

▼ Imports

```
1 import pandas as pd
2 import matplotlib.pyplot as plt
3 import numpy as np
```

▼ Data Processing

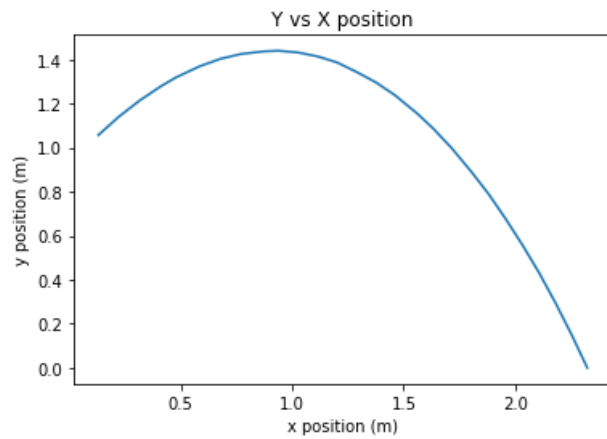
```
1 # Import Data
2 df = pd.read_csv('car_launch_data_5.txt', header=1, delim_whitespace=True)
3
4
5 # Clean data
6 # Recenter y data; Translate Y data, shift lowest y to 0
7 m = min(df['y'])
8 df['y'] = df['y'].apply(lambda x: x-m)
9
10 # Make t start at 0
11 # t_offset = min(df['x'])
12 # df['t'] = df['t'].apply(lambda x: x-t_offset)
13
14 print(df)
```

	t	x	y
0	0.000	0.132	1.057
1	0.033	0.223	1.140
2	0.067	0.317	1.214
3	0.100	0.411	1.278
4	0.133	0.480	1.319
5	0.167	0.585	1.369
6	0.200	0.678	1.403
7	0.233	0.770	1.425
8	0.267	0.872	1.437
9	0.300	0.938	1.440
10	0.333	1.027	1.432
11	0.367	1.115	1.414
12	0.400	1.201	1.387
13	0.433	1.290	1.343
14	0.467	1.376	1.296
15	0.500	1.461	1.238
16	0.533	1.564	1.152
17	0.567	1.634	1.085
18	0.600	1.715	0.998
19	0.633	1.803	0.891
20	0.667	1.881	0.789
21	0.700	1.957	0.677
22	0.733	2.033	0.557
23	0.767	2.108	0.430
24	0.800	2.181	0.294
25	0.833	2.253	0.150
26	0.867	2.323	0.000

▼ Initial Plots

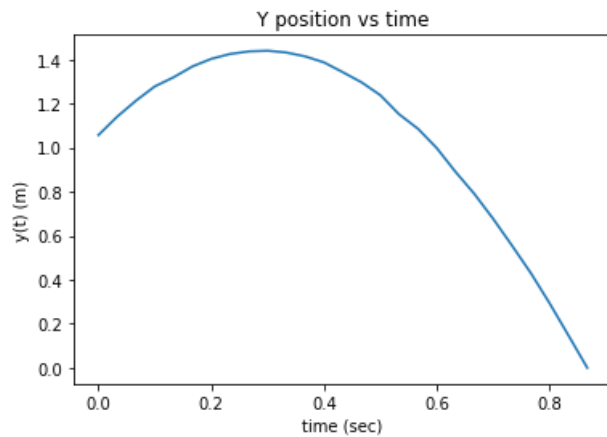
```
1 # Initial visualization of the data
2 plt.plot(df['x'], df['y'])
3 plt.xlabel("x position (m)")
```

```
4 plt.ylabel("y position (m)")
5 plt.plot(df['x'], df['y'])
6 plt.text(0.5, 1.0, 'Y vs X position')
```



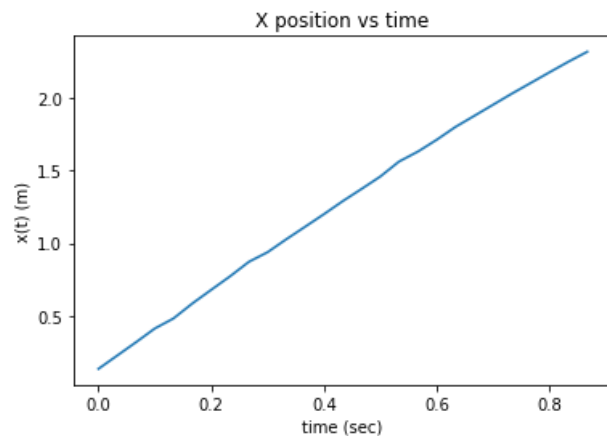
```
1 plt.plot(df['t'], df['y'])
2 plt.xlabel("time (sec)")
3 plt.ylabel("y(t) (m)")
4 plt.title("Y position vs time")

Text(0.5, 1.0, 'Y position vs time')
```



```
1 plt.plot(df['t'], df['x'])
2 plt.xlabel("time (sec)")
3 plt.ylabel("x(t) (m)")
4 plt.title("X position vs time")

Text(0.5, 1.0, 'X position vs time')
```



▼ Simulation Helpers

```
1 # Constants
2 g = 9.8 # gravity (m/s^2)
3
4 t0 = 0 # initial Time (s)
5 dt = .033 # time resolution
6
7 tf = 0.867 # total time to simulate
8 nsteps = int(tf/dt) + 1 # number of time steps

1 # Calculate initial conditions based on two given (early) coordinate points
2 def initial_conditions(coord1, coord2):
3     t = coord2[0] - coord1[0]
4     x = coord2[1] - coord1[1]
5     y = coord2[2] - coord1[2]
6     return [x/t, y/t, coord1[1], coord1[2]]
7
8
```

▼ Euler Method

```
1 # Initial Conditions
2
3 # Getting initial velocity components
4 i_coord = initial_conditions(df.loc[0], df.loc[1])
5
6 print("X vel:", i_coord[0], "m/s")
7 print("Y vel:", i_coord[1], "m/s")
8 print("X pos:", i_coord[2], "m")
9 print("Y pos:", i_coord[3], "m")
10
11 y0 = i_coord[3] # initial y (m)
12 x0 = i_coord[2] # initial x (m)
13
14 vy0 = i_coord[1] # initial y velocity (m/s)
15 vx0 = i_coord[0] # initial x velocity (m/s)

X vel: 2.7575757575757573 m/s
Y vel: 2.515151515151514 m/s
X pos: 0.132 m
Y pos: 1.057 m

1 # Euler method setting up - time evolution
2 t = np.linspace(t0, tf, nsteps)
3
4 vx = np.zeros([nsteps])
5 vy = np.zeros([nsteps])
6 y = np.zeros([nsteps])
7 x = np.zeros([nsteps])
8
9 y[0] = y0
10 x[0] = x0
11 vy[0] = vy0
12 vx[0] = vx0
13
14 ## time evolving
```

```

15 for i in range(0, nsteps-1):
16     y[i+1] = y[i] + dt * vy[i]
17     vy[i+1] = vy[i] - dt * g
18
19     x[i+1] = x[i] + dt * vx[i]
20     vx[i+1] = vx[i]
21

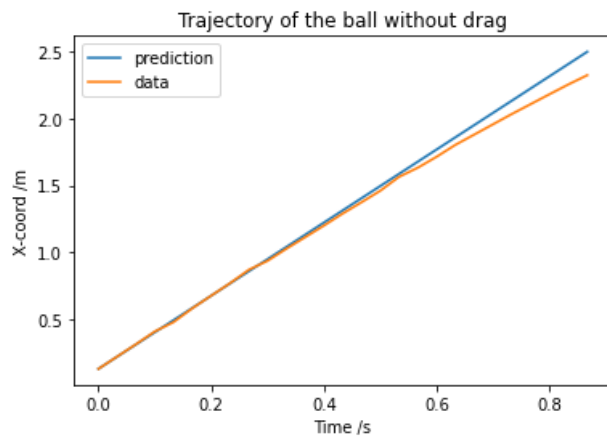
```

▼ Prediction Plots

```

1 plt.plot(t, x, label="prediction")
2 plt.plot(df['t'], df['x'], label="data")
3 plt.plot()
4 plt.title('Trajectory of the ball without drag')
5 plt.xlabel('Time /s')
6 plt.ylabel('X-coord /m')
7 plt.legend()
8 plt.show()

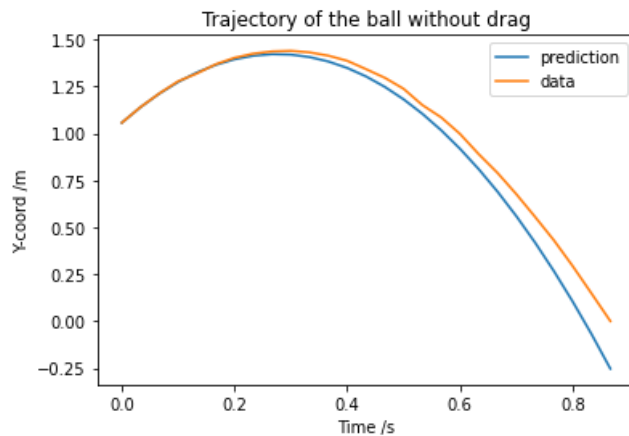
```



```

1 plt.plot(t, y, label="prediction")
2 plt.plot(df['t'], df['y'], label = "data")
3 plt.title('Trajectory of the ball without drag')
4 plt.xlabel('Time /s')
5 plt.ylabel('Y-coord /m')
6 plt.legend()
7 plt.show()

```



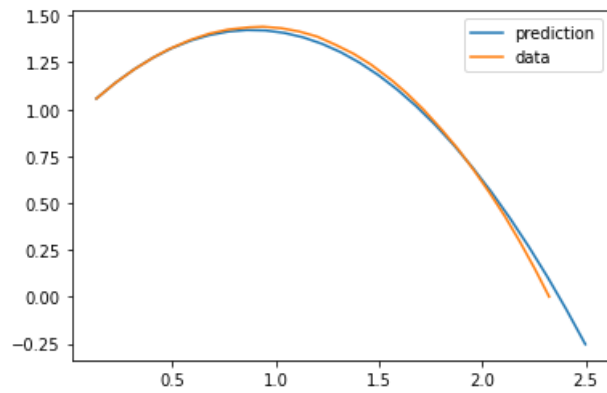
```

1 plt.plot(x,y, label="prediction")
2 plt.plot(df['x'], df['y'], label="data")

```

```
3 plt.legend()
```

```
<matplotlib.legend.Legend at 0x7f608ced79a0>
```



▼ Error

```
1 def err(actual, expected):
2     if not len(actual) == len(expected):
3         return "number of data points do not match"
4     else:
5         return [round(abs(actual[i] - expected[i]), 5) if actual[i] > 0 and expected[i] > 0 else 0 for
```

```
1 x_data = df['x'].values
2 x_pred = x
3
4 y_data = df['y'].values
5 y_pred = y
6
7 print(len(x_data), len(x_pred))
8 print(x_data[-1], x_pred[-1])
9
10 print(len(y_data), len(y_pred))
11 print(y_data[-1], y_pred[-1])
```

```
27 27
2.323 2.4980000000000007
27 27
0.0 -0.2534650000000009
```

```
1 x_err = err(x_data, x_pred)
2 y_err = err(y_data, y_pred)
3 print(y_err)
```

```
[0.0, 0.0, 0.00167, 0.00402, 0.00597, 0.00372, 0.00808, 0.01112, 0.01482, 0.0202, 0.02525, 0.03097, 0.
```

▼ Error Plots

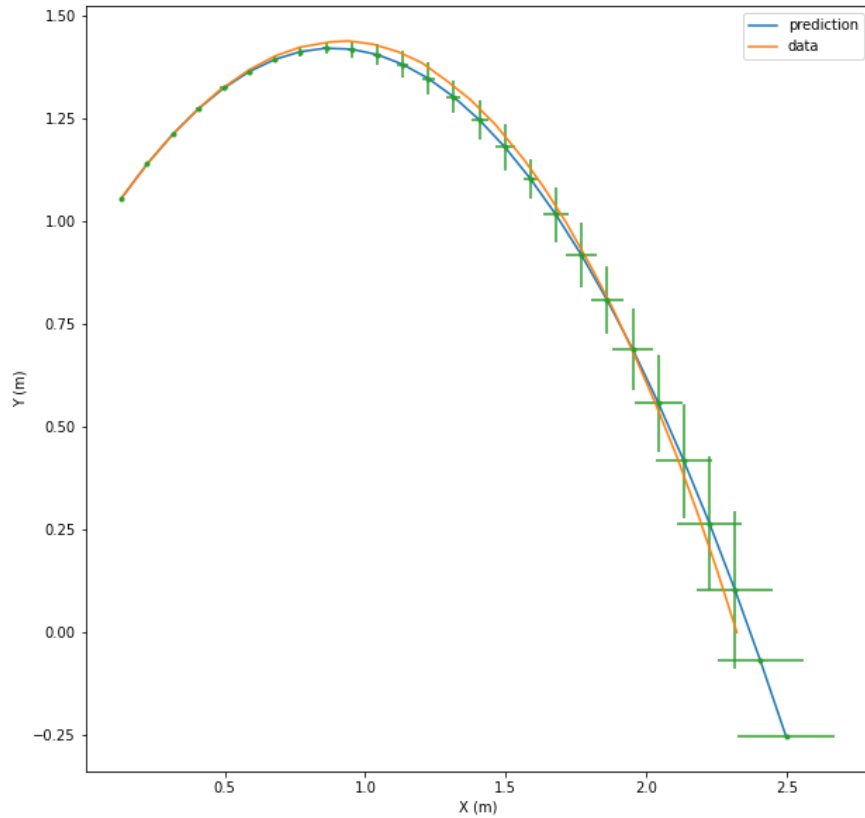
```
1 fig = plt.figure(figsize=(10, 10))
2 plt.plot(x,y, label="prediction")
3 plt.plot(df['x'], df['y'], label="data")
4 plt.errorbar(x, y,
5             xerr = x_err,
6             yerr = y_err,
7             fmt='.')
8 plt.xlabel('X (m)')
```

```
9 plt.ylabel('Y (m)')
```

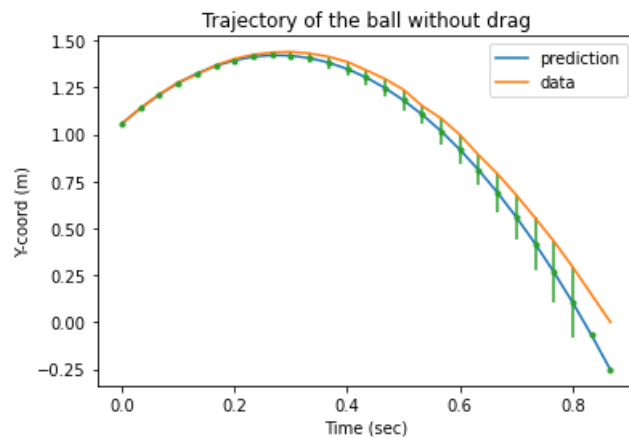
```
10
```

```
11 plt.legend()
```

```
<matplotlib.legend.Legend at 0x7f608ce46d00>
```



```
1 plt.plot(t, y, label="prediction")
2 plt.plot(df['t'], df['y'], label = "data")
3 plt.title('Trajectory of the ball without drag')
4 plt.xlabel('Time (sec)')
5 plt.ylabel('Y-coord (m)')
6 plt.legend()
7 plt.errorbar(t, y, yerr=y_err, fmt='.')
8 plt.show()
```

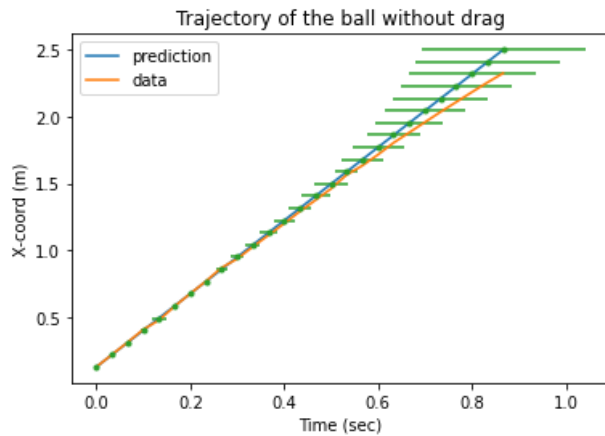


```
1 plt.plot(t, x, label="prediction")
2 plt.plot(df['t'], df['x'], label="data")
```

```

3 plt.plot()
4 plt.title('Trajectory of the ball without drag')
5 plt.xlabel('Time (sec)')
6 plt.ylabel('X-coord (m)')
7 plt.errorbar(t, x, xerr=x_err, fmt='.')
8 plt.legend()
9 plt.show()

```



▼ Bashforth-Adams

The bashforth-adams algorithm is apparently an improvement on the Euler algorithm so we will make an attempt at comparing them here

For a given position component, we have:

$$y_{n+1} = y_n + \frac{3}{2}\delta t f(t_n, y_n) - \frac{1}{2}\delta t f(t_{n-1}, y_{n-1})$$

It seems that most people are defining the function f as the derivative of the position component.

```

1 # BA Initial Conditions
2
3 i_coord0 = initial_conditions(df.loc[0], df.loc[1])
4 i_coord1 = initial_conditions(df.loc[1], df.loc[2])
5
6 y0 = i_coord0[3]      # initial y (m)
7 x0 = i_coord0[2]      # initial x (m)
8 vy0 = i_coord0[1]     # iniital y velocity (m/s)
9 vx0 = i_coord0[0]     # initial x velocity (m/s)
10
11 y1 = i_coord1[3]      # initial +1 y (m)
12 x1 = i_coord1[2]     # initial +1 x (m)
13 vy1 = i_coord1[1]    # iniital +1 y velocity (m/s)
14 vx1 = i_coord1[0]    # initial +1 x velocity (m/s)
15
16 print(i_coord0,i_coord1)
17
[2.7575757575757573, 2.515151515151514, 0.132, 1.057] [2.764705882352941, 2.1764705882352957, 0.223, 1

```

```

1 # Bashforth-Adams method setting up - time evolution
2 t = np.linspace(t0, tf, nsteps)
3

```

```

4 vy_ba = np.zeros([nsteps])
5 vx_ba = np.zeros([nsteps])
6 y_ba = np.zeros([nsteps])
7 x_ba = np.zeros([nsteps])
8
9 # Init time 0 and 1
10 y_ba[0] = y0
11 x_ba[0] = x0
12 y_ba[1] = y1
13 x_ba[1] = x1
14 vy_ba[0] = vy0
15 vx_ba[0] = vx0
16 vy_ba[1] = vy1
17 vx_ba[1] = vx1

1 # Time evolving
2 for i in range(0, nsteps-2):
3     y_ba[i+2] = y_ba[i+1] + dt * (3/2*vy_ba[i+1] - 1/2*vy_ba[i])
4     vy_ba[i+2] = vy_ba[i+1] - dt * g
5
6     x_ba[i+2] = x_ba[i+1] + dt * (3/2*vx_ba[i+1] - 1/2*vx_ba[i])
7     vx_ba[i+2] = vx_ba[i]
8

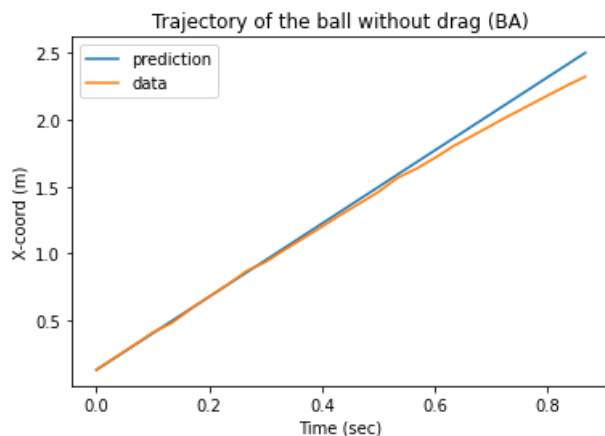
```

▼ Prediction Plots

```

1 plt.plot(t, x_ba, label="prediction")
2 plt.plot(df['t'], df['x'], label="data")
3 plt.plot()
4 plt.title('Trajectory of the ball without drag (BA)')
5 plt.xlabel('Time (sec)')
6 plt.ylabel('X-coord (m)')
7 plt.legend()
8 plt.show()

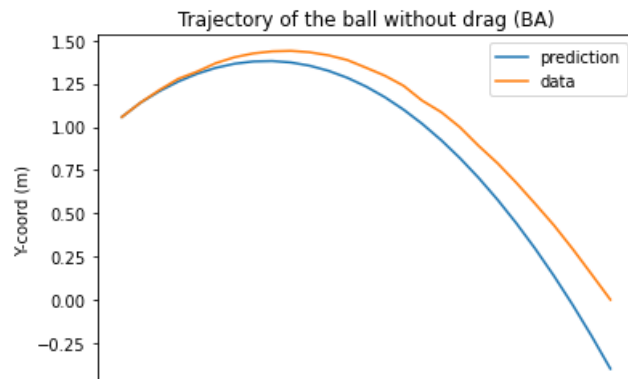
```



```

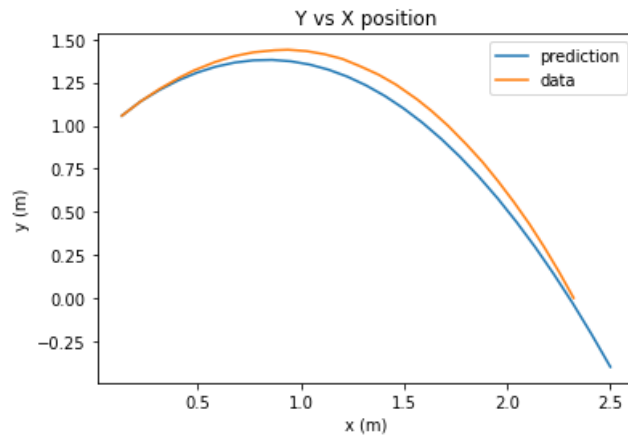
1 plt.plot(t, y_ba, label="prediction")
2 plt.plot(df['t'], df['y'], label = "data")
3 plt.title('Trajectory of the ball without drag (BA)')
4 plt.xlabel('Time (sec)')
5 plt.ylabel('Y-coord (m)')
6 plt.legend()
7 plt.show()

```

```
1 plt.plot(x_ba,y_ba, label="prediction")
2 plt.plot(df['x'], df['y'], label="data")
3 plt.xlabel("x (m)")
4 plt.ylabel("y (m)")
5 plt.title("Y vs X position")
6 plt.legend()
```

<matplotlib.legend.Legend at 0x7f608cea7640>



1