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1 Modelling Solar Cells

- In [1]: %matplotlib notebook
 # For matplotlib animations
 - executed in 2.00s, finished 16:29:30 2021-09-17
- In [2]: import numpy as np
 - import matplotlib.pyplot as plt
 - import matplotlib.animation as animation
 - executed in 193ms, finished 16:29:31 2021-09-17
- In [3]: import Grid as GridClass
 - GridS = GridClass.GridS
 - GridD = GridClass.GridD
 - executed in 7.16s, finished 16:29:39 2021-09-17
- In [4]: import ipyvolume as ipv # Does not come with default anaconda #Used to show 3D graphs
- executed in 6ms, finished 16:29:40 2021-09-17

1.1 Equivalent Circuit

1.2 Classical Analytical Expressions

1.3 Numerical Drift-Diffusion Models

1.3.1 Drift Diffusion Equations

Take as 1 Dimension

$$\frac{\partial n}{\partial t} = G_n - R_n + \frac{1}{q} \frac{\partial J_n}{\partial x}$$

$$\frac{\partial p}{\partial t} = G_p - R_p - \frac{1}{q} \frac{\partial J_p}{\partial x}$$

$$J_n = q u_n F n + q D_n \frac{\partial n}{\partial x}$$

$$J_p = q u_p F p - q D_p \frac{\partial p}{\partial x}$$

$$\frac{\partial F}{\partial x} = \frac{q}{\epsilon_0 \epsilon_r} (p - n + c)$$

With $u_n=qD_n\beta$, $u_p=qD_p\beta$ with $\beta=\frac{1}{k_bT}$, with q being the fundamental charge.

If wanted we could include the last equation (4.22), but it is negligible. Only take into account when building Dyson Spheres around stars larger than ours.

1.3.1.1 Steady State Solution

We can set $\frac{\partial n}{\partial t} = \frac{\partial p}{\partial t} = 0$ to get a steady state solution, this gives us a coupled ODE which should be easy to solve.

$$\frac{dJ_n}{dx} = q(R_n - G_n)$$

$$\frac{dJ_p}{dx} = q(G_p - R_p)$$

$$\frac{dn}{dx} = \frac{J_n - qu_n Fn}{qD_n}$$

$$\frac{dp}{dx} = \frac{qu_p Fp - J_p}{qD_p}$$

$$\frac{dF}{dx} = \frac{q}{\epsilon_0 \epsilon_r} (p - n + c)$$

Tells us the steady state related to the boundary condition.

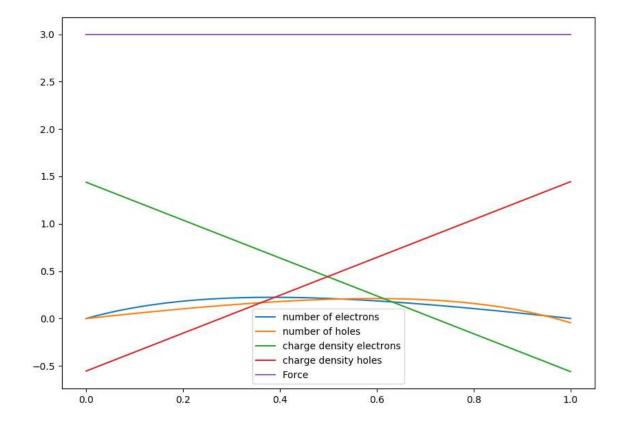
We use the fact

$$y_{n+1} = y_n + \frac{dy_n}{dx} dx$$

```
In [5]: # Constants
        T = 1
        k_b = 1
        B = 1/(T*k_b)
        q = 1
        G_p, R_p = 2, 0
        G_n, R_n = 2, 0
        D_n, D_p = 1, 1
        u_n = q*D_n*B
        u_p = q*D_p*B
        A = 1
        C = 1
        dx = 0.001
        x_f = 1
        N = int((x_f)//dx) + 1
        # Could put in 2D grid but easier to read like this
        J_n_ = np.zeros(N)
        J_p_ = np.zeros(N)
        n_ = np.zeros(N)
        p_ = np.zeros(N)
        F_{\underline{}} = np.zeros(N)
        # Initial conditions at x_0
        p_{0} = 0
        n_{0} = 0
        F_{0} = 3
        J_n[0] = 1.4380931547599094*D_n*q
        J_p[0] = -0.55404*D_p*q
        def Calc_Next(J_n, J_p, n, p, F):
            # n prefix stands for new value
            nJ_n = J_n + (q*(R_n-G_n))*dx
            nJ_p = J_p + (q*(G_p-R_p))*dx
            nn = n + ((J_n - q*u_n*F*n)/(q*D_n))*dx
            n_p = p + ((-J_p + q*u_p*F*p)/(q*D_p))*dx # We need np extension
            nF = F \# + A*(p-n+C)*dx
            return nJ_n, nJ_p, nn, n_p, nF
        for i in range(N-1):
            J_n[i+1], J_p[i+1], n[i+1], p[i+1], F[i +
                                                         1] = Calc_Next(J_n_[i], J_p_[i], n_[i], p_[i], F_[i])
        executed in 49ms, finished 16:29:44 2021-09-17
```

```
In [6]: x = np.linspace(0, x_f, N)
    plt.figure(figsize=(10, 7))
    plt.plot(x, n_, label="number of electrons")
    plt.plot(x, p_, label="number of holes")
    plt.plot(x, J_n_, label="charge density electrons")
    plt.plot(x, J_p_, label="charge density holes")
    plt.plot(x, F_, label="Force")
    plt.legend()
    plt.show()

    executed in 74ms, finished 16:29:50 2021-09-17
```



1.3.1.3 High Temperature Solution

1.3.1.4 Constant Force n(x)

When the force is constant, the p and n equations do not depend on each other. We will look at n.

```
\frac{\partial n}{\partial t} = G_n - R_n + u_n F \frac{\partial n}{\partial x} + D_n \frac{\partial^2 n}{\partial x^2}
```

 $G_n - R_n$ is a concentration of electrons source,

$$n_N = n_O + \delta t (G_n - R_n + \frac{u_n F}{2h} (n_E - n_W) + \frac{D_n}{h^2} (n_E - 2n_O + n_W))$$

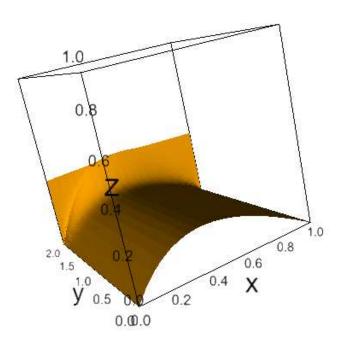
```
In [7]: # Constants
        T = 1
        k_b = 1
        B = 1/(T*k_b)
        q = 1
        G_n, R_n = 2, 0
        D_n = 1
        u_n = q*D_n*B
        F = 3
        dx = 0.05
        x_f = 1
        N_x = int((x_f)/dx) + 1
        dt = 0.001
        T = 2
        N_t = int((T)/dt) + 1
        Grid = np.zeros([N_t, N_x])
        # BC's
        # zero do nothing
        # Grid[:,0] = np.zeros(len(Grid))+10. #Run if we want a cluster of electrons at the positive node
        # IC
        \#n_0 = np.array([i*dx for i in range(0,N_x)])
        Grid[len(Grid)-1] = np.zeros(len(Grid[0]))+0.4
        Grid = GridS(Grid)
        n_{-} = Grid.u_{-}
        def new_n_(i, j):
            j -= 1
            temp = G_n - R_n + ((u_n*F)/(2*0.05))*(n_(i+1, j) - n_(i-1, j)) + 
                ((D_n)/(0.05**2))*(n_(i+1, j)-2*n_(i, j)+n_(i-1, j))
            temp *= dt
            return n_(i, j) + temp
        for j in range(1, N_t):
            for i in range(1, N_x-1):
                Grid.Set_u_(i, j, new_n_(i, j))
        executed in 476ms, finished 16:30:05 2021-09-17
```

```
In [8]: np.max(Grid.Grid[0]) executed in 14ms, finished 16:30:07 2021-09-17
```

Out[8]: 0.2239453283460387

```
In [ ]:
In [9]:
    a = np.arange(0.0, x_f+0.00001, dx)
    b = np.arange(0.0, T+0.00001, dt)
    U, V = np.meshgrid(a, b)
    X = U
    Y = V # The y-axis is the wrong way around
    Z = Grid.Grid
    ipv.figure()
    ipv.plot_surface(X, Y, Z, color="orange")
    #ipv.plot_wireframe(X, Y, Z, color="red")
    ipv.show()
    executed in 327ms, finished 16:30:12 2021-09-17
```





Electrons travel in opposite direction of force as they are negatively charged. Force is the direction away from the positive electrode. Since F > 0 positive node at x = 0. And the electrons want to be next to the positive node. But why is there zero at the positive node? (No it was because the is what we set our BC's) to be Maybe as soon as it touches the positive electrode the electron get taken into the circuit. (Only have a curve as we have a positive generation rate)

Analytic Solution, for n(x)

To Prove Steady State is correct, we will solve

$$D_n \frac{\partial^2 n}{\partial x^2} + u_n F \frac{\partial n}{\partial x} = R_n - G_n$$

Very simple ODE with BC's 0. Solved solution is

$$n(x) = \frac{R_n - G_n}{u_n F} \left[x - \frac{exp(-\frac{u_n F}{D_n} x) - 1}{exp(-\frac{u_n F}{D_n}) - 1} \right]$$

```
In [10]:
    def n(x):
        temp = np.exp(-x*(u_n*F)/(D_n)) - 1
        temp /= np.exp(-(u_n*F)/(D_n)) - 1
        temp = x - temp
        temp /= u_n*F
        return (R_n - G_n)*temp

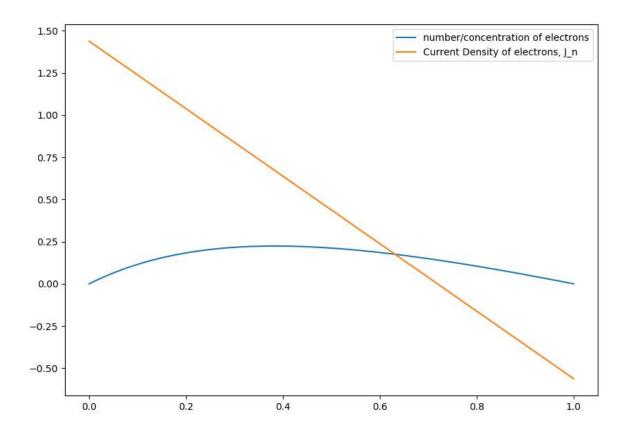
def grad_n(x):
    # Could also calculate analytically
    d = 0.00000001
    return (n(x+d)-n(x))/d

#Now calculate Current desity J_n
    def J_n(x):
        return q*u_n*F*n(x)+q*D_n*grad_n(x)
    executed in 25ms, finished 16:30:26 2021-09-17
```

```
In [11]: #Now calculate Current desity J_n
executed in 6ms, finished 16:30:28 2021-09-17
```

```
In [12]: x = np.linspace(0, 1, 100)
    plt.figure(figsize=(10, 7))
    plt.plot(x, n(x), label="number/concentration of electrons")
    plt.plot(x, J_n(x), label="Current Density of electrons, J_n")
    plt.legend()
    plt.show()

    executed in 65ms, finished 16:30:30 2021-09-17
```



The steady state is the same !!:

1.3.1.5 Constant Force p(x)

When the force is constant, the p and n equations do not depend on each other. We will look at p.

$$\frac{\partial p}{\partial t} = G_p - R_p - u_p F \frac{\partial p}{\partial x} + D_p \frac{\partial^2 p}{\partial x^2}$$

 $G_p - R_p$ is a concentration of electrons source,

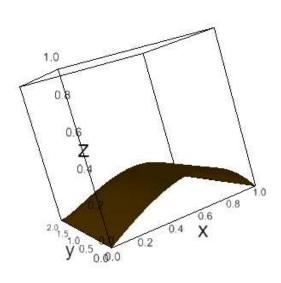
```
p_N = p_O + \delta t (G_p - R_p - \frac{u_p F}{2h} (p_E - p_W) + \frac{D_p}{h^2} (p_E - 2p_O + p_W))
```

```
In [13]: G_p, R_p = 2, 0
          D_p = 1
          u_p = q*D_p*B
          Grid = np.zeros([N_t, N_x])
          Grid = GridS(Grid)
          p_{-} = Grid.u_{-}
          def new_p_(i, j):
              j -= 1
              temp = G_p - R_p - ((u_p*F)/(2*0.05))*(p_(i+1, j) - p_(i-1, j)) + 
                  ((D_p)/(0.05**2))*(p_(i+1, j)-2*p_(i, j)+p_(i-1, j))
              temp *= dt
              return p_(i, j) + temp
          for j in range(1, N_t):
              for i in range(1, N_x-1):
                  Grid.Set_u_(i, j, new_p_(i, j))
          executed in 453ms, finished 16:30:56 2021-09-17
```

```
In [14]:
    a = np.arange(0.0, x_f+0.00001, dx)
    b = np.arange(0.0, T+0.00001, dt)
    U, V = np.meshgrid(a, b)
    X = U
    Y = V # The y-axis is the wrong way around
    Z = Grid.Grid

    ipv.figure()
    ipv.plot_surface(X, Y, Z, color="orange")
    #ipv.plot_wireframe(X, Y, Z, color="red")
    ipv.show()
    executed in 85ms, finished 16:31:00 2021-09-17
```





1.3.1.6 Coupled Solution With Non Constant Force

We rearrange the drift diffusion equations, and remember that p, n, F depend on x, t

$$\frac{\partial n}{\partial t} = G_n - R_n + D_n \frac{\partial^2 n}{\partial x^2} + u_n F \frac{\partial n}{\partial x} + u_n n A(p - n + c)$$

$$\frac{\partial p}{\partial t} = G_p - R_p + D_p \frac{\partial^2 p}{\partial x^2} - u_p F \frac{\partial p}{\partial x} - u_p p A(p - n + c)$$

$$F(s,t) = F(s_0,t) + Ac(s - s_0) + A \int_{s_0}^{s} (p - n) dx$$

Where $A=\frac{q}{\epsilon_0\epsilon_r}$, the last part comes from the fact

$$\frac{\partial F(x,t)}{\partial x} = A(p-n+c)$$

$$\int_{F(s_0,t)}^{F(s,t)} 1\partial F = \int_{s_0}^{s} A(p-n+c)\partial x$$

$$F(s,t) = F(s_0,t) + Ac(s-s_0) + A\int_{s_0}^{s} (p-n)dx$$

Discretise the equations.

$$n_{i,j+1} = n_{i,j} + \delta t \left(G_n - R_n + \frac{D_n}{h^2} (n_{i+1,j} - 2n_{i,j} + n_{i-1,j}) + \frac{u_n F_{i,j}}{2h} (n_{i+1,j} - n_{i-1,j}) + u_n n_{i,j} A(p_{i,j} - n_{i,j} + c) \right)$$

$$p_{i,j+1} = p_{i,j} + \delta t \left(G_p - R_p + \frac{D_p}{h^2} (p_{i+1,j} - 2p_{i,j} + p_{i-1,j}) - \frac{u_p F_{i,j}}{2h} (p_{i+1,j} - p_{i-1,j}) - u_p p_{i,j} A(p_{i,j} - n_{i,j} + c) \right)$$

$$F_{i,j} = F_{i_0,j} + Ach(i - i_0) + A \sum_{k=i_0}^{i} (p_{k,j} - n_{k,j})$$

For here we will take $i_0 = 0$

```
In [15]: i_0 = 0 #Has to be zero due to for loop, can be changed
          #Constants
          T = 1
          k_b = 1
          B = 1/(T*k_b)
          q = 1
          G_p, R_p = 2, 0
          G_n, R_n = 2, 0
          D_n, D_p = 1, 1
          u_n = q*D_n*B
          u_p = q*D_p*B
          A = 1
          C = 1
          h = 0.05 \# dx
          x_f = 1
          dt = 0.0005
          t_f = 1
          Num_x = int((x_f)//h) + 1
          Num_t = int((t_f)//dt) + 1
          # Create Grids
          Grid_n = np.zeros([Num_t,Num_x])
          Grid_p = np.zeros([Num_t,Num_x])
          Grid_F = np.zeros([Num_t,Num_x])
          #Initial (IC's)
          Grid_n[-1] = np.ones([Num_x])
          Grid_p[-1] = np.ones([Num_x])
          Grid_F[-1] = np.ones([Num_x])*3
          # BC's
          Grid_F[:,0] = np.ones([Num_t])*3
          #Grid_F[:,-1]
          Grid_n = GridS(Grid_n)
          Grid_p = GridS(Grid_p)
          Grid_F = GridS(Grid_F)
          n_{-} = Grid_n.u_{-}
          p_{-} = Grid_p.u_{-}
          F_{-} = Grid_F.u_{-}
          def new_n_(i,j):
              j -= 1
              partA = G_n - R_n
              partB = n_{(i+1,j)-2*n_{(i,j)+n_{(i-1,j)}}
partB *= (D_n)/(h^{**2})
              partC = n_{i+1,j}-n_{i-1,j}
              partC *= (u_n*F_(i,j))/(2*h)
              partD = p_{(i,j)}-n_{(i,j)}+C
              partD *= u_n*n_(i,j)*A
              return n_(i,j) + dt*(partA + partB + partC + partD)
          def new_p_(i,j):
              j -= 1
              partA = G_n - R_n
              partB = p_{(i+1,j)-2}p_{(i,j)+p_{(i-1,j)}}
              partB *= (D_p)/(h^{**2})
              partC = p_{(i+1,j)}-p_{(i-1,j)}
              partC *= (u_p*F_(i,j))/(2*h)
              partD = p_{(i,j)}-n_{(i,j)}+C
              partD *= u_p*p_(i,j)*A
              return p_(i,j) + dt*(partA + partB - partC - partD)
          def new_F_(i,j):
              temp = 0
```

```
for k in range(i_0,i+1):
        temp += p_(k,j)+n_(k,j)
    temp *= A
    temp += F_{(i_0, j)} + A*C*h*(i-i_0)
    return temp
def solve_n(j):
    #Not left or right boundary
    for i in range(1,Num_x-1):
        Grid_n.Set_u_(i,j,new_n_(i,j))
def solve_p(j):
    #Not left or right boundary
    for i in range(1,Num_x-1):
        Grid_p.Set_u_(i,j, new_p_(i,j))
def solve_F(j):
    #Not the i_0 element
    for i in range(1,Num_x):
        Grid_F.Set_u_(i,j,new_F_(i,j))
for j in range(1, Num_t):
    solve_n(j)
    solve_p(j)
    solve_F(j)
```

executed in 2.89s, finished 16:31:19 2021-09-17

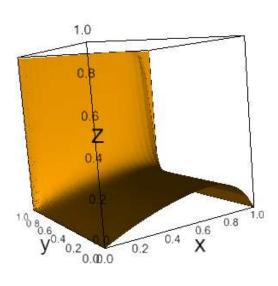
executed in 101ms, finished 16:31:26 2021-09-17

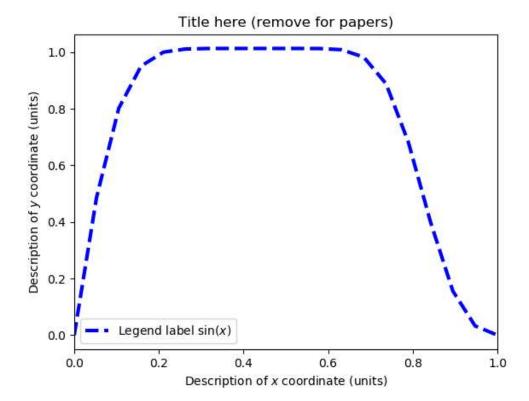
```
In [16]: a = np.linspace(0, x_f, num=Num_x, endpoint=True)
b = np.linspace(0, t_f, num=Num_t, endpoint=True)

U, V = np.meshgrid(a, b)
X = U
Y = V # The y-axis is the wrong way around
Z = Grid_p.Grid

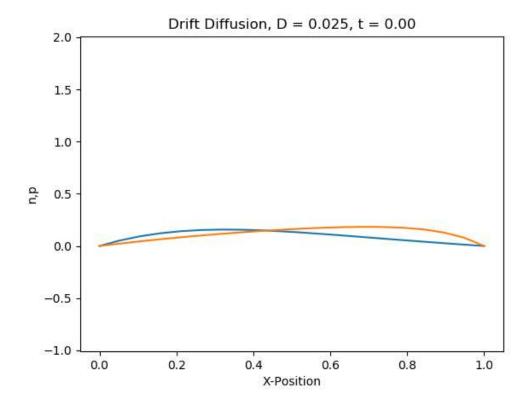
ipv.figure()
ipv.plot_surface(X, Y, Z, color="orange")
#ipv.plot_wireframe(X, Y, Z, color="red")
ipv.show()
```



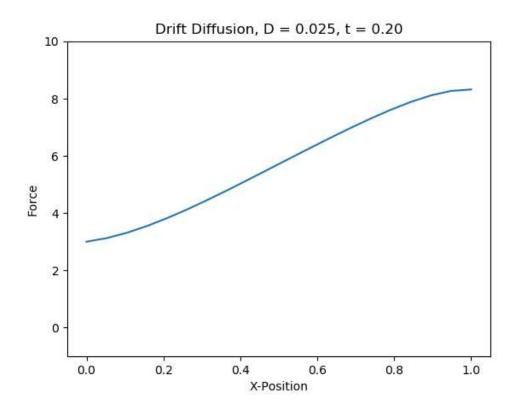




```
In [18]: def dataLD(i, y_n,y_p, line_n, line_p):
             t = i*dt
             y_n = Grid_n.Grid[-i]
             y_p = Grid_p.Grid[-i]
             ax.clear()
             ax.set_ylim(-1.01, 2.01)
             ax.set_title('Drift Diffusion, D = 0.025, t = ' + "{:.2f}".format(t))
             line_n = ax.plot(x, y_n)
             line_p = ax.plot(x, y_p)
             ax.set_ylabel('n,p')
             ax.set_xlabel('X-Position')
             #fig.colorbar(line, shrink=0.5, aspect=5)
             return line,
         fig = plt.figure()
         ax = fig.add_subplot(111)
         ax.set_title('Drift Diffusion, D = 0.025, t = 0.00')
         x = np.linspace(0, 1, num=Num_x, endpoint=True)
         y_n = Grid_n.Grid[-1]
         y_p = Grid_p.Grid[-1]
         line_n = ax.plot(x, y_n)
         line_p = ax.plot(x, y_p)
         ax.set_ylim(-1.01, 2.01)
         ax.set_ylabel('n,p')
         ax.set_xlabel('X-Position')
         #fig.colorbar(line, shrink=0.5, aspect=10)
         ani = animation.FuncAnimation(fig, dataLD, fargs=(
             y_n, y_p, line_n, line_p), frames=np.arange(0, 500), interval=20, blit=True)
         plt.show()
         executed in 57ms, finished 16:32:28 2021-09-17
```



```
In [19]: def dataLD(i, line_F):
             t = i*dt
             y_F = Grid_F.Grid[-i]
             ax.clear()
             ax.set_ylim(-1.01, 10.01)
             ax.set_title('Drift Diffusion, D = 0.025, t = ' + "{:.2f}".format(t))
             line_F = ax.plot(x, y_F)
             ax.set_ylabel('Force')
             ax.set_xlabel('X-Position')
             #fig.colorbar(line, shrink=0.5, aspect=5)
             return line_F,
         fig = plt.figure()
         ax = fig.add_subplot(111)
         ax.set_title('Drift Diffusion, D = 0.025, t = 0.00')
         x = np.linspace(0, 1, num=Num_x, endpoint=True)
         y_F = Grid_F.Grid[-1]
         line_F = ax.plot(x, y_F)
         ax.set_ylim(-1.01, 10.01)
         ax.set_ylabel('Force')
         ax.set_xlabel('X-Position')
         #fig.colorbar(line, shrink=0.5, aspect=10)
         ani = animation.FuncAnimation(fig, dataLD, fargs=(line_F), frames=np.arange(0, 500), interval=20, blit=Tru
         plt.show()
         executed in 65ms, finished 16:32:42 2021-09-17
```



1.3.1.7 Drift Diffusion with Energy as 2nd Dimension

We have the equations, which include energy

$$\frac{\partial \tilde{n}}{\partial t} = \frac{v_0}{\sqrt{2\pi\tilde{\sigma}}} \exp(-\frac{1}{2\tilde{\sigma}^2} (\epsilon - E_c - \lambda)^2) \left(\tilde{K} \left(-\frac{\beta g_1(x)}{2} \frac{\partial \phi}{\partial x} + \frac{\partial g_1(x)}{\partial x} \right) \frac{\partial \tilde{n}}{\partial x} + \frac{\tilde{K}g_1(x)}{2} \frac{\partial^2 \tilde{n}}{\partial x^2} + Cg_1(x) \bar{E}_n \frac{\partial \tilde{n}}{\partial \epsilon} \right)$$

$$\frac{\partial \tilde{p}}{\partial t} = \frac{v_0}{\sqrt{2\pi\tilde{\sigma}}} \exp(-\frac{1}{2\tilde{\sigma}^2} (\epsilon - E_c + \lambda)^2) \left(\tilde{K} \left(\frac{\beta g_1(x)}{2} \frac{\partial \phi}{\partial x} + \frac{\partial g_1(x)}{\partial x} \right) \frac{\partial \tilde{p}}{\partial x} + \frac{\tilde{K}g_1(x)}{2} \frac{\partial^2 \tilde{p}}{\partial x^2} + Cg_1(x) \bar{E}_p \frac{\partial \tilde{p}}{\partial \epsilon} \right)$$

$$\frac{\partial \phi}{\partial x}|_{x=s} = \frac{\partial \phi}{\partial x}|_{x=0} + A \int_0^s (\tilde{p} - \tilde{n}) dx$$

Key

$$\tilde{n}(x, \epsilon, t) = \tilde{n}_{O}$$

$$\tilde{n}(x + \delta x, \epsilon, t) = \tilde{n}_{E}$$

$$\tilde{n}(x - \delta x, \epsilon, t) = \tilde{n}_{W}$$

$$\tilde{n}(x, \epsilon + \delta \epsilon, t) = \tilde{n}_{N}$$

$$\tilde{n}(x, \epsilon - \delta \epsilon, t) = \tilde{n}_{S}$$

$$\tilde{n}(x, \epsilon, t + \delta t) = \tilde{n}_{U}$$

$$\tilde{n}(x, \epsilon, t - \delta t) = \tilde{n}_{D}$$

By using expansions(Taylor) Calculate gradient of g_1

$$\frac{\partial g_1}{\partial x} = \frac{1}{h} (g_1(x+h) - g_1(x))$$

Calculate partial for \tilde{n}

$$\begin{split} \frac{\partial \tilde{n}}{\partial t} &= \frac{1}{\delta t} (\tilde{n}_U - \tilde{n}_O) \rightarrow \tilde{n}_U = \tilde{n}_O + \delta t \frac{\partial \tilde{n}}{\partial t} \\ \frac{\partial \tilde{n}}{\partial \epsilon} &= \frac{1}{2\delta \epsilon} (\tilde{n}_N - \tilde{n}_S) \\ \frac{\partial \tilde{n}}{\partial x} &= \frac{1}{2\delta x} (\tilde{n}_E - \tilde{n}_W) \\ \frac{\partial^2 \tilde{n}}{\partial x^2} &= \frac{1}{(\delta x)^2} (\tilde{n}_E - 2\tilde{n}_O + \tilde{n}_W) \end{split}$$

Calculate partial for \tilde{n}

$$\begin{split} \frac{\partial \tilde{p}}{\partial t} &= \frac{1}{\delta t} (\tilde{p}_{U} - \tilde{p}_{O}) \rightarrow \tilde{p}_{U} = \tilde{p}_{O} + \delta t \frac{\partial \tilde{p}}{\partial t} \\ \frac{\partial \tilde{p}}{\partial \epsilon} &= \frac{1}{2\delta \epsilon} (\tilde{p}_{N} - \tilde{p}_{S}) \\ \frac{\partial \tilde{p}}{\partial x} &= \frac{1}{2\delta x} (\tilde{p}_{E} - \tilde{p}_{W}) \\ \frac{\partial^{2} \tilde{p}}{\partial x^{2}} &= \frac{1}{(\delta x)^{2}} (\tilde{p}_{E} - 2\tilde{p}_{O} + \tilde{p}_{W}) \end{split}$$

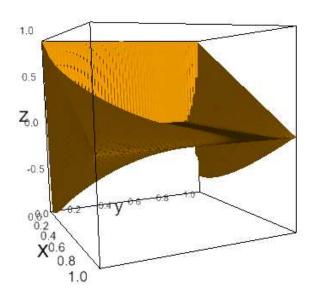
We sub these into the equation above, we find the value of many constants in the gdm document.

Compute On GPU (CUDA)

Compute done on university cluster and output to file which is copied across.

```
In [20]: | name = "nOutput"
         #name = "pOutput"
         xDim = 101
         yDim = 101
         def read_file(filename):
             f = open(filename,"r")
             return np.array(f.read().splitlines()).astype(float)
         x = np.linspace(0.0,1.0,xDim)
         y = np.linspace(0.0,1.0,yDim)
         U, V = np.meshgrid(x, y)
         Y = V # The y-axis is the wrong way around
         Z = read_file(name)
          ''' For single Z
         ipv.figure()
         ipv.plot_surface(X, Y, Z, color="orange")
         #ipv.plot_wireframe(X, Y, Z, color="red")
         ipv.show()
         ipv.figure()
         s = ipv.plot_surface(X, Y, Z.reshape(len(Z)//(xDim*yDim),1,-1), color="orange")
         #ipv.plot_wireframe(X, Y, Z, color="red")
         ipv.animation_control(s, add = True, interval=200)#, sequence_length=2)
         ipv.show()
         executed in 19.1s, finished 16:33:47 2021-09-17
```





1199.00

1.4 Numerical Gaussian Disorder Models

1.4.1 1D GDM (Energy and x_pos)

```
In [21]: def F_Normal(x, u=0, o=1):
    # returns P(N<x/u,o)
    temp = 0
    ds = 0.01
    for s in np.arange(-10,x,ds):
        temp += np.exp(-(s-u)**2/(2*o**2))*ds
    temp /= (2*np.pi*o)**0.5

#Can include last step, as may be incorrect size
    return temp

executed in 14ms, finished 16:34:17 2021-09-17</pre>
```

```
In [22]: E_u, E_o = 5,1
         x_u, x_o = 0,1
         x_L, x_U = 0,2
         x_uniform = True
         Num = 200#Number of sites
         #Node Positions
         Up Node = 1.9
         Low_Node = 0.1 #Outside of this range is where the nodes are (electrons enter/leave to form current)
         v_0 = 1
         Gamma = 1
         k_B = 1
         T = 1
         beta = 1/(k_B*T)
         lam = 1
         e = -1
         F = 6
         def v_1(i,j):
             return v_0*np.exp(-2*Gamma*abs(x[i]-x[j]))
         def v_2_mar(i,j): #Marcus jump rates
             #from i to j
             exp = -(beta)/(4*lam)
             exp *= (E[j]-E[i]+lam)**2
             temp = (beta/(4*np.pi*lam))**0.5
             temp *= np.exp(exp)
             return temp
         def v_2_mil(i,j): #Miller-Abrahms type
             #from i to j
             temp = 1
             if E[j]>E[i]:
                 #Force only effects jumps from low energy to high energy
                 #Force in direction of hop site
                 temp = np.exp(-(E[j]-E[i]-e*F*(x[j]-x[i]))/(k_B*T))
             return temp
         def v_(i,j): #Jump Rate
             if i == j:
                 return 0
             return v_1(i,j)*v_2_mil(i,j)
         def pick_index(a):
             input
             vector(np.array): a
             vector of probs sum up to one
             returns
             returns the index of the element that gets chosen randomly depending on probs in vector
             1.1.1
             CDF = np.array([sum(a[:n+1]) for n in range(len(a))])
             x = np.random.uniform(low = 0, high = 1, size = 1)
             for index in range(len(a)):
                 if x <= CDF[index]:</pre>
                     return index
             return "Error"
```

```
x = np.random.normal(loc=x_u, scale=x_o, size=Num)
         if x_uniform:
             x = np.random.uniform(low=x_L, high=x_U, size=Num)
          executed in 19ms, finished 16:34:26 2021-09-17
In [24]: #?we can simulate the force by changing the values of the energy?
         #Calculate Jump Rates
         jump_rates = np.array([[v_(i,j) for j in range(Num)] for i in range(Num)])
          #Calculate Jump Probs
         jump_probs = np.array([jump_rates[i]/sum(jump_rates[i]) for i in range(Num)])
         executed in 431ms, finished 16:34:29 2021-09-17
In [25]:
         #Start at E[0], x[0]
          #Calculate path
          steps = 25
         path = np.zeros(steps) #RandomLy gen path
         path_ML = np.zeros(steps) # Most likely path not random
         #IC's start on
         path[0] = x.argmax()
         path_ML[0] = x.argmax()
          #Path Calculated Randomly
         for i in range(1,steps):
              path[i] = pick_index(jump_probs[int(path[i-1])])
          #Path_ML takes most likly path
         for i in range(1,steps):
              path_ML[i] = jump_probs[int(path_ML[i-1])].argmax()
         executed in 205ms, finished 16:34:34 2021-09-17
In [26]: num samples = 200
         paths = [[np.random.randint(0, len(x))] for path index in range(num samples)]
          steps = np.zeros(num_samples)
         for path_index in range(num_samples):
              step = 0
```

paths[path index].append(pick index(jump probs[int(paths[path index][-1])]))

if x[int(paths[path index][-1])] < Low Node:</pre>

steps[path_index] = step

In [23]: E = np.random.normal(loc=E_u, scale=E_o, size=Num)

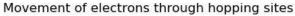
while True:

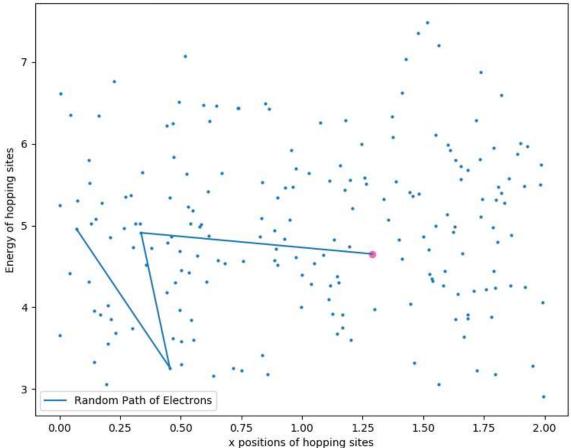
print(np.mean(steps))

break

executed in 8.27s, finished 16:34:46 2021-09-17

```
In [27]: #Plot Grid
         plt.figure(figsize=(9, 7))
         plt.scatter(x,E, s = 4)
         plt.xlabel(r"x positions of hopping sites")
         plt.ylabel(r"Energy of hopping sites")
         plt.title(r"Movement of electrons through hopping sites")
         #Plot random path of electrons
         E_path = np.array([E[int(path_val)] for path_val in paths[0]])
         x_path = np.array([x[int(path_val)] for path_val in paths[0]])
         plt.plot(x_path, E_path, label = "Random Path of Electrons")
         . . .
         #Plot most likly path of electrons
         E_path_ML = np.array([E[int(path_val)] for path_val in path_ML])
         x_path_ML = np.array([x[int(path_val)] for path_val in path_ML])
         plt.plot(x_path_ML, E_path_ML, label = "Most Likely Path Of Electrons")
         #Make start of path a different colour (Red = Start)
         plt.scatter(x_path[0], E_path[0], color = 'hotpink')
         plt.legend(loc="lower left")
         plt.show()
         executed in 75ms, finished 16:34:53 2021-09-17
```





- 1.5 Monte Carlo Simulations
- 1.6 Quantum Mechanical/Chemical Calculations