本文使用高斯牛顿法实现曲线拟合

目的

一方面出于学习的目的完成这个小项目,另一方面出于为其他的同学提供一个相关的 代码学习资料

拟合方程

$$y = e^{ax^3 + bx^2 + cx + d} + w$$

数据产生 (随机数产生,误差服从高斯分布)

$$w\left(0,\sigma^2\right)$$

误差定义

$$er_i = y_i - e^{ax_i^3 + bx_i^2 + cx_i + d}$$

求偏导(最关键)

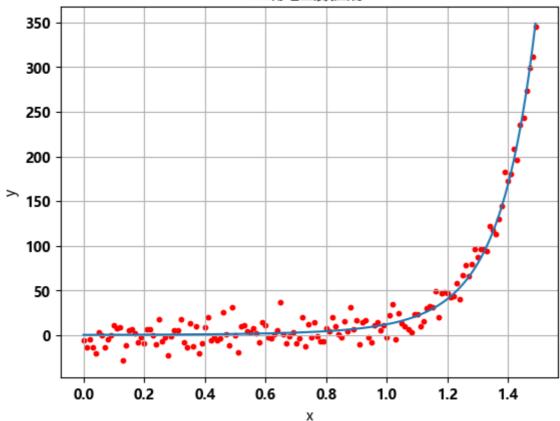
$$\left\{egin{array}{l} rac{\partial er_i}{\partial a} &= -x_i^3 e^{ax_i^3 + bx_i^2 + cx_i + d} \ rac{\partial er_i}{\partial a} &= -x_i^2 e^{ax_i^3 + bx_i^2 + cx_i + d} \ rac{\partial er_i}{\partial c} &= -x_i e^{ax_i^3 + bx_i^2 + cx_i + d} \ rac{\partial er_i}{\partial d} &= -e^{ax_i^3 + bx_i^2 + cx_i + d} \end{array}
ight.$$

```
1  # 产生固定间隔的数组
2  import numpy as np
3  
4  start = 0  # 起始值
5  stop = 1.5  # 结束值 (不包含在数组中)
6  step = 0.01  # 间隔
7
```

```
arr = np.arange(start, stop, step)
9
    # print(arr)
10
11
12
    mean = 0 # 均值
13
    std = 10 # 标准差
14
    size = int((stop-start)/step) # 数组大小
15
16
    noise = np.random.normal(mean, std, size)
17
    # print(noise)
18
19
20
    # 生成观测数据x,y
21
22
    # 定义函数参数
23
    a = 0.5
24
    b = 1
25
    c = 2.0
26
    d = -1
27
    x=arr
28
    obs_y=np.exp(a*x**3 + b*x**2 + c*x + d)+noise
```

```
1
    import matplotlib.pyplot as plt
2
    import matplotlib
3
    matplotlib.rc("font",family='MicroSoft YaHei',weight="bold")
4
    # 绘制图像
5
6
    # 定义函数
7
    def f(x,A):
8
        return np.exp(A[0]*x**3 + A[1]*x**2 + A[2]*x + A[3])
9
10
    # 计算对应的 y 值
11
    y = f(x,[a,b,c,d])
12
13
     plt.plot(x, y)
14
     plt.scatter(x, obs_y, color='red', s=10,label='Scatter Points')
15
    plt.xlabel('x')
16
    plt.ylabel('y')
17
     plt.title('正确地函数图像')
18
     plt.grid(True)
     plt.show()
19
```

正确地函数图像



高斯牛顿迭代进行曲线拟合 (我理解为间接平差升级版)

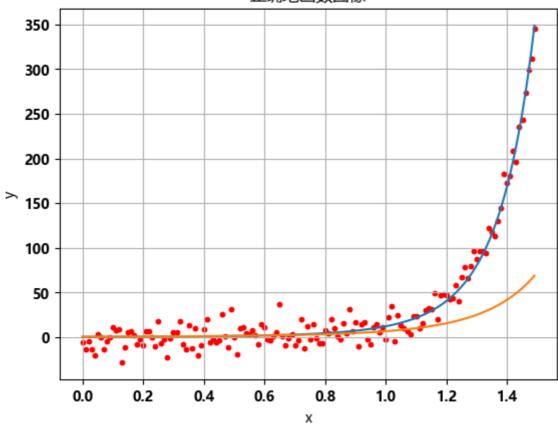
直接求得解析解

$$egin{align*} V = Bx - L \ x = \left(B^T B
ight) * \left(B^T L
ight) \ & \left[egin{align*} er_1 \ er_2 \ dots \ er_{n-1} \ er_n \ \end{array}
ight] = oldsymbol{B} \left[egin{align*} \Delta a \ \Delta b \ \Delta c \ \Delta d \ \end{array}
ight] - oldsymbol{L} \ & \left[egin{align*} e^{a_0 x_1^3 + b_0 x_1^2 + c_0 x_1 + d_0} & \cdots & -e^{a_0 x_1^3 + b_0 x_1^2 + c_0 x_1 + d_0} \ -x_2^3 e^{a_0 x_2^3 + b_0 x_2^2 + c_0 x_2 + d_0} & dots & -e^{a_0 x_1^3 + b_0 x_1^2 + c_0 x_2 + d_0} \ dots & & dots & dots \ -e^{a_0 x_{n-1}^3 + b_0 x_{n-1}^2 + c_0 x_{n-1} + d_0} \ -x_{n-1}^3 e^{a_0 x_{n-1}^3 + b_0 x_{n-1}^2 + c_0 x_n + d_0} & dots & -e^{a_0 x_{n-1}^3 + b_0 x_{n-1}^2 + c_0 x_{n-1} + d_0} \ -x_n^3 e^{a_0 x_n^3 + b_0 x_n^2 + c_0 x_n + d_0} & dots & -e^{a_0 x_{n-1}^3 + b_0 x_{n-1}^2 + c_0 x_n + d_0} \ \end{bmatrix}$$

```
L = egin{bmatrix} e^{a_0x_1^3+b_0x_1^2+c_0x_1+d_0} - y_1 \ e^{a_0x_2^3+b_0x_2^2+c_0x_2+d_0} - y_2 \ dots \ e^{a_0x_{n-1}^3+b_0x_{n-1}^2+c_0x_{n-1}+d_0} - y_{n-1} \ e^{a_0x_n^3+b_0x_n^2+c_0x_n+d_0} - y_n \end{bmatrix}
```

```
# 定义a,b,c,d的初始值(随便定)
 2
    a0=0.3
 3
    b0=0.7
4
    c0=1.6
 5
    d0 = -0.7
 6
7
    # 定义函数, A为参数数组
8
    def f0(x,A):
9
        return np.exp(A[0]*x**3 + A[1]*x**2 + A[2]*x + A[3])
10
11
    # 计算对应的 y 值
12
    y0 = f0(x,[a0,b0,c0,d0])
13
14
    plt.plot(x, y)
15
     plt.plot(x, y0)
16
     plt.scatter(x, obs_y, color='red', s=10,label='Scatter Points')
17
     plt.xlabel('x')
18
     plt.ylabel('y')
19
    plt.title('正确地函数图像')
20
     plt.grid(True)
21
     plt.show()
```

正确地函数图像



```
1
     # 进行间接平差求解
 2
     # 构建B矩阵
 3
     a0_temp=a0
 4
     b0_temp=b0
 5
     c0_temp=c0
 6
     d0_{temp}=d0
 7
8
     approx=np.array([a0_temp,b0_temp,c0_temp,d0_temp])
9
10
     # 迭代次数
11
     n=10
12
     dieDaiResult = np.zeros((n, 4))
13
     # 用一个
14
15
     for j in range(n):
16
             B=np.zeros((size,4))
17
             L=np.zeros((size,1))
18
             for i in range(size):
19
                 B[i,0]=-x[i]**3*f0(x[i],approx)
20
                 B[i,1]=-x[i]**2*f0(x[i],approx)
21
                 B[i,2]=-x[i]*f0(x[i],approx)
22
                 B[i,3]=-f0(x[i],approx)
23
                 L[i,0]=f0(x[i],approx)-obs_y[i]
24
```

```
25
             arr_B=np.array(B)
26
             arr_L=np.array(L)
27
             tem1=np.linalg.inv(np.dot(np.transpose(arr_B),arr_B))
28
             tem2=np.dot(np.transpose(arr_B),arr_L)
29
30
             delta_x=np.dot(tem1,tem2)
31
             # 这里进行下一次迭代
32
             approx=approx+delta_x.flatten()
33
             # list_approx=approx.
34
             dieDaiResult[j,:]=approx
35
36
     fig = plt.figure(figsize=(10, 5))
37
     plt.subplot(1, 2, 1)
38
     # 标准曲线
39
     plt.plot(x, y)
40
     # 初始曲线
41
     plt.plot(x, y0)
42
     # 迭代曲线
43
     for i in range(n-8,n):
44
          xishu=dieDaiResult[i]
45
          print(xishu)
46
          tem_y=f0(x,xishu)
47
           plt.plot(x, tem_y,color='#aa00ff',linestyle='--')
48
     # plt.plot(x, y1,color='#aa00ff')
49
     plt.scatter(x, obs_y, color='red', s=10,label='Scatter Points')
50
     # plt.plot(x, y1)
51
     plt.xlabel('x')
52
     plt.ylabel('y')
53
     plt.title('正确地函数图像')
54
     plt.grid(True)
55
56
     # 创建第二个子图
57
     plt.subplot(1, 2, 2)
58
     # 标准曲线
59
     plt.plot(x, y)
60
     # 初始曲线
61
     plt.plot(x, y0)
62
     # 迭代曲线
63
     for i in range(n-8,n):
64
          xishu=dieDaiResult[i]
65
          print(xishu)
66
          tem_y=f0(x,xishu)
67
           plt.plot(x, tem_y,color='#aa00ff',linestyle='--')
68
     # plt.plot(x, y1,color='#aa00ff')
69
     plt.scatter(x, obs_y, color='red', s=10,label='Scatter Points')
```

```
70
    # plt.plot(x, y1)
71
    plt.xlabel('x')
72
    plt.ylabel('y')
73
    plt.title('局部放大图')
74
75
    # 设置局部放大范围
76
    plt.xlim(1.32, 1.44) # 设置 x 轴范围
77
    plt.ylim(100, 250) # 设置 y 轴范围
78
    plt.grid(True)
79
    plt.show()
80
```

```
1
     [ 15.18948191 -48.88914189 58.18225421 -21.87081973]
 2
     [ 11.06123187 -37.62392869 48.70588659 -19.61002625]
 3
     [ 2.98965953 -9.4558721 16.33702568 -7.39086396]
 4
     [-0.25717518 2.86430852 0.85419677 -0.95130386]
 5
     [ 2.28260343 -6.65929933 12.58667395 -5.68529249]
 6
     [-0.53772553 \quad 3.94032672 \quad -0.50831067 \quad -0.38253999]
 7
     [ 2.22309202 -6.36924235 12.12862682 -5.44958593]
 8
     [-0.55472966 \quad 4.00668407 \quad -0.59381963 \quad -0.34621364]
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```

