(Custom CSS files are not reliable for controlling Jupyter font style. To establish the same appearance as the original notebook, depend on the browser to control the font, by setting the desired font faces in the browser settings. For example, Chrome 135 or Firefox 134 can do this. In this notebook series, Bookerly font is for markdown and Monaco is for code.)

## Chapter 31-5: Elliptic PDEs Using the Monte Carlo Method.

Monte Carlo Methods are a class of computational solutions that can be applied to various problems which rely on repeated random sampling to provide generally approximate solutions. Monte Carlo is a stochastic approach, in which a series of simulations (trials), representing the analyzed problem, with randomly selected input values, are performed. Among these trials, a specified number of properly defined successes is achieved. The ratio between the number of success trials to the number of all trials, scaled by dimensional quantity (e.g., area or function value) allows for the estimation of the unknown solution, providing the number of trials is large enough.

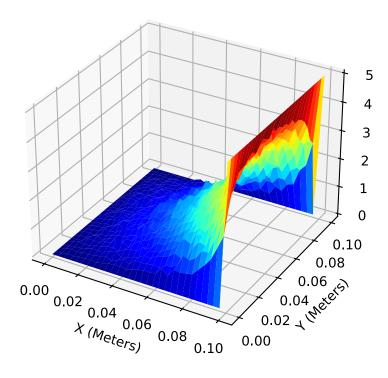
A capacitor is constructed out of two infinite metal plates spaced 10 cm apart from one another. One plate is placed at 5V potential with respect to the other and the space between the plates is free of charges. The number of random walks is taken to be 400 and the number of lattice points is taken to be 30.

(The description of the problem and the code solution was taken from the repository of sankur.)

```
In [1]:
        All calculations in SI units
        import random
        from mpl_toolkits.mplot3d import axes3d
        import matplotlib.pyplot as plt
        from matplotlib import cm
        import numpy as np
        %config InlineBackend.figure_formats = ['svg']
        h = 10e-2 # Distance between plates = 10cm
        lattice_points = 30 # Number of Points in lattice
        d = h / lattice_points # Lattice size
        boundary_voltage_high = 5.0 # 5 Volts at Positive Plate
        boundary_voltage_low = 0.0 # 0 Volts at Negative Plate
        epsilon_naught = 8.854e-12 # Permittivity of Vaccum
        charge_density = 6e-16 # Coulomb per meter cube
        N = 400 # Number of Random Walks
        def f(x):
             # The Function \nabla^2(phi) = f
             # For Laplace f = 0
             return 0
        def g(x):
             # Two Dimensional Boundary Conditions: two parallel metal plates at x=0, x=h
             # the plate at x=h is at high potential and x=0 is low potential
             # Assume that there are metal plates along y=0 and y=h (uncharged)
             # this is because I dont know how to simulate open boundry conditions
             if x[0] \leftarrow 0:
                 return boundary_voltage_low
             if x[0] >= h:
                 return boundary_voltage_high
            if x[1] \le 0 or x[1] >= h:
                 return boundary_voltage_low
        \mathbf{u} \cdot \mathbf{u} \cdot \mathbf{u}
        def f_2(x):
            # Alternative charge distribution: A charged Sphere in the centre of metal box
            if (h / 2 - x[0]) ** 2 + (h / 2 - x[1]) ** 2 <= (h / 5) ** 2:
                 return -charge_density * 5 / epsilon_naught
            else:
                 return 0
        def g_2(x):
            # Two Dimentional Alternative Boundary Conditions: uncharged metal box
            return 0
         11 11 11
        def f_3(x):
             # Alternative charge distribution: TWO charged Sphere in the centre of metal box
             if (h / 3 - x[0]) ** 2 + (h / 2 - x[1]) ** 2 \leftarrow (h / 5) ** 2:
                 return -charge_density * 5 / epsilon_naught
             if (2 * h / 3 - x[0]) ** 2 + (h / 2 - x[1]) ** 2 <= (h / 5) ** 2:
                 return charge_density * 5 / epsilon_naught
             else:
                 return 0
        @np.vectorize
        def poisson_approximation_fixed_step(*A):
             # Returns the Value of Potential Feild at a given point A with N random walks
             result = 0
             F = 0
             for i in range(N):
                 x = list(A)
                 while True:
                     if x[0] \leftarrow 0 or x[0] \rightarrow h or x[1] \leftarrow 0 or x[1] \rightarrow h:
                     random_number = random.randint(0, 3)
                     if random_number == 0:
                         x[0] += d
                     elif random_number == 1:
                         x[0] -= d
                     elif random_number == 2:
                         x[1] += d
                     elif random_number == 3:
                         x[1] -= d
                     F += \bar{f}(x) * h ** 2
                 result += g(x) / N
             result = result - F
             return result
        def plot(x, y, z):
             # Function for plotting the potential
             fig = plt.figure()
```

```
ax = fig.add_subplot(111, projection="3d")
    ax.plot_surface(x, y, np.array(z), cmap=cm.jet, linewidth=0.1)
    plt.xlabel("X (Meters)")
    plt.ylabel("Y (Meters)")
    ax.set_zlabel("Potential (Volts)")
    plt.show()
if __name__ == "__main__":
    # Experiment E: 2D Capacitor
    print(
        f"Calculating Monte Carlo with {lattice_points}x{lattice_points} lattice points and {N}
    lattice_x, lattice_y = np.mgrid[
        0 : h : lattice_points * 1j, 0 : h : lattice_points * 1j
    z = poisson_approximation_fixed_step(lattice_x.ravel(), lattice_y.ravel()).reshape(
        lattice_x.shape
    plot(lattice_x, lattice_y, z)
    # Experiment F: Metal box with positively charged metal ball inside
   f = f2
    g = g2
   print(
        f"Calculating Monte Carlo with {lattice_points}x{lattice_points} lattice points and {N}
    lattice_x, lattice_y = np.mgrid[
        0 : h : lattice_points * 1j, 0 : h : lattice_points * 1j
    z = poisson_approximation_fixed_step(lattice_x.ravel(), lattice_y.ravel()).reshape(
        lattice_x.shape
    plot(lattice_x, lattice_y, z)
    # Experiment G: Metal Box with two spheres (positive and negative)
    f = f_3
    g = g_2
    print(
        f"Calculating Monte Carlo with {lattice_points}x{lattice_points} lattice points and {N}
    lattice_x, lattice_y = np.mgrid[
        0 : h : lattice_points * 1j, 0 : h : lattice_points * 1j
    z = poisson_approximation_fixed_step(lattice_x.ravel(), lattice_y.ravel()).reshape(
        lattice_x.shape
    plot(lattice_x, lattice_y, z)
```

Calculating Monte Carlo with 30x30 lattice points and 400 random walks



	~\AppData\Local\Temp/ipykernel_11260/8079670 116 # g = g2 117 print(	th {lattice_points}x{lattice_points} lattice point
	<pre>120     lattice_x, lattice_y = np.mgrid  NameError: name 'laplace' is not defined</pre>	-
In [ ]:		
In []:	: 	