```
In [ ]: import numpy as np
        import matplotlib.pyplot as plt
        import matplotlib as mpl
        from scipy.special import gamma
        import tqdm
        from maxpy.makegif import make gif
        def cback(data, name):
            plt.clf()
            plt.xlim(-2, 2)
            plt.ylim(-2, 2)
            plt.scatter(data[0][:, 0], data[0][:, 1], c=data[1])
            plt.savefig(name)
        # two different kernels, a Gaussian and a spline
        # Gaussian kernel
        def W(x, y, z, h):
            0.00
            Gausssian kernel in 3D
                         list of positions in x, y, z
                x,y,z
                          smoothing length
                return smoothing function
            0.00
            r = np.sqrt(x**2 + y**2 + z**2)
            return (1.0 / (h * np.sqrt(np.pi))) ** 3 * np.exp(-(r**2) / h**2)
        # derivative of Gaussian kernel
        def grad W(x, y, z, h):
            Gradient of the Gausssian kernel W
            x,y,z list of positions in x, y, z
                      smoothing length
            wx,wy,wz gradient of W
            r = np.sqrt(x**2 + y**2 + z**2)
            n = -2.0 * np.exp(-(r**2) / h**2) / h**5 / (np.pi) ** (3.0 / 2.0)
            dwx = n * x
            dwy = n * y
            dwz = n * z
            return dwx, dwy, dwz
        def compute_pairwise_distances(ri, rj):
            compute pairwise separations between 2 sets of coordinates
            ri is an M x 3 matrix of positions
                 is an N \times 3 matrix of positions
            dx, dy, dz
                       are M x N matrices of separations
            0.00
            M = ri.shape[0]
```

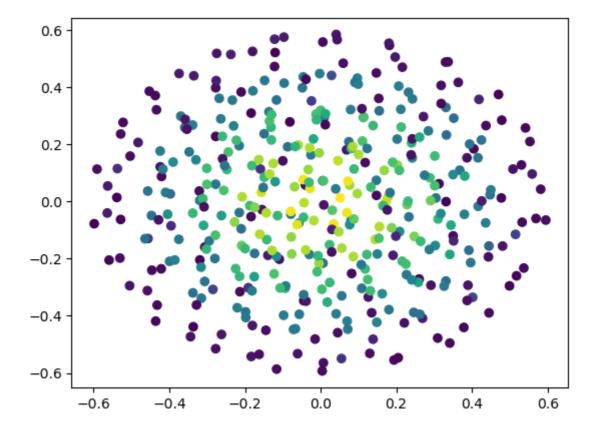
```
N = rj.shape[0]
    # positions ri = (x, y, z)
    rix = ri[:, 0].reshape((M, 1))
    riy = ri[:, 1].reshape((M, 1))
    riz = ri[:, 2].reshape((M, 1))
    # other set of points positions rj = (x,y,z)
    rjx = rj[:, 0].reshape((N, 1))
    rjy = rj[:, 1].reshape((N, 1))
    rjz = rj[:, 2].reshape((N, 1))
    # matrices that store all pairwise particle separations: r i - r j
    dx = rix - rjx.T
    dy = riy - rjy.T
    dz = riz - rjz.T
    return dx, dy, dz
def compute density(r, pos, m, h):
    Compute density at sampling loctions from SPH particle distribution
          is an M x 3 matrix of sampling locations
          is an N x 3 matrix of SPH particle positions
    pos
          is the particle mass
          is the smoothing length
    rho
          is M x 1 vector of densities
    M = r.shape[0]
    dx, dy, dz = compute pairwise distances(r, pos)
    rho = np.sum(m * W(dx, dy, dz, h), 1).reshape((M, 1))
    return rho
def compute_pressure(rho, k, n):
    equation of state
         vector of densities
    rho
          constant for the equation of state
          polytropic index
    return k * rho ** (1 + 1 / n)
def compute_accelerations(pos, vel, m, h, k, n, Fext, nu):
    calculate the acceleration on each SPH particle
         is an N \times 3 matrix of positions
    vel
         is an N x 3 matrix of velocities
          is the particle mass
          is the smoothing length
    h
          constant for the equation of state
          polytropic index
    n
    Fext external force constant
```

```
nu viscosity
      is N x 3 matrix of accelerations
а
N = pos.shape[0]
# Calculate densities at the position of the particles
rho = compute density(pos, pos, m, h)
# Get the pressures
P = compute_pressure(rho, k, n)
# Get pairwise distances and gradients
dx, dy, dz = compute pairwise distances(pos, pos)
dWx, dWy, dWz = grad W(dx, dy, dz, h)
# Calculate the pressure gradient
tmp = P / rho**2
pj, pi = np.meshgrid(tmp, tmp)
# Add Pressure contribution to accelerations
ax = -np.sum(m * (pi + pj) * dWx, axis=1)
ay = -np.sum(m * (pi + pj) * dWy, axis=1)
az = -np.sum(m * (pi + pj) * dWz, axis=1)
# pack together the acceleration components
a = np.vstack((ax, ay, az)).T
# Add external potential force
a -= Fext * pos
# Add viscosity
a -= nu * vel
return a
```

```
In [ ]:
        def run simulation(tEnd=12, dt=0.04, N=400, h=0.1, nu=1):
            """parameters for the main simulation funcion
            tEnd : end time
                 : time step
            dt
                   : number of particles
                  : globally constant smoothing length
            nu
                  : artificial viscosity
            0.00
            # Main simulation parameters
            t = 0 # current time of the simulation
            M = 2 # star mass
            R = 0.75 # star radius
            k = 0.1 # equation of state constant
            n = 1 # polytropic index
            # Generate Initial Conditions
            np.random.seed(42) # set the random number generator seed
            Fext = (
                2 * k * (1 + n) * np.pi ** (-3 / (2 * n)) * (M * gamma(5 / 2 + n))
            ) # ~ 2.01
            m = M / N # single particle mass
            pos = np.random.randn(N, 3) # randomly selected positions and veloci
```

```
vel = np.zeros(pos.shape)
    # calculate initial gravitational accelerations
    acc = compute accelerations(pos, vel, m, h, k, n, Fext, nu)
    # number of timesteps
    Nt = int(np.ceil(tEnd / dt))
    result = []
    resolution = 128
    lin = np.linspace(-2, 2, resolution)
    x, y = np.meshgrid(lin, lin)
    r = np.array([x.flatten(), y.flatten(), np.zeros(x.shape).flatten()])
    # Simulation Main Loop
    for i in tqdm.tqdm(range(Nt)):
        # kick-drift-kick
        # Kick
        vel += 0.5 * acc * dt
        # Drift
        pos += vel * dt
        # Calculate new acceleration
        acc = compute accelerations(pos, vel, m, h, k, n, Fext, nu)
        # Kick
        vel += 0.5 * acc * dt
        # update time
        t += dt
        # get density for plotting
        rho = compute_density(pos, pos, m, h)
        # get density for image
        # rho plot = compute density(r, pos, m, h).reshape((resolution, r
        result.append([pos.copy(), rho.copy(), vel.copy()])
        # result.append(rho_plot.copy())
    return result
# create final figure with positions of particles and density coloour cod
plt.scatter(result[-1][0][:, 0], result[-1][0][:, 1], c=result[-1][1])
plt.show()
```

```
In [ ]: result = run_simulation(tEnd=12.0, dt=0.04, N=400, h=0.1, nu=1.0)
                     | 0/300 [00:00<?, ?it/s]100%| | 300/300 [00:01<00:
      00, 217.10it/s]
```



In [ ]: make\_gif(cback, result, "sph\_nul.gif")

Save Images...

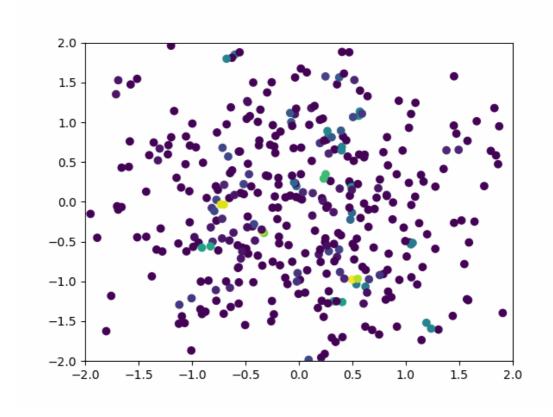
100%| 300/300 [00:00<00:00, 348.35it/s]

Done.

Make Gif...

100%| 300/300 [00:00<00:00, 444.49it/s]

Done.

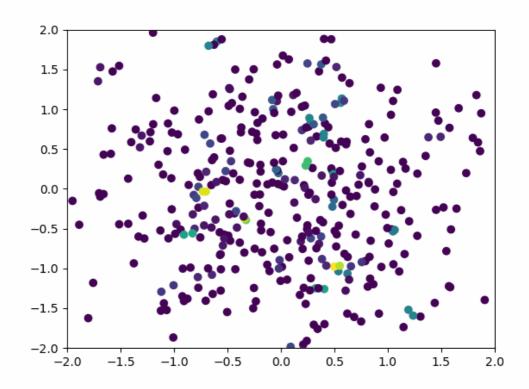


```
In [ ]: result = run simulation(tEnd=12.0, dt=0.04, N=400, h=0.1, nu=0.0)
        # create final figure with positions of particles and density coloour cod
        plt.scatter(result[-1][0][:, 0], result[-1][0][:, 1], c=result[-1][1])
        plt.show()
       100%|
                      | 300/300 [00:01<00:00, 158.23it/s]
         0.4
         0.3
         0.2
         0.1
         0.0
       -0.1
       -0.2
       -0.3
       -0.4
                 -0.4
                             -0.2
                                           0.0
                                                        0.2
                                                                     0.4
In [ ]: make gif(cback, result, "sph nu0.gif")
       Save Images...
       100%
                     300/300 [00:00<00:00, 337.40it/s]
       Done.
       Make Gif...
```

| 300/300 [00:00<00:00, 399.27it/s]

100%

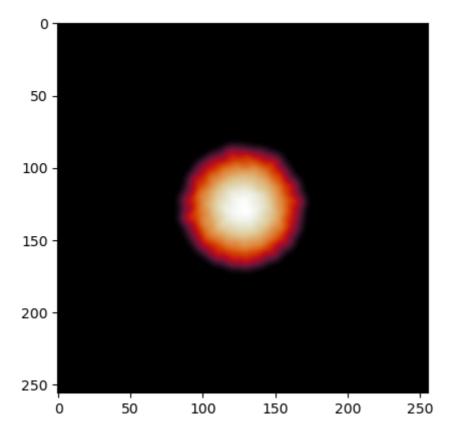
Done.



```
In [ ]: from scipy.signal import find peaks
        from scipy.optimize import curve fit
        viscosities = [0.5, 1.0, 2.0]
        for nu in viscosities:
            result = run simulation(tEnd=20.0, dt=0.04, N=400, h=0.1, nu=nu)
            # Root mean square velocity
            vels = np.array([np.sum(result[i][2] ** 2, axis=1) for i in range(len
            rms = np.sqrt(np.mean(vels, axis=1))
            # Plot the final positions
            fig, ax = plt.subplots(1, 2, figsize=(10, 5))
            ax[0].scatter(result[-1][0][:, 0], result[-1][0][:, 1], c=result[-1][
            ax[0].set_title(f"nu = {nu}")
            ax[0].set_xlabel("x")
            ax[0].set_ylabel("y")
            # Find peaks in the rms data
            peaks, _ = find_peaks(rms)
            # Define the exponential decay function
            def exponential_decay(x, a, b, c):
                return a * np.exp(-b * x) + c
            # Fit the exponential decay curve onto the peaks
            x_data = np.arange(len(rms))[peaks]
            y_data = rms[peaks]
            popt, pcov = curve_fit(exponential_decay, x_data, y_data, p0=[2.5, 0.
            xlin = np.linspace(0, len(rms), 1000)
            # Plot the rms data and the fitted curve
            ax[1].plot(np.arange(len(rms)), rms, "black", label="Velocity RMS")
```

```
# Plot the peaks
       ax[1].scatter(peaks, rms[peaks], color="red", label="Peaks")
       ax[1].plot(
            xlin,
            exponential_decay(xlin, *popt),
            color="orange",
            label="Exponential Decay with $\lambda=$\{:.3f\}".format(popt[1]),
       )
       ax[1].set xlabel("t")
       ax[1].set ylabel("rms")
       ax[1].set_title(f"nu = {nu}")
       ax[1].set yscale("log")
       ax[1].legend()
       plt.show()
100%|
                      500/500 [00:02<00:00, 217.10it/s]
(21,)
                                                                      nu = 0.5
                       nu = 0.5
   0.6
                                                   10<sup>0</sup>
   0.4
   0.2
                                                rms
   0.0
                                                  10^{-1}
  -0.2
  -0.4
                                                            Velocity RMS
                                                             Peaks
                                                             Exponential Decay with \lambda = 0.010
  -0.6
                                                  10
       -0.6
             -0.4
                          0.0
                                0.2
                                      0.4
                                            0.6
                                                              100
                                                                     200
                                                                             300
                                                                                    400
                                                                                           500
100%|
                      500/500 [00:02<00:00, 220.28it/s]
(16,)
                       nu = 1.0
                                                                      nu = 1.0
   0.6
                                                                  Velocity RMS
                                                                  Peaks
                                                                  Exponential Decay with \lambda = 0.021
                                                   10<sup>0</sup>
   0.4
   0.2
                                                ¥ 10<sup>−1</sup>
   0.0
  -0.2
  -0.4
                                                  10^{-2}
  -0.6
       -0.6
             -0.4
                          0.0
                                0.2
                                      0.4
                                            0.6
                                                              100
                                                                     200
                                                                             300
                                                                                    400
                                                                                           500
100%|
                      500/500 [00:02<00:00, 218.41it/s]
```

```
(7,)
                            nu = 2.0
                                                                       nu = 2.0
                                                                    Velocity RMS
           0.6
                                                                    Peaks
                                                      10<sup>0</sup>
                                                                    Exponential Decay with \lambda = 0.040
           0.4
           0.2
                                                  § 10⁻¹
           0.0
          -0.2
                                                     10^{-2}
          -0.4
          -0.6
                                          0.4
                                                0.6
                                                                100
                                                                      200
                                                                             300
                                                                                   400
                                                                                          500
                                    0.2
              -0.6
                               0.0
In [ ]: from maxpy.makegif import make_gif
         def cback(data, name):
              plt.clf()
              plt.imshow(data[0], cmap="cmr.sunburst")
              plt.savefig(name)
         make gif(cback, result)
        Save Images...
                        | 150/150 [00:00<00:00, 321.87it/s]
        100%|
        Done.
        Make Gif...
        100%|
                         | 150/150 [00:00<00:00, 405.41it/s]
        Done.
In [ ]: import cmasher as cmr
         print(result[-1][0].shape)
         plt.imshow(result[-1][0], cmap="cmr.sunburst")
        (256, 256)
Out[]: <matplotlib.image.AxesImage at 0x7f8ebf6864d0>
```



In [ ]: