

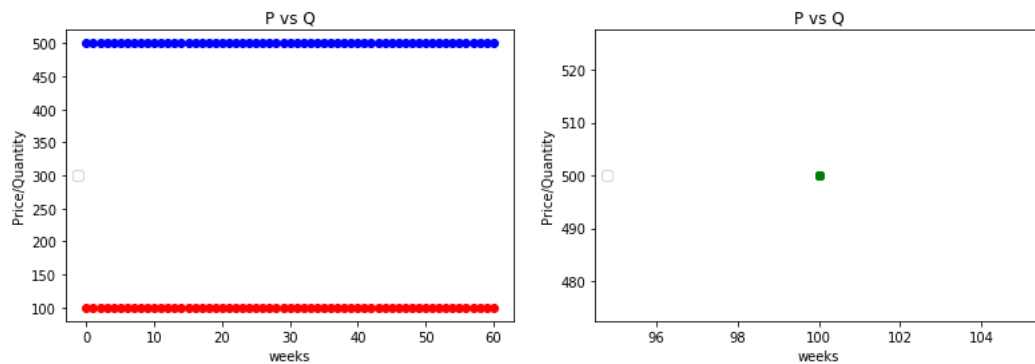
Foundations of Applied Math  
HW #4 Due Friday, Oct. 9

**Reminder** You need to turn in a .zipped folder that contains your .tex file, your image files, your python file and the tex file must compile. Rename the .tex file: HW4\_YourLastName.tex and call the folder which you will compress: HW4\_YourLastName

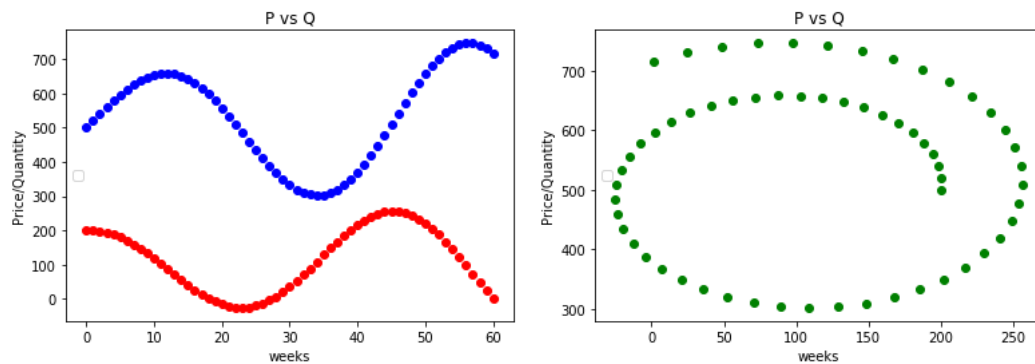
Please make sure you learn the examples in 1.4 before you do this HW.

- 1.4) Do #6b this time using python. For each of the four cases please make sure you generate a plots like the ones we generated in HW 3 using Excel. (On Friday I will show you how to generate plots using arrays in python and how to generate a legend). Include your python file so that I can run it and paste your 8 graphs here. (Each case should have 2 graphs.)

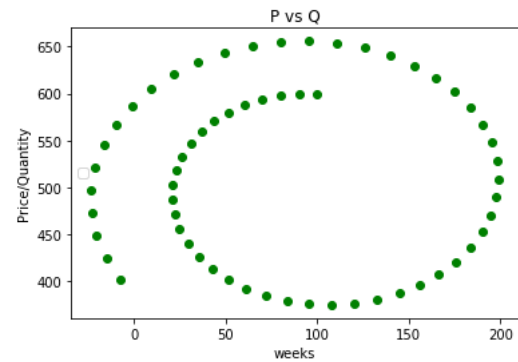
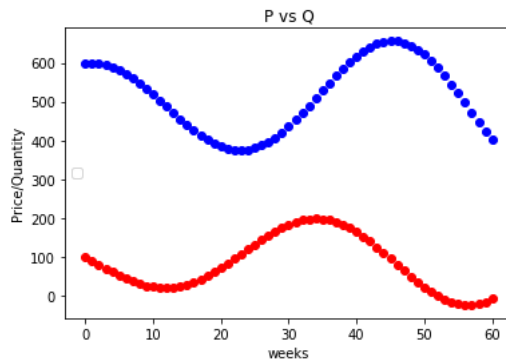
(a) Case A:



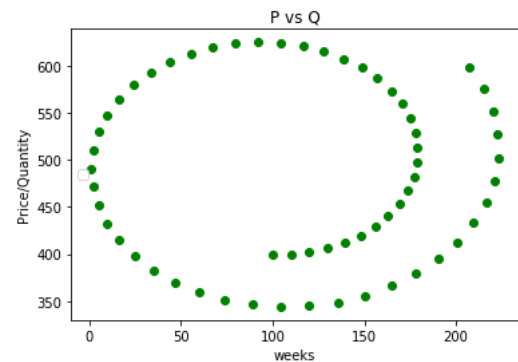
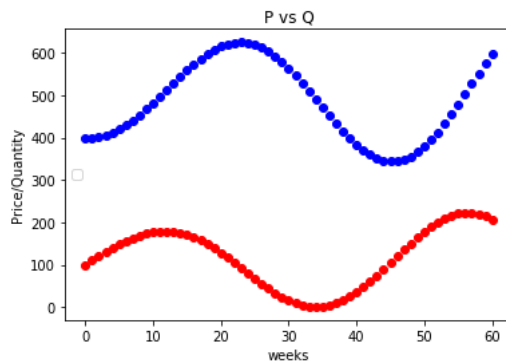
(b) Case B:



(c) Case C:



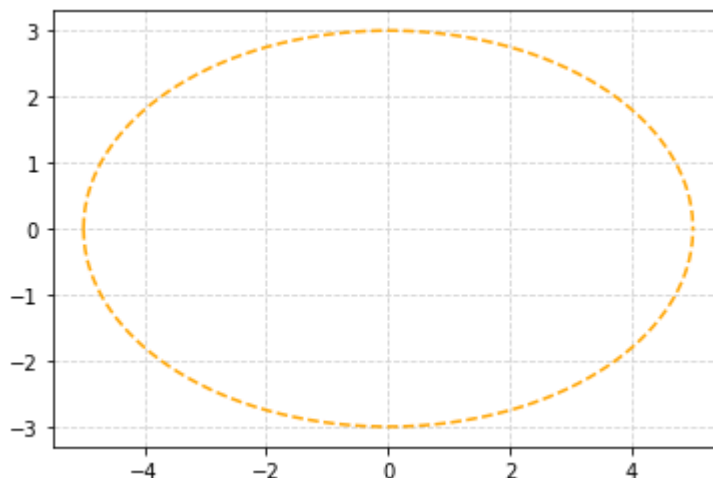
(d) Case D:



2. 1.4) Using arrays plot  $\frac{x^2}{25} + \frac{y^2}{9} = 1$  using python. Make sure your plot is one color curve. Have your last two lines be:

```
plt.grid(color='lightgray',linestyle='--')
plt.show()
```

Include your python file so that I can run it and paste your graph here. Hint- you will need to solve for  $y$ .



Python file is included in the zipped folder.

3. Consider the following model for two species.

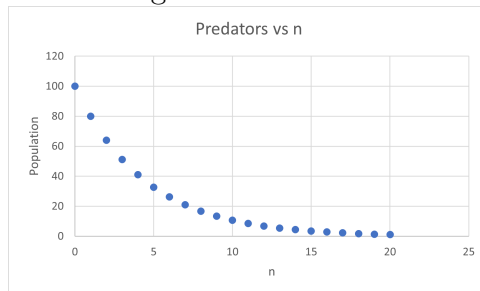
$$\begin{aligned}x_{n+1} &= 0.8x_n + 0.001x_ny_n \\y_{n+1} &= y_n + 0.0001y_n(3000 - y_n) - 0.004x_ny_n\end{aligned}$$

- a) This model is a predator-prey model. Explain why. Also state which species is the predator.

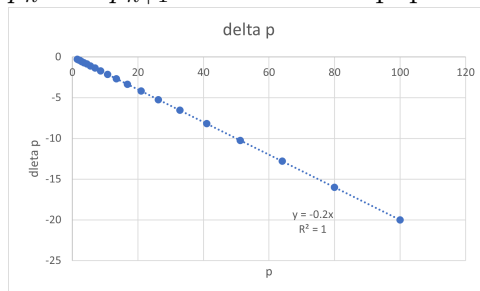
This model shows that the change in  $x_{n+1}$  increases depending on how many interactions there are between  $x$  and  $y$ . However,  $y$  decreases depending on how many interactions there are between  $x$  and  $y$ . Therefore,  $x$  is the predator and  $y$  is the prey.

- b) According to the model, what happens to the population of the predator if there is no prey present? Support your answer with a graph of predators vs.  $n$ .

The population of the predators will only decrease if there are no prey. Specifically, only 80% of the predator population will survive in the next iteration if there are no prey. Looking at the graph of  $n$  vs  $p$ , we see that the population of the predators do indeed go down:



Plotting the  $\Delta p_n$  vs  $p$  graph, we also see that the change is indeed  $-20\%$  between  $p_n$  and  $p_{n+1}$  or 80% of the population.



- c) According to the model, will the population of the prey experience unlimited population growth if there are no predators present?

The population of the prey will not experience unlimited population growth because of the second term. This logistic term will force the maximum population to be 3000.

4. 2.2, #3

By Hooke's law:

$$\begin{aligned}F &= kS \\14 &= k(.37) \\k &= \frac{14}{.37} \\k &\approx 37.83784\end{aligned}$$

Therefore, for 9-lb on the same spring:

$$\begin{aligned}F &= kS \\S &= \frac{F}{k} \\S &= \frac{9}{37.83784} \\S &\approx .23786\end{aligned}$$

So a 9-lb weight will stretch our spring by about .23876 inches.

For a 22-lb force:

$$\begin{aligned}F &= kS \\S &= \frac{F}{k} \\S &= \frac{22}{37.83784} \\S &\approx .58143\end{aligned}$$

So a 22-lb weight will stretch our spring by about .58143 inches.

## 5. 2.2 #6

By taking our data and fitting an OLS regression, we find through our summary:

OLS Regression Results						
=====						
Dep. Variable:	z^1/2	R-squared:	0.998			
Model:	OLS	Adj. R-squared:	0.997			
Method:	Least Squares	F-statistic:	1492.			
Date:	Wed, 07 Oct 2020	Prob (F-statistic):	3.82e-05			
Time:	11:55:07	Log-Likelihood:	9.8452			
No. Observations:	5	AIC:	-15.69			
Df Residuals:	3	BIC:	-16.47			
Df Model:	1					
Covariance Type:	nonrobust					
=====						
	coef	std err	t	P> t	[0.025	0.975]
-----						
const	0.0590	0.076	0.774	0.495	-0.183	0.301
y	0.4822	0.012	38.621	0.000	0.442	0.522
=====						
Omnibus:	nan	Durbin-Watson:		1.823		
Prob(Omnibus):	nan	Jarque-Bera (JB):		0.506		
Skew:	0.237	Prob(JB):		0.777		
Kurtosis:	1.516	Cond. No.		24.4		
=====						

where we find that our P value is much greater than .05. Therefore, we fail to reject the null hypothesis that the y-intercept is zero. Since our  $R^2$  value is also quite close to 1, we say that the data is indeed proportional. Therefore, running our model again forcing the y-intercept to be zero:

OLS Regression Results						
=====						
Dep. Variable:	z^1/2	R-squared (uncentered):		1.000		
Model:	OLS	Adj. R-squared (uncentered):		1.000		
Method:	Least Squares	F-statistic:		2.629e+04		
Date:	Thu, 08 Oct 2020	Prob (F-statistic):		8.68e-09		
Time:	09:47:52	Log-Likelihood:		9.3897		
No. Observations:	5	AIC:		-16.78		
Df Residuals:	4	BIC:		-17.17		
Df Model:	1					
Covariance Type:	nonrobust					
=====						
	coef	std err	t	P> t	[0.025	0.975]
-----						
y	0.4915	0.003	162.157	0.000	0.483	0.500
=====						
Omnibus:	nan	Durbin-Watson:		1.673		
Prob(Omnibus):	nan	Jarque-Bera (JB):		0.367		
Skew:	-0.553	Prob(JB):		0.832		
Kurtosis:	2.264	Cond. No.		1.00		
=====						

We find from the model that our proportionality constant is .4915, therefore, our slope  $k = .4915$

6. 2.2 #10 Using the data given and fitting an OLS regression:

OLS Regression Results						
Dep. Variable:	y	R-squared:	1.000			
Model:	OLS	Adj. R-squared:	1.000			
Method:	Least Squares	F-statistic:	7.348e+04			
Date:	Thu, 08 Oct 2020	Prob (F-statistic):	3.84e-17			
Time:	10:53:30	Log-Likelihood:	-10.931			
No. Observations:	10	AIC:	25.86			
Df Residuals:	8	BIC:	26.47			
Df Model:	1					
Covariance Type:	nonrobust					
	coef	std err	t	P> t	[0.025	0.975]
const	2.3333	0.396	5.888	0.000	1.420	3.247
x	2.1342	0.008	271.072	0.000	2.116	2.152
Omnibus:	0.894	Durbin-Watson:	2.146			
Prob(Omnibus):	0.640	Jarque-Bera (JB):	0.455			
Skew:	-0.483	Prob(JB):	0.797			
Kurtosis:	2.602	Cond. No.	78.2			

we see that the p-value is less than .05 therefore, we can conclude that in this case,  $y$  is not proportional to  $x^2$ .

7. 2.3), #2 Since we assume density is constant, we know weight is proportional to the volume:

$$\begin{aligned}
 W &= kV \\
 W &= k(L^3) \\
 \frac{W}{L^3} &= k \\
 k &= \frac{20}{3^3} \\
 k &= \frac{20}{9} \\
 k &\approx 2.2222
 \end{aligned}$$

Therefore, for a 100-lb flamingo,

$$\begin{aligned}
 100 &= 2.2222V \\
 \frac{100}{2.2222} &= V \\
 \frac{100}{2.2222} &= L^3 \\
 L &= \sqrt[3]{\frac{100}{2.2222}} \\
 L &\approx 3.5569
 \end{aligned}$$

Which shows that a 100-lb flamingo would have length 3.5569ft. Since the leg height is also proportional to the length of the flamingo:

$$\begin{aligned}
 L_{height} &= mL_{leg} \\
 3 &= 2m \\
 \frac{3}{2} &= m
 \end{aligned}$$

and using our new height:

$$\begin{aligned}
 3.5569 &= \frac{3}{2}L_{leg} \\
 3.5569\frac{2}{3} &= L_{leg} \\
 L_{leg} &\approx 2.3713
 \end{aligned}$$

So we see height of the 100-lb flamingo is about 3.5569ft and the leg height of the 100-lb flamingo is 2.3713ft.