

# 计算物理第八次作业

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## 第一题

### 理论

$$\nabla^2 = (\hat{i} \frac{\partial}{\partial x} + \hat{j} \frac{\partial}{\partial y}) = \frac{\partial^2}{\partial x^2} + \frac{\partial^2}{\partial y^2}$$

二阶偏微分差分格式为:

$$\frac{\partial^2 V}{\partial x^2} = \frac{V(i+1, j) + V(i-1, j) - 2V(i, j)}{\Delta x^2}$$

由此差分Laplace方程得:

$$V(i, j) = \frac{V(i+1, j) + V(i-1, j) + V(i, j+1) + V(i, j-1)}{4}$$

### 代码实现

首先导入库:

```
import numpy as np
import numba
from numba import jit
from mpl_toolkits.mplot3d import Axes3D
import matplotlib.pyplot as plt
from matplotlib import cm
from matplotlib.ticker import LinearLocator, FormatStrFormatter
```

然后实现 `Capcitorsolver` 类:

```
class Capcitorsolver():
    def __init__(self, geometry = [(-1,-1),(1,1)], meshLeng = 0.001, relaxError = 0.00001):
        self.geo = geometry
        self.size = (int((self.geo[1][0] - self.geo[0][0])/meshLeng),
int((self.geo[1][1] - self.geo[0][1])/meshLeng))
        self.dl = meshLeng
        self.v = np.zeros(self.size)
        self.r = relaxError
    @jit()
    def __boundaryCondition(self):
        self.v[0,:] = 0
        self.v[-1,:] = 0
        self.v[:,0] = 0
        self.v[:, -1] = 0
        self.v[int(self.size[0]/3), int(self.size[1]/4):int(self.size[1]*3/4)] =
1
        self.v[int(self.size[0]*2/3), int(self.size[1]/4):int(self.size[1]*3/4)]
= -1
```

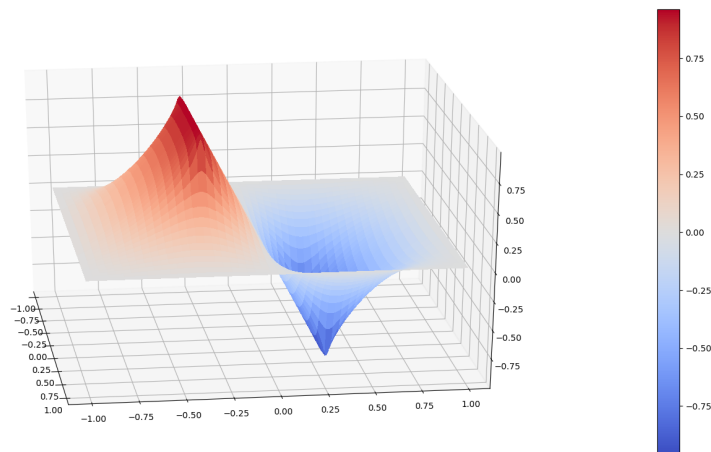
```

@jit()
def __update(self):
    new = (self.v[0:-2, 1:-1] + self.v[2:, 1:-1] + self.v[1:-1, 2:] +
self.v[1:-1, 0:-2])/4
    delta_v = np.mean(np.abs(new - self.v[1:-1, 1:-1]))
    print(delta_v)
    self.v[1:-1, 1:-1] = new
    if delta_v > self.r:
        return False
    else:
        return True
def calc(self):
    flag = False
    while not flag:
        self.__boundaryCondition()
        flag = self.__update()
def show(self):
    X = np.arange(self.geo[0][0], self.geo[1][0], self.dl)
    Y = np.arange(self.geo[0][1], self.geo[1][1], self.dl)
    X, Y = np.meshgrid(X, Y)
    fig = plt.figure()
    ax = fig.gca(projection='3d')
    surf = ax.plot_surface(X, Y, self.v, cmap=cm.coolwarm, linewidth=0,
antialiased=False)
    fig.colorbar(surf)
    plt.show()

```

对于 `__boundaryCondition` 和 `__update` 方法,我们使用 `numba` 库进行 `jit` 编译,以提升运行速度.

## 结果展示



可见,计算结果还算准确.

## 第二题

## 问题分析

首先,在我们的算法中,较多的用到矩阵的切片,而不需要用到矩阵之间的加法和乘法,因此,我们选用 `lil_matrix` (List of lists format)

关于 `scipy` 中各种不同稀疏矩阵储存方式的区别,请参见:<https://blog.csdn.net/jeffery0207/article/details/100064602>

我们可以直接继承自第一题中的 `CapcitorSolver` 类,但是我们需要重写 `__init__` 方法和 `show` 方法:

- 在 `__init__` 方法中,我们需要重写,以将电势转化为稀疏矩阵储存
- 在 `show` 方法中,我们需要重写,因为 `matplotlib` 不支持稀疏矩阵形式的输入,我们需要调用稀疏矩阵的 `toarray` 方法,来实现成功的绘图.

## 代码实现

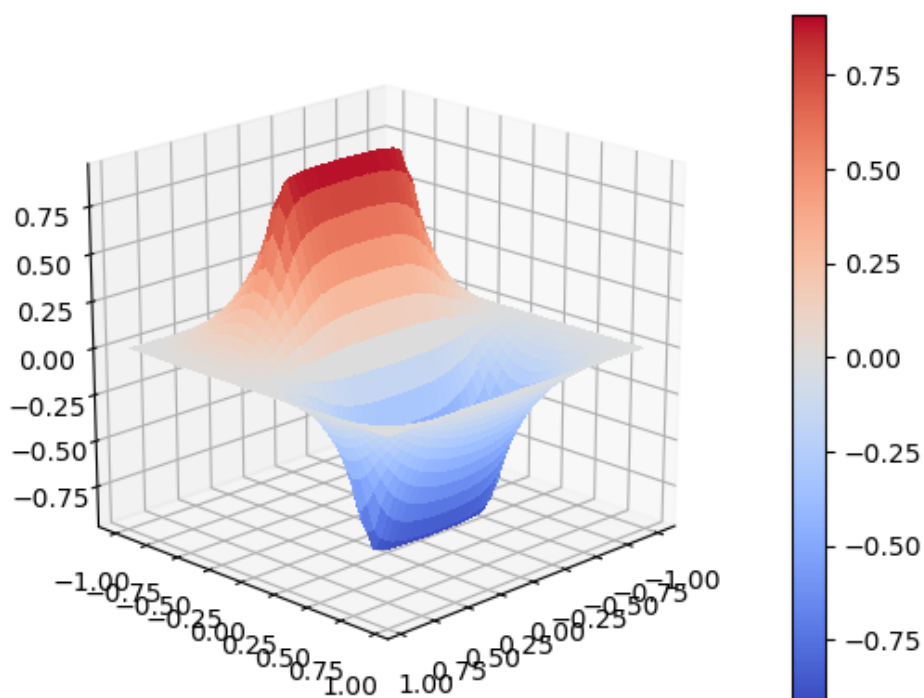
引入相关库

```
import numpy as np
from Q1 import CapcitorSolver
from scipy.sparse import lil_matrix
from mpl_toolkits.mplot3d import Axes3D
import matplotlib.pyplot as plt
from matplotlib import cm
from matplotlib.ticker import LinearLocator, FormatStrFormatter
```

继承并重写类:

```
class sparseMatrixCapcitorSolver(CapcitorSolver):
    def __init__(self, geometry = [(-1,-1),(1,1)], meshLeng = 0.05, relaxError = 0.003):
        super().__init__(geometry, meshLeng, relaxError)
        print("Common matrix occupied {} bytes of
memory.".format(self.v.__sizeof__()))
        self.v[0,:] = 0
        self.v[-1,:] = 0
        self.v[:,0] = 0
        self.v[:, -1] = 0
        self.v[int(self.size[0]/3), int(self.size[1]/4):int(self.size[1]*3/4)] = 1
        self.v[int(self.size[0]*2/3), int(self.size[1]/4):int(self.size[1]*3/4)] = -1
        self.v = lil_matrix(self.v)
        print("Sparse matrix occupied {} bytes of
memory.".format(self.v.__sizeof__()))
    def show(self):
        X = np.arange(self.geo[0][0], self.geo[1][0], self.dl)
        Y = np.arange(self.geo[0][1], self.geo[1][1], self.dl)
        X, Y = np.meshgrid(X, Y)
        fig = plt.figure()
        ax = fig.gca(projection='3d')
        surf = ax.plot_surface(X, Y,
self.v.toarray(), cmap=cm.coolwarm, linewidth=0, antialiased=False)
        fig.colorbar(surf)
        plt.show()
```

## 结果展示



计算出来的结果自然是相同的.但是值得注意的是空间占用情况.在代码中,我们调用了 python 内置的 `__sizeof__` 方法来获取对象的内存占用,可以看出,对于网格长度为0.05时,

```
Common matrix occupied 12912 bytes of memory.  
Sparse matrix occupied 32 bytes of memory.
```

二者占用空间已存在较大差距.

如果进一步,将网格长度缩小到0.0001时:

```
Common matrix occupied 3200000112 bytes of memory.  
Sparse matrix occupied 32 bytes of memory.
```

可见,此时非稀疏矩阵的内存占用远远大于稀疏矩阵.