计算物理第一次作业

祝茗 2024202020022

Bessel 函数

Bessel 方程为

$$x^{2} \frac{\mathrm{d}^{2} y}{\mathrm{d}x^{2}} + x \frac{\mathrm{d}x}{\mathrm{d}y} + (x^{2} - n^{2})y = 0$$

其中 n 被称为其对应 Bessel 方程的阶数。实际应用中最常见的情形为 n 是整数,对应解 称为 n 阶 Bessel 函数 $J_n(x)$

Bessel 函数 $J_n(x)$ 之间有递推关系

$$J_{n+1}(x) - \frac{2n}{r}J_n(x) + J_{n-1}(x) = 0$$
 (1)

和一个求和恒等式

$$[J_0(x)]^2 + 2\sum_{n=1}^{\infty} [J_n(x)]^2 = 1 \tag{2}$$

根据这两个关系,我们可以来绘制,当 x=1,10 时 Bessel 函数 $J_n(x)$ 作为 n 的函数 图像。

可以预见的是

$$|J_n(x)|_{n o\infty} o 0$$

考虑到数值计算的精度,当 n 大到一定程度 (例如 n>M 时),都小于计算机浮点数可以允许的精度

$$|J_n(x)|_{n\geq M}\leq eps$$

所以我们可以做一个假设

$$J_M(x) = 0, \quad J_{M-1} = 1$$

然后根据递推关系,递推地算出其余 $J_n(x)$ 的相对大小,由于是齐次的线性的关系式,最后根据求和恒等式,就可以得到需要的答案。

导入前置的包并设置基本的参数

In [37]: import numpy as np
 import matplotlib.pyplot as plt
 from scipy.special import jn

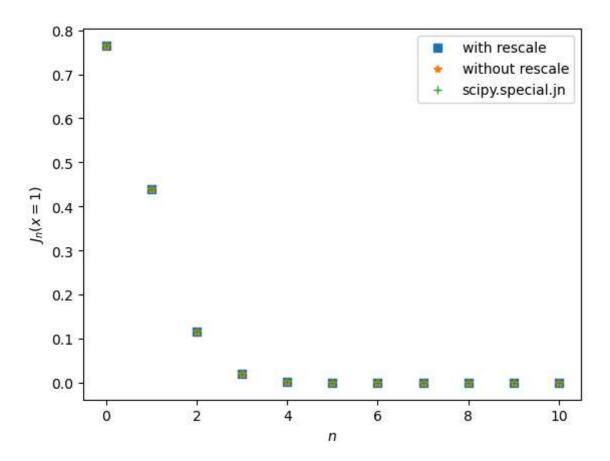
计算 Bessel 函数 $J_n(x)$

```
In [38]: | def BesselJ_with_rescale(x: float | np.ndarray, M: int) -> np.ndarray:
            """计算 Bessel 函数 J_n(x) 的递推值,基于递推关系式。
            对 J n(x) 进行归一化处理, 防止数值溢出。
            Args:
                x (float): 自变量
                M (int): 返回的阶数
            Returns:
                np.ndarray: J_n(x) 的数组
            Example:
                >>> x = 1
                >>> Jn = BesselJ(x, 10)
                >>> plt.plot(x, Jn.T)
                >>> plt.show()
            Jn = np.zeros(M+1)
            Jn[M] = 0 # 假设 <math>J_M(x) = 0
            Jn[M-1] = 1 # 假设 J_{M-1}(x) = 1
            for n in range(M-2, -1, -1):
                Jn[n] = (2*(n+1)/x) * Jn[n+1] - Jn[n+2]
                #整体缩放,防止递推过程中数值溢出
                if abs(Jn[n]) > 1/epsi:
                    Jn *= epsi
            normalization_factor = np.sqrt(Jn[0]**2 + 2 * np.sum(Jn[1:]**2))
            Jn /= normalization_factor
            return Jn
        def BesselJ without rescale(x: float | np.ndarray, M: int) -> np.ndarray:
             """计算 Bessel 函数 J_n(x) 的递推值,基于递推关系式。
            Args:
                x (float): 自变量
                M (int): 返回的阶数
            Returns:
                np.ndarray: J_n(x) 的数组
            Example:
                >>> x = np.linspace(0, 20, 100)
                >>> Jn = BesselJ(x, 10)
                >>> plt.plot(x, Jn.T)
                >>> plt.show()
            Jn = np.zeros(M+1)
            Jn[M] = 0 # 假设 <math>J_M(x) = 0
            Jn[M-1] = 1 # 假设 J_{M-1}(x) = 1
```

函数效果直观比较

$$x = 1, M = 20$$

```
In [73]: x = 1
          M = 20
          J1 = BesselJ_with_rescale(x, M)
          J2 = BesselJ_without_rescale(x, M)
          J3 = np.array([jn(n, 1) for n in range(M+1)])
          # 前 m 个点的比较
          m = 11
          J1 = J1[:m]
          J2 = J2[:m]
          J3 = J3[:m]
          fig = plt.figure()
          plt.xlabel(f'$n$')
          plt.ylabel(f'$J_n(x=1)$')
          plt.plot(J1, 's', label='with rescale')
plt.plot(J2, '*', label='without rescale')
          plt.plot(J3, '+', label='scipy.special.jn')
          plt.legend()
          plt.show()
```

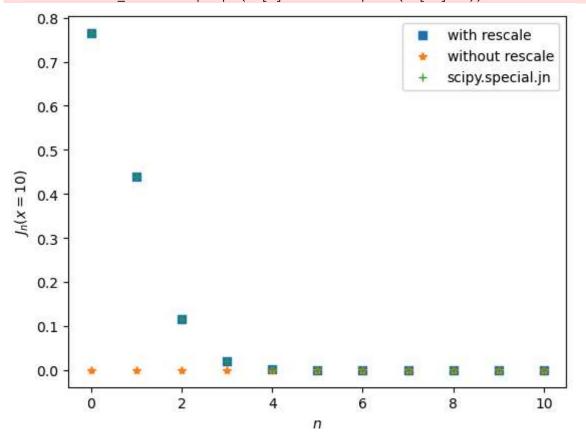


M 较小的时候,是否在递推过程中整体放缩影响不大。

$$x = 1, M = 90$$

```
In [75]: x = 1
         M = 90
         J1 = BesselJ_with_rescale(x, M)
         J2 = BesselJ_without_rescale(x, M)
         J3 = np.array([jn(n, 1) for n in range(M+1)])
         # 前 m 个点的比较
         m = 11
         J1 = J1[:m]
         J2 = J2[:m]
         J3 = J3[:m]
         fig = plt.figure()
         plt.xlabel(f'$n$')
         plt.ylabel(f'$J_n(x=10)$')
         plt.plot(J1, 's', label='with rescale')
         plt.plot(J2, '*', label='without rescale')
         plt.plot(J3, '+', label='scipy.special.jn')
         plt.legend()
         plt.show()
```

```
/tmp/ipykernel_13122/2824819278.py:62: RuntimeWarning: overflow encountered in sc
alar power
  normalization_factor = np.sqrt(Jn[0]**2 + 2 * np.sum(Jn[1:]**2))
/tmp/ipykernel_13122/2824819278.py:62: RuntimeWarning: overflow encountered in sq
uare
  normalization_factor = np.sqrt(Jn[0]**2 + 2 * np.sum(Jn[1:]**2))
```

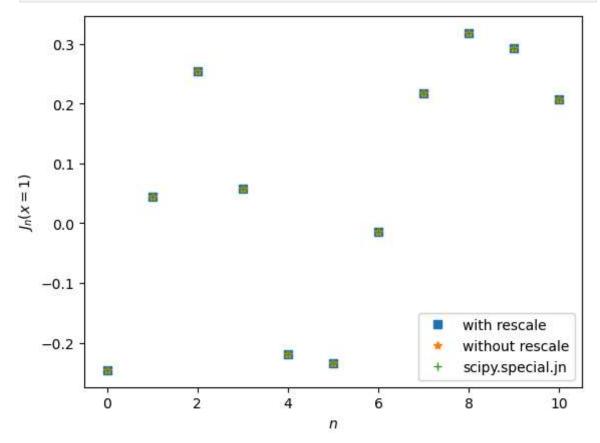


M 较大的时候,是否在递推过程中,如果不整体放缩,确实会对结果产生影响。

$$x = 10, M = 20$$

```
In [76]: x = 10
         M = 20
         J1 = BesselJ with rescale(x, M)
         J2 = BesselJ_without_rescale(x, M)
         J3 = np.array([jn(n, 10) for n in range(M+1)])
         # 前 m 个点的比较
         m = 11
         J1 = J1[:m]
         J2 = J2[:m]
         J3 = J3[:m]
         fig = plt.figure()
         plt.xlabel(f'$n$')
         plt.ylabel(f'$J_n(x=1)$')
         plt.plot(J1, 's', label='with rescale')
         plt.plot(J2, '*', label='without rescale')
         plt.plot(J3, '+', label='scipy.special.jn')
```

```
plt.legend()
plt.show()
```

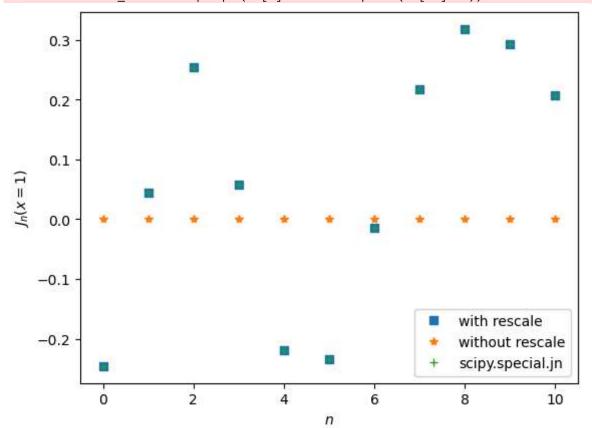


M 较小的时候,是否在递推过程中整体放缩影响不大。

$$x = 10, M = 200$$

```
In [81]: x = 10
          M = 200
          J1 = BesselJ_with_rescale(x, M)
          J2 = BesselJ_without_rescale(x, M)
          J3 = np.array([jn(n, 10) for n in range(M+1)])
          # 前 m 个点的比较
          m = 11
          J1 = J1[:m]
          J2 = J2[:m]
          J3 = J3[:m]
          fig = plt.figure()
          plt.xlabel(f'$n$')
          plt.ylabel(f'$J_n(x=1)$')
          plt.plot(J1, 's', label='with rescale')
          plt.plot(J2, '*', label='without rescale')
plt.plot(J3, '+', label='scipy.special.jn')
          plt.legend()
          plt.show()
```

```
/tmp/ipykernel_13122/2824819278.py:62: RuntimeWarning: overflow encountered in sc
alar power
  normalization_factor = np.sqrt(Jn[0]**2 + 2 * np.sum(Jn[1:]**2))
/tmp/ipykernel_13122/2824819278.py:62: RuntimeWarning: overflow encountered in sq
uare
  normalization_factor = np.sqrt(Jn[0]**2 + 2 * np.sum(Jn[1:]**2))
```



M 较大的时候,是否在递推过程中,如果不整体放缩,确实会对结果产生影响。

函数效果数值比较

之后部分只使用函数 BesselJ with rescale

```
In [40]: import os
    import pandas as pd

# 读取数据文件
    data_x1 = np.loadtxt(os.path.join('source', 'bessel_x1.dat'))
    data_x10 = np.loadtxt(os.path.join('source', 'bessel_x10.dat'))

# dateframe
    df_x1 = pd.DataFrame(data_x1, columns=['n', 'J_n(x=1)'])
    df_x10 = pd.DataFrame(data_x10, columns=['n', 'J_n(x=10)'])

In [41]: cal_J1 = BesselJ_with_rescale(x=1, M=13)
    df_x1['cal_J_n(x=1)'] = cal_J1
    df_x1['error'] = df_x1['cal_J_n(x=1)'] - df_x1['J_n(x=1)']
    print(df_x1)
```

```
J n(x=1) cal J n(x=1)
              n
                                                   error
                 7.651977e-01 7.651977e-01 -4.203349e-11
       0
            0.0
       1
            1.0 4.400506e-01 4.400506e-01 4.493350e-11
       2
            2.0 1.149035e-01 1.149035e-01 3.190047e-11
       3
            3.0 1.956335e-02 1.956335e-02 2.668404e-12
       4
                2.476639e-03 2.476639e-03 1.099550e-13
            4.0
       5
            5.0 2.497577e-04 2.497577e-04 1.123445e-14
            6.0 2.093834e-05 2.093834e-05 2.389266e-15
       6
       7
            7.0 1.502326e-06 1.502326e-06 4.368077e-16
       8
            8.0 9.422344e-08 9.422344e-08 -3.961428e-18
       9
            9.0 5.249250e-09 5.249250e-09 -1.904972e-19
       10
           10.0 2.630615e-10 2.630615e-10 -1.867515e-18
           11.0 1.198007e-11 1.198003e-11 -3.661981e-17
       11
       12
          12.0 4.999718e-13 4.991680e-13 -8.038662e-16
       13 13.0 1.925617e-14 0.000000e+00 -1.925617e-14
In [42]: cal_J10 = BesselJ_with_rescale(x=10, M=34)
         df_x10['cal_J_n(x=10)'] = cal_J10
         df x10['error'] = df x10['cal J n(x=10)'] - df x10['J n(x=10)']
         print(df_x10)
                    J n(x=10) cal J n(x=10)
              n
            0.0 -2.459358e-01 -2.459358e-01 4.865169e-11
       0
       1
            1.0 4.347275e-02
                               4.347275e-02 -1.138492e-12
       2
            2.0 2.546303e-01 2.546303e-01 -1.487943e-11
       3
            3.0 5.837938e-02
                              5.837938e-02 -4.813268e-12
       4
                              -2.196027e-01 -2.008560e-12
            4.0 -2.196027e-01
       5
            5.0 -2.340615e-01 -2.340615e-01 1.320641e-11
       6
            6.0 -1.445884e-02 -1.445884e-02 -4.785042e-12
       7
            7.0 2.167109e-01
                             2.167109e-01 -1.494843e-11
                               3.178541e-01 4.385720e-11
       8
            8.0 3.178541e-01
       9
                              2.918557e-01 -3.487999e-11
            9.0 2.918557e-01
       10
           10.0 2.074861e-01
                              2.074861e-01 3.335879e-11
           11.0 1.231165e-01
                              1.231165e-01 1.597597e-12
       11
           12.0 6.337025e-02
                              6.337025e-02 1.559863e-13
       13
           13.0 2.897208e-02
                              2.897208e-02 -3.223248e-12
       14 14.0 1.195716e-02
                              1.195716e-02 -5.364268e-13
          15.0 4.507973e-03
                              4.507973e-03 -2.787484e-13
       15
           16.0 1.566756e-03
                               1.566756e-03 -2.998205e-13
       16
       17
           17.0 5.056467e-04
                               5.056467e-04 1.932460e-14
       18 18.0 1.524425e-04
                               1.524425e-04 4.552419e-14
       19
           19.0 4.314628e-05
                               4.314628e-05 4.562519e-15
          20.0 1.151337e-05
       20
                               1.151337e-05 -2.186612e-15
          21.0 2.907199e-06
                               2.907199e-06 -3.089684e-16
       22 22.0 6.968685e-07
                               6.968685e-07 -1.105346e-17
       23
          23.0 1.590220e-07
                               1.590220e-07 -1.966734e-17
       24 24.0 3.463263e-08
                               3.463263e-08 5.837778e-19
       25 25.0 7.214635e-09
                               7.214635e-09 4.694730e-19
       26 26.0 1.440545e-09
                               1.440545e-09 -2.364181e-19
       27
           27.0 2.762005e-10
                               2.762005e-10 1.153141e-21
       28
          28.0 5.093755e-11
                               5.093755e-11 -2.735481e-20
           29.0 9.049767e-12
       29
                               9.049767e-12 -1.213401e-19
       30
           30.0 1.551096e-12
                               1.551095e-12 -6.798177e-19
           31.0 2.568095e-13
       31
                               2.568055e-13 -3.959366e-18
          32.0 4.112271e-14
                               4.109885e-14 -2.386780e-17
       33 33.0 6.375893e-15
                               6.227098e-15 -1.487946e-16
       34 34.0 9.581766e-16
                               0.000000e+00 -9.581766e-16
```