

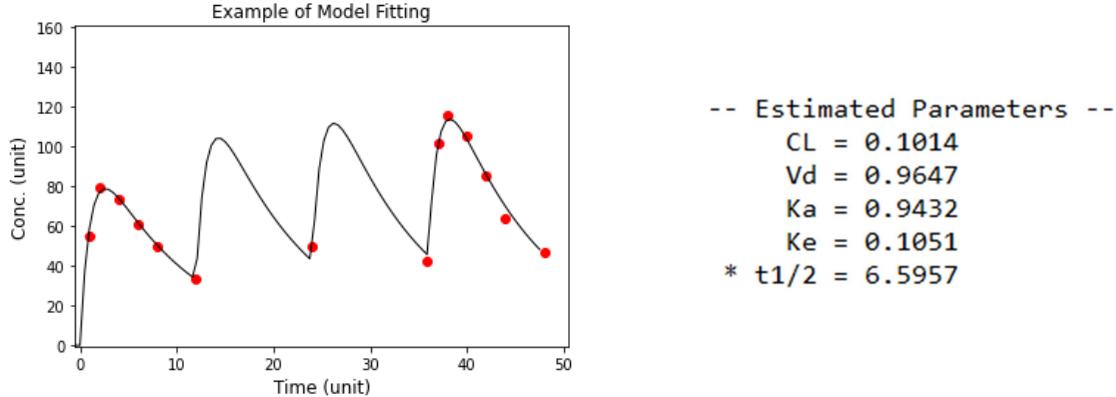
薬物動態～経口投与反復 1-コンのあてはめ計算

薬物動態、経口反復投与 1-コンパートメントモデルのあてはめ計算 (prog02_12.py)

* 前回までの記事を確認した上で下記を参照のこと。

- (1) 投与量 (DOSE)、薬物動態パラメータ (クリアランス (CL)、分布容積 (VD)、吸収速度定数 (KA))。反復投与は一定用量・一定間隔 (TAU) とした。
- (2) minimize 関数を用いた。残差平方和を関数 obj_func() に、モデル式を model_eq121() に定義。
- (3) 薬物動態パラメータはリスト (P[]) とし、グローバル変数とした。

以上の内容で作成した例を次ページ「prog02_12.py」に示した。出力結果は次のとおり。



```

final_simplex: (array([[0.10138255, 0.96471875, 0.94317568],
[0.10137955, 0.96474343, 0.9432075 ],
[0.10138128, 0.96465381, 0.94311458],
[0.10137819, 0.96471284, 0.94310344]]), array([125.69856873, 125.69857812,
125.69860529, 125.69861858]))
    fun: 125.69856872941324
message: 'Optimization terminated successfully.'
    nfev: 98
    nit: 52
    status: 0
success: True
    x: array([0.10138255, 0.96471875, 0.94317568])

```

以上

```

"""# Program 02-12 "prog02_12.py"""

# Fitting to One-Compartment po model after multiple dosing
import numpy as np
import matplotlib.pyplot as plt
import scipy.optimize as opt

# Global varialbes
DOSE = 100.
TAU = 12.0 # Dosing Interval

Nparam = 3
P = [0 for j in range(Nparam)]
P[0] = 0.1 # CL
P[1] = 1.0 # Vd (Vd/F)
P[2] = 1.5 # Ka
Npoint = 14
x = [0 for j in range(Npoint)]
y = [0 for j in range(Npoint)]

#
def set_data():
    # Data input in this program
    data_x = np.array([1.0, 2.0, 4.0, 6.0, 8.0, 11.9, 23.9, 35.9,
                      37, 38, 40.0, 42.0, 44.0, 48.0])
    data_y = np.array([55.1, 79.2, 73.5, 60.7, 49.5, 33.2, 49.5, 42.1,
                      101.2, 115.3, 105.2, 85.3, 63.9, 46.8])
    return data_x, data_y

def obj_func(P):
    SS = 0.0
    for i in range(Npoint):
        xx = x[i]
        Cp = model_eq121(xx, i, P)
        SS = SS + (y[i] - Cp) ** 2
    return SS

def model_eq121(xx, i, P):
    # 1-Comp. po model (no lag time)

```

```

tn = 0
x2 = 0

while xx > tn * TAU:

    tn = tn + 1
    x2 = xx - (tn - 1) * TAU

CL = P[0]
VD = P[1]
ka = P[2]
ke = CL / VD

E1 = 1 - np.exp(-ke * tn * TAU)
E2 = 1 - np.exp(-ke * TAU)
E3 = 1 - np.exp(-ka * tn * TAU)
E4 = 1 - np.exp(-ka * TAU)
E5 = DOSE / VD * ka / (ka - ke)

Cp = E5 * (E1 / E2 * np.exp(-ke * x2) - E3 / E4 * np.exp(-ka * x2))

return Cp

```

```

def draw_plot_obs(data_x, data_y):
    # Plotting observed data
    plt.title("Example of Model Fitting", fontsize = 12)
    plt.xlim(-0.5, 50.5)
    plt.ylim(-0.5, 160.5)
    plt.xlabel("Time (unit)", fontsize = 12)
    plt.ylabel("Conc. (unit)", fontsize = 12)
    plt.plot(data_x, data_y, "o", color = "red")
    #plt.show()

```

```

def draw_plot_pred():
    # Simulation of predicted curve
    tx = [0 for j in range(100)]
    cy = [0 for j in range(100)]
    for i in range(100):
        tx[i] = (i - 1) * (48.0 - 0.0) / (100 - 1)
        cy[i] = model_eq121(tx[i], i, P)
    plt.plot(tx, cy, linestyle = "solid", linewidth = "1", color = "black")
    plt.show()

```

```

def print_output():

```

```

# Output

print ("-- Estimated Parameters --")
print ('    CL = {:.4f}'.format(CL))
print ('    Vd = {:.4f}'.format(VD))
print ('    Ka = {:.4f}'.format(KA))
print ('    Ke = {:.4f}'.format(CL / VD))
print (' * t1/2 = {:.4f}'.format(np.log(2) / (CL / VD)))

#
if __name__ == '__main__':
    x, y = set_data()
    draw_plot_obs(x, y)
    results = opt.minimize(obj_func, P, method='Nelder-Mead')
    print (results)
    P[0] = results.x[0]
    P[1] = results.x[1]
    P[2] = results.x[2]
    CL = P[0]
    VD = P[1]
    KA = P[2]
    draw_plot_pred()
    print_output()

```