

## 第二次作业解答

### 1. Cholesky 消元法解方程组

第一题复用了作业一中的[matrixCalc.py](#)，得到的结果如下：

```
(base) D:\PKU\notes\Computational Physics>python homework-2/T1.py
Cholesky消元法得到的矩阵：
+0.22360680 +0.31304952 +0.26832816 +0.22360680
+0.00000000 +0.04472136 -0.08944272 -0.00000000
+0.00000000 +0.00000000 +0.14142136 +0.21213203
+0.00000000 +0.00000000 +0.00000000 +0.07071068

[0.9999999999998886, 1.0000000000000666, 1.000000000000003, 0.999999999999819]
[1.0000000000001632, 0.999999999999899, 0.999999999999633, 1.000000000000222]
```

其中下方分别是通过GEM和Cholesky方法得到的方程组的解。注意到由于浮点数精度问题，得到的解并不严格等于1，而是存在很小的误差。

### 2. 三次样条插值

设 $f(x) = \cos(x^2)$ ，设 $[x_0, x_1]$ 与 $[x_1, x_2]$ 两端的三次样条函数分别为：

$$\hat{f}(x) = \begin{cases} a_1(x - x_0)^3 + b_1(x - x_0)^2 + c_1(x - x_0) + d_1 & x \in [x_0, x_1] \\ a_2(x - x_1)^3 + b_2(x - x_1)^2 + c_2(x - x_1) + d_2 & x \in [x_1, x_2] \end{cases}$$

#### (a) 已知端点二阶导数为零

可以列出约束方程：

$$\begin{aligned} \hat{f}''(x_0) &= \hat{f}''(x_2) = 0 \\ \hat{f}''(x_1 + 0) &= \hat{f}''(x_1 - 0) \\ \hat{f}'(x_1 + 0) &= \hat{f}'(x_1 - 0) \\ \hat{f}(x_0) &= f(x_0) \\ \hat{f}(x_1 - 0) &= \hat{f}(x_1 + 0) = f(x_1) \\ \hat{f}(x_2) &= f(x_2) \end{aligned}$$

得到：

$$\begin{aligned} b_1 &= 0 \\ 6a_2h_1 + 2b_2 &= 0 \end{aligned}$$

$$\begin{aligned}
6a_1h_0 + 2b_1 &= 2b_2 \\
3a_1h_0^2 + 2b_1h_0 + c_1 &= c_2 \\
d_1 &= f(x_0) \\
a_1h_0^3 + b_1h_0^2 + c_1h_0 + d_1 &= f(x_1) \\
d_2 &= f(x_1) \\
a_2h_1^3 + b_2h_1^2 + c_2h_1 + d_2 &= f(x_2)
\end{aligned}$$

其中

$$h_i = x_{i+1} - x_i$$

为步长,

化简得到:

$$\begin{aligned}
b_1 &= 0 \\
d_1 &= f(x_0) \\
d_2 &= f(x_1) \\
a_1 &= \frac{b_2}{3h_0} \\
a_2 &= -\frac{b_2}{3h_1} \\
\frac{b_2h_0^2}{3} + c_1h_0 &= f(x_1) - f(x_0) \\
\frac{-b_2}{3}h_1^2 + b_2h_1^2 + c_2h_1 &= f(x_2) - f(x_1) \\
b_2h_0 + c_1 &= c_2
\end{aligned}$$

解方程组, 并将 $f(x_i)$ 简记为

$$y_i = f(x_i)$$

得到:

$$\begin{aligned}
b_2 &= \frac{-3h_0y_1 + 3h_0y_2 + 3h_1y_0 - 3h_1y_1}{2h_0^2h_1 + 3h_0h_1^2 - h_1^3} \\
c_1 &= \frac{h_0^3y_1 - h_0^3y_2 - 3h_0^2h_1y_0 + 3h_0^2h_1y_1 - 3h_0h_1^2y_0 + 3h_0h_1^2y_1 + h_1^3y_0 - h_1^3y_1}{2h_0^3h_1 + 3h_0^2h_1^2 - h_0h_1^3} \\
c_2 &= \frac{-2h_0^3y_1 + 2h_0^3y_2 - 3h_0h_1^2y_0 + 3h_0h_1^2y_1 + h_1^3y_0 - h_1^3y_1}{2h_0^3h_1 + 3h_0^2h_1^2 - h_0h_1^3}
\end{aligned}$$

代入数值计算得到(见[T2.ipynb](#)):

$$\begin{aligned}
b_2 &= -1.19082 \\
c_1 &= 0.13132
\end{aligned}$$

$$c_2 = -0.58316$$

$$b_1 = 0.00000$$

$$d_1 = 1.00000$$

$$d_2 = 0.93590$$

$$a_1 = -0.66156$$

$$a_2 = 1.32313$$

## (b) 已知端点一阶导数

与上面的方程相比，该种情形下的约束方程只需要替换前两个式子即可：

$$\hat{f}'(x_0) = f'(x_0)$$

$$\hat{f}'(x_2) = f'(x_2)$$

$$\hat{f}''(x_1 + 0) = \hat{f}''(x_1 - 0)$$

$$\hat{f}'(x_1 + 0) = \hat{f}'(x_1 - 0)$$

$$\hat{f}(x_0) = f(x_0)$$

$$\hat{f}(x_1 - 0) = \hat{f}(x_1 + 0) = f(x_1)$$

$$\hat{f}(x_2) = f(x_2)$$

得到：

$$c_1 - t_0 = 0$$

$$3a_2h_1^2 + 2b_2h_1 + c_2 - t_2 = 0$$

$$6a_1h_0 + 2b_1 = 2b_2$$

$$3a_1h_0^2 + 2b_1h_0 + c_1 = c_2$$

$$d_1 = f(x_0)$$

$$a_1h_0^3 + b_1h_0^2 + c_1h_0 + d_1 = f(x_1)$$

$$d_2 = f(x_1)$$

$$a_2h_1^3 + b_2h_1^2 + c_2h_1 + d_2 = f(x_2)$$

其中

$$t_0 = f'(0) = -2x \sin x^2|_{x=0} = 0$$

$$t_2 = f'(0.9) \approx -1.30372$$

$$a_1 = -0.62916$$

$$a_2 = -1.12487$$

$$b_1 = 0.19943$$

$$b_2 = -0.93305$$

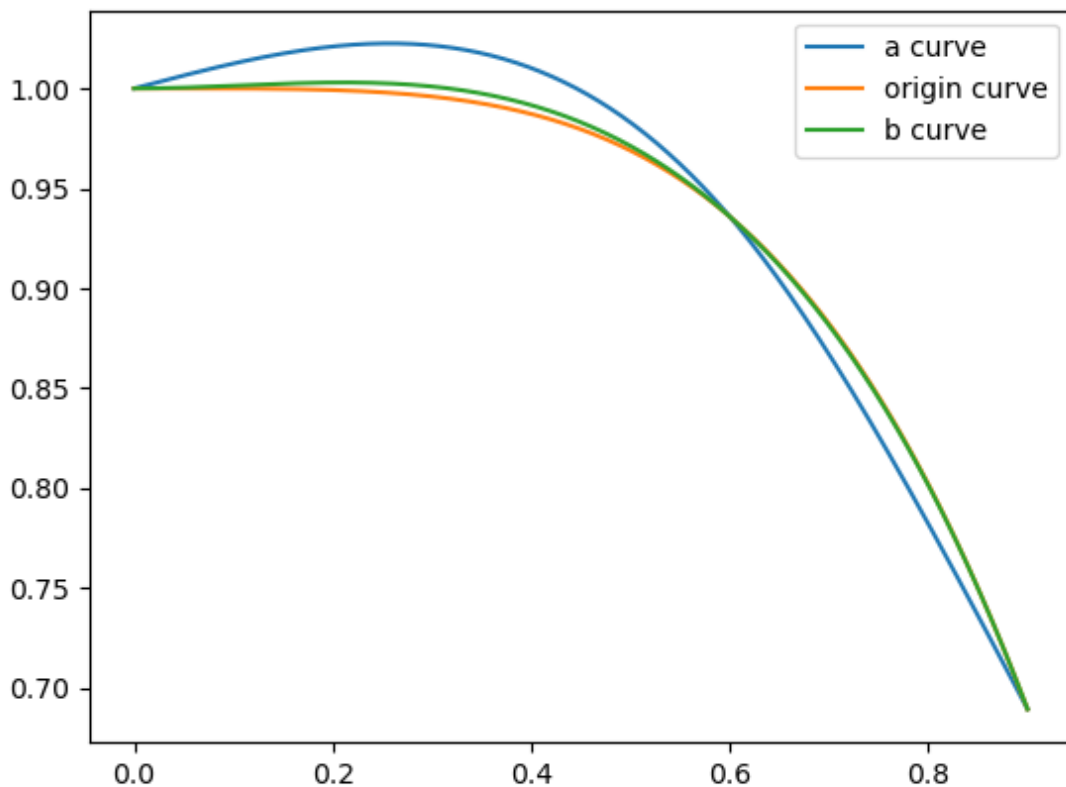
$$c_1 = 0.00000$$

$$c_2 = -0.44017$$

$$d_1 = 1.00000$$

$$d_2 = 0.93590$$

通过画图比较可知，第二种的拟合效果要更好。



### 3.Chebyshev多项式展开

展开过程：首先将 $\log_2(x)$ ,  $x \in [1, 2]$ 转换为 $[0, 1]$ 之间的函数：通过变量代换： $x' = 2\frac{x-1}{2-1} - 1 = 2x - 3$ ，得到新的函数为：

$$f(x') = f_0(x) = \log_2 \frac{x' + 3}{2}$$

$f(x') = \sum_{n=0}^{\infty} c_n T_n(x')$ ，根据正交归一条件得到系数 $c_n$ 为：

$$c_n = \int_{-1}^1 \log_2 \frac{x' + 3}{2} \frac{T_n(x')}{\sqrt{1-x'^2}} dx' \frac{1}{\frac{\pi}{2}(1 + \delta_{n0})}$$

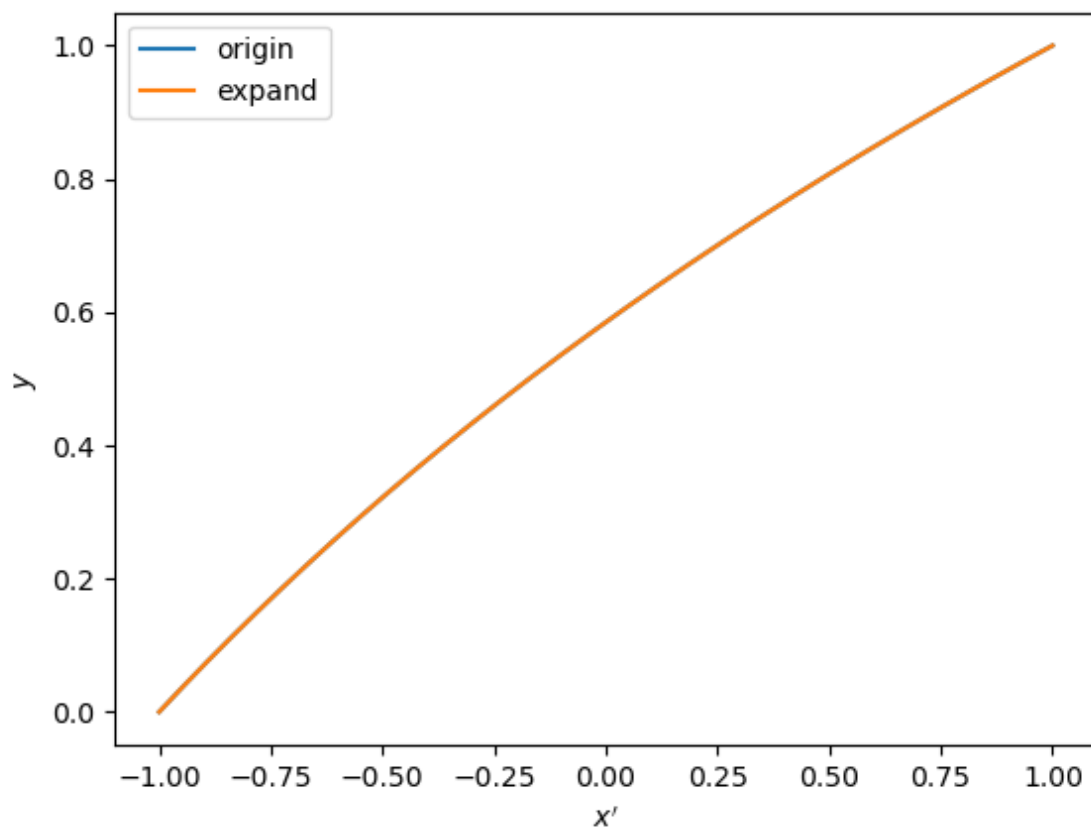
作代换： $x' = \cos \theta$ ，由 $T_n(x) = \cos n \arccos x$ ，

$$c_n = \int_0^\pi f(\cos \theta) \cos n\theta d\theta \frac{(1 - \delta_{n0})}{\pi}$$

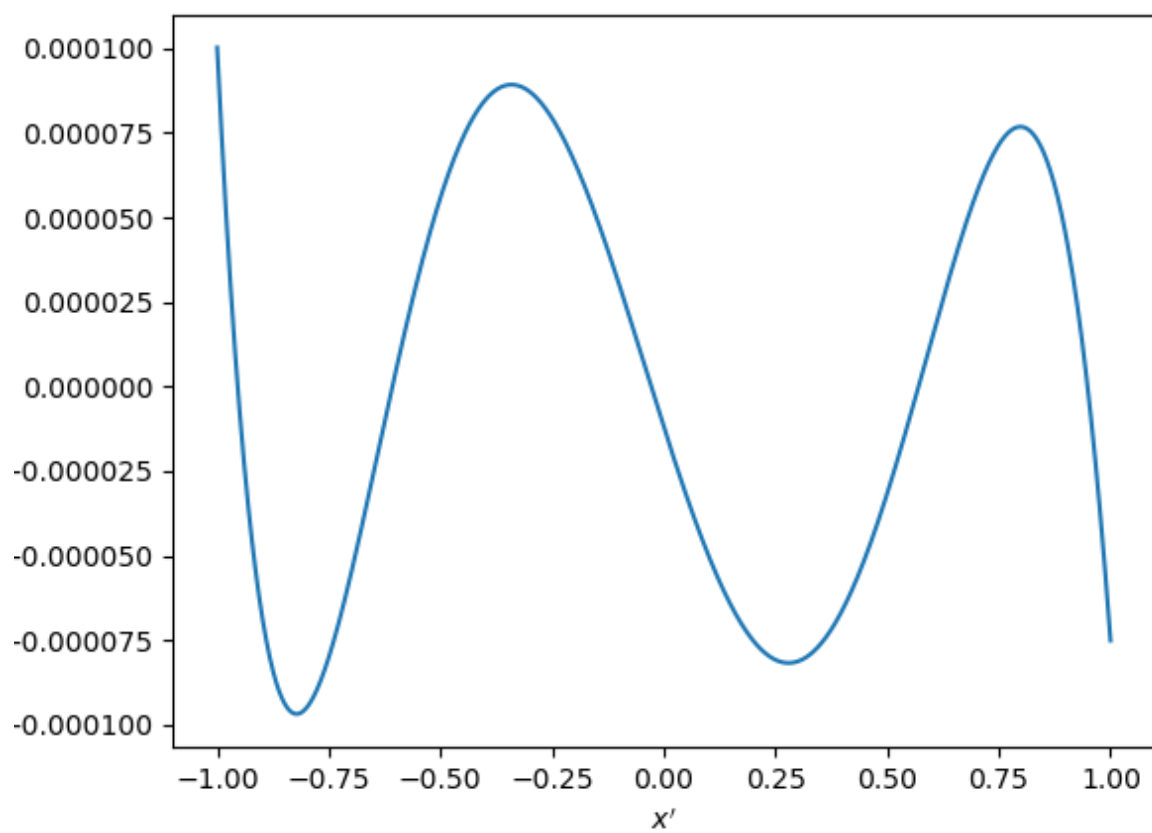
取上式的离散近似，并截断求和为N项： $\theta_k = \frac{\pi(k+1/2)}{N}, \Delta\theta_k = \pi/N$

$$c_n \approx c_{n,N} = \frac{2 - \delta_{0m}}{N} \sum_{k=0}^{N-1} f\left(\cos \frac{\pi(k+1/2)}{N}\right) \cos m\pi \frac{k+1/2}{N}$$

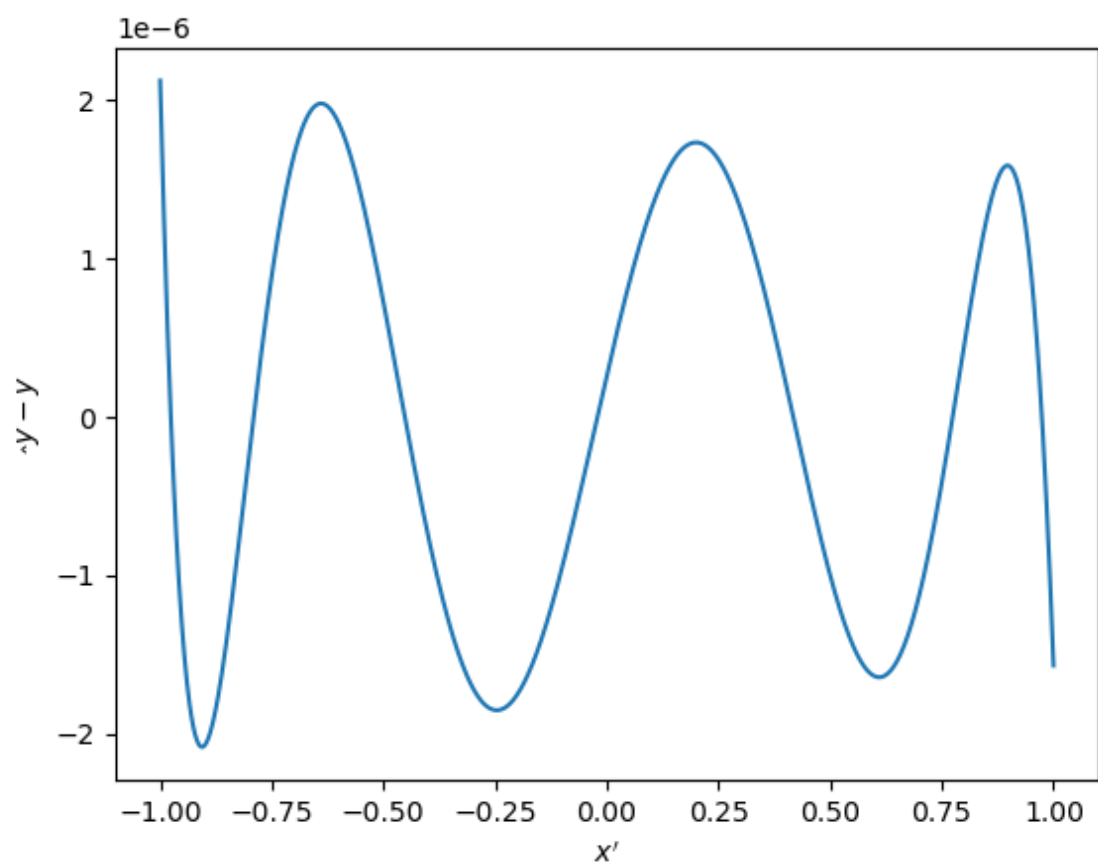
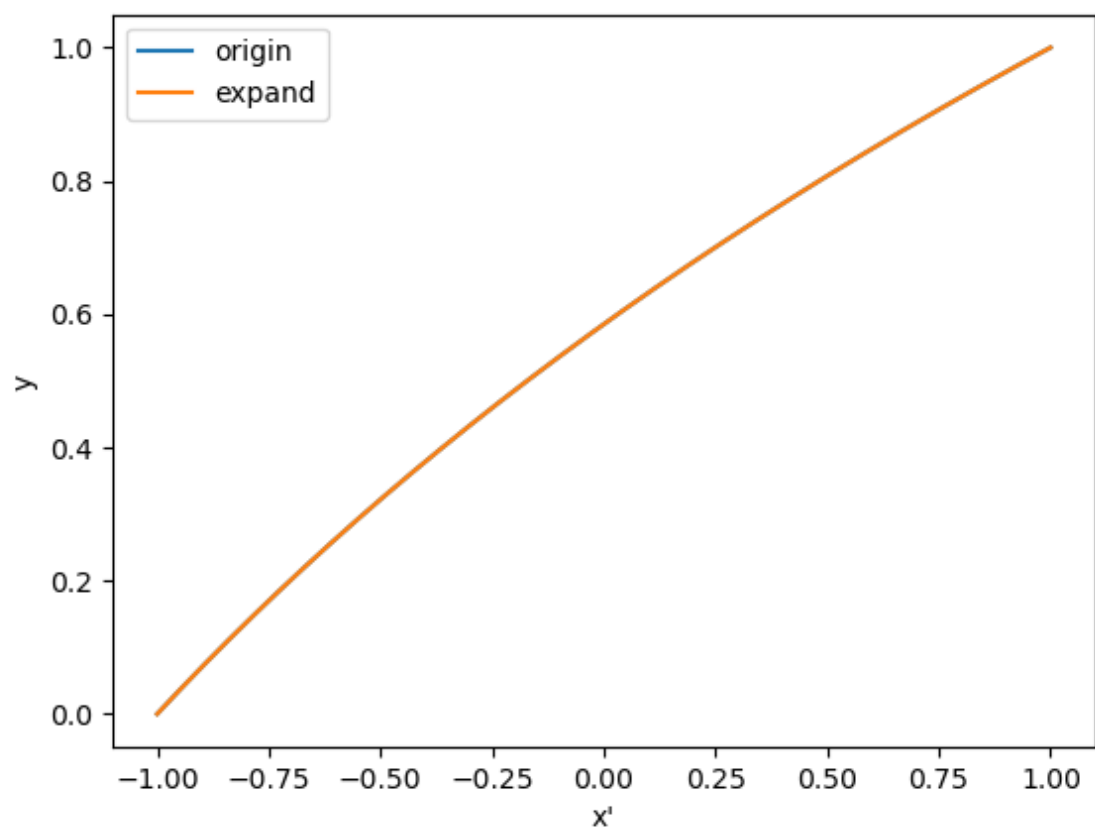
取0-4阶进行比较作图，可以看到二者的曲线十分接近：



作差分析误差，可以看到误差在 $\sim 10^{-5}$ 量级，而且在0两侧接近对称分布：



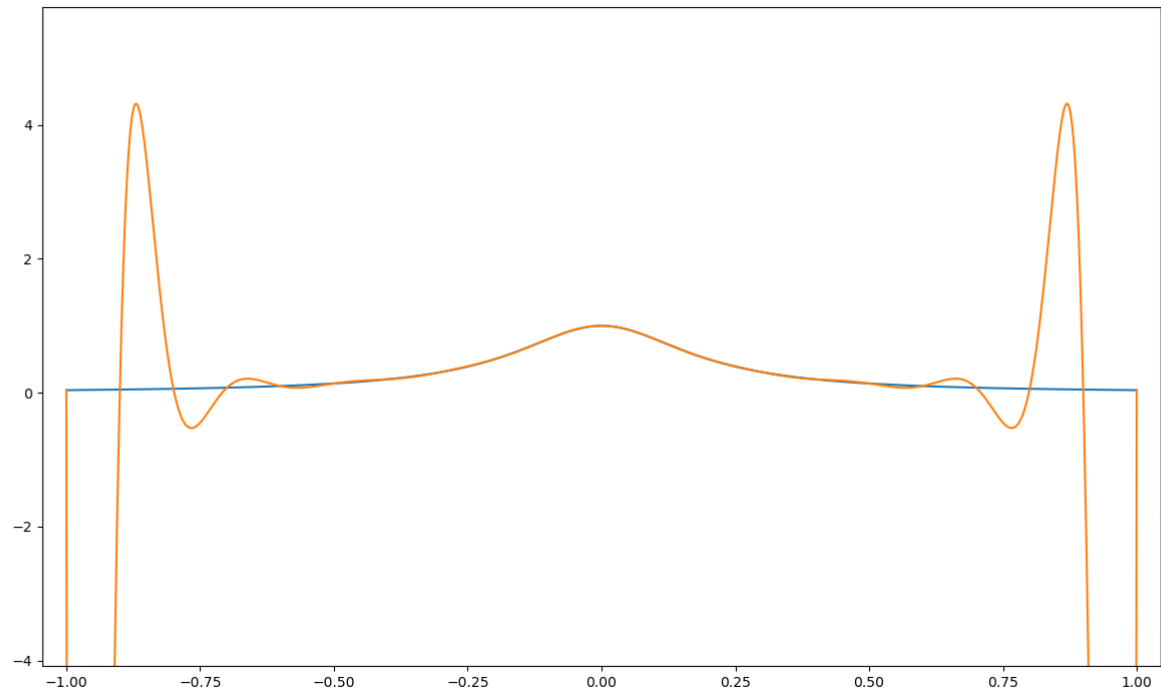
对0-6阶进行同样的分析，得到的误差将变得更小：



# 4.Runge效应

## a. Lagrange插值

通过拉格朗日插值法得到 $P_{20}(x)$ ，作图得到：



拉格朗日插值法在插值区域中心处近似比较好，但在两端会出现剧烈震荡。

做表格如下所示：

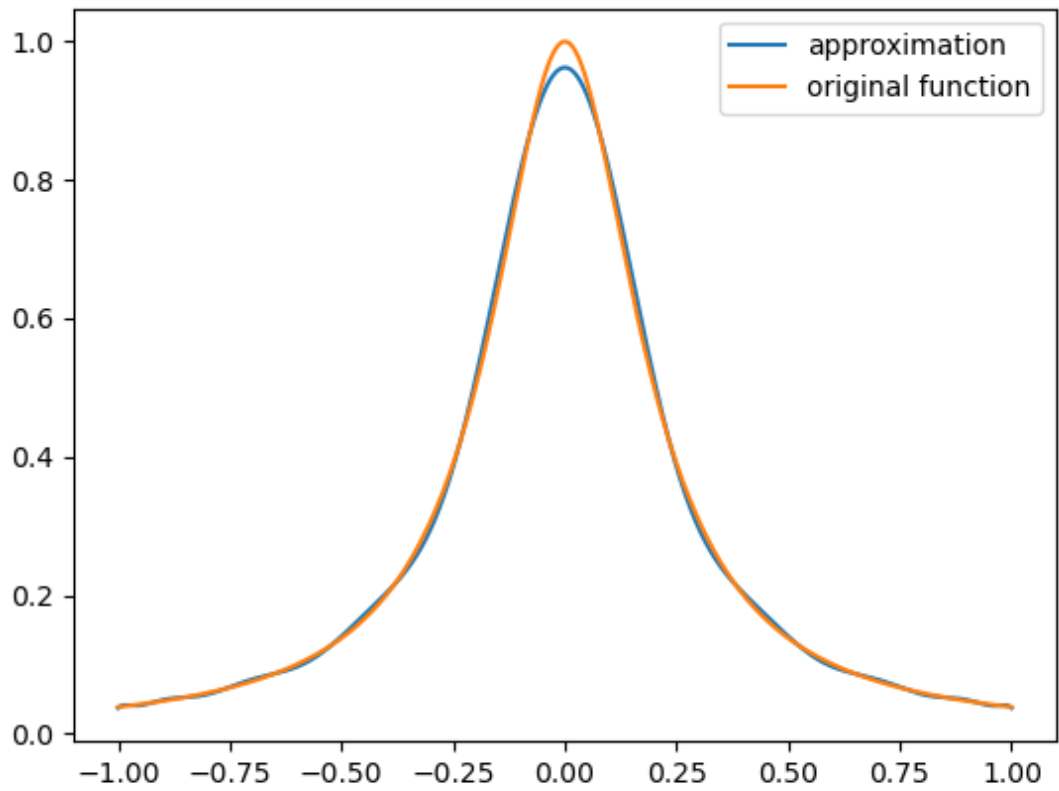


x	origin	Lagrange	abs
-1.0000	+0.0385	+0.0385	+0.0000
-0.9500	+0.0424	-39.9524	+39.9949
-0.9000	+0.0471	+0.0471	+0.0000
-0.8500	+0.0525	+3.4550	+3.4025
-0.8000	+0.0588	+0.0588	+0.0000
-0.7500	+0.0664	-0.4471	+0.5134
-0.7000	+0.0755	+0.0755	+0.0000
-0.6500	+0.0865	+0.2024	+0.1159
-0.6000	+0.1000	+0.1000	+0.0000
-0.5500	+0.1168	+0.0807	+0.0361
-0.5000	+0.1379	+0.1379	+0.0000
-0.4500	+0.1649	+0.1798	+0.0148
-0.4000	+0.2000	+0.2000	+0.0000
-0.3500	+0.2462	+0.2384	+0.0077
-0.3000	+0.3077	+0.3077	+0.0000
-0.2500	+0.3902	+0.3951	+0.0048
-0.2000	+0.5000	+0.5000	+0.0000
-0.1500	+0.6400	+0.6368	+0.0032
-0.1000	+0.8000	+0.8000	+0.0000
-0.0500	+0.9412	+0.9425	+0.0013
0.0000	+1.0000	+1.0000	+0.0000
0.0500	+0.9412	+0.9425	+0.0013
0.1000	+0.8000	+0.8000	+0.0000
0.1500	+0.6400	+0.6368	+0.0032
0.2000	+0.5000	+0.5000	+0.0000
0.2500	+0.3902	+0.3951	+0.0048
0.3000	+0.3077	+0.3077	+0.0000
0.3500	+0.2462	+0.2384	+0.0077
0.4000	+0.2000	+0.2000	+0.0000
0.4500	+0.1649	+0.1798	+0.0148
0.5000	+0.1379	+0.1379	+0.0000
0.5500	+0.1168	+0.0807	+0.0361
0.6000	+0.1000	+0.1000	+0.0000
0.6500	+0.0865	+0.2024	+0.1159
0.7000	+0.0755	+0.0755	+0.0000
0.7500	+0.0664	-0.4471	+0.5134
0.8000	+0.0588	+0.0588	+0.0000
0.8500	+0.0525	+3.4550	+3.4025
0.9000	+0.0471	+0.0471	+0.0000
0.9500	+0.0424	-39.9524	+39.9949

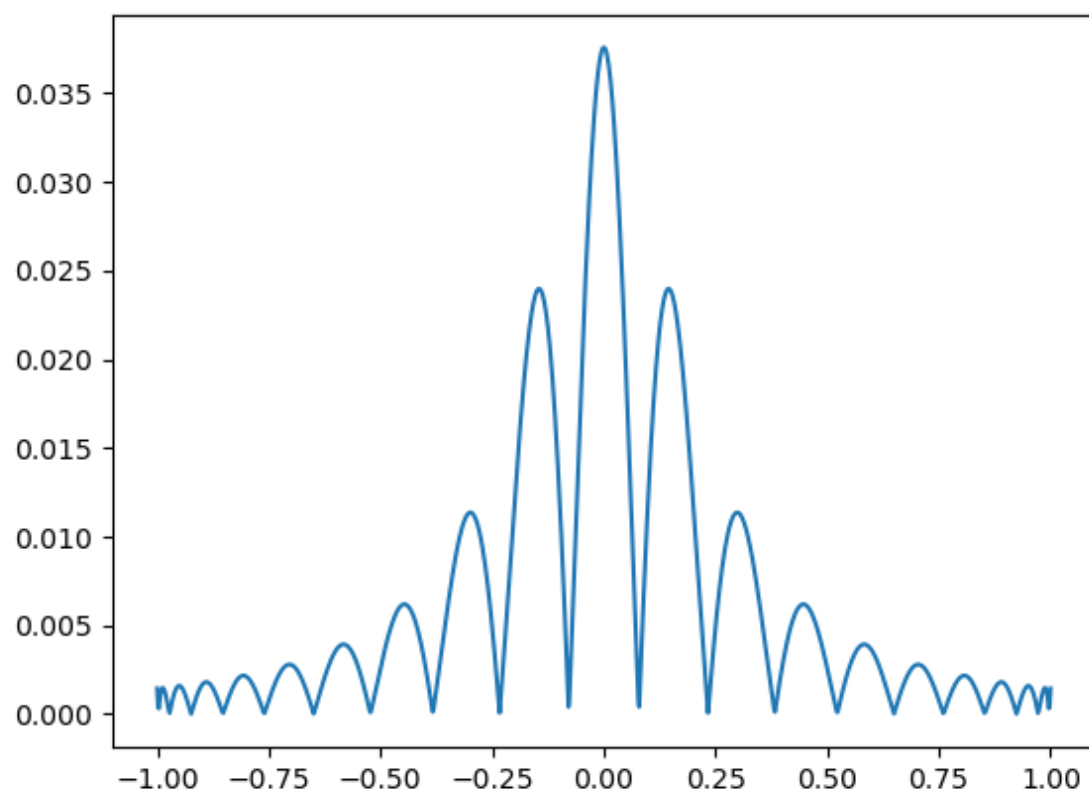
```
0.9500 +0.0424 +39.9524 +39.9549
1.0000 +0.0385 +0.0385 +0.0000
```

b. Chebyshev多项式插值

进行Chebyshev多项式插值，得到的结果如下图所示：



计算二者差的绝对值，作图得到：



或用数据表表示：

x	origin	Chebyshev	abs
-----			
-1.0000	+0.0385	+0.0370	+0.0014
-0.9500	+0.0424	+0.0408	+0.0016
-0.9000	+0.0471	+0.0487	+0.0016
-0.8500	+0.0525	+0.0523	+0.0002
-0.8000	+0.0588	+0.0567	+0.0021
-0.7500	+0.0664	+0.0672	+0.0008
-0.7000	+0.0755	+0.0783	+0.0028
-0.6500	+0.0865	+0.0865	+0.0000
-0.6000	+0.1000	+0.0964	+0.0036
-0.5500	+0.1168	+0.1141	+0.0027
-0.5000	+0.1379	+0.1405	+0.0026
-0.4500	+0.1649	+0.1711	+0.0062
-0.4000	+0.2000	+0.2028	+0.0028
-0.3500	+0.2462	+0.2402	+0.0060
-0.3000	+0.3077	+0.2963	+0.0114
-0.2500	+0.3902	+0.3853	+0.0049
-0.2000	+0.5000	+0.5119	+0.0119
-0.1500	+0.6400	+0.6639	+0.0239
-0.1000	+0.8000	+0.8126	+0.0126
-0.0500	+0.9412	+0.9221	+0.0191
0.0000	+1.0000	+0.9624	+0.0376
0.0500	+0.9412	+0.9221	+0.0191
0.1000	+0.8000	+0.8126	+0.0126
0.1500	+0.6400	+0.6639	+0.0239
0.2000	+0.5000	+0.5119	+0.0119
0.2500	+0.3902	+0.3853	+0.0049
0.3000	+0.3077	+0.2963	+0.0114
0.3500	+0.2462	+0.2402	+0.0060
0.4000	+0.2000	+0.2028	+0.0028
0.4500	+0.1649	+0.1711	+0.0062
0.5000	+0.1379	+0.1405	+0.0026
0.5500	+0.1168	+0.1141	+0.0027
0.6000	+0.1000	+0.0964	+0.0036
0.6500	+0.0865	+0.0865	+0.0000
0.7000	+0.0755	+0.0783	+0.0028
0.7500	+0.0664	+0.0672	+0.0008
0.8000	+0.0588	+0.0567	+0.0021
0.8500	+0.0525	+0.0523	+0.0002
0.9000	+0.0471	+0.0487	+0.0016
0.9500	+0.0424	+0.0408	+0.0016

```
0.9988 10.0424 10.0408 10.0010
1.0000 +0.0385 +0.0370 +0.0014
```

可以看到，切比雪夫近似在 $x = 0$ 附近的误差最大，相对误差远小于拉格朗日插值近似

### c. 三次样条函数插值

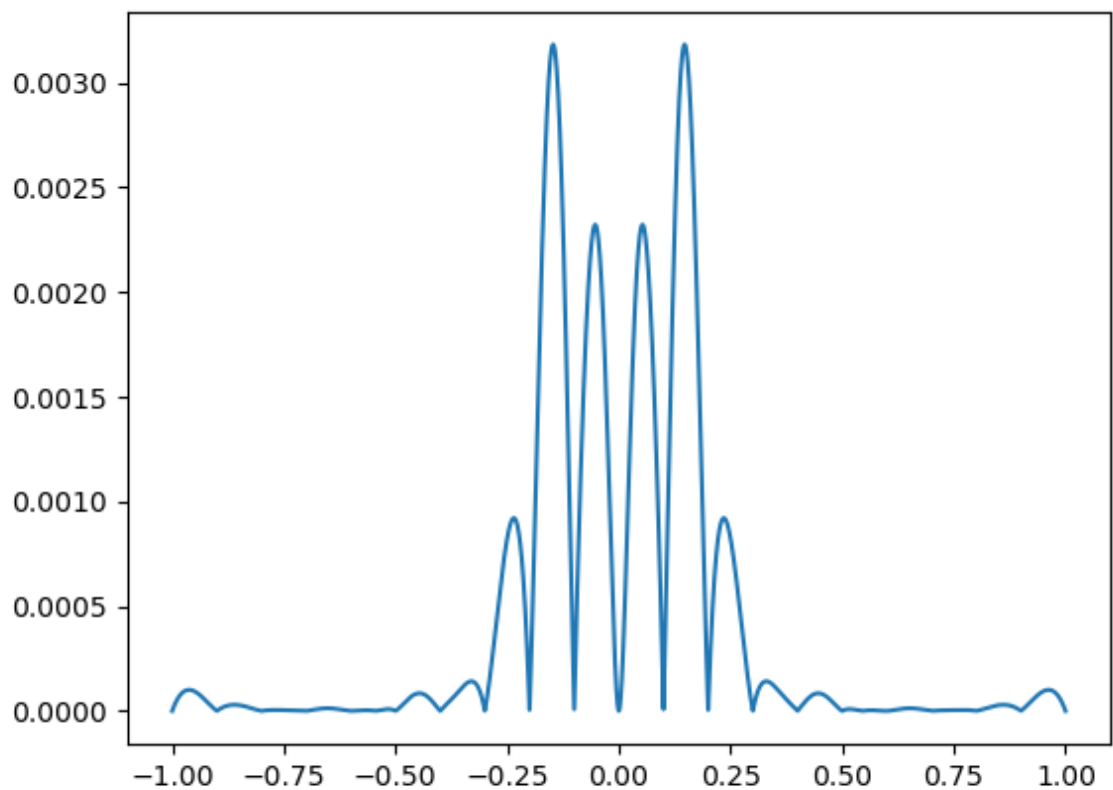
编写三次样条插值程序[spline\\_interpolation.py](#)和自动测试程序[autotest.py](#)，得到三次样条函数插值曲线（由于该种插值方法与原函数贴合紧密，无法看出区别，故只作出二者绝对值随 $x$ 变化）：

数据表：

```
(base) D:\PKU\notes\Computational Physics\homework-2>python T4_cubic_spline.py
```

```
-----  
-1.0000 +0.0385 +0.0385 +0.0000  
-0.9500 +0.0424 +0.0425 +0.0001  
-0.9000 +0.0471 +0.0471 +0.0000  
-0.8500 +0.0525 +0.0524 +0.0000  
-0.8000 +0.0588 +0.0588 +0.0000  
-0.7500 +0.0664 +0.0664 +0.0000  
-0.7000 +0.0755 +0.0755 +0.0000  
-0.6500 +0.0865 +0.0865 +0.0000  
-0.6000 +0.1000 +0.1000 +0.0000  
-0.5500 +0.1168 +0.1168 +0.0000  
-0.5000 +0.1379 +0.1379 +0.0000  
-0.4500 +0.1649 +0.1649 +0.0001  
-0.4000 +0.2000 +0.2000 +0.0000  
-0.3500 +0.2462 +0.2463 +0.0001  
-0.3000 +0.3077 +0.3077 +0.0000  
-0.2500 +0.3902 +0.3894 +0.0008  
-0.2000 +0.5000 +0.5000 +0.0000  
-0.1500 +0.6400 +0.6432 +0.0032  
-0.1000 +0.8000 +0.8000 +0.0000  
-0.0500 +0.9412 +0.9389 +0.0023  
0.0000 +1.0000 +1.0000 +0.0000  
0.0500 +0.9412 +0.9389 +0.0023  
0.1000 +0.8000 +0.8000 +0.0000  
0.1500 +0.6400 +0.6432 +0.0032  
0.2000 +0.5000 +0.5000 +0.0000  
0.2500 +0.3902 +0.3894 +0.0008  
0.3000 +0.3077 +0.3077 +0.0000  
0.3500 +0.2462 +0.2463 +0.0001  
0.4000 +0.2000 +0.2000 +0.0000  
0.4500 +0.1649 +0.1649 +0.0001  
0.5000 +0.1379 +0.1379 +0.0000  
0.5500 +0.1168 +0.1168 +0.0000  
0.6000 +0.1000 +0.1000 +0.0000  
0.6500 +0.0865 +0.0865 +0.0000  
0.7000 +0.0755 +0.0755 +0.0000  
0.7500 +0.0664 +0.0664 +0.0000  
0.8000 +0.0588 +0.0588 +0.0000  
0.8500 +0.0525 +0.0524 +0.0000  
0.9000 +0.0471 +0.0471 +0.0000  
0.9500 +0.0424 +0.0425 +0.0001  
1.0000 +0.0385 +0.0385 +0.0000
```

图像:



可以看到，二者最大误差在0.003左右，是切比雪夫多项式的拟合最大误差的1/10。

## 5. 样条函数在计算机绘图中的应用

(a)

选取 $t = 0, 1, 2, \dots, 8$ , 给出 $x_t = r(\phi) \cos \phi$ 和 $y_t = r(\phi) \sin \phi$ 的数值。将数值作为精确的数值列到一个表中。

$t$	$x_t$	$y_t$
0	0	0
1	$\frac{\sqrt{2}\left(1-\frac{\sqrt{2}}{2}\right)}{2}$	$\frac{\sqrt{2}\left(1-\frac{\sqrt{2}}{2}\right)}{2}$
2	0	1
3	$-\frac{\sqrt{2}\left(\frac{\sqrt{2}}{2}+1\right)}{2}$	$\frac{\sqrt{2}\left(\frac{\sqrt{2}}{2}+1\right)}{2}$
4	-2	0

$t$	$x_t$	$y_t$
5	$-\frac{\sqrt{2}\left(\frac{\sqrt{2}}{2}+1\right)}{2}$	$\frac{\sqrt{2}\left(\frac{\sqrt{2}}{2}+1\right)}{2}$
6	0	-1
7	$\frac{\sqrt{2}\left(1-\frac{\sqrt{2}}{2}\right)}{2}$	$-\frac{\sqrt{2}\left(1-\frac{\sqrt{2}}{2}\right)}{2}$

b)给出三次样条函数 $S_{\Delta}(X;t)$ 和 $S_{\Delta}(Y;t)$

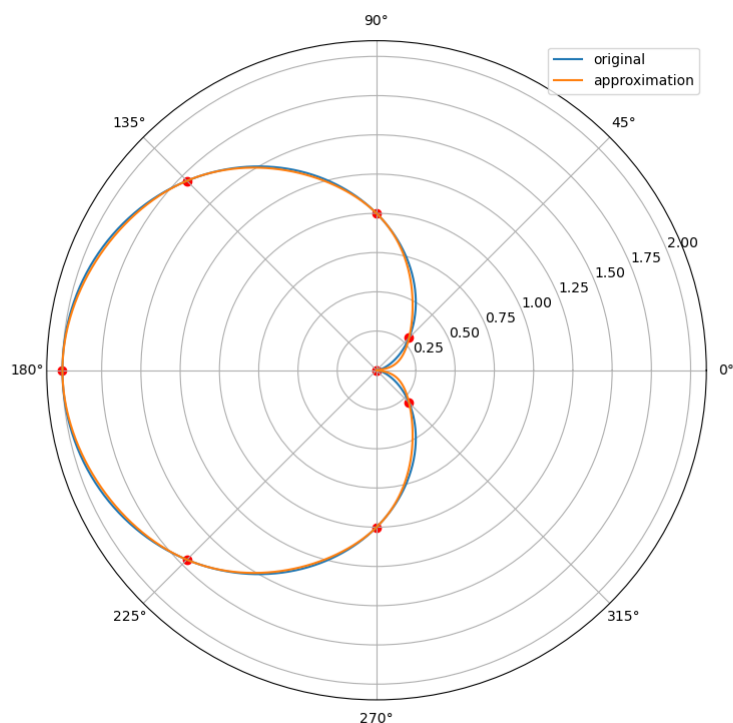
两个样条函数的表达式比较繁琐，各段的系数 $a, b, c, d$ 请通过运行[T5.ipynb](#)获得，从而可以写出其分段表达式：

$$S_{\Delta}(X, t) = a_t + b_t(x - x_t) + c_t(x - x_t)^2 + d_t(x - x_t)^3, \quad x \in [x_t, x_{t+1}]$$

$$S_{\Delta}(Y, t) = a_t + b_t(y - y_t) + c_t(y - y_t)^2 + d_t(y - y_t)^3, \quad y \in [y_t, y_{t+1}]$$

(c)画出曲线，比较

作图如下所示。插值点已经在图中标出。



(d)



该算法可以平滑连接所有点，是因为在该算法中， $x$ 和 $y$ 都是关于参数 $t$ 的三次样条插值函数，三次样条插值是二阶导数连续的，因此当参数 $t$ 连续变化时，它们形成的点 $(x(t), y(t))$ 也将构成一条平滑曲线。