Homework 7 writeups

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Section: AMATH 301 B

Problem 1

```
In []:
    import numpy as np
    import time
    import scipy.integrate

s = 77.27
w = 0.161
q = 1

y1_prime = lambda y1, y2, y3: s*(y2 - y1*y2 + y1 - q*y1**2)
y2_prime = lambda y1, y2, y3: 1/s* (-y2 - y1 * y2 + y3)
y3_prime = lambda y1, y2, y3: w * (y1 - y3)

y1_0 = 1
y2_0 = 2
y3_0 = 3
y0 = np.array([y1_0, y2_0, y3_0])
```

Part a - Timing RK45 and BDF

```
In [ ]: # Logarithmically spaced points for q
        qspan = np.logspace(0, -5, 10)
        # Time values for each q value
        time rk45 = np.zeros(len(qspan))
        time bdf = np.zeros(len(qspan))
        # RK45 timer for each q
        for index, q in enumerate(qspan):
             y1_prime = lambda y1, y2, y3: s*(y2 - y1*y2 + y1 - q*y1**2)
             odefun = lambda t, y: np.array([y1_prime(y[0], y[1], y[2]), y2_prime(y[0],
             time 0 = time.perf counter()
             sol = scipy.integrate.solve ivp(odefun, [0, 30], y0)
             time i = time.perf counter() - time 0
             time rk45[index] = time i
        print("RK45 time values at each q:", time rk45)
        # BDF timer for each q
        for index, q in enumerate(qspan):
             y1 \text{ prime} = 1 \text{ambda} \ y1, \ y2, \ y3: \ s*(y2 - y1*y2 + y1 - q*y1**2)
             odefun = lambda t, y: np.array([y1 prime(y[0], y[1], y[2]), y2 prime(y[0],
```

```
time_0 = time.perf_counter()
    sol = scipy.integrate.solve_ivp(odefun, [0, 30], y0, method="BDF")
    time_i = time.perf_counter() - time_0
    time_bdf[index] = time_i

print("BDF time values at each q:", time_bdf)

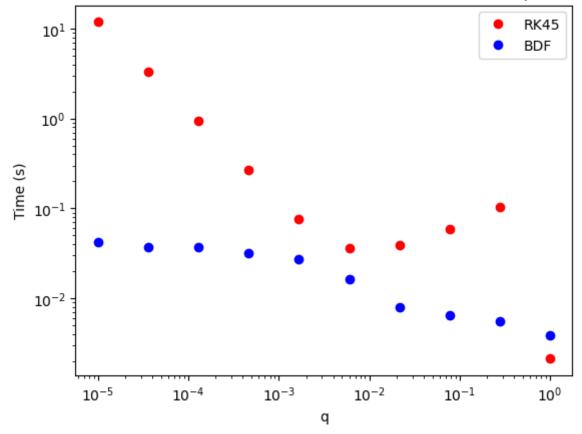
RK45 time values at each q: [2.12387500e-03 1.03477334e-01 5.90362920e-02 3.91
462090e-02
    3.63556670e-02 7.67165420e-02 2.68152083e-01 9.45407500e-01
    3.34381533e+00 1.19257095e+01]
BDF time values at each q: [0.00385038 0.00551083 0.00638638 0.00797275 0.0164
6154 0.02717754
    0.03180287 0.03739267 0.03702042 0.0419315 ]
```

Part b - Create a loglog plot

```
In []: plt.figure()
   plt.title("Time taken for RK45 and BDF methods at each q")
   plt.loglog(qspan, time_rk45, 'or', label="RK45")
   plt.loglog(qspan, time_bdf, 'ob', label="BDF")
   plt.xlabel("q")
   plt.ylabel("Time (s)")
   plt.legend()
```

Out[]: <matplotlib.legend.Legend at 0x7f9afaa53760>

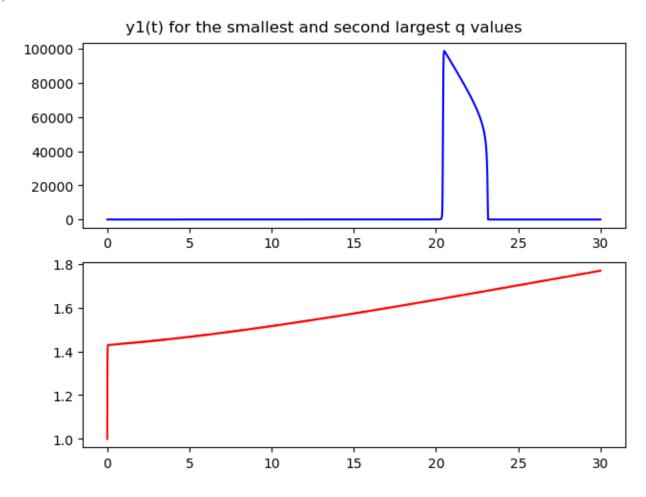
Time taken for RK45 and BDF methods at each q



Part c - Create a 2 panel figure.

```
smallest_q = qspan[-1]
In [ ]:
        second_largest_q = qspan[1]
        y1_prime = lambda y1, y2, y3: s*(y2 - y1*y2 + y1 - smallest_q*y1**2)
        odefun = lambda t, y: np.array([y1_prime(y[0], y[1], y[2]), y2_prime(y[0], y[1]
        sol = scipy.integrate.solve_ivp(odefun, [0, 30], y0)
        y1\_smallest\_q = sol.y[0, :]
        y1_prime = lambda y1, y2, y3: s*(y2 - y1*y2 + y1 - second_largest_q*y1**2)
        odefun = lambda t, y: np.array([y1_prime(y[0], y[1], y[2]), y2_prime(y[0], y[1]
        sol2 = scipy.integrate.solve_ivp(odefun, [0, 30], y0)
        y1_second_largest_q = sol2.y[0, :]
        x = np.linspace(-1, 1, 100)
        fig, ax = plt.subplots(2, 1, constrained layout=True)
        fig.suptitle("y1(t) for the smallest and second largest q values")
        ax[0].plot(sol.t, y1 smallest q, 'b')
        ax[1].plot(sol2.t, y1_second_largest_q, 'r')
```

Out[]: [<matplotlib.lines.Line2D at 0x7f9ad8ea2af0>]



Part d - Comment on what we see.

Part (i) - Compare the two methods

Generally, the BDF method is better as it follows a linear trend and takes less time to compute than RK45. The RK45 steadily decreases the time taken as q value increases, but

increases time again before suddenly dropping below the BDF for the largest q value. At the largest q value, the RK45 method turns out to be better.

Part (ii) - Time as q increases

The time taken for RK45 and decreasing q values forms to a quadratic trend with the exception of q = 1. When we see the data point at q = 1 as an outlier, we can see the red dots in the graph forming a parabola.

Part (iii) - What makes calculation slower for RK45?

When we use an explicit method like RK45, very small step sizes are needed to solve the problem, minimizing the error. At smaller q values, these step sizes become prohibitively small, which is why they lead to a higher runtime.

Part (iv) - Is this equation stiff? How do we know?

Yes, the equation is still for small q as the vector field rapidly changes and goes all the way vertically straight up from being horizontal earlier around the time = 20 secs.

Problem 2

Part a - Ratio of points, RK45 to BDF.

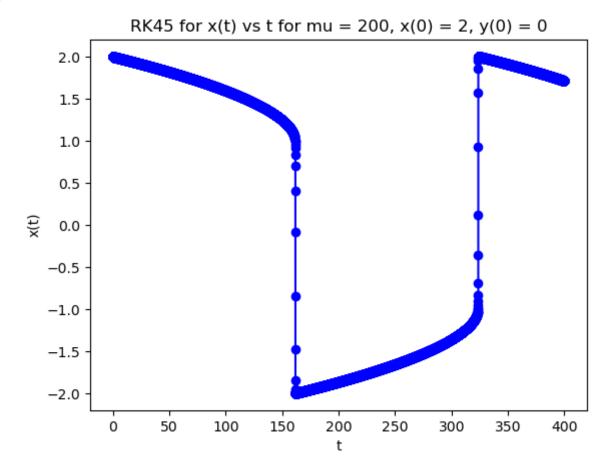
```
In []: mu = 200
        x0 = 2
        y0 = 0
        z0 = np.array([x0, y0])
        dxdt = lambda x, y: y
        dydt = lambda x, y: mu * (1 - x**2) * y - x
        ode = lambda t, z: np.array([dxdt(z[0], z[1]), dydt(z[0], z[1])])
        sol = scipy.integrate.solve ivp(ode, [0, 400], z0)
        sol bdf = scipy.integrate.solve ivp(ode, [0, 400], z0, method="BDF")
        p rk45 = len(sol.t)
        p bdf = len(sol bdf.t)
        print("RK45 number of points:", p rk45)
        print("BDF number of points:", p bdf)
        ratio = p rk45 / p bdf
        print("Ratio of RK45 points to BDF points:", ratio)
        difference = p rk45 - p bdf
        print("Difference between RK45 points and BDF points:", difference)
        RK45 number of points: 46401
        BDF number of points: 298
        Ratio of RK45 points to BDF points: 155.70805369127515
```

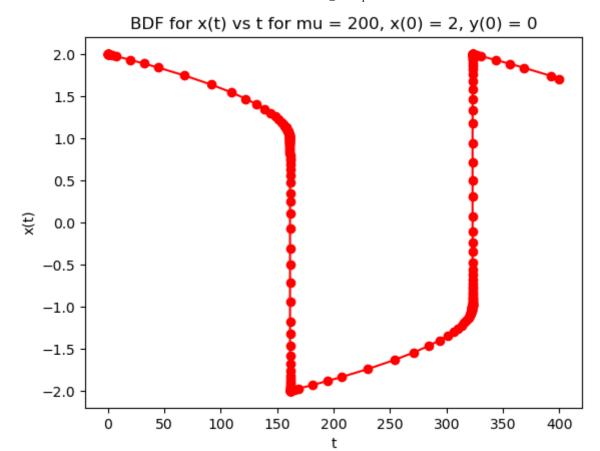
Part b - Plot solution, x(t)

Difference between RK45 points and BDF points: 46103

```
In []: mu = 200
        dxdt = lambda x, y: y
        dydt = lambda x, y: mu * (1 - x**2) * y - x
        ode = lambda t, z: np.array([dxdt(z[0], z[1]), dydt(z[0], z[1])])
        x0 = 2
        y0 = 0
        z0 = np.array([x0, y0])
        sol = scipy.integrate.solve_ivp(ode, [0, 400], z0)
        A8 = sol.y[0, :]
        sol_bdf = scipy.integrate.solve_ivp(ode, [0, 400], z0, method="BDF")
        A9 = sol_bdf.y[0, :]
        plt.figure()
        plt.plot(sol.t, A8, '-ob')
        plt.xlabel("t")
        plt.ylabel("x(t)")
        plt.title("RK45 for x(t) vs t for mu = 200, x(0) = 2, y(0) = 0")
        plt.figure()
        plt.plot(sol_bdf.t, A9, '-or')
        plt.xlabel("t")
        plt.ylabel("x(t)")
        plt.title("BDF for x(t) vs t for mu = 200, x(0) = 2, y(0) = 0")
```

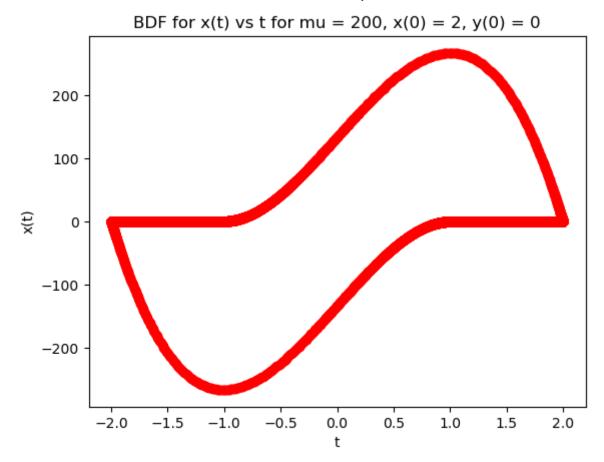
Out[]: Text(0.5, 1.0, 'BDF for x(t) vs t for mu = 200, x(0) = 2, y(0) = 0')





Part c - Plot x(t) vs. y(t) (y(t) on vertical axis)

```
In []: sol_bdf = scipy.integrate.solve_ivp(ode, [0, 400], z0, method="BDF", t_eval=np.
    plt.figure()
    plt.plot(sol_bdf.y[0, :], sol_bdf.y[1, :], '-or')
    plt.xlabel("t")
    plt.ylabel("x(t)")
    plt.title("BDF for x(t) vs t for mu = 200, x(0) = 2, y(0) = 0")
Out[]: Text(0.5, 1.0, 'BDF for x(t) vs t for mu = 200, x(0) = 2, y(0) = 0')
```



Part d - Discussion

This ODE is stiff because the RK45 method requires a significantly higher number of points than the BDF method to obtain a smooth curve. Also, the shape for the RK45 method is much steeper and changes more rapidly than the shape of the BDF method. As per my understanding of vector fields, this ODE will be stiff because the vector field is rapidly changing in a small region. This also implies that solving this problem will require a very small step size to accurately capture the trend and minimize the error when using explicit methods like the RK45 or the BDF.