Michael Wasserstein - Module 2

1/26/2021

Excercises Completed

- 1. Getting Started and Modifying Code
- 2. Explaining Error
- 3. Exact and Approximate Trajectories in 2D
- 4. Adding air resistance to the 1D free fall model
- 5. Compare exact and approximate numerical solutions for 1D free fall with air resistance
- 6. Euler Simulations of Drag in 2D
- 7. Physics Extension 4 Including atmospheric density variation
- 8. Physics Extension 3 Is 45° still the maximum range?

I collaborated with Julian, one of my groupmates, on this module. We worked mostly on Exercise 3.

```
In [1]: #Import the numpy and matplotlib packages
import numpy as np
import matplotlib.pyplot as plt
```

(Required) Exercise 1:

Getting Started and Modifying Code

Copy and modify the code above to make the simulation plot the position versus time and velocity versus time curves for an object thrown vertically upward (as opposed to dropped) at 80 m/s. The time axis should extend at least as long as it takes for the ball to return to its starting height.

Projectile Motion

Free Fall in 1D

Two key equations

```
1. $$ y(t) = y_0 + v_{y0} t + \frac{1}{2} a t^2 $$
2. $$v(t) = v_{y0} + a t$$
```

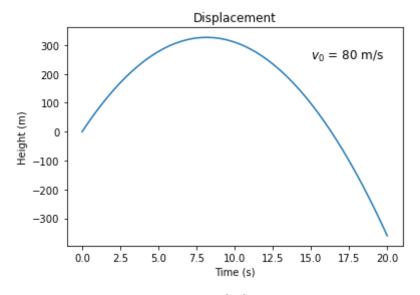
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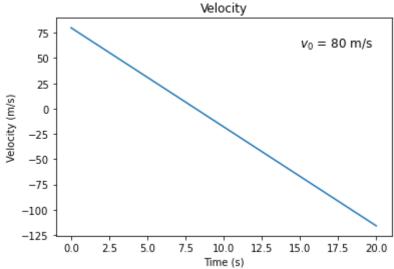
```
y0 - initial position (m) - float or int
v0 - initial velocity (m/s) - float or int
a - constant acceleration (m/s**2) - float or int
t - time (s) - float, int, or NumPy array
'''
#if t is an array, then y is an array with the same number of points
y = y0 + v0*t + 0.5*a*t**2
return y
```

```
#Now generate the data to plot from the analytic solution
In [4]:
         #Input parameters for model
         g = 9.8
                   #acceleration due to gravity (m/s^2)
                    #initial position (m)
         y \ 0 = 0
                      #initial vertical velocity (m/s)
         v \ v0 = 80
         t fin = 20
                      #Final time (s)
         #Define the time points to plot.
         t = np.linspace(0, t fin) #t is an array of 50 points (by default) from 0 to 10
         #Use the analytic solutions to find the position and velocity as a function of t
         y = KinematicEq1(y 0, v y0, -g, t)
         v = KinematicEq2(v y0, -g, t)
```

```
In [5]:
        #Now plot the position and velocity as functions of time.
         #The way I'm doing it here plots the two in separate figures
         figy, axy = plt.subplots() #Create a new figure object and its associated axe
         axy.plot(t, y)
                                      #Plot position versus t
         axy.set xlabel('Time (s)') #Label the x-axis
         axy.set_ylabel('Height (m)') #Label the y-axis
         axy.text(15, 250, r'$v 0$ = 80 m/s', fontsize=12) # Label v_0
         axy.set title("Displacement") #Title the plot
         #Repeat this for the velocity plot
         figv, axv = plt.subplots()
         axv.plot(t, v)
         axv.set xlabel('Time (s)')
         axv.set ylabel('Velocity (m/s)')
         axv.set title('Velocity')
         axv.text(15, 60, r'$v 0$ = 80 m/s', fontsize=12)
         plt.show()
```

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(Required) Exercise 2:

Explaining Error

```
# The following cells are code for Excercise 2
In [6]:
         #Lets define a function to encapsulate this algorithm
In [7]:
         #It will return the data describing the time, position, and velocity of the 1D t
         #We can then manipulate or plot that data as we see fit
         def EulerFreeFall1D(y0, vy0, a, t_fin, dt):
             Performs the Euler Method to solve of the 1D trajectory of an object in free
             Parameters:
             y0 - Initial height (m)
             v0 - Initial velocity (m/s)
             t_fin - Final time of simulation (s)
             dt - Time Step Size (s)
             Returns:
             A tuple of NumPy arrays
             (time, vy, y)
```

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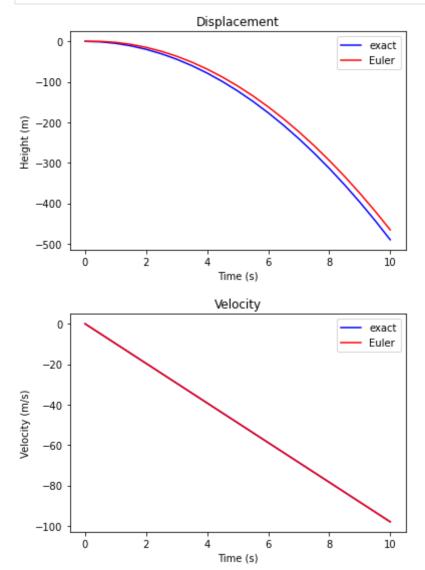
```
# The Euler algorithm requires us to discretize time, and evolve the position
# of the object in discree time steps
t_steps = int(t_fin/dt) + 1
                               #number of steps necessary (always rounding
                                # Int makes integer, add 1 to round up
# Defines the 1D arrays to be used in the computation and
# sets all values in the arrays to zero
t = np.zeros(t_steps)
y = np.zeros(t steps)
vy = np.zeros(t steps)
# Initial conditions. The first index of an array is zero.
t[0] = 0
y[0] = y0
vy[0] = vy0
# Euler algorithm and evaluation of exact solutions for v and y
# Use for loop to iterate over the range of time values we want
for i in range(0, t_steps-1):
   t[i+1] = t[i] + dt
                                                #Time increases linearly
    vy[i+1] = vy[i] + a * dt
                                                #First update the velocity
                                                #Then update the position
    y[i+1] = y[i] + vy[i] * dt
#We include a return statement that retusn our three NumPy data arrays group
return (t, vy, y)
```

```
#Now we call the function we just write to get the data
In [8]:
         #The utility of putting it into a function was that we could call it over and ov
         #initial conditions if we wanted to, without copying and pasting that large sect
         #over again
         #Physical Parameters to pass into our functions
                   #acceleration due to gravity (m/s^2)
                    #initial position (m)
         y \ 0 = 0
         v y 0 = 0
                    #initial vertical velocity (m)
         t fin = 10
                      #Final time (s)
         #Time steps for the Euler method
         dt = 0.5
         #Generate data using the Euler Method
         (time, vy Euler, y Euler) = EulerFreeFall1D(y 0, v y0, -g, t fin, dt)
         #Generate data from the analytic solutions for the same time point
         y Exact = KinematicEq1(y 0, v y0, -g, time)
         v Exact = KinematicEq2(v y0, -g, time)
```

```
In [9]:
        #Now we can plot both the position and velocity, comparing the exact solution to
         figy, axy = plt.subplots()  #Create a new figure object and its associated axe
         axy.plot(time, y Exact, 'b-', label='exact')
                                                                   #Plot exact position
         axy.plot(time, y_Euler, 'r-', label='Euler')
                                                        #Add Euler solution to the plo
         axy.set xlabel('Time (s)')
                                                          #Label the horizontal axis
         axy.set ylabel('Height (m)')
                                                          #Label the vertical axis
         axy.set title("Displacement")
                                                          #Title the plot
         axy.legend()
                                                          #Put a legend on the plot
         #Repeat for the velocity plot
```

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```
figv, axv = plt.subplots()
axv.plot(time, v_Exact, 'b-', label='exact')
axv.plot(time, vy_Euler, 'r-', label='Euler')
axv.set_xlabel('Time (s)')
axv.set_ylabel('Velocity (m/s)')
axv.set_title('Velocity')
axv.legend()
plt.show()
```



Hmmm. Seems like for the default value of step size I have used, the exact and Euler solution don't agree that well in the position versus time graph. Why do you think that is? Change the step size and find one for which the exact and Euler solutions agree reasonably well through the entire simulation. Produce plots of \$v(t)\$ and \$y(t)\$ where each has as least three different curves calculated using the Euler method with different values of \$\Delta\$ t\$. At least one should be in "good agreement" with the analytic solution. What did you define as "good agreement"?

As shown in the first plot, for the timestep \$\Delta t\$ = 0.5 s, the exact and Euler solutions don't agree well for position vs. time. I think this occurs because \$\Delta t\$ is too large to accurately simulate the exact solution using the Euler method. The smaller the timestep is, the closer the result will be to the exact solution, because we will be "nudging" the height to agree with what is expected more frequently.

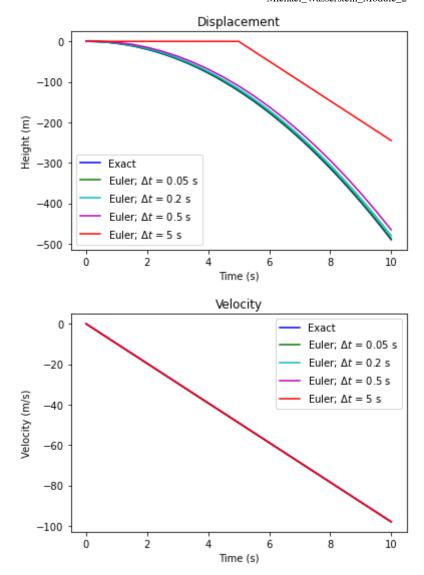
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If we make the timestep be smaller by a factor of 10 so that \$\Delta t\$ = 0.05 s, we have a the exact and Euler solutions will agree reasonably well.

```
#Now we call the function we just write to get the data
#The utility of putting it into a function was that we could call it over and ov
#initial conditions if we wanted to, without copying and pasting that large sect
#over again
#Physical Parameters to pass into our functions
g = 9.8 #acceleration due to gravity (m/s^2)
y_0 = 0
          #initial position (m)
v y 0 = 0
           #initial vertical velocity (m)
             #Final time (s)
t fin = 10
#Time steps for the Euler method
dt1 = 0.05
dt2 = 0.5
dt3 = 5
dt4 = 0.2
#Generate data using the Euler Method for each timestep
(time1, vy_Euler1v, y_Euler1v) = EulerFreeFall1D(y_0, v_y0, -g, t_fin, dt1)
(time2, vy_Euler2v, y_Euler2v) = EulerFreeFall1D(y_0, v_y0, -g, t_fin, dt2)
(time3, vy Euler3v, y Euler3v) = EulerFreeFall1D(y 0, v y0, -q, t fin, dt3)
(time4, vy_Euler4v, y_Euler4v) = EulerFreeFall1D(y_0, v_y0, -g, t_fin, dt4)
#Generate data from the analytic solutions for the same time point
y_Exact = KinematicEq1(y_0, v_y0, -g, time1)
v Exact = KinematicEq2(v y0, -g, time1)
```

```
#Now we can plot both the position and velocity, comparing the exact solution to
In [11]:
          figy, axy = plt.subplots() #Create a new figure object and its associated axe
          axy.plot(time1, y_Exact, 'b-', label='Exact')
                                                                          #Plot exact position
          axy.plot(time1, y Euler1v, 'g-', label=r'Euler; $\Delta t$ = 0.05 s')
          axy.plot(time4, y_Euler4v, 'c-', label=r'Euler; $\Delta t$ = 0.2 s')
                                                                                        #Add E
          axy.plot(time2, y_Euler2v, 'm-', label=r'Euler; $\Delta t$ = 0.5 s')
                                                                                        #Add E
          axy.plot(time3, y Euler3v, 'r-', label=r'Euler; $\Delta t$ = 5 s')
          axy.set xlabel('Time (s)')
                                                               #Label the horizontal axis
          axy.set ylabel('Height (m)')
                                                               #Label the vertical axis
          axy.set title("Displacement")
                                                               #Title the plot
          axy.legend()
                                                               #Put a legend on the plot
          #Repeat for the velocity plot
          figv, axv = plt.subplots()
          axv.plot(time1, v Exact, 'b-', label='Exact')
          axv.plot(time1, vy_Euler1v, 'g-', label=r'Euler; $\Delta t$ = 0.05 s')
axv.plot(time4, vy_Euler4v, 'c-', label=r'Euler; $\Delta t$ = 0.2 s')
          axv.plot(time2, vy_Euler2v, 'm-', label=r'Euler; $\Delta t$ = 0.5 s')
          axv.plot(time3, vy_Euler3v, 'r-', label=r'Euler; $\Delta t$ = 5 s')
          axv.set xlabel('Time (s)')
          axv.set ylabel('Velocity (m/s)')
          axv.set title('Velocity')
          axv.legend()
          plt.show()
```

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As we can see in the height vs. time plot, the light blue, green, and blue lines are practically on top of each other. Because this is the case, I can reasonly conclude that that when $\Delta = 0.2$ s or less (the light blue line), the Euler solution is in "good agreement" with the analytic one. My definition of "good agreement" is if the two lines appear on top of each other. Additionally, the light blue color, where $\Delta = 0.2$ s is the largest $\Delta = 0.2$ s where I can conclude that the Euler and analytic solutions are in good agreement.

Challenge: Why does the Euler solution on the velocity versus time graph agree with the exact solution, regardless of time step?

The Euler solution on the velocity versus time graph always agrees with the exact solution because our code looks like vy[i+1] = vy[i] + a * dt. We have a for loop for the range of timesteps, meaning that our for loop is longer if \$\Delta t\$ (dt) is smaller. However, as shown in vy[i+1] = vy[i] + a * dt, we always add the acceleration multiplied by the timestep to generate the next item in our array. This means that for a small timestep, we may see velocities like [0, -1, -2, -3, -4, -5, -6, -7, -8, -9, -10] in our array, whereas for a larger timestep, we may have something like [0, -2, -4, -6, -8, -10]. But when plotted vs. time in the x-axis, on the same plot, these two arrays will appear exactly the same, as we see in the velocity vs. time plot above.

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Coding Challenge/Extension: Can some parts of the for loop in EulerFreeFall1D be readily implemented with NumPy vector/array operations? Why or why not? If so, implement them this way.

Yes! The time part of the for loop in EulerFreeFall1D can be easily implemented with NumPy array operations. Here is code that will do it:

```
In [12]: t_fin = 10
    dt = 0.5
    t_steps = int(t_fin/dt) + 1
    t = np.linspace(0, t_fin, t_steps)
```

In the above code, we use a final time, a timestep \$\Delta t\$ and an inital time (0 s). We use the np.linspace() function to generate an array of all times used in the Euler method.

I do not believe that the velocity and height parts of the for loop in EulerFreeFall1D can be easily implemented with NumPy array operations because they do not increase linearly, as time does.

(Required) Exercise 3:

Exact and Approximate Trajectories in 2D

Make plots of the trajectory of a projectile in free fall launched in 2-dimensions. So you must now, for each value of time, keep track of both the x-position and y-position, as well as the two components of the velocity vector \$v_x\$ and \$v_y\$.

```
#In this cell we define a funciton which will caculate the analytic

#expression for y(t), using the trajectory equation

def tragectory(theta_0, v0, a, x):

Returns the displacement in y direction of an object subject to constant acc

Parameters:

theta_0 - initial angle (degrees) - float or int

v0 - initial velocity (m/s) - float or int

a - constant acceleration (m/s**2) - float or int

x - position in x direction (m) - float, int, or NumPy array

'''

#if x is an array, then y is an array with the same number of points

y = np.tan(theta_0) * x + (a / (2 * (v0 * np.cos(theta_0))) ** 2) * x ** 2

return y
```

```
#Lets define a function to encapsulate this algorithm
#It will return the data describing the time, position, and velocity of the 1D t
#We can then manipulate or plot that data as we see fit
def EulerFreeFalltrag(y0, vy0, x0, vx0, a, t_fin, dt):

Performs the Euler Method to solve of the 1D trajectory of an object in free

Parameters:
y0 - Initial height (m)
x0 - Initial displacement (m)
vx0 - Initial y velocity (m/s)
vy0 - Initial y velocity (m/s)
t_fin - Final time of simulation (s)
```

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```
dt - Time Step Size (s)
Returns:
A tuple of NumPy arrays
(time, vy, y, vx, x)
# The Euler algorithm requires us to discretize time, and evolve the position
# of the object in discree time steps
t steps = int(t fin/dt) + 1
                                #number of steps necessary (always rounding
                                # Int makes integer, add 1 to round up
# Defines the 1D arrays to be used in the computation and
# sets all values in the arrays to zero
t = np.zeros(t steps)
y = np.zeros(t_steps)
vy = np.zeros(t_steps)
vx = np.zeros(t_steps)
x = np.zeros(t steps)
# Initial conditions. The first index of an array is zero.
t[0] = 0
y[0] = y0
vy[0] = vy0
x[0] = x0
vx[0] = vx0
# Euler algorithm and evaluation of exact solutions for v and y
# Use for loop to iterate over the range of time values we want
for i in range(0, t steps-1):
    t[i+1] = t[i] + dt
                                                #Time increases linearly
                                   #First update the y velocity
    vy[i+1] = vy[i] + a * dt
    y[i+1] = y[i] + vy[i] * dt
                                    # Update height
    vx[i+1] = vx[i]
                                    # Update x velocity (remains same)
    x[i+1] = x[i] + vx[i] * dt
#We include a return statement that returns our five NumPy data arrays group
return (t, vy, y, vx, x)
```

```
#Now generate the data to plot from the analytic solution
In [15]:
          #Input parameters for model
          q = 9.8
                    #acceleration due to gravity (m/s^2)
          theta 0 = 70 # initial angle (degrees)
          theta 0 rad = np.radians(theta 0) # Converts our initial angle to radians
          x 0 = 0
                       #initial x position (m)
                       #initial y position (m)
          y \ 0 = 0
          v0 = 10
                       # initial velocity (m/s)
          v_y0 = v0 * np.sin(theta_0_rad) # initial y velocity (m/s)
          v_x0 = v0 * np.cos(theta_0_rad) # initial x velocity (m/s)
          t fin = 10
                                      #Final time (s)
          #Time steps for the Euler method
          dt1 = 0.05
```

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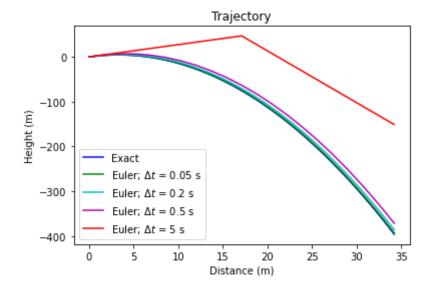
```
dt2 = 0.5
dt3 = 5
dt4 = 0.2

#Define the time points to plot.
t = np.linspace(0, t_fin, 50)

#Generate data using the Euler Method
(time1, vy_Euler1, y_Euler1, vx_Euler1, x_Euler1) = EulerFreeFalltrag(y_0, v_y0, (time2, vy_Euler2, y_Euler2, vx_Euler2, x_Euler2) = EulerFreeFalltrag(y_0, v_y0, (time3, vy_Euler3, y_Euler3, vx_Euler3, x_Euler3) = EulerFreeFalltrag(y_0, v_y0, (time4, vy_Euler4, y_Euler4, vx_Euler4, x_Euler4) = EulerFreeFalltrag(y_0, v_y0, #Use the analytic solutions to find the height and velocity as a function of x d y_Exact = KinematicEq1(y_0, v_y0, -g, t)
x_Exact = KinematicEq1(x_0, v_x0, 0, t)
vy_Exact = KinematicEq2(v_y0, -g, t)
vx_Exact = KinematicEq2(v_x0, 0, t)
```

Here we have a plot of height in the vertical vs. displacement in the horizont In [16]: #Create a new figure object and its associated axe figy, axy = plt.subplots() axy.plot(x_Exact, y_Exact, 'b-', label='Exact') #Plot exact positi axy.plot(x_Euler1, y_Euler1, 'g-', label=r'Euler; \$\Delta t\$ = 0.05 s') axy.plot(x_Euler4, y_Euler4, 'c-', label=r'Euler; \$\Delta t\$ = 0.2 s') axy.plot(x_Euler2, y_Euler2, 'm-', label=r'Euler; \$\Delta t\$ = 0.5 s') axy.plot(x_Euler3, y_Euler3, 'r-', label=r'Euler; \$\Delta t\$ = 5 s') #Label the horizontal axis axy.set_xlabel('Distance (m)') axy.set_ylabel('Height (m)') #Label the vertical axis #Title the plot axy.set title("Trajectory") axy.legend() #Put a legend on the plot

Out[16]: <matplotlib.legend.Legend at 0x11e620ee0>

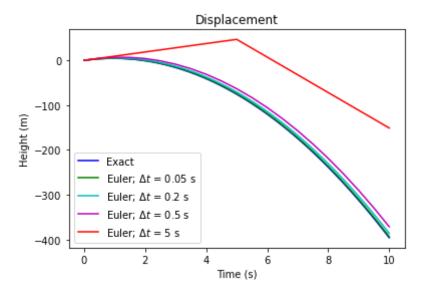


```
In [17]: # Here we have a plot of height vs. time
figy, axy = plt.subplots() #Create a new figure object and its associated axe
axy.plot(t, y_Exact,'b-', label='Exact') #Plot exact position ver
axy.plot(time1, y_Euler1, 'g-', label=r'Euler; $\Delta t$ = 0.05 s')
axy.plot(time4, y_Euler4, 'c-', label=r'Euler; $\Delta t$ = 0.2 s')
axy.plot(time2, y_Euler2, 'm-', label=r'Euler; $\Delta t$ = 0.5 s')
axy.plot(time3, y_Euler3, 'r-', label=r'Euler; $\Delta t$ = 5 s')
axy.set_xlabel('Time (s)') #Label the horizontal axis
```

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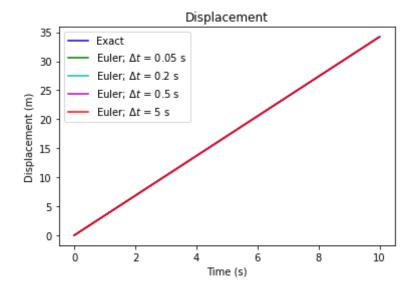
```
axy.set_ylabel('Height (m)')  #Label the vertical axis
axy.set_title("Displacement")  #Title the plot
axy.legend()  #Put a legend on the plot
```

Out[17]: <matplotlib.legend.Legend at 0x11e778c10>



```
# Here we have a plot of x displacement vs. time
In [18]:
          figy, axy = plt.subplots()
                                        #Create a new figure object and its associated axe
          axy.plot(t, x_Exact, 'b-', label='Exact')
                                                                   #Plot exact position ver
          axy.plot(time1, x_Euler1, 'g-', label=r'Euler; $\Delta t$ = 0.05 s')
          axy.plot(time4, x_Euler4, 'c-', label=r'Euler; $\Delta t$ = 0.2 s')
          axy.plot(time2, x_Euler2, 'm-', label=r'Euler; $\Delta t$ = 0.5 s')
          axy.plot(time3, x_Euler3, 'r-', label=r'Euler; $\Delta t$ = 5 s')
          axy.set xlabel('Time (s)')
                                                             #Label the horizontal axis
          axy.set ylabel('Displacement (m)')
                                                                   #Label the vertical axis
          axy.set title("Displacement")
                                                             #Title the plot
          axy.legend()
                                                             #Put a legend on the plot
```

Out[18]: <matplotlib.legend.Legend at 0x11e930460>

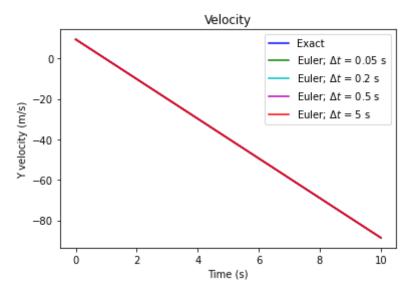


```
In [19]: # Here we have a plot of y velocity vs. time
figy, axy = plt.subplots() #Create a new figure object and its associated axe
axy.plot(t, vy_Exact,'b-', label='Exact') #Plot exact position ve
axy.plot(time1, vy_Euler1, 'g-', label=r'Euler; $\Delta t$ = 0.05 s')
axy.plot(time4, vy_Euler4, 'c-', label=r'Euler; $\Delta t$ = 0.2 s')
```

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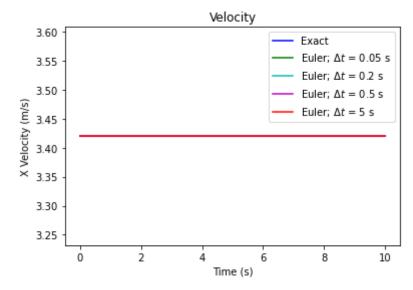
```
axy.plot(time2, vy_Euler2, 'm-', label=r'Euler; $\Delta t$ = 0.5 s')
axy.plot(time3, vy_Euler3, 'r-', label=r'Euler; $\Delta t$ = 5 s')
axy.set_xlabel('Time (s)')  #Label the horizontal axis
axy.set_ylabel('Y velocity (m/s)')  #Label the vertical axis
axy.set_title("Velocity")  #Title the plot
axy.legend()  #Put a legend on the plot
```

Out[19]: <matplotlib.legend.Legend at 0x11e48c430>



```
# Here we have a plot of x velocity vs. time
In [20]:
          figy, axy = plt.subplots()
                                        #Create a new figure object and its associated axe
          axy.plot(t, vx Exact, 'b-', label='Exact')
                                                                    #Plot exact position ve
          axy.plot(time1, vx_Euler1, 'g-', label=r'Euler; $\Delta t$ = 0.05 s')
          axy.plot(time4, vx Euler4, 'c-', label=r'Euler; $\Delta t$ = 0.2 s')
          axy.plot(time2, vx Euler2, 'm-', label=r'Euler; $\Delta t$ = 0.5 s')
          axy.plot(time3, vx Euler3, 'r-', label=r'Euler; $\Delta t$ = 5 s')
          axy.set_xlabel('Time (s)')
                                                             #Label the horizontal axis
          axy.set ylabel('X Velocity (m/s)')
                                                                   #Label the vertical axis
          axy.set title("Velocity")
                                                         #Title the plot
                                                             #Put a legend on the plot
          axy.legend()
```

Out[20]: <matplotlib.legend.Legend at 0x11e275160>



(Required) Exercise 4:

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Adding air resistance to the 1D free fall model

Add the drag force to your numerical simualtion using the Euler Method. Plot the position versus time and velocity versus time graphs. Include the trajectories in vaccum (no air resitance) on the same plots. You should observe the object reaches a terminal velocity. You'll need to consider a real object, so here are some ideas of air resitantce that can be calculated for spherical objects (i.e. balls). My recommendation is you define a function that computes the drag strength \$b\$ from these parameters.

For a sphere $C_D = 0.5$. At atmostpheric pressure $\rho = 1.2~\text{kg/m}^3$ at sea level (\$20^\circ\$ and 1 atm of pressure).

For a baseball m = 0.145 kg, and d = 74 mm (according to Wikipedia)

For an NBA basketball m = 0.620 kg, and $d = 75/(2\pi)$ cm (according to Wikipedia))

We will first write code to compute drag strength \$b\$:

```
In [22]:
          #Lets define a function to encapsulate this algorithm
          #It will return the data describing the time, position, and velocity of the 1D t
          #We can then manipulate or plot that data as we see fit
          def EulerFreeFallDrag(y0, vy0, a, t fin, dt, C D, rho, radius, mass):
              Performs the Euler Method to solve of the 1D trajectory of an object in free
              Parameters:
              y0 - Initial height (m)
              x0 - Initial displacement (m)
              vx0 - Initial y velocity (m/s)
              vy0 - Initial y velocity (m/s)
              t fin - Final time of simulation (s)
              dt - Time Step Size (s)
              C D - drag coefficient (0.5 for a sphere); depends on shape of object (dimen
              rho - density of air (generally 1.2 kg/m<sup>3</sup>)
              radius - radius of object
              mass -mass of object (kg)
              Returns:
              A tuple of NumPy arrays
              (time, vy, y, vx, x)
```

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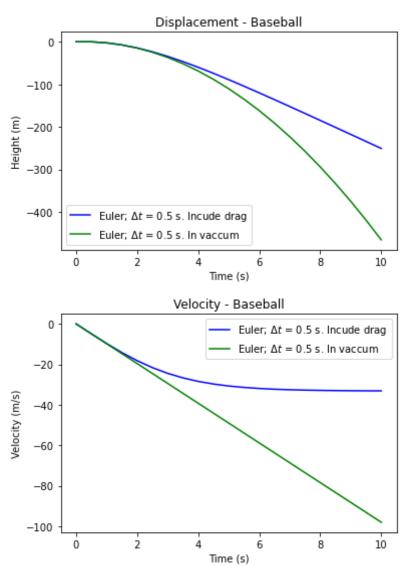
```
111
b = drag(C_D, rho, radius)
m = mass
# The Euler algorithm requires us to discretize time, and evolve the position
# of the object in discree time steps
t steps = int(t fin/dt) + 1
                              #number of steps necessary (always rounding
                                # Int makes integer, add 1 to round up
# Defines the 1D arrays to be used in the computation and
# sets all values in the arrays to zero
t = np.zeros(t_steps)
y = np.zeros(t_steps)
vy = np.zeros(t_steps)
# Initial conditions. The first index of an array is zero.
t[0] = 0
y[0] = y0
vy[0] = vy0
# Euler algorithm and evaluation of exact solutions for v and y
# Use for loop to iterate over the range of time values we want
for i in range(0, t_steps-1):
   t[i+1] = t[i] + dt
                                                #Time increases linearly
    vy[i+1] = vy[i] + a * dt - (b / m) * vy[i] ** 2 * np.sign(vy[i]) * dt
    y[i+1] = y[i] + vy[i] * dt
                                                #Then update the position
#We include a return statement that retusn our three NumPy data arrays group
return (t, vy, y)
```

```
#Now we call the function we just write to get the data
In [23]:
          #The utility of putting it into a function was that we could call it over and ov
          #initial conditions if we wanted to, without copying and pasting that large sect
          #over again
          #Physical Parameters to pass into our functions
          g = 9.8 #acceleration due to gravity (m/s^2)
                     #initial position (m)
          y \ 0 = 0
                     #initial vertical velocity (m)
          y y 0 = 0
          t fin = 10
                       #Final time (s)
          #Time steps for the Euler method
          dt1 = 0.05
          dt2 = 0.5
          dt3 = 5
          dt.4 = 0.2
          # Use a baseball
                         # (kg)
          mass = 0.145
          radius = (74 / 1000) / 2 \# (m)
          C D = 0.5
          rho = 1.2
          #Generate data using the Euler Method
          (time1, vy Eulerd1, y Eulerd1) = EulerFreeFallDrag(y 0, v y0, -g, t fin, dt1, C
          (time2, vy_Eulerd2, y_Eulerd2) = EulerFreeFallDrag(y_0, v_y0, -g, t_fin, dt2, C_
          (time3, vy_Eulerd3, y_Eulerd3) = EulerFreeFallDrag(y_0, v_y0, -g, t_fin, dt3, C_
          (time4, vy Eulerd4, y Eulerd4) = EulerFreeFallDrag(y_0, v_y0, -g, t_fin, dt4, C_
```

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```
# Here we have a plot of height in the vertical vs. time
In [24]:
          figy, axy = plt.subplots()
                                      #Create a new figure object and its associated axe
          axy.plot(time2, y_Eulerd2,'b-', label=r'Euler; $\Delta t$ = 0.5 s. Incude drag')
          axy.plot(time2, y Euler2v, 'g-', label=r'Euler; $\Delta t$ = 0.5 s. In vaccum')
          axy.set_xlabel('Time (s)')
                                                             #Label the horizontal axis
          axy.set ylabel('Height (m)')
                                                             #Label the vertical axis
          axy.set_title("Displacement - Baseball")
                                                                        #Title the plot
                                                             #Put a legend on the plot
          axy.legend()
          # Here we have a plot of velocity in the vertical vs. time
          figy, axy = plt.subplots()
                                        #Create a new figure object and its associated axe
          axy.plot(time2, vy_Eulerd2,'b-', label=r'Euler; $\Delta t$ = 0.5 s. Incude drag'
          axy.plot(time2, vy_Euler2v, 'g-', label=r'Euler; $\Delta t$ = 0.5 s. In vaccum')
          axy.set_xlabel('Time (s)')
                                                             #Label the horizontal axis
          axy.set_ylabel('Velocity (m/s)')
                                                                 #Label the vertical axis
          axy.set_title("Velocity - Baseball")
                                                                    #Title the plot
          axy.legend()
                                                             #Put a legend on the plot
```

Out[24]: <matplotlib.legend.Legend at 0x11dee89d0>



(Required) Exercise 5:

Compare exact and approximate numerical solutions for 1D free fall with air resistance

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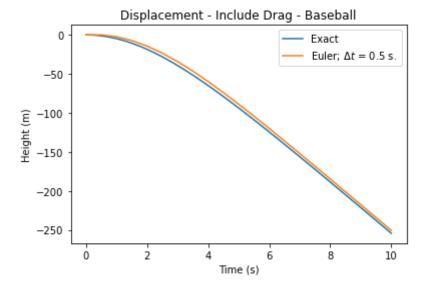
Now produce a plot where you compare the exact functions given above to the approximate numerical solution. Indicate the value of \$v_{\textrm{ter}}\$ on your plot of velocity using a horizontal line placed at the appropriate value. Play around a bit with your step size \$\Delta t\$ to find the largest value where the simulation and exact solution agree. How does this value of \$\Delta t\$ compared to that when you compared the exact and Euler solution without air resistance in Exercise 2?

```
In [28]: #Now generate the data to plot from the analytic solution
```

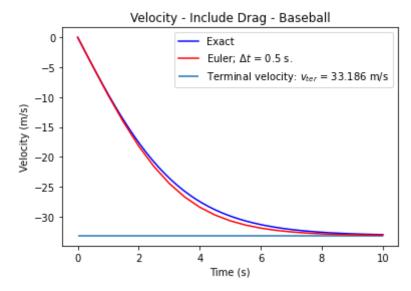
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```
#Input parameters for model
q = 9.8
            #acceleration due to gravity (m/s^2)
t fin = 10
              #Final time (s)
# Use a baseball
mass = 0.145
                 # (kg)
radius = (74 / 1000) / 2
C D = 0.5
rho = 1.2
v ter = v ter func(mass, C D, rho, radius)
#Define the time points to plot.
t = np.linspace(0, t_fin) #t is an array of 50 points (by default) from 0 to 10
#Use the analytic solutions to find the position and velocity as a function of t
v = ExactDragV(v ter, -g, t)
y = ExactDragY(v_ter, -g, t)
```

```
#Now plot the position and velocity as functions of time.
In [29]:
          #The way I'm doing it here plots the two in separate figures
          figy, axy = plt.subplots()
                                       #Create a new figure object and its associated axe
          axy.plot(t, y, label='Exact')
                                                       #Plot position versus t
          axy.plot(time2, y_Eulerd2,label = r'Euler; $\Delta t$ = 0.5 s.')
          axy.set xlabel('Time (s)')
                                      #Label the x-axis
          axy.set_ylabel('Height (m)') #Label the y-axis
          axy.set title("Displacement - Include Drag - Baseball") #Title the plot
          axy.legend()
          #Repeat for the velocity plot
          figv, axv = plt.subplots()
          axv.plot(t, v, 'b-', label='Exact')
          axv.plot(time2, vy Eulerd2, 'r-', label=r'Euler; $\Delta t$ = 0.5 s.')
          axv.set xlabel('Time (s)')
          axv.set ylabel('Velocity (m/s)')
          axv.set_title('Velocity - Include Drag - Baseball')
          axv.hlines(-v_ter, 0, 10, label = r'Terminal velocity: $v_{ter}$ = 33.186 m/s')
          axv.legend()
          plt.show()
```

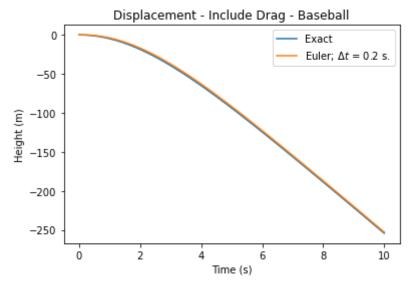


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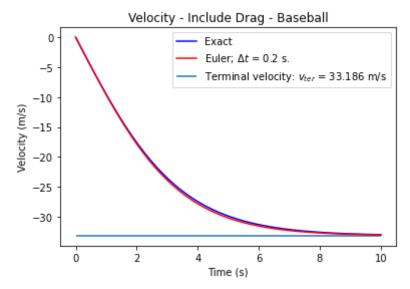


Now I repeat the process with $\Delta t = 0.2$ s, which appears to be the largest value of $\Delta t = 0.2$ s, which appears to be the largest value of $\Delta t = 0.2$ s, which appears to be the largest value of $\Delta t = 0.2$ s, which appears to be the largest value of $\Delta t = 0.2$ s, which appears to be the largest value of $\Delta t = 0.2$ s, which appears to be the largest value of $\Delta t = 0.2$ s, which appears to be the largest value of $\Delta t = 0.2$ s, which appears to be the largest value of $\Delta t = 0.2$ s.

```
In [30]:
          #Now plot the position and velocity as functions of time.
          #The way I'm doing it here plots the two in separate figures
          figy, axy = plt.subplots()
                                        #Create a new figure object and its associated axe
          axy.plot(t, y, label='Exact')
                                                       #Plot position versus t
          axy.plot(time4, y_Eulerd4,label = r'Euler; $\Delta t$ = 0.2 s.')
          axy.set xlabel('Time (s)')
                                       #Label the x-axis
          axy.set_ylabel('Height (m)') #Label the y-axis
          axy.set title("Displacement - Include Drag - Baseball") #Title the plot
          axy.legend()
          #Repeat for the velocity plot
          figv, axv = plt.subplots()
          axv.plot(t, v, 'b-', label='Exact')
          axv.plot(time4, vy_Eulerd4, 'r-', label=r'Euler; $\Delta t$ = 0.2 s.')
          axv.set xlabel('Time (s)')
          axv.set ylabel('Velocity (m/s)')
          axv.set_title('Velocity - Include Drag - Baseball')
          axv.hlines(-v ter, 0, 10, label = r'Terminal velocity: $v {ter}$ = 33.186 m/s')
          axv.legend()
          plt.show()
```



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In the above plots, \$\Delta t\$ = 0.2 s, which appears to be the largest value where the simulation and exact solution agree. This is the exact same largest \$\Delta t\$ as I had without air resistance in Exercise 2.

(Required) Exercise 6:

Euler Simulations of Drag in 2D

A) Write down the four differential equations that you need to solve numerically to ultimately plot a projectile trajectory \$y\$ vs \$x\$. You can still do that with pencil and paper, but if you want to learn how to typeset math, you can copy and past some of the math typed into the markdown cells here to typeset your equations into LaTex (a widely used math typesetting language supported in Markdown). Here is a quick guide to LaTex commands. All the code you see in between \$ \$ in the Markdown cells is Latex.

Thanks to Prof. Hess for some guidance on this part!

```
1. \frac{dv_y}{dt} = -g - \frac{b v_y}{m}{\sqrt{v_x}^2 + {v_y}^2}
```

- 2. $\frac{dv_x}{dt} = -\frac{b v_x}{m}{\sqrt{v_x}^2 + {v_y}^2}$
- 3. $frac{dy}{dt} = v_y$
- 4. $frac{dx}{dt} = v_x$

B) Now add drag to your simulation of 2D trajectories. The components of the velocity unit vector you can calculate by noting that a unit vector is, by definition $\frac{v}{v} = \frac{v}{v}$ { $|v| = \frac{v_x}{|v|} \cdot \frac{v_y}{|v|} \cdot \frac{v$

```
In [32]: def v_hat(vx, vy):
```

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Returns the unit vector components for a particular vector

v_hat = vx / magnitude(vx, vy) + vy / magnitude(vx, vy)

return v_hat

```
#Lets define a function to encapsulate this algorithm
In [33]:
          #It will return the data describing the time, position, and velocity of the 1D t
          #We can then manipulate or plot that data as we see fit
          def EulerFreeFallDrag2D(y0, vy0, x0, vx0, a, t_fin, dt, C_D, rho, radius, mass):
              Performs the Euler Method to solve of the 1D trajectory of an object in free
              Parameters:
              y0 - Initial height (m)
              x0 - Initial displacement (m)
              vx0 - Initial y velocity (m/s)
              vy0 - Initial y velocity (m/s)
              t fin - Final time of simulation (s)
              dt - Time Step Size (s)
              C_D - drag coefficient (0.5 for a sphere); depends on shape of object (dimen
              rho - density of air (generally 1.2 kg/m^3)
              radius - radius of object
              mass -mass of object (kg)
              Returns:
              A tuple of NumPy arrays
              (time, vy, y, vx, x)
              b = drag(C D, rho, radius)
              m = mass
              # The Euler algorithm requires us to discretize time, and evolve the position
              # of the object in discree time steps
              t steps = int(t fin/dt) + 1
                                              #number of steps necessary (always rounding
                                              # Int makes integer, add 1 to round up
              # Defines the 1D arrays to be used in the computation and
              # sets all values in the arrays to zero
              t = np.zeros(t_steps)
              y = np.zeros(t steps)
              vy = np.zeros(t steps)
              vx = np.zeros(t steps)
              x = np.zeros(t steps)
              # Initial conditions. The first index of an array is zero.
              t[0] = 0
              y[0] = y0
              vy[0] = vy0
              x[0] = x0
              vx[0] = vx0
              # Euler algorithm and evaluation of exact solutions for v and y
              # Use for loop to iterate over the range of time values we want
              for i in range(0, t steps-1):
                  t[i+1] = t[i] + dt
                                                               #Time increases linearly
```

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```
vy[i+1] = vy[i] + a * dt - (b / m) * magnitude(vx[i], vy[i]) * vy[i] * d
y[i+1] = y[i] + vy[i] * dt  #Then update the position
vx[i+1] = vx[i] - (b / m) * magnitude(vx[i], vy[i]) * vx[i] * dt
x[i+1] = x[i] + vx[i] * dt

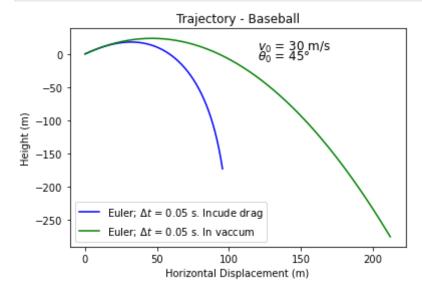
#We include a return statement that return our three NumPy data arrays group
return (t, vy, y, vx, x)
```

C) Produce a plot of the modified trajectory of a projectile launched at \$45^{\circ}\$ on the same plot as the tractory in vacuum (no air resistance). Pick a launch speed for a particular projecile that shows a clear difference between the two trajectories. Make sure your simulation runs for enough time that you can see the object return to the ground.

```
#Now generate the data to plot from the analytic solution
In [34]:
          #Input parameters for model
          g = 9.8 #acceleration due to gravity (m/s^2)
          theta_0 = 45  # initial angle
          theta 0 rad = np.radians(theta 0)
                       #initial x position (m)
          x \ 0 = 0
          v0 = 30
                         # initial velocity (m/s)
          v_y0 = v0 * np.sin(theta_0_rad)
          v_x0 = v0 * np.cos(theta_0_rad)
          y \ 0 = 0
          t fin = 10
          x fin = 10
                       #Final x position (m)
          #Time steps for the Euler method
          dt1 = 0.05
          dt2 = 0.5
          dt3 = 5
          #Define the time points to plot.
          x = np.linspace(x 0, x fin) #t is an array of 50 points (by default) from 0 to 1
          t = np.linspace(0, t fin, 50)
          # Use a baseball
                           # (kg)
          mass = 0.145
          radius = (74 / 1000) / 2
                                     \# (m)
          C D = 0.5
          rho = 1.2
          #Generate data using the Euler Method
          # With Drag first
          (time1, vy Euler1d, y Euler1d, vx Euler1d, x Euler1d) = EulerFreeFallDrag2D(y 0,
          # Now in vacuum
          (time1, vy Euler1, y Euler1, vx Euler1, x Euler1) = EulerFreeFalltrag(y 0, v y0,
```

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```
axy.text(120, -10, r'$\theta_0$ = 45°', fontsize=12)
plt.show()
```



I referred to this link to get text in my plot:

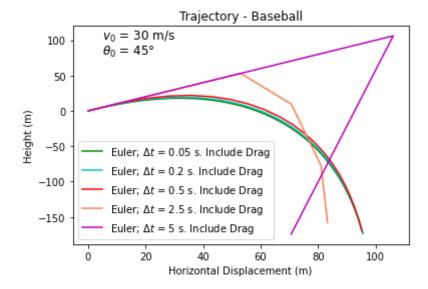
https://matplotlib.org/3.1.0/tutorials/text/text_intro.html

D) Double check the simulation for a few different values of \$\Delta t\$ and notice the plot changes if \$\Delta t\$ gets too large. Produce a plot with at least four different trajectories calculated with the same initial conditions, but different values of \$\Delta t\$. What value of \$\Delta t\$ is "just small enough" to not notice any difference between plots, and how does this value compare to what you found in Exercises 2 and 5?

```
In [36]:
          #Now generate the data to plot from the analytic solution
          #Input parameters for model
          g = 9.8
                      #acceleration due to gravity (m/s^2)
          theta 0 = 45
                         # initial angle
          x 0 = 0
                        #initial x position (m)
          v0 = 30
                         # initial velocity (m/s)
          theta 0 rad = np.radians(theta 0)
          v y0 = v0 * np.sin(theta 0 rad)
          v_x0 = v0 * np.cos(theta_0_rad)
          y_0 = 0
          t fin = 10
          x fin = 10
                        #Final x position (m)
          #Time steps for the Euler method
          dt1 = 0.05
          dt2 = 0.5
          dt3 = 5
          dt4 = 0.2
          dt5 = 2.5
          #Define the time points to plot.
          x = np.linspace(x 0, x fin) #t is an array of 50 points (by default) from 0 to 1
          t = np.linspace(0, t_fin, 50)
          # Use a baseball
          mass = 0.145
                           \# (kq)
          radius = (74 / 1000) / 2
                                     # (m)
```

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```
# Here we have a plot of x displacement vs. time
In [37]:
          figy, axy = plt.subplots()
                                        #Create a new figure object and its associated axe
          axy.plot(x_Euler1d, y_Euler1d, 'g-', label=r'Euler; $\Delta t$ = 0.05 s. Include
          axy.plot(x_Euler4d, y_Euler4d, 'c-', label=r'Euler; $\Delta t$ = 0.2 s. Include
          axy.plot(x_Euler2d, y_Euler2d, 'r-', label=r'Euler; $\Delta t$ = 0.5 s. Include
          axy.plot(x_Euler5d, y_Euler5d, 'coral', label=r'Euler; $\Delta t$ = 2.5 s. Inclu
          axy.plot(x_Euler3d, y_Euler3d, 'm-', label=r'Euler; $\Delta t$ = 5 s. Include Dr
          axy.set xlabel('Horizontal Displacement (m)')
                                                                                #Label the
          axy.set ylabel('Height (m)')
                                                             #Label the vertical axis
          axy.set_title("Trajectory - Baseball")
                                                                      #Title the plot
          axy.legend()
                                                             #Put a legend on the plot
          axy.text(5, 100, r'$v_0$ = 30 m/s', fontsize=12)
          axy.text(5, 80, r'\$\theta 0\$ = 45°', fontsize=12)
          plt.show()
```



From the above plot, we can see that a value for \$\Delta t\$ of \$\Delta t\$ = 0.2 s is just large enough to not notice a difference between plots. It turns out that this is the exact same value I found in Exercises 2 and 5.

I had an interesting discussion with Prof. Herdman about the pink line in the above plot, which is physically impossible, but a function of the large timestep ($\$ Delta t\$ = 5 s) that I used. With each new timestep, the height gets "nudged" in a specific direction, and we have a large "nudge" here. In the for loop I have vertical velocity as vx[i+1] = vx[i] - (b / m) * magnitude(vx[i], vy[i]) * vx[i] * dt. The minus sign next to (b / m) causes

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vx[i] to be less than vx[i+1] and in the cause of the pink line, it is so much less that it causes the horizontal displacement to decrease. This is because the dt term in the code is so large. Physically, though, this would not happen.

E) Comment on what drag has done to the shape of the trajectory.

Drag has caused the trajectory to have a lower maximum hight and a smaller maximum horizontal displacement. Another way to put it is that it has decreased the radius of the curve produced by the projectile

F) Plot the magnitude of the velocity over enough time to observe the object approach a terminal velocity.

```
#Now generate the data to plot from the analytic solution
In [38]:
          #Input parameters for model
          g = 9.8 #acceleration due to gravity (m/s^2)
          theta_0 = 45  # initial angle
         x_0 = 0 #initial x position (m)
          v0 = 30
                       # initial velocity (m/s)
          theta_0_rad = np.radians(theta 0)
          v_y0 = v0 * np.sin(theta_0_rad)
          v_x0 = v0 * np.cos(theta_0_rad)
         y \ 0 = 0
          t fin = 30
          x \text{ fin} = 10 #Final x position (m)
          #Time steps for the Euler method
          dt4 = 0.2
          #Define the time points to plot.
          x = np.linspace(x_0, x_fin) #t is an array of 50 points (by default) from 0 to 1
          t = np.linspace(0, t fin)
          # Use a baseball
         mass = 0.145 # (kg)
         radius = (74 / 1000) / 2
                                   # (m)
          C_D = 0.5 # for sphere
          rho = 1.2
                         # atmosphere
          #Generate data using the Euler Method
          # With Drag first
          (time4, vy_Euler4d, y_Euler4d, vx_Euler4d, x_Euler4d) = EulerFreeFallDrag2D(y_0,
                                                                                     t fi
```

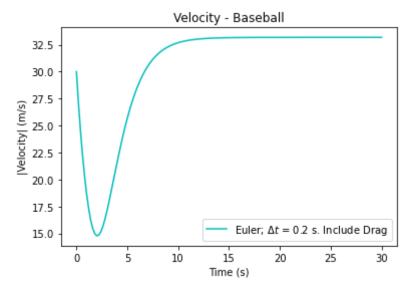
```
In [39]: mag_v = magnitude(vx_Euler4d, vy_Euler4d)
    print(mag_v[-1])

# Here we have a plot of x displacement vs. time
    figy, axy = plt.subplots() #Create a new figure object and its associated axe
    axy.plot(time4, mag_v, 'c-', label=r'Euler; $\Delta t$ = 0.2 s. Include Drag')
    axy.set_xlabel('Time (s)') #Label the horizontal axis
    axy.set_ylabel('|Velocity| (m/s)') #Label the vertical axis
    axy.set_title("Velocity - Baseball") #Title the plot
    axy.legend() #Put a legend on the plot
```

Out[39]: <matplotlib.legend.Legend at 0x11df12a30>

33.186366371209864

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In the above plot, we can see the ball slow down initially, an then reach a terminal velocity as time progresses. The magnitude of this terminal velocity is approximately $|v_{ter}| = 33.1864$ m/s, as determined from my print statement $print(mag_v[-1])$.

Physics Extension 4

Including atmospheric density variation

To get an accurate trajectory for a projectile one must take into account several complications which can be accounted for in the numerical framework. For example, if a projectile goes very high then we have to allow for the reduction in air resistance as atmospheric density decreases. To illustrate this, consider an iron connonball (diameter 15 cm, density 7.8 g/cm\$^3\$) that is fired with intial velocity 300 m/s at 50 degrees above the horizontal. Assume the atmospheric density falls off expoentially with height \$\$\rho(y) = \rho_0 \exp(-y/\lambda)\$\$ where \$\rho_0 = 1.2~\textrm{kg/m}^3\$ is the atmospheric pressure at ground level, and \$\lambda \approx 10,000~\textrm{m}\$\$ is the empirical decay coefficient of the pressure with increasing height.

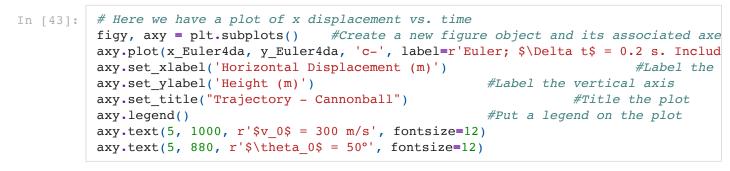
A) Modify your simulation to account for this changing atmospheric pressure, and and solve numerically for x(t) and y(t) for $0 \leq 35$ s. Plot the ball's trajectory and find its horizontal range.

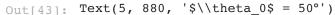
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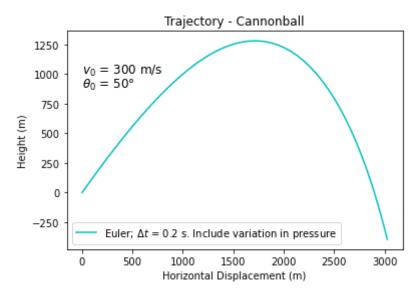
```
x0 - Initial displacement (m)
vx0 - Initial y velocity (m/s)
vy0 - Initial y velocity (m/s)
t_fin - Final time of simulation (s)
dt - Time Step Size (s)
C D - drag coefficient (0.5 for a sphere); depends on shape of object (dimen
rho - density of air (generally 1.2 kg/m<sup>3</sup>)
radius - radius of object
mass -mass of object (kg)
Returns:
A tuple of NumPy arrays
(time, vy, y, vx, x)
m = mass
# The Euler algorithm requires us to discretize time, and evolve the position
# of the object in discree time steps
t_steps = int(t_fin/dt) + 1
                                #number of steps necessary (always rounding
                                # Int makes integer, add 1 to round up
# Defines the 1D arrays to be used in the computation and
# sets all values in the arrays to zero
t = np.zeros(t_steps)
y = np.zeros(t steps)
vy = np.zeros(t steps)
vx = np.zeros(t steps)
x = np.zeros(t steps)
rho = np.zeros(t steps)
b = np.zeros(t steps)
# Initial conditions. The first index of an array is zero.
t[0] = 0
y[0] = y0
vy[0] = vy0
x[0] = x0
vx[0] = vx0
rho[0] = 1.2
b[0] = drag(C D, rho[0], radius)
# Euler algorithm and evaluation of exact solutions for v and y
# Use for loop to iterate over the range of time values we want
for i in range(0, t steps-1):
    t[i+1] = t[i] + dt
    y[i+1] = y[i] + vy[i] * dt
    x[i+1] = x[i] + vx[i] * dt
    rho[i+1] = rho func(y[i])
    b[i+1] = drag(C D, rho[i], radius)
    vy[i+1] = vy[i] + a * dt - (b[i] / m) * magnitude(vx[i], vy[i]) * vy[i]
    vx[i+1] = vx[i] - (b[i] / m) * magnitude(vx[i], vy[i]) * vx[i] * dt
#We include a return statement that retusn our three NumPy data arrays group
return (t, vy, y, vx, x)
```

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```
#Now generate the data to plot from the analytic solution
In [42]:
          #Input parameters for model
          g = 9.8
                      #acceleration due to gravity (m/s^2)
          theta 0 = 50
                         # initial angle
          theta_0_rad = np.radians(theta_0)
          x 0 = 0
                        #initial x position (m)
          v0 = 300
                          # initial velocity (m/s)
          v y0 = v0 * np.sin(theta 0 rad)
          v_x0 = v0 * np.cos(theta_0_rad)
          y \ 0 = 0
          t_fin = 35
          #Time steps for the Euler method
          dt4 = 0.2
          #Define the time points to plot.
          t = np.linspace(0, t_fin, 50)
          # Use an iron cannonball (diameter 15 cm, density 7.8 g/cm^3)
          radius = (15 / 100) / 2
                                   # (m)
          volume = (4/3) * np.pi * radius ** 3 # (m ^3)
          density = 7.8 * 1000
                                  \#(kg/m^3)
          mass = volume * density
          C D = 0.5
          #Generate data using the Euler Method
          # With Drag first
          (time4, vy_Euler4da, y_Euler4da, vx_Euler4da, x_Euler4da) = EulerFreeFallDrag2DA
                                                                                        t fi
```







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```
# Code to determine range of projectile
idx = np.argwhere(np.diff(np.sign(y_Euler4d - np.zeros(len(x_Euler4d))))).flatte
x_Euler4d[idx[1]]
```

```
Out[44]: 58.656136020960666
```

From the above code, we can see that the range of the projectile is 2783.3934394341422 m.

For inspiration for writing the above code, I referred to a stack overflow page: https://stackoverflow.com/questions/28766692/intersection-of-two-graphs-in-python-find-the-x-value

Esentially, we see where the difference between y values in the Euler y position array and 0 are 0, and that is the index in the Euler simulation corresponding with the range of the projectile. Then I just use that index to find the value in the x position array corresponding with that index.

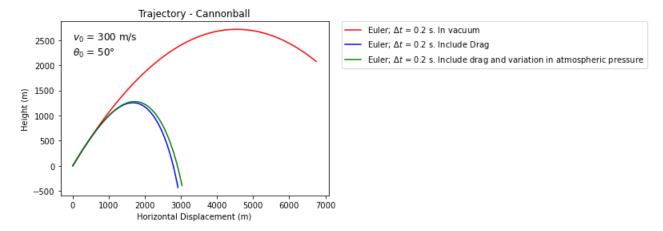
B) Do the same calculation ignoring the variation in atmospheric density (keep the density fixed at \$\rho_0\$), and yet again ignoring air resistance entirely. Plot all three trajectories for appropriate time intervals on the same graph. You will find that in this case air resistance makes a huge difference and that the variation of air resistance makes a small, but not negligible, difference.

```
#Now generate the data to plot from the analytic solution
In [45]:
          #Input parameters for model
          q = 9.8
                    #acceleration due to gravity (m/s^2)
          theta 0 = 50 # initial angle
          theta 0 rad = np.radians(theta 0)
                      #initial x position (m)
         v0 = 300
                         # initial velocity (m/s)
          v y0 = v0 * np.sin(theta 0 rad)
          v x0 = v0 * np.cos(theta 0 rad)
          y 0 = 0
          t fin = 35
          #Time steps for the Euler method
          dt4 = 0.2
          #Define the time points to plot.
          t = np.linspace(0, t fin, 50)
          # Use an iron cannonball (diameter 15 cm, density 7.8 g/cm^3)
          radius = (15 / 100) / 2
                                  # (m)
          volume = (4/3) * np.pi * radius ** 3 # (m ^3)
          density = 7.8 * 1000
                                \#(kg/m^3)
          mass = volume * density
         C D = 0.5
          #Generate data using the Euler Method
          # With Drag first
          (time4, vy Euler4d, vx Euler4d, xx Euler4d) = EulerFreeFallDrag2D(y 0,
                                                                                     CD,
          # Now in vacuum
          (time4, vy Euler4, y Euler4, vx Euler4, x Euler4) = EulerFreeFalltrag(y 0, v y0,
          # With Drag and variation in atmospheric pressure
          (time4, vy Euler4da, y Euler4da, vx Euler4da, x Euler4da) = EulerFreeFallDrag2DA
                                                                                     t fi
```

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```
# Here we have a plot of trajectory for a vacuum, drag, drag with variation in a
In [46]:
          figy, axy = plt.subplots()
                                        #Create a new figure object and its associated axe
          axy.plot(x_Euler4, y_Euler4, 'r-', label=r'Euler; $\Delta t$ = 0.2 s. In vacuum'
          axy.plot(x_Euler4d, y_Euler4d, 'b-', label=r'Euler; $\Delta t$ = 0.2 s. Include
          axy.plot(x_Euler4da, y_Euler4da, 'g-', label=r'Euler; $\Delta t$ = 0.2 s. Includ
          axy.set xlabel('Horizontal Displacement (m)')
                                                                                #Label the
          axy.set_ylabel('Height (m)')
                                                            #Label the vertical axis
          axy.set_title("Trajectory - Cannonball")
                                                                       #Title the plot
          axy.legend(bbox to anchor=(1.05, 1), loc='upper left', borderaxespad=0.)
          axy.text(5, 2500, r'$v_0$ = 300 m/s', fontsize=12)
          axy.text(5, 2200, r'$\theta_0$ = 50°', fontsize=12)
```

Out[46]: Text(5, 2200, '\$\\theta_0\$ = 50°')



For some help with formatting the legend in the above plot, I used this link:

https://matplotlib.org/3.3.3/gallery/userdemo/simple_legend01.html#sphx-glr-gallery-userdemo-simple-legend01-py

We can see that including drag causes a huge variation in the trajectory, but including variation in atmospheric pressure only causes a slight additional variation.

Physics Extension 3:

Is \$45^{\circ}\$ still the maximum range?

For a projectile launched in vacuum, the maximium range (i.e. horizontal distance when it returns to its starting height) is achieved with a launch angle of \$45^\circ\$. Is this still true when you introduce air resistance?

For a fixed launch speed and drag strength, answer this question. A **good** answer should show a plot of the range as a function of a launch angle, and numerically find where the maximum value is. A **great** answer might check this for a few different values of \$b\$, to see whether their result depdends on the strength of the drag. An **exceptional** solution might even show a plot of maximum range angle versus \$b\$, having collected data by looping over a range of values of \$b\$.

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```
v_0 - initial velocity (m/s)
theta_0 - initial angle (degrees)

returns:
range - (m)
'''
theta_0_rad = np.radians(theta_0)
r = (v_0 ** 2) * np.sin(2 * theta_0_rad) / 9.8
return r
```

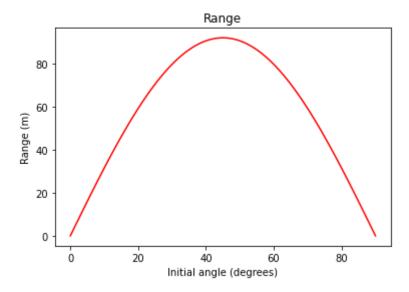
I found the above range equation from this resource: https://openstax.org/books/university-physics-volume-1/pages/4-3-projectile-motion

```
In [48]: thetas = np.linspace(0, 90, 91)
v_0 = 30  # We'll keep v_0 consistent at 30 m/s
ranges = traj_range(v_0, thetas)  # find ranges for different angles
```

First, I decided to plot range as a function of angle in a vacuum.

```
figa, axa = plt.subplots()  #Create a new figure object and its associated axe
axa.plot(thetas, ranges, 'r-')
axa.set_xlabel(r'Initial angle (degrees)')  #Label the hor
axa.set_ylabel('Range (m)')  #Label the vertical axis
axa.set_title("Range")  #Title the plot
```

Out[49]: Text(0.5, 1.0, 'Range')



As expected, the maximum range occurs when \$\theta_0\$ = 45°. Now we'll take a look at what drag does.

```
In [50]: #Now generate the data to plot from the analytic solution
    #Input parameters for model

thetas = np.linspace(0, 90, 91)  # creates an array of all integer angles from
    thetas_0_rad = np.radians(thetas) # convert angles to radians

g = 9.8  # acceleration due to gravity (m/s^2)
    x_0 = 0  # initial x position (m)
    v0 = 30  # initial velocity (m/s)
    y_0 = 0  # initial y position (m)
    t_fin = 10  # final time for simulation (s)
```

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```
#Time steps for the Euler method
dt1 = 0.05
#Define the time points to plot.
t = np.linspace(0, t_fin, 50)
# Use a baseball
mass = 0.145
                # (kg)
radius = (74 / 1000) / 2
                          # (m)
C D = 0.5
                          # for sphere
rho = 1.2
                           # atmospheric
#Generate data using the Euler Method
#### Note: I tried to do this using numpy arrays, but it just didn't work
#### So I resorted to using lists
# Drag first
data_listx = []  # Empty list for all x values
data_listy = []  # Empty list for all y values
for i in thetas_0_rad: # Loop through all angles
    v y0 = v0 * np.sin(i) # Get y velocity based on theta
                           # Get x velocity based on theta
    v x0 = v0 * np.cos(i)
    # Conduct simulations and add to lists
    # We will have two lists that contain 90 arrays for all angles
    data listx.append(EulerFreeFallDrag2D(y_0, v_y0, x_0, v_x0, -g, t_fin, dt1,
    data_listy.append(EulerFreeFallDrag2D(y_0, v_y0, x_0, v_x0, -g, t_fin, dt1,
```

```
In [51]: # Now we find index where data has y value changing from positive to negative
   indexes = []
   for angle in data_listy:
        count = 1  # we start at one because otherwise
        while count != -1:
        if np.sign(angle[count]) != np.sign(angle[count + 1]):
            indexes.append(count)
            count = -1
        else:
            count += 1
```

```
In [52]: # From what we learned above, get range for each
    ranges = []
    count = 0
    # Loop through x values for each angle
    for theta in data_listx:
        ranges.append(theta[indexes[count]]) # Use the index corresponding to t
        count += 1 # In order to cycle through our li

# Find the angle corresponding to the maximum range and print it; must add 1 sin
# we want angle
    max_angle_range = ranges.index(max(ranges)) + 1
    print(max_angle_range)
```

42

The integer value of an angle that gives the maximum range when we include drag for a baseball is 42°.

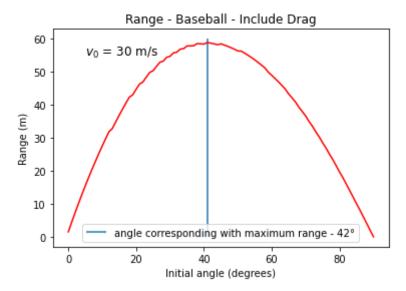
```
In [53]: # Convert to np array
```

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```
ranges_array =np.array(ranges)
```

```
figa, axa = plt.subplots()  #Create a new figure object and its associated axe
axa.plot(thetas, ranges_array, 'r-')
axa.set_xlabel(r'Initial angle (degrees)')  #Label the hor
axa.set_ylabel('Range (m)')  #Label the vertical axis
axa.vlines(41, 0, 60, label = 'angle corresponding with maximum range - 42°')
axa.set_title("Range - Baseball - Include Drag")  #Title the
axa.legend()
axa.text(5, 55, r'$v_0$ = 30 m/s', fontsize=12)
```

```
Out[54]: Text(5, 55, '$v_0$ = 30 m/s')
```



Rather than the maximum range being when $\theta = 45^{\circ}$, when we include drag for a baseball it is 42°.

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