

CEE 6736: HW #1

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[ ]: # !sudo apt install cm-super dvipng texlive-latex-extra texlive-latex-recommended
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Problem 1

Please show all work (i.e. include source code, and include thoughtful, analytical discussions):

- (1) Pick a population (e.g. New York City in March of 2020, or a city in the state of Florida in August 2021, etc.) and find the COVID data from the relevant health department. Assume defensible values for X and I in your populations. Use these data to test our COVID transmission model. Discuss the results.

Solution:

Recall the COVID transmission model discussed in class. First defining: $N_i \equiv$ Number of infections on a given day i ; $X \equiv$ Expected numbers of daily contacts per infected person; $I \equiv$ Probability of infection for each contact. The infection for the next day writes:

$$N_{i+1} = (IX + 1)N_i$$

This model can be further expressed as, at a certain day i , the infectious number follows:

$$N_i = (1 + IX)^i N_0$$

where N_0 denotes the initial values at $t = 0$. Promoting this to a continuous model:

$$\frac{dN}{dt} = (IX) N$$

where IX is identified as the \mathcal{R} factor.

This is a special case that follows the linear ODE form when the bias term equals zero. Hence, the solution ought to follow the basic form: $N = Ce^{\mathcal{R}t}$, where C is a constant. Based on the data fitting, it seems like a good fit for \mathcal{R} is 0.05. Assuming an (reasonable) I of 0.001, the X is hence 50.

We now apply the data of Tompkins county (from July 2021 to Oct 2021) to test the model, starting from loading the data from .txt file:

```
[ ]: import numpy as np
import matplotlib.pyplot as plt
import matplotlib as mpl
```

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from scipy.integrate import odeint

t_discretized = np.linspace(0,90,90)
N_real = np.exp(.05*t_discretized)
N_real_2 = np.exp(.075*t_discretized[0:60])

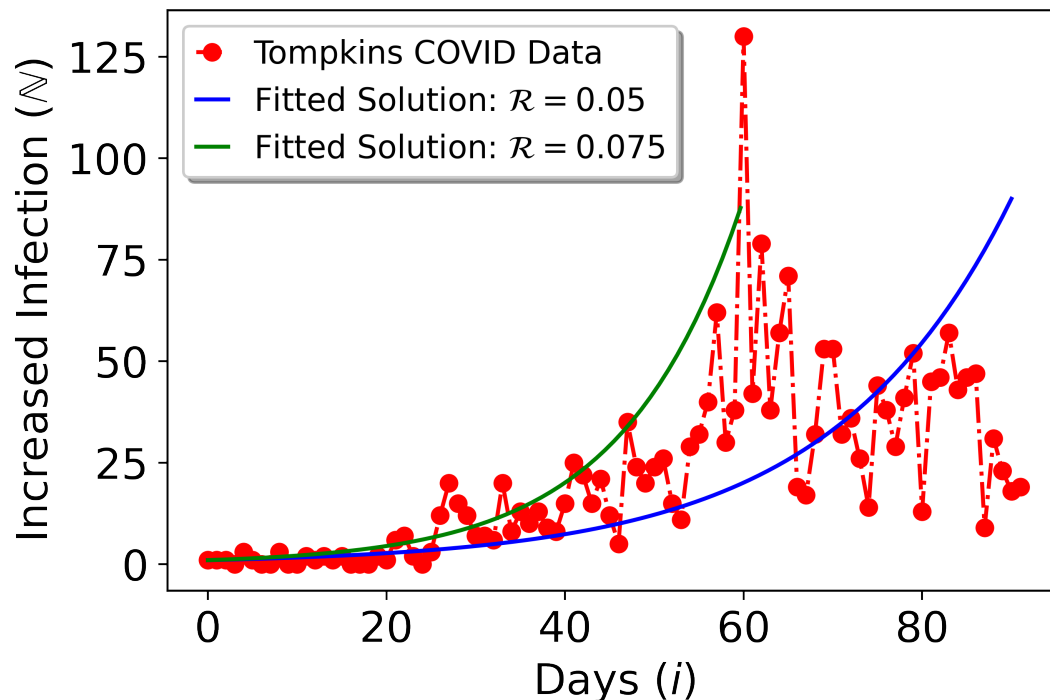
tompkins_data = np.loadtxt("tompkins_data.txt")
num_infec = np.flip(tompkins_data.T[2])
daily_text = np.flip(tompkins_data.T[1]);daily_text_identity = daily_text/np.
    ↳mean(daily_text)
total_infec = np.flip(tompkins_data.T[3]);total_infec_identity = total_infec/np.
    ↳mean(total_infec)

plt.plot(num_infec,'ro-.',label='Tompkins COVID Data')
plt.plot(t_discretized,N_real,'b-',label='Fitted Solution:  $\mathcal{R} = 0.05$ ')
plt.plot(t_discretized[0:60],N_real_2,'g-',label='Fitted Solution:  $\mathcal{R}_{\text{L}} \rightarrow 0.075$ ')
plt.xlabel("Days ( $i$ )")
plt.ylabel("Increased Infection ( $\mathbb{N}$ )")

plt.legend(shadow=True, handlelength=1, fontsize=12)

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[]: <matplotlib.legend.Legend at 0x7f445026c390>



Problem 2

Determine if the equation of a line ($y = mx + b$) satisfies the *Axioms of Linearity*. Discuss the results.

Solution:

Let $f(x) = y = mx + b$, recall the definition of axioms of linearity,

$$cf(x_1 + x_2) = cf(x_1) + cf(x_2)$$

Substitute it into $f(x)$ we have:

$$\text{LHS} = c(m(x_1 + x_2) + b) = cmx_1 + cmx_2 + cb \quad \text{RHS} = c(mx_1 + b) + c(mx_2 + b) = cmx_1 + cmx_2 + 2cb$$

It is observed that $\text{LHS} \neq \text{RHS}$. Hence the line does not satisfy the axiom of linearity. What's more, it can be determined if the bias term b is eliminated, the line hence obeys the axiom of linearity.

Problem 3

Use the ODE solution method discussed in class (i.e. $\frac{d}{dt} \ln|\cdot|$) to exactly solve the drone ODE from HW1 using mass, $m = 10\text{kg}$ and the drag coefficient, $\gamma = 2\text{kg/sec}$ and the initial condition: $v(0) = 0$. Please plot the exact solution within the vector plot from HW1 and discuss.

Solution:

Recall the governing equation of the drone free fall, the velocity writes:

$$\frac{dv(t)}{dt} = -g - \frac{\gamma}{m}v$$

where

$$\begin{aligned}\gamma &\equiv \text{drag of coefficient} \\ \gamma v &\equiv \text{wind of resistance} \\ m &\equiv \text{mass of quad copter}\end{aligned}$$

written in the standard form

$$\frac{dv(t)}{dt} + \frac{\gamma}{m}v = -g$$

applying the integration factor $\mu(t) = e^{\frac{\gamma}{m}t}$, the equation can be rewrite in the form given the fact that $\frac{d\mu(t)}{dt} = \frac{\gamma}{m}\mu$:

$$\frac{d}{dt} \left[e^{\frac{\gamma}{m}t} v(t) \right] = -e^{\frac{\gamma}{m}t} g$$

the solution further writes:

$$v(t) = e^{-\frac{\gamma}{m}t} \int_{t_0}^t e^{\frac{\gamma}{m}s} (-g) \cdot ds + C e^{-\frac{\gamma}{m}t}$$

skipping the detailed derivation, we directly land in the final solution form of $v(t)$:

$$v(t) = \frac{1}{\mu(t)} \left[\int_{t_0}^t \mu(s) (-g) \cdot ds + C \right]$$

To determine the constant C , we substitute the given initial values: $v(0) = 0 \rightarrow C = 5g$. We, therefore, obtain the solution:

$$v(t) = \frac{1}{e^{\frac{\gamma}{m}t}} \left[-g \int_{t_0}^t e^{\frac{\gamma}{m}s} \cdot ds + 5g \right] = \frac{1}{e^{\frac{\gamma}{m}t}} \left[-g \frac{m}{\gamma} e^{\frac{\gamma}{m}t} + 5g \right] = \frac{1}{e^{0.2t}} [-5ge^{0.2t} + 5g]$$

We, therefore, plot the graph:

```
[ ]: import numpy as np
import matplotlib.pyplot as plt
import matplotlib as mpl
from scipy.integrate import odeint

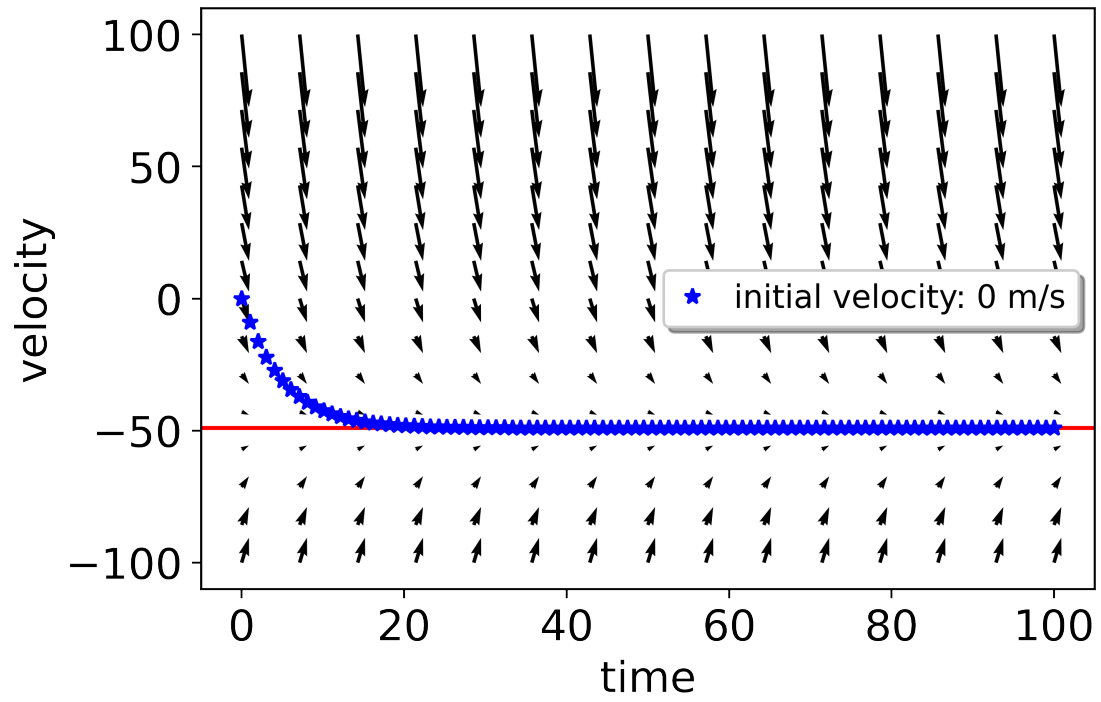
t,v = np.meshgrid(np.linspace(0,100,15),np.linspace(-100,100,15)) # generate mesh

# define parameters
gamma = 2
m = 10

# define equations
drag = (gamma/m) * v
g = 9.8
RHS = - g - drag

sol_exact = (1/np.exp(0.2*t_dis)) * (-5*g*np.exp(0.2*t_dis) + 5*g)
# print(np.shape(sol_exact))
# solving the equation(s) using odeint
t_dis = np.linspace(0,100,100)

# # plotting
plt.quiver(t,v,3,RHS)
plt.axhline(y=-49, color='r', linestyle='-')
plt.plot(t_dis, sol_exact, 'b*', label="initial velocity: 0 m/s")
plt.xlabel('time')
mpl.rcParams.update({'font.size': 16})
plt.ylabel('velocity')
plt.legend(shadow=True, handlelength=1, fontsize=12)
plt.rcParams['figure.dpi'] = 500
plt.show()
plt.figure(figsize=(5, 3))
```



[]: <Figure size 2500x1500 with 0 Axes>

<Figure size 2500x1500 with 0 Axes>