PS8 | Muath Hamidi

```
In [1]:
       2 # Course: PHYS 512
     3 # Problem: PS8
     5 # By: Muath Hamidi
     6 # Email: muath.hamidi@mail.mcgill.ca
     7 # Department of Physics, McGill University
      # November 2022
     10
       # Libraries
     11
     13
       import numpy as np # For math
       import matplotlib.pyplot as plt # For graphs
     14
     15
```

Problem 1

The solution is a complex exponential and has the form;

$$f(x,t) = \xi^t exp(ikx)$$

Substitute in the leapfrog scheme

$$\frac{exp(ikx)(\xi^{t+dt} - \xi^{t-dt})}{2dt} = \frac{exp(ikx)\xi^{t}(\xi^{+dt} - \xi^{-dt})}{2dt}$$

Take this as the LHS. Now, the RHS;

$$-v\frac{exp(ikx)\xi^{t}(exp(+ikdx) - exp(-ikdx))}{2dx}$$

Match the both sides;

$$\frac{exp(ikx)\xi^{t}(\xi^{+dt} - \xi^{-dt})}{2dt} = -v\frac{exp(ikx)\xi^{t}(exp(+ikdx) - exp(-ikdx))}{2dx}$$

So,

$$\to \xi^{+dt} - \xi^{-dt} = v \frac{dt}{dx} (2isin(kdx))$$

Multiply both sides by ξ^{+dt}

$$\rightarrow \xi^{2dt} - 2iv \frac{dt}{dx} \xi^{dt} sin(kdx) - 1 = 0$$

So, the solution is:

This is a complex solution. If $v \frac{dt}{dx} \le 1$ then $|\xi^{dt}|^2 = 1$. If $v \frac{dt}{dx} > 1$ then the solution is not in our interest.

Problem 2

First, let's use the average neighbours method to find the potential;

```
In [137]:
          # Part a
        # Functions
        4
          5
          def average_neighbors(mat):
        6
        7
             out=0*mat
        8
             out=out+np.roll(mat,1,0)
        9
             out=out+np.roll(mat,-1,0)
       10
             out=out+np.roll(mat,1,1)
       11
             out=out+np.roll(mat,-1,1)
             return out/4
       12
       13
          #-----
       14
       15
          # Average Neighbours Method
       17 | size = 200 |
       18 s2 = int(abs(size/2))
       19 | V = np.zeros([size,size]) # V array
       20 V[s2, s2] = 1 # origion
       21 iterations = 10000
       22
       23
          for i in np.arange(iterations):
            V = average_neighbors(V)
       24
       25
            V[s2, s2] +=1 # origin
       26
       27
       28
         scale = 1 - V[s2, s2]
          V = V + scale
       29
       30
       31
          32 # Print Chosen Values
       r0 = V[s2, s2] + average_neighbors(V)[s2, s2]
       35 \ v0 = V[s2, s2]
       36 \text{ v1} = \text{V[s2 + 1, s2]}
       37 \ v2 = V[s2 + 2, s2]
       38 v5 = V[s2 + 5, s2]
       39 print("V[0,0] = ", v0)
       40 print("rho[0,0] = ",r0)
       41 print("V[1,0] = ", v1)
       42 print("V[2,0] = ", v2)
          print("V[5,0] = ", v5)
       43
       44
       46 # Plot - Greens' Function
       48 plt.imshow(V, cmap="Blues_r")
       49 plt.colorbar()
       50 plt.title("Potential")
       V[0,0] = 1.0
       rho[0,0] = 1.0
```

.

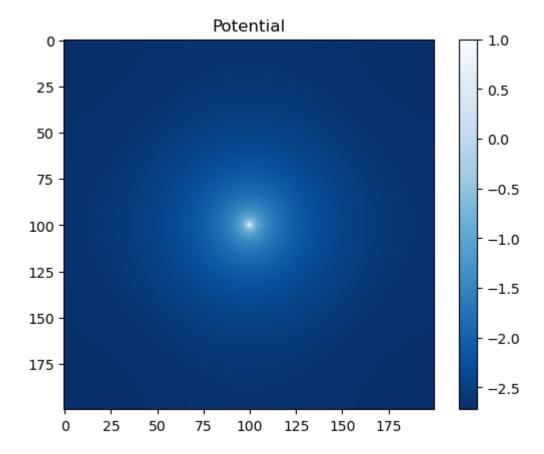
```
rho[0,0] = 1.0

V[1,0] = 0.0

V[2,0] = -0.4533841134643257

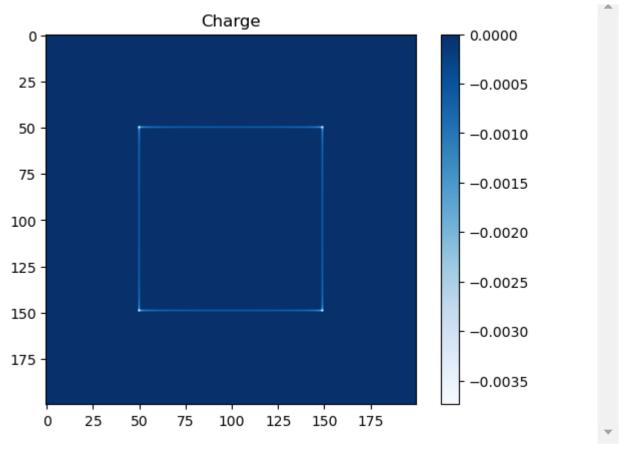
V[5,0] = -1.050788709610979
```



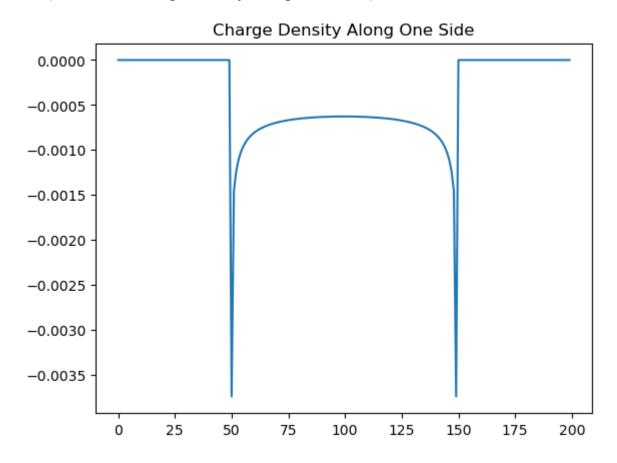


Nice and realistic, with true potential values as in the question.

```
In [138]:
          #-----
          # Part b
        3
          # Functions
        4
          dim = s2 # the square (shape) mask dimension
         7
          dim2 = int(dim/2)
          mask = np.zeros([size, size], dtype = bool)
        9
          mask[dim - dim2 : dim + dim2, dim - dim2 : dim + dim2] = 1
        10
        11
          # Functions
        12
        13
          #-----
          def MaskConv(rho, mask):
        14
             rho mask = np.zeros(mask.shape)
        15
        16
             rho_mask[mask] = rho
        17
             return np.fft.fftshift(np.fft.irfft2(np.fft.rfft2(V) * np.fft.rfft2(rho
        18
        19
          def conjgrad(A, b, mask, x): # Conjugate gradient
             iterarions = 5000
        20
        21
             r = b - A(x, mask)
        22
             p = r.copy()
        23
             rtr = np.sum(r**2)
        24
        25
             for i in range(iterarions):
                Ap=A(p, mask)
        26
                pAp=np.sum(p*Ap)
        27
        28
                alpha=rtr/pAp
        29
                x=x+alpha*p
                r=r-alpha*Ap
        30
        31
                rtr new=np.sum(r**2)
                beta=rtr new/rtr
        32
        33
                p=r+beta*p
                rtr=rtr_new
        34
        35
             return x
        36
        37
          38
          # Distribution
        39
          rho = conjgrad(MaskConv, mask[mask], mask, mask[mask])
        40
        41
          mask rho = np.zeros(mask.shape)
        42
        43
          mask rho[mask] = rho
        44
        45
          46 # PLots
        47
          48
          plt.imshow(mask rho, cmap="Blues")
          plt.title("Charge")
        49
        50 plt.colorbar()
        51 plt.show()
        52 plt.close()
        53
        54 | plt.plot(mask_rho[:, s2 + dim2 - 1])
          plt.title("Charge Density Along One Side")
```

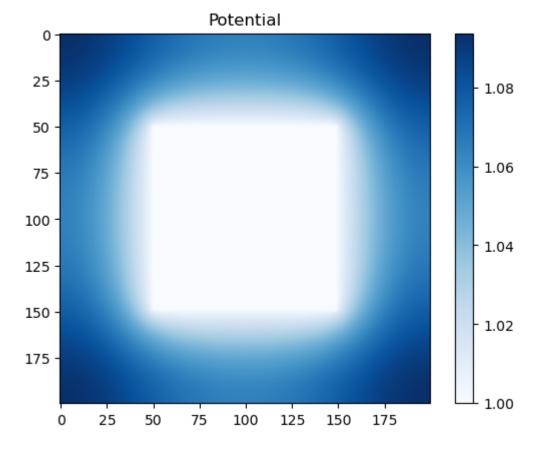


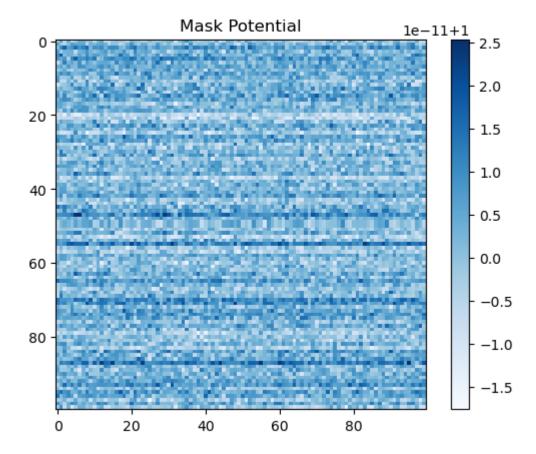
Out[138]: Text(0.5, 1.0, 'Charge Density Along One Side')



This is expected as the charge density will be maximum at the corners due to their repulsive force.

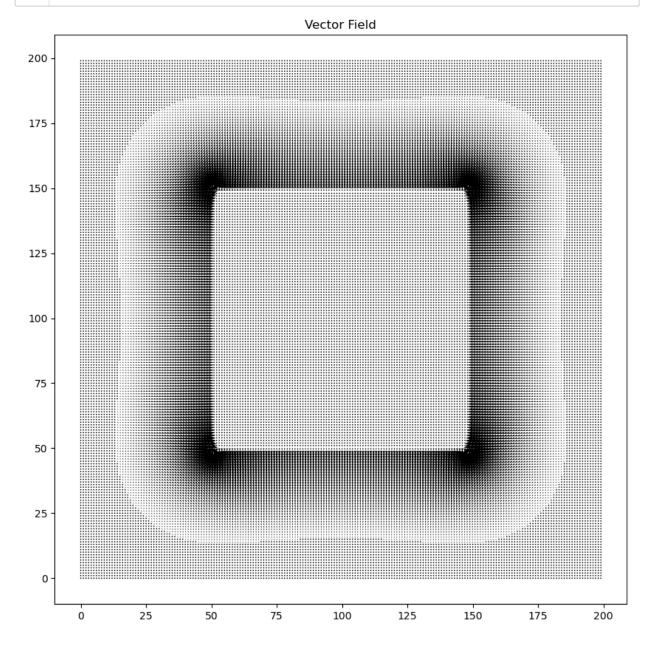
```
In [139]:
         # Part c
        4 # Plot Potential
         6 Pot = np.fft.fftshift(np.fft.irfft2(np.fft.rfft2(V) * np.fft.rfft2(mask_rho)
        7
         plt.imshow(Pot, cmap="Blues")
       8 plt.title("Potential")
       9
         plt.colorbar()
       10 plt.show()
       11 plt.close()
       12
       14
         # Plot - Mask Potential
       16 plt.imshow(Pot[50:150, 50:150], cmap="Blues")
         plt.title("Mask Potential")
       17
       18
         plt.colorbar()
       19
       21 | # Calculations - Maximum potential difference on the mask
       23 vmin = np.min(Pot[50:150, 50:150])
       24 vmax = np.max(Pot[50:150, 50:150])
       25 | print("Potential mean = ", Pot[50:150, 50:150].mean())
       26 print("Maximum potential difference on the mask = ", vmax - vmin)
```



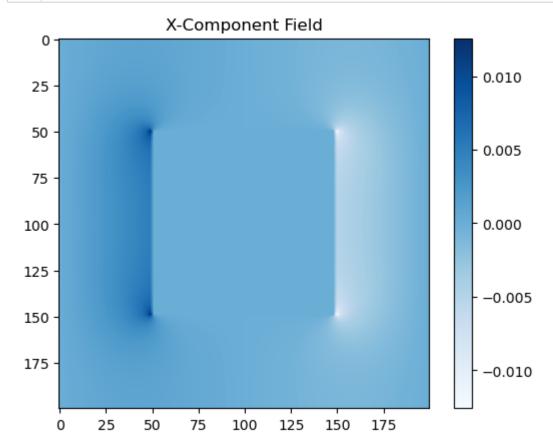


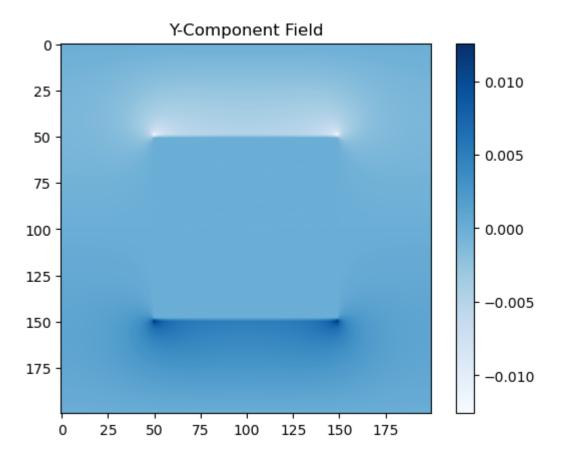
As you can see, the maximum potential difference on the mask is in order of 10^{-11} , so the potential in the interior is really close to constant (=1).

```
In [174]:
             # Plot - Field Strength
           2
           3
             #-----
             Ex = (np.roll(Pot,1,axis=1) - np.roll(Pot,-1,axis=1))
             Ey = (np.roll(Pot,-1,axis=0) - np.roll(Pot,1,axis=0))
           5
           6
             # Vector Field
           7
             Vectors = np.zeros(Pot.shape)
             X = np.arange(size)
           9
            Y = np.arange(size)
          10
          11
             plt.figure(figsize=(10,10))
             plt.quiver(X,Y,Ex,Ey, linewidth=0.5)
          12
          13 | plt.title("Vector Field")
             plt.show()
          14
             plt.close()
          15
          16
```

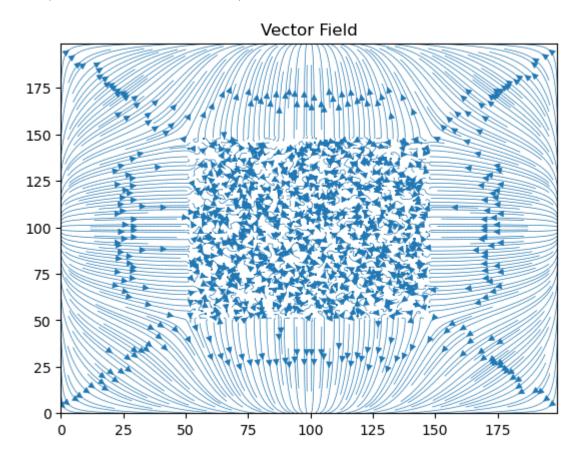


```
In [175]:
           1 # x - component
           plt.imshow(Ex, cmap="Blues")
           3 plt.colorbar()
           4 plt.title("X-Component Field")
            5 plt.show()
            6
              plt.close()
            7
           8 # y - component
           9 plt.imshow(Ey, cmap="Blues")
           10 plt.colorbar()
           11 plt.title("Y-Component Field")
           12 plt.show()
           13 plt.close()
           14
           15 # Vector Field - Direction
           16 Vectors = np.zeros(Pot.shape)
           17 X = np.arange(size)
           18 Y = np.arange(size)
          19 plt.streamplot(X,Y,Ex,Ey, density=4, linewidth=0.5)
           20 plt.title("Vector Field")
```





Out[175]: Text(0.5, 1.0, 'Vector Field')



This is what we expect. Strong field near the corners.

In []: 1