t Ester_gl
                                  Ester_mM
                                                         Amox_gl
                                                                    Amox_mM \
                                                Amox_t
      0
            9.859813
                      6.133333 34.040034
                                             14.719626 0.883436
                                                                   2.417722
            20.017523
                      5.866667
                                 32.560033
                                             29.906542 1.262440
                                                                   3.454954
      1
      2
           29.827687
                      5.733333 31.820032
                                             39.953271 1.546012
                                                                   4.231013
      3
           39.986371
                     5.511111
                                30.586697
                                             59.929907
                                                       2.110429
                                                                   5.775669
      4
                     5.288889
                                 29.353363
           50.145055
                                             85.864486 2.797546
                                                                   7.656119
      5
           59.973715 6.000000 33.300033
                                            103.037383 3.067485
                                                                   8.394868
      6
           70.132399 5.777778
                                 32.066699
                                            112.149533 3.042945
                                                                   8.327709
      7
           80.292056 5.600000 31.080031
                                            141.939252 3.558282
                                                                   9.738047
      8
           90.103193 5.511111
                                30.586697
                                            155.257009 3.975460
                                                                  10.879749
      9
          100.261877 5.288889
                                 29.353363 181.191589 4.073620
                                                                  11.148384
      10
          110.091511 6.044444
                                33.546700
                                            191.004673
                                                       4.368098
                                                                  11.954292
      11
          120.251168 5.866667
                                32.560033 213.084112 4.736196
                                                                  12.961676
      12
          130.061332 5.733333
                                31.820032
                                            223.247664
                                                       5.030675
                                                                  13.767583
      13
          140.221963 5.600000
                                 31.080031
                                            266.004673
                                                                  15.043603
                                                       5.496933
                                                                  16.185305
          150.031153
                       5.422222
                                30.093363
                                            276.168224
                                                        5.914110
      15
          160.210280
                      6.133333
                                34.040034
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                      5.955556 33.053366
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          170.019470
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          180.180101
                      5.822222
                                32.313366
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          189.989291 5.644444 31.326698
      18
                                                   NaN
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          200.149922 5.511111 30.586697
                                                   NaN
                                                             NaN
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```

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21
           220.140187
                        6.133333
                                   34.040034
                                                      NaN
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       22
           230.299844
                        5.955556
                                   33.053366
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       23
           240.109034
                        5.777778
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           250.269665
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           260.079829
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                                   30.586697
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       26
           270.240460
                        5.377778
                                   29.846697
                                                      NaN
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                                       APA mM
                                                   POHPG_t
                                                            POHPG gl
                                                                        POHPG mM
                APA t
                           APA gl
       0
            10.163551
                        15.409471
                                    71.254373
                                                 15.070093 -0.024540
                                                                        -0.146805
       1
                        15.186630
                                    70.223941
            19.976636
                                                 29.789720
                                                             0.368098
                                                                         2.202071
       2
            30.140187
                        15.008357
                                    69.399596
                                                 39.953271
                                                             0.417178
                                                                         2.495680
       3
            39.953271
                        15.498607
                                    71.666546
                                                 59.929907
                                                             0.736196
                                                                         4.404142
       4
            49.766355
                        15.320334
                                    70.842200
                                                 86.214953
                                                             0.073620
                                                                         0.440414
       5
                                    70.223941
            59.929907
                        15.186630
                                                103.037383
                                                             1.079755
                                                                         6.459408
       6
            69.742991
                        15.008357
                                    69.399596
                                                111.799065
                                                             0.858896
                                                                         5.138165
       7
            79.906542
                        15.498607
                                    71.666546
                                                142.289720
                                                             0.809816
                                                                         4.844556
       8
            90.070093
                        15.364903
                                    71.048287
                                                155.257009
                                                             1.055215
                                                                         6.312603
       9
           100.233645
                        15.186630
                                    70.223941
                                                181.542056
                                                             1.128834
                                                                         6.753017
       10
           110.046729
                        15.097493
                                    69.811768
                                                191.355140
                                                             1.398773
                                                                         8.367869
       11
           120.210280
                        15.543175
                                    71.872632
                                                213.084112
                                                             1.472393
                                                                         8.808283
       12
           130.023364
                        15.454039
                                    71.460460
                                                223.247664
                                                             1.742331
                                                                       10.423135
       13
           139.836449
                        15.275766
                                    70.636114
                                                266.004673
                                                             2.159509
                                                                       12.918816
                                                             2.233129
       14
           150.000000
                        15.186630
                                                276.518692
                                                                       13.359230
                                    70.223941
       15
           160.163551
                        15.008357
                                    69.399596
                                                       NaN
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       16
           169.976636
                        15.498607
                                    71.666546
                                                       NaN
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                                    71.048287
       17
           180.140187
                        15.364903
                                                       NaN
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       18
           189.953271
                        15.275766
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                                    70.636114
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       19
           199.766355
                        15.142061
                                    70.017855
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       20
           209.929907
                        15.052925
                                    69.605682
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       21
           219.742991
                        15.632312
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                                    72.284805
                                                       NaN
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       22
           230.257009
                        15.454039
                                    71.460460
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                                                                  NaN
                        15.320334
       23
           240.420561
                                    70.842200
                                                       NaN
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                                                                              NaN
       24
           249.883178
                        15.275766
                                    70.636114
                                                       NaN
                                                                  NaN
                                                                              NaN
       25
           260.046729
                        15.097493
                                    69.811768
                                                       NaN
                                                                              NaN
                                                                  NaN
       26
           269.859813
                        15.052925
                                    69.605682
                                                       NaN
                                                                  NaN
                                                                              NaN
      CAB_medida = df.loc[:,'Ester_mM'].dropna().to_numpy()
[153]:
       CAN_medida = df.loc[:,'Amox_mM'].dropna().to_numpy()
       CNH_medida = df.loc[:,'APA_mM'].dropna().to_numpy()
       CAOH medida = df.loc[:,'POHPG mM'].dropna().to numpy()
       t_cab = df.loc[:,'Ester_t'].dropna().to_numpy().round(0)
       t_can = df.loc[:,'Amox_t'].dropna().to_numpy().round(0)
       t_cnh = df.loc[:,'APA_t'].dropna().to_numpy().round(0)
       t_caoh = df.loc[:,'POHPG_t'].dropna().to_numpy().round(0)
       tempo_substratos = t_cab
       tempo_produtos = t_can
```

20

210.309579

5.333333

29.600030

NaN

NaN

NaN

```
CAB medida t = np.hstack([CAB medida.reshape(-1,1),t_cab.reshape(-1,1)])
       CAN medida t = np.hstack([CAN medida.reshape(-1,1),t_can.reshape(-1,1)])
       CNH medida t = np.hstack([CNH medida.reshape(-1,1),t_cab.reshape(-1,1)])
       CAOH_medida t = np.hstack([CAOH_medida.reshape(-1,1),t_can.reshape(-1,1)])
       medidas_exp_list = [CAB_medida_t,CAN_medida_t,CNH_medida_t,CAOH_medida_t]
       t_total = np.unique(np.sort(np.hstack([t_can,t_cab])))
[154]: fed_ab = np.array([5,10,15,21])
       fed_ab_t = np.array([t_cab[fed_ab[0]],
                            t_cab[fed_ab[1]],
                            t_cab[fed_ab[2]],
                            t_cab[fed_ab[3]])
       fed_ab_c = np.array([CAB_medida[fed_ab[0]],
                            CAB medida[fed ab[1]],
                            CAB medida[fed ab[2]],
                            CAB_medida[fed_ab[3]]])
       fed_ab = np.vstack([fed_ab_t,fed_ab_c])
       fed_nh = np.array([3,7,11,16,21])
       fed_nh_t = np.array([t_cnh[fed_nh[0]],
                            t_cnh[fed_nh[1]],
                            t_cnh[fed_nh[2]],
                            t_cnh[fed_nh[3]],
                            t_cnh[fed_nh[4]],
                            ])
       fed_nh_c = np.array([CNH_medida[fed_nh[0]],
                            CNH_medida[fed_nh[1]],
                            CNH_medida[fed_nh[2]],
                            CNH medida[fed nh[3]],
                            CNH_medida[fed_nh[4]]])
       fed_nh = np.vstack([fed_nh_t,fed_nh_c])
       print(fed_nh)
      [[ 40.
                                  120.
                                                170.
                                                             220.
       [71.66654602 71.66654602 71.87263242 71.66654602 72.28480522]]
[155]: def enzymic_amox(t, y, kcat1, kcat2, Km1, Km2, Tmax, Ken, kAB, kAN, kAOH, kNH):
           FAB = 0
           FNH = 0
           CAB = y[0]
```

CAN = y[1] CNH = y[2]

```
CAOH = y[3]
           Cez = 1 # Assuming a constant value for Cez if not provided
           # Consumo de ester
           VAB = (kcat1 * CAB * Cez) / ((Km1 * (1 + (CAN/kAN) + (CAOH/kAOH))) + CAB)
           # Hidrolise de amoxicilina
           VAN = (kcat2 * CAN * Cez) / ((Km2 * (1 + (CAB/kAB) + (CNH/kNH) + (CAOH/kNH) + (CA
⇒kAOH))) + CAN)
           # Enzima saturada com 6-apa
           X = CNH / (Ken + CNH)
           # Sintese enzimatica
           VS = VAB * Tmax * X
           # Hidrolise de ester
           Vh1 = (VAB - VS)
           dy = np.zeros(4)
           # C. ester
           dy[0] = (-(VS - VAN) - (Vh1 + VAN)) + FAB
           # C. amox
           dy[1] = VS - VAN
           # C. 6-apa
           dy[2] = -(VS - VAN) + FNH
           # C. POHPG
           dy[3] = Vh1 + VAN
           return dy
```

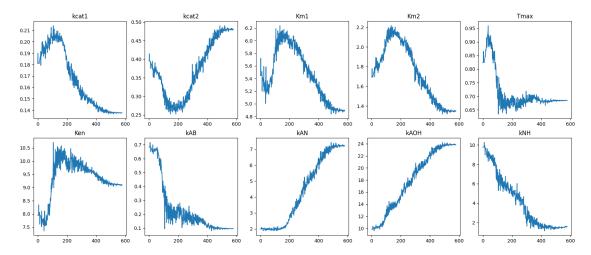
```
fed_ab = np.vstack([fed_ab_t,fed_ab_c])
       fed_nh = np.array([3,7,11,16,21])
       fed_nh_t = np.array([t_cnh[fed_nh[0]],
                            t_cnh[fed_nh[1]],
                             t_cnh[fed_nh[2]],
                             t_cnh[fed_nh[3]],
                             t_cnh[fed_nh[4]],
       fed_nh_c = np.array([CNH_medida[fed_nh[0]],
                            CNH_medida[fed_nh[1]],
                             CNH_medida[fed_nh[2]],
                             CNH_medida[fed_nh[3]],
                            CNH_medida[fed_nh[4]]])
       fed_nh = np.vstack([fed_nh_t,fed_nh_c])
       print(fed_nh)
                                   120.
      [[ 40.
                       80.
                                                170.
                                                              220.
       [71.66654602 71.66654602 71.87263242 71.66654602 72.28480522]]
[157]: kcat1
                    = 0.181
       kcat2
                    = 0.395
       Km1
                    = 5.449
       Km2
                    = 1.694
       Tmax
                    = 0.824
       Ken
                    = 7.947
       kAB
                    = 0.682
       kAN
                    = 1.989
       kAOH
                    = 9.856
       kNH
                    = 9.763
       P = np.zeros(10)
       P[0]
             = kcat1
       P[1]
              = kcat2
       P[2]
              = \text{Km1}
       P[3]
             = \text{Km2}
       P[4]
             = Tmax
       P[5]
              = Ken
       P[6]
            = kAB
       P[7]
             = kAN
       P[8]
            = kAOH
       P[9] = kNH
```

Np = len(P)

```
[158]: def ode15s_amox_fed(P,t):
           fed_t = np.unique(np.sort(np.hstack([fed_ab[0],fed_nh[0]])))
           CI_ode = np.
        →array([CAB_medida[0],CAN_medida[0],CNH_medida[0],CAOH_medida[0]])
           all t = []
           for i,fed in enumerate(fed_t):
               all_t.insert(i,[])
               count = 0
               for p in t:
                   all_t[i].append(p)
                   count += 1
                   if p == fed:
                       break
               t = t[count-1:]
           all_t.append(t)
           final Y = np.zeros([4,1])
           for t in all_t:
               Y = scipy.integrate.
        solve_ivp(enzymic_amox,t_span=(t[0],t[-1]),t_eval=t,y0=CI_ode,method='BDF',args=P)
               Y = Y.v
               CI_ode = Y[:,-1]
               if t[-1] in fed_ab[0]:
                   CI_ode[0] = fed_ab[1][list(fed_ab[0]).index(t[-1])]
               if t[-1] in fed_nh[0]:
                   CI_ode[2] = fed_nh[1][list(fed_nh[0]).index(t[-1])]
               final_Y = np.hstack([final_Y,Y[:,:-1]])
           final_Y = np.hstack([final_Y,Y[:,-1].reshape(-1,1)])
           final_Y = np.delete(final_Y,0,1)
           return final Y
[159]: def calculate_error(model_matrix, experimental_data_list):
           # Extract model time points
           model_time = model_matrix[:, -1]
           total_error = 0
           for var_index, experimental_data in enumerate(experimental_data_list):
               # Extract variable model data
               model_variable_data = model_matrix[:, var_index]
               # Match time points between model and experimental data
               exp_time = experimental_data[:, -1]
               exp_values = experimental_data[:, 0]
               for t, exp_value in zip(exp_time, exp_values):
                   if not np.isnan(exp_value): # Skip NaN values in experimental data
                       # Find corresponding model value at time t
                       model_value = model_variable_data[model_time == t]
                       if model_value.size > 0: # If there's a matching time point
```

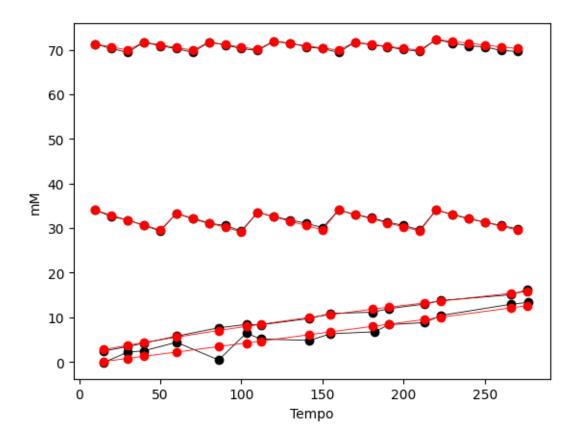
```
total_error += (exp_value - model_value[0]) ** 2
           print(total_error,end="\r")
           return total_error
[173]: def objective_function(params):
           kcat1, kcat2, Km1, Km2, Tmax, Ken, kAB, kAN, kAOH, kNH = params
           record.append(params)
           sol = ode15s_amox_fed(params,t_total)
           model_output = sol.T
           model_output = np.hstack([model_output,t_total.reshape(-1,1)])
           return calculate_error(model_output,medidas_exp_list)
[249]: options = {'maxiter':400}
       record = []
       bounds = [(0.001, None) for _ in range(len(P))]
       result = minimize(objective_function, P, __
        omethod='Nelder-Mead', bounds=bounds, options=options)
       record = np.array(record)
       # Optimized parameters
       optimized_parameters = result.x
       print("Optimized Parameters:", optimized_parameters)
       P_new = optimized_parameters
      Optimized Parameters: [ 0.13774187  0.47860872  4.89771429  1.35408369
      0.68420837 9.10175497
        0.09727782 7.18711047 23.81970143 1.62049931]
[250]: fig,ax = plt.subplots(2,5,figsize=[20,8])
       ax = ax.flatten()
       labels = ['kcat1',
       'kcat2',
       'Km1',
       'Km2',
       'Tmax',
       'Ken',
       'kAB',
       'kAN',
       'kAOH',
       'kNH']
       print(record.shape)
       for idx,g in enumerate(ax):
           g.plot(record[:,idx])
           g.set_title(labels[idx])
       fig.savefig('scipy_methods/NelderMead.png',dpi=400)
```

(579, 10)

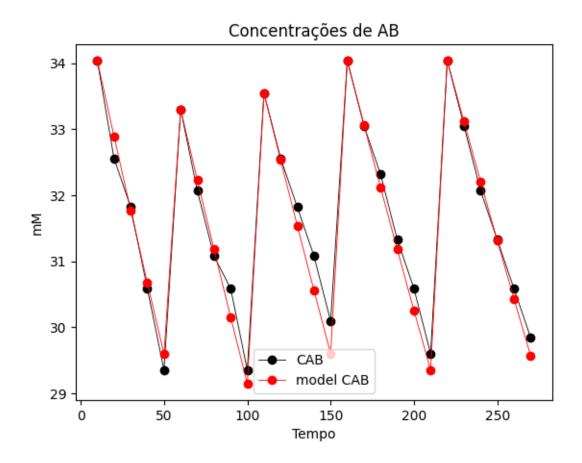


```
[252]: sol = ode15s_amox_fed(P_new,t_total)
       pidx = np.isin(t_total, tempo_produtos)
       sidx = np.isin(t_total, tempo_substratos)
       CAB = sol[0][sidx]
       CAN = sol[1][pidx]
       CNH = sol[2][sidx]
       CAOH = sol[3][pidx]
       t_p = t_total[pidx]
       t_s = t_total[sidx]
[253]: fig = plt.figure()
       plt.plot(t_cab, CAB_medida,'-ok',linewidth=0.6)
       plt.plot(t_can, CAN_medida,'-ok',linewidth=0.6)
       plt.plot(t_cnh, CNH_medida,'-ok',linewidth=0.6)
       plt.plot(t_caoh,CAOH_medida,'-ok',linewidth=0.6)
       plt.plot(t_s, CAB,'-or',linewidth=0.6)
       plt.plot(t_p, CAN,'-or',linewidth=0.6)
       plt.plot(t_s, CNH,'-or',linewidth=0.6)
       plt.plot(t_p, CAOH, '-or', linewidth=0.6)
       plt.xlabel('Tempo')
       plt.ylabel('mM')
```

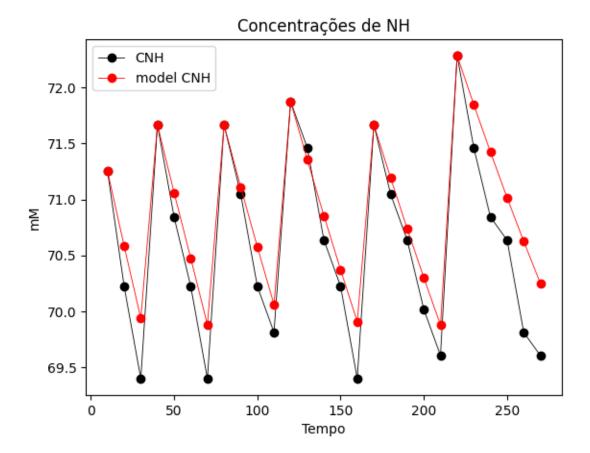
[253]: Text(0, 0.5, 'mM')



```
[254]: fig = plt.figure()
  plt.plot(t_cab, CAB_medida,'-ok',linewidth=0.6)
  plt.plot(t_s, CAB,'-or',linewidth=0.6)
  plt.title(f'Concentrações de AB')
  plt.legend(['CAB','model CAB'])
  plt.xlabel('Tempo')
  plt.ylabel('mM');
```

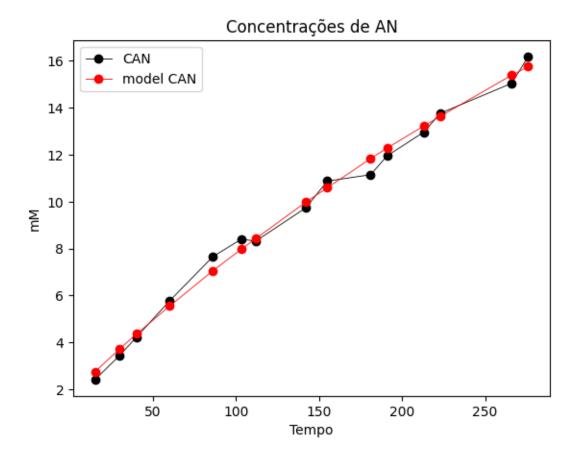


```
[255]: fig = plt.figure()
  plt.plot(t_cnh, CNH_medida,'-ok',linewidth=0.6)
  plt.plot(t_s, CNH,'-or',linewidth=0.6)
  plt.title(f'Concentrações de NH')
  plt.legend(['CNH','model CNH']);
  plt.xlabel('Tempo')
  plt.ylabel('mM');
```



```
[256]: fig = plt.figure()

plt.plot(t_can, CAN_medida,'-ok',linewidth=0.6)
plt.plot(t_p, CAN,'-or',linewidth=0.6)
plt.title(f'Concentrações de AN')
plt.legend(['CAN','model CAN']);
plt.xlabel('Tempo')
plt.ylabel('mM');
```



```
[257]: fig = plt.figure(figsize=(10,6))

plt.plot(t_caoh,CAOH_medida,'-ok',linewidth=0.6)
plt.plot(t_p, CAOH,'-or',linewidth=0.6)
plt.title(f'Concentrações de AOH')
plt.legend(['CAOH','model CAOH'])
plt.xlabel('Tempo')
plt.ylabel('mM');
```

