Problem Set 8

Problem 1

Plot the phase portrait and classify the fixed points for the following systems:

(a)
$$\dot{x} = -3x + 2y$$
, $\dot{y} = x - 2y$;

(b)
$$\dot{x} = 5x + 10y, \quad \dot{y} = -x - y,$$

(c)
$$\dot{x}=y,~~\dot{y}=-x-2y.$$

Solution

 $\begin{bmatrix} \frac{dx}{dt} \\ \frac{dy}{dt} \end{bmatrix} = A \begin{bmatrix} x \\ y \end{bmatrix}$

We'll skip the process of finding the eigenvalues and eigenvectors of the 2×2 matrix at this point, the e-vals are discussed at the end of the problem.

```
import numpy as np
import sympy as sp
import scipy as scp
import matplotlib.pyplot as plt
%config InlineBackend.figure_format = 'jpg'
%matplotlib inline
```

```
a, b, c, d = sp.symbols('a, b, c, d')
t = sp.symbols('t')
x = sp.Function('x')(t)
y = sp.Function('y')(t)

A = sp.Matrix([
      [a, b],
      [c, d]
])
u = sp.Matrix([x, y])
```

(a), (b), (c)

$$A_a = \begin{bmatrix} -3 & 2 \\ 1 & -2 \end{bmatrix}, \ \ A_b = \begin{bmatrix} 5 & 10 \\ -1 & -1 \end{bmatrix}, \ \ A_c = \begin{bmatrix} 0 & 1 \\ -1 & -2 \end{bmatrix}.$$

The solution to these ODE systems are:

```
subs1 = {a: -3, b: 2, c: 1, d: -2}
subs2 = {a: 5, b: 10, c: -1, d: -1}
subs3 = {a: 0, b: 1, c: -1, d: -2}
subs1 = sp.solvers.ode.systems.dsolve_system(u.diff(t) - A.subs(subs1)*u, funcs=[x, y], t=t)
sol2 = sp.solvers.ode.systems.dsolve_system(u.diff(t) - A.subs(subs2)*u, funcs=[x, y], t=t)
sol3 = sp.solvers.ode.systems.dsolve_system(u.diff(t) - A.subs(subs3)*u, funcs=[x, y], t=t)
u1 = sp.Matrix([i.rhs for i in sol1[0]])
u2 = sp.Matrix([i.rhs for i in sol2[0]])
u3 = sp.Matrix([i.rhs for i in sol3[0]])
```

• System (a):

u1

$$\begin{bmatrix} -2C_1e^{-4t} + C_2e^{-t} \\ C_1e^{-4t} + C_2e^{-t} \end{bmatrix}$$

• System (b):

u2

$$\begin{bmatrix} (C_1 + 3C_2)e^{2t}\sin{(t)} - (3C_1 - C_2)e^{2t}\cos{(t)} \\ C_1e^{2t}\cos{(t)} - C_2e^{2t}\sin{(t)} \end{bmatrix}$$

• System (c):

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$$\begin{bmatrix} C_2 t e^{-t} + (C_1 + C_2) e^{-t} \\ -C_1 e^{-t} - C_2 t e^{-t} \end{bmatrix}$$

Now let's plot the phase portraits:

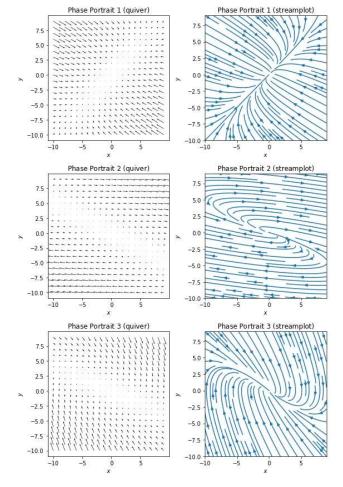
```
Aa = np.array([
        [-3, 2],
[1, -2]
 Ab = np.array([
        [5, 10],
[-1, -1]
])
x = np.arange(-10, 10, 1)
y = np.arange(-10, 10, 1)
X, Y = np.meshgrid(x, y)
ni, nj = X.shape
U1 = np.zeros(X.shape)
V1 = np.zeros(Y.shape)
U2 = np.zeros(X.shape)
V2 = np.zeros(Y.shape)
U3 = np.zeros(X.shape)
V3 = np.zeros(Y.shape)
for i in range(ni):
    for j in range(nj):
        U1[i][j], V1[i][j] = Aa.dot(np.array([X[i][j], Y[i][j]]))
        U2[i][j], V2[i][j] = Ab.dot(np.array([X[i][j], Y[i][j]]))
        U3[i][j], V3[i][j] = Ac.dot(np.array([X[i][j], Y[i][j]]))
Us = [U1, U2, U3]
Vs = [V1, V2, V3]
 fig, (ax1, ax2, ax3) = plt.subplots(3, 2)
for i, ax in enumerate(axes):

ax[0].quiver(X, Y, Us[i], Vs[i])

ax[1].streamplot(X, Y, Us[i], Vs[i])
         for a in ax:
    a.set_xlabel('$x$')
         a.set_ylabel('$y$')

ax[0].set_title('Phase Portrait ${}$ (quiver)'.format(i+1))

ax[1].set_title('Phase Portrait ${}$ (streamplot)'.format(i+1))
fig.set_size_inches(8, 12)
plt.tight_layout()
```



Now let's analyze the fixed points for each system of ODEs:

- (a) This is a stable node, as the solution tends to 0 as t o 0. x decreases as fast as y. This is also seen as $\lambda_{1,2} < 0 \in \mathbb{R}$.
- **(b)** Spiral. The eigenvalues are complex conjugate (as we have both trigonometric and exponential functions in the solution, also checked via direct e-val computation). y and x are a combionation of oscillations with increasing amplitude over time.
- (c) Degenerate node. The eigenvalues $\lambda_1=\lambda_2=1$ are real and equal.

Problem 2

Suppose the relationship between Romeo and Juliet is such that

$$\dot{R} = aR + bJ, \quad \dot{J} = -bR - aJ$$

with positive a and b. Describe the type of the relationship and explain its fate depending on the initial conditions.

Solution

Let's write down the matrix equation:

$$\frac{d}{dt}\begin{bmatrix}R\\J.\end{bmatrix} = \begin{bmatrix}a & b\\-b & -a\end{bmatrix}\begin{bmatrix}R\\J\end{bmatrix}$$

While Romeo is sure about his feelings (aR) and responds to Juliet's attitude (bJ), Juliet is being a real woman - she hates him when he loves her and vice versa (-bR), also not knowing what to do with herself (-aJ). Once a wise man said, "Чем меньше женщину мы любим, тем больше меньше мы, чем тем".

The eigenvalues can be represented with trace $\boldsymbol{\tau}$

$$\tau = a - a = 0$$

and determinant Δ

$$\Delta = -a^2 + b^2$$

as follows:

$$\lambda_{1,2} = \frac{1}{2}(\tau \pm \sqrt{\tau^2 - 4\Delta}) = \frac{\pm 1}{2}\sqrt{-4(b^2 - a^2)} = \pm \sqrt{a^2 - b^2}$$

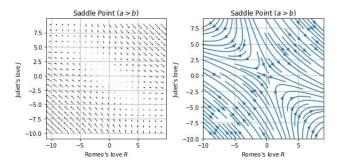
with a>0, b>0.

The (non-normalized) eigenvectors are:

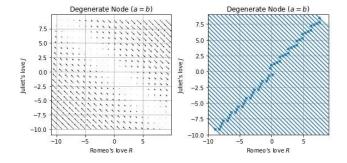
$$s_{1,2} = \left[-rac{a}{b} \pm rac{\sqrt{(a-b)(a+b)}}{b}
ight]$$

• a>b, then $\lambda_1=-\lambda_2$ and $\lambda_{1,2}\in\mathbb{R}$. This will match the saddle case. Depending on the intial conditions J_0 and R_0 , their relation will "follow" one of the main directions (eigenvectors). The eigenvectors will point to 2^{nd} and 4^{th} quadrants with corresponding eigenvalues being positive and negative, respectively. Thus, the love point will tend to 2^{nd} or 4^{th} quadrant, meaning that eventually either Juliet hates Romeo, but he keeps loving her more and more (poor guy = (), or vice versa (poor girl = ().

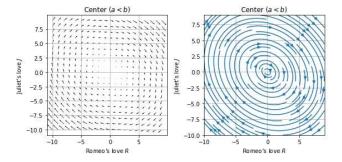
```
RJ = np.array([
       [5, 3],
[-3, -5]
])
x = np.arange(-10, 10, 1)
      np.arange(-10, 10, 1)
X, Y = np.meshgrid(x, y)
ni, nj = X.shape
U = np.zeros(X.shape)
V = np.zeros(Y.shape)
for i in range(ni):
    for j in range(nj):
             U[i][j], V[i][j] = RJ.dot(np.array([X[i][j], Y[i][j]]))
fig, (ax1, ax2) = plt.subplots(1, 2)
ax1.quiver(X, Y, U, V)
ax2.streamplot(X, Y, U, V)
ax1.grid()
axi.grid()
axi.set_title('Saddle Point ($a > b$)')
axi.set_xlabel('Romeo\'s love $R$')
axi.set_ylabel('Juliet\'s love $J$')
ax2.grid()
ax2.grid()
ax2.set_title('Saddle Point ($a > b$)')
ax2.set_xlabel('Romeo\'s love $R$')
ax2.set_ylabel('Juliet\'s love $J$')
fig.set_size_inches(8, 4)
plt.tight_layout()
```



• a=b, thus, $\lambda_1=\lambda_2=0$. This is a singular matrix (rg(A)=0). This case is a degenerate node. There is one eigenvector: $(1,1)^T$, which is the nullspace of the matrix. Romeo values his feeling as much as Juliet's: aR+aJ. Juliet hesitates both in her state and in Romeo's suspicious attention to her equally: -aR-aJ. Thus, when R=-J, their relationship will forever remain the same: one loves and another hates. Other than that case, whether R>J, Romeo will keep fallin in love while Juliet keep hating him more. Opposite in the other case.



• $a < b - \lambda_1 = \lambda_2^*$, both are purely imaginary. This represents a center. With imaginary eigenvalues the matrix A represents pure rotation, thus, every change vector is perpendicular to the current love vector. The relationship will go roundabout an elliptic trajectory with certain radiuses, dependent on initial conditions.



Note

Poor guys =(

Problem 3

For the system

$$\dot{x} = xy - 1, \quad \dot{y} = x - y^3$$

find the fixed points, classify them, sketch the neighbouring trajectories and try to fill in the rest of the phase plane.

Solution

Starting with the fixed points:

$$egin{aligned} xy &= 1 \ x &= y^3 \end{aligned}$$

We get 4 different fixed points:

$$y=[1,-1,i,-i] \ x=[1,-1,-i,i] \ p_1=(1,1), \ \ p_2=(-1,-1), \ \ p_3=(i,-i), \ \ p_4=(-i,i).$$

Assuming that \boldsymbol{x} and \boldsymbol{y} are real, we are left with the following two:

$$p_1 = (1, 1)$$

 $p_2 = (-1, -1).$

To classify them, let's linearize the equations:

$$\frac{dx}{dt} = f(x, y, t),$$

$$f(x, y, t) = f(x_0, y_0, t) + f'_x(x_0, y_0, t)(\Delta x) + f'_y(x_0, y_0, t)(\Delta y) + \dots$$

$$\frac{dy}{dt} = g(x, y, t),$$

$$g(x, y, t) = f(x_0, y_0, t) + g'_x(x_0, y_0, t)(\Delta x) + g'_y(x_0, y_0, t)(\Delta y) + \dots$$

This gives us

$$\dot{x} = x_0 \Delta x + y_0 \Delta y$$

 $\dot{y} = \Delta x + -3y_0^2 \Delta y$

$$\frac{d}{dt} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} x_0 & y_0 \\ 1 & -3y_0^2 \end{bmatrix} \begin{bmatrix} \Delta x \\ \Delta y \end{bmatrix}.$$

Note

Here and afterwards x,y in the right hand side of the linearized equation are actually x' and y' that can be obtained as $x'=x_0+\Delta x, \ y'=y_0+\Delta y.$

• Starting with the first point, (1,1):

$$A = \begin{bmatrix} 1 & 1 \\ 1 & -3 \end{bmatrix}$$

```
A = sp.Matrix([
    [1, 1],
    [1, -3]
])
A.eigenvects()
```

The eigenvalues are

$$\lambda_{1,2} = -1 \pm \sqrt{5}$$

and the eigenvectors:

$$s_{1,2} = egin{bmatrix} 2 \pm \sqrt{5} \ 1 \end{bmatrix}$$

This gives us the saddle fixed point.

• For the second point (-1, -1):

$$A = egin{bmatrix} -1 & -1 \ 1 & 3 \end{bmatrix}$$

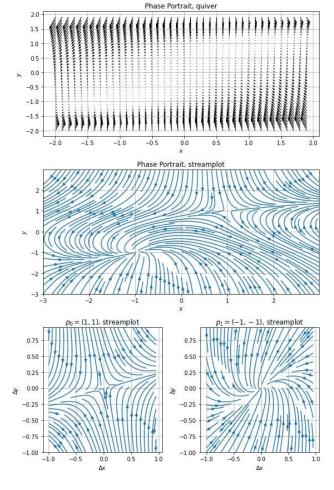
```
A = sp.Matrix([
     [-1, -1],
     [1, -3]
])
A.eigenvects()
```

```
[(-2,
2,
[Matrix([
[1],
[1]])])]
```

We have the eigenvalue $\lambda=2$ with multiplicity 2 and the eigenvector $s=\begin{bmatrix}1\\1\end{bmatrix}$, which yields the degenerate node.

Now let's sketch the phase portrait. By the powers of **python** we can avoid sketching only the neighbourhoods and sketch the whole portrait:

```
def dx(x, y, t):
                       return x*y - 1
   def dy(x, y, t):
                       return x - y**3
   A1 = np.array([
                      [1, -3]
   A2 = np.array([
                     [-1, -1],
[1, -3]
  x = np.arange(-2, 2, 0.1)
y = np.arange(-2, 2, 0.1)
 The state of 
   for i in range(ni):
    for j in range(nj):
 U[i][j] = dx(X[i][j], Y[i][j], 0)
V[i][j] = dy(X[i][j], Y[i][j], 0)
plt.subplot(3, 1, 1)
plt.quiver(X, Y, U, V)
plt.title('Phase Portrait, quiver')
plt.xlabel('$x$')
   plt.ylabel('$y$')
   plt.grid()
   #streamplot
  x = np.arange(-3, 3, 0.01)
y = np.arange(-3, 3, 0.01)
X, Y = np.meshgrid(x, y)
  ni, nj = X.shape
U = np.zeros(X.shape)
V = np.zeros(X.shape)
   for i in range(ni):
for i in range(ni):
    for j in range(nj):
        U[i][j] = dx(X[i][j], Y[i][j], 0)
        V[i][j] = dy(X[i][j], Y[i][j], 0)
plt.subplot(3, 1, 2)
plt.streamplot(X, Y, U, V, density=2)
plt.title('Phase Portrait, streamplot')
plt.xlabel('$x$')
plt.ylabel('$x$')
plt.ylabel('$y$')
nlt.grid()
   plt.grid()
   #point (1, 1)
  #point (1, 1)
x = np.arange(-1, 1, 0.05)
y = np.arange(-1, 1, 0.05)
X, Y = np.meshgrid(x, y)
  ni, nj = X.shape
U = np.zeros(X.shape)
   V = np.zeros(X.shape)
 V = np.Zeros(X.shape)
for i in range(ni):
    for j in range(nj):
        U[i][j], V[i][j] = A1.dot(np.array([X[i][j], Y[i][j]]))
plt.subplot(3, 2, 5)
plt.streamplot(X, Y, U, V)
plt.title('$p_0 = (1, 1)$, streamplot')
plt.xlabel('$\Delta x$')
plt.ylabel('$\Delta x$')
plt.ylabel('$\Delta y$')
nlt grid()
   plt.grid()
 #point (-1, -1)
x = np.arange(-1, 1, 0.05)
y = np.arange(-1, 1, 0.05)
X, Y = np.meshgrid(x, y)
 ni, nj = X.shape
U = np.zeros(X.shape)
V = np.zeros(X.shape)
for i in range(ni):
                     for j in range(nj):
    U[i][j], V[i][j] = A2.dot(np.array([X[i][j], Y[i][j]]))
 U[1][], V[1][]] = A2.dot(np.arra)
plt.subplot(3, 2, 6)
plt.streamplot(X, Y, U, V)
plt.title('$p_1 = (-1, -1)$, streamplot')
plt.xlabel('$\Delta x$')
   plt.ylabel('$\Delta y$')
   plt.grid()
   fig = plt.gcf()
  fig.set_size_inches(8, 12)
plt.tight_layout()
```



Problem 4

For the following model of rabbits and sheep, find the fixed points, investigate their stability and draw the phase portrait. Indicate the basins of attraction of any stable fixed point:

$$\dot{x} = x(3 - 2x - 2y), \quad \dot{y} = y(2 - x - y).$$

Solution

As always, we start with finding the fixed points:

```
x, y = sp.symbols('x, y')
eq1 = x*(3 - 2*x - 2*y)
eq2 = y*(2 - x - y)
eqs = [eq1, eq2]
sol = sp.solve(eqs, x, y)
sol

[(0, 0), (0, 2), (3/2, 0)]
```

Fixed points are:

$$p_1 = (0,0) \ p_2 = (0,2) \ p_3 = (rac{3}{2},0)$$

Now let's linearize the equations around the fixed points:

$$egin{bmatrix} \dot{x} \ \dot{y} \end{bmatrix} = egin{bmatrix} 3 - 4x_0 & -2x_0 \ -y_0 & 2 - 2y_0 \end{bmatrix} egin{bmatrix} \Delta x \ \Delta y \end{bmatrix}$$

And find the types of the points:

A.subs(subs1).eigenvects()

```
[(2,

1,

[Matrix([

[0],

[1]])]),

(3,

1,

[Matrix([

[1],

[0]])])]
```

A.subs(subs2).eigenvects()

```
[(-2,

1,

[Matrix([

[0],

[1]])]),

(3,

1,

[Matrix([

[-5/2],

[ 1]]))]]
```

A.subs(subs3).eigenvects()

```
[(-3,

1,

[Matrix([

[1],

[0]])]),

(2,

1,

[Matrix([

[-3/5],

[ 1]])])]
```

• p_1 = (0, 0): unstable node

$$\lambda_1=2,\ \ \lambda_2=3,$$

$$s_1 = egin{bmatrix} 1 \ 0 \end{bmatrix}, \;\; s_2 = egin{bmatrix} 0 \ 1 \end{bmatrix}$$

• p_2 = (0, 2): saddle

$$\lambda_1=-2,~~\lambda_2=3,$$

$$s_1 = egin{bmatrix} 0 \ 1 \end{bmatrix}, \;\; s_2 = egin{bmatrix} -rac{5}{2} \ 1 \end{bmatrix}$$

• p_3 = $(\frac{3}{2}, 0)$: saddle

$$\lambda_1=-3,~~\lambda_2=2$$

$$s_1 = egin{bmatrix} 1 \ 0 \end{bmatrix}, \;\; s_2 = egin{bmatrix} -rac{3}{5} \ 1 \end{bmatrix}$$

Now let's draw the phase portrait:

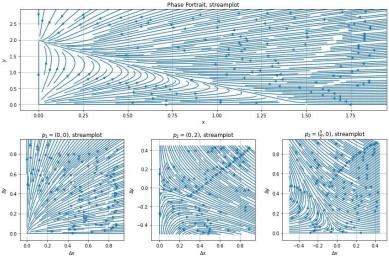
Note

From the formulation of the problem we are only interested in positive values of x and y, thus, phase portrait will only be drawn only for x > 0, y > 0.

Note

Point (0,0) on the phase portrait around the fixed point i coincides with the (x_0^i,y_0^i) point on the main phase portrait.

```
def dx(x, y):
       return x*(3 - 2*x - 2*y)
def dy(x, y):
    return y*(2 - x - y)
A1 = np.array(A.subs(subs1), dtype=np.float64)
A2 = np.array(A.subs(subs2), dtype=np.float64)
A3 = np.array(A.subs(subs3), dtype=np.float64)
x = np.arange(0, 2, 0.05)
y = np.arange(0, 3, 0.05)
X, Y = np.meshgrid(x, y)
ni, nj = X.shape
U = np.zeros(X.shape)
V = np.zeros(X.shape)
plt.title('Phase Portrait, streamplot')
plt.xlabel('$x$')
 plt.ylabel('$y$')
plt.grid()
x = np.arange(0, 1, 0.05)
y = np.arange(0, 1, 0.05)
X, Y = np.meshgrid(x, y)
ni, nj = X.shape
U = np.zeros(X.shape)
V = np.zeros(X.shape)
for i in range(ni):
for i in range(nj):
    for j in range(nj):
        U[i][j], V[i][j] = A1.dot(np.array([X[i][j], Y[i][j]]))
plt.subplot(2, 3, 4)
plt.streamplot(X, Y, U, V, density=2)
plt.title('$p_1 = (0, 0)$, streamplot')
plt.xlabel('$\text{Velta x$'})
plt.xlabel('$\text{Velta x$'})
plt.ylabel('$\Delta y$')
plt.grid()
x = np.arange(0, 1, 0.05)
y = np.arange(-0.5, 0.5, 0.05)
X, Y = np.meshgrid(x, y)
ni, nj = X.shape
U = np.zeros(X.shape)
V = np.zeros(X.shape)
for i in range(ni):
    for j in range(nj):
        U[i][j], V[i][j] = A2.dot(np.array([X[i][j], Y[i][j]]))
plt.subplot(2, 3, 5)
plt.streamplot(X, Y, U, V, density=2)
plt.title('$p_2 = (0, 2)$, streamplot')
plt.xlabel('$\Delta x$')
plt.ylabel('$\Delta x$')
plt.ylabel('$\Delta y$')
nlt.prid()
plt.grid()
x = np.arange(-0.5, 0.5, 0.05)
y = np.arange(0, 1, 0.05)
X, Y = np.meshgrid(x, y)
ni, ni = X.shape
U = np.zeros(X.shape)
V = np.zeros(X.shape)
plt.ylabel('$\Delta y$')
plt.grid()
fig = plt.gcf()
fig.set_size_inches(12, 8)
plt.tight_layout()
                                                                            Phase Portrait, streamplot
```



As for the basins of attraction: neither of these points is a stable fixed point. However, we can predict the behaviour of the solutions at the large t.

For p_1 , there is no basin of the attraction, as the node is unstable - every solution diverges from the point.

For p_2 , we have two eigenvectors and eigenvalues with different signs, the first being negative and the second positive. This means that solutions will tend towards the second eigenvector $s_2 = \begin{bmatrix} -1 \\ \frac{2}{\epsilon} \end{bmatrix}$.

▲ Warning

I don't quite understand what is going on on the plot for $p_2=(0,2)$, as it should point in the opposite direction.

For p_3 the same logic is applied with its eigenvectors. For y=0, the solutions tend to the point $(\frac{3}{2},0)$. With $y_0>0$ they tend towards the eigenvector $s_2=\begin{bmatrix} -\frac{3}{5}\\1 \end{bmatrix}$.

Problem 5

Consider the system

$$\dot{x} = -y - x^3, \quad \dot{y} = x.$$

Show that the origin is a spiral, although the linearization predicts a center.

Solution

Fixed points:

$$-y - x^3 = 0, \quad x = 0$$

$$\Rightarrow x = 0, \quad y = 0.$$

Linearization:

$$\frac{d}{dt} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} -3x_0^2 & -1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} \Delta x \\ \Delta y \end{bmatrix}$$

Let's find eigenvalues of the matrix $A|_{(0,0)}$:

```
A = np.array([
       [0, -1],
       [1, 0]
])
sp.Matrix(A).eigenvects()
```

```
[(-I,

1,

[Matrix([

[-I],

[ 1]])]),

(I,

1,

[Matrix([

[I],

[I])])]
```

Pure imaginary λ 's indeed tell us this should be a center (pure rotation of the vectors). However, let's plot the phase portrait aroun (0,0):

```
def dx(x, y):
    return -y - x**3

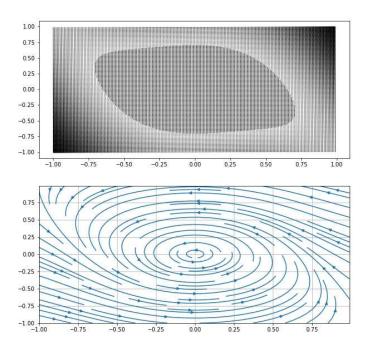
def dy(x, y):
    return x

x = np.arange(-1, 1, 0.01)
y = np.arange(-1, 1, 0.01)
X, Y = np.meshgrid(x, y)
ni, nj = X.shape

U = np.zeros(X.shape)
V = np.zeros(X.shape)

for i in range(ni):
    for j in range(nj):
        U[i][j] = dx(X[i][j], Y[i][j])
        V[i][j] = dy(X[i][j], Y[i][j])

plt.subplot(211)
plt.quiver(X, Y, U, V)
plt.subplot(212)
plt.streamplot(X, Y, U, V)
plt.grid()
plt.gcf().set_size_inches(10, 10)
```



To explain this behaviour, let's move to polar coordinates:

$$r^2 = x^2 + y^2,$$

$$x = r\cos\varphi, \quad y = r\sin\varphi.$$

Then

$$\dot{r}\cos\varphi - r\dot{\varphi}\sin\varphi = -r\sin\varphi - r^3\cos^3\varphi,$$
$$\dot{r}\sin\varphi + r\dot{\varphi}\cos\varphi = r\cos\varphi$$

and

$$r\dot{arphi}=r+r^3\cos^3arphi\sinarphi$$

Assuming we start not from the fixed point (i.e. $r \neq 0$), we get

$$\dot{arphi}=1+r^2\cos^3arphi\sinarphi.$$

Substitute this to the first equation:

$$\dot{r}\sin\varphi\cos\varphi - r\dot{\varphi}\sin^2\varphi = -r\sin^2\varphi - r^3\cos^3\varphi\sin\varphi,$$

$$\dot{r} = -r^3 \cos^4 \varphi.$$

From here we see, recalling our assumption $r \neq 0$, that at $\forall \varphi \neq \pm \frac{\pi}{2}$ the \dot{r} term is not zero, thus, the radius will change, meaning the fixed point is not center anymore (center means rotational movement). At $\varphi = \pm \frac{\pi}{2}$ $\dot{r} = 0$, which is seen on the phase portrait as the motion perpendicular to OY.

Problem 6

The Kermack-McKendrick model of an epidemic describes the population of helathy people x(t) and sick people y(t) in terms of the equations

$$\dot{x}=-kxy,\ \dot{y}=kxy-ly,$$

where k, l > 0. Here, l is the death rate of the sick people in equation for \dot{y} implies that people get sick at a rate proportional to their encounters (which itself is proportional to the product of the number of sick people y and healthy people x). The parameter k measures the probability of transmission of the disease during encounters.

- (a) Find and classify the fixed points.
- (b) Sketch the nullclines and the vector field.
- (c) Find a conserved quantity for the system (hintL form an ODE for dy/dx and integrate it).
- (d) Plot the phase portrait. What happens as $t o \infty$?
- (e) Let (x_0, y_0) be the initial condition. Under what conditions on (x_0, y_0) will the epidemic occur? (Epidemic occurs if y(t) increases initially).

Solution

(a)

The fixed points are:

$$\begin{cases} -kxy = 0, \\ y(kx - l) = 0 \end{cases}$$

In this case there is only one possibility: $y=0, \Rightarrow x \in \mathbb{R}$ - this is a line of fixed points in X direction.

Linearization gives us the following equation:

$$\frac{d}{dt} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 0 & -kx \\ 0 & kx - l \end{bmatrix} \begin{bmatrix} \Delta x \\ \Delta y \end{bmatrix}$$

The matrix \boldsymbol{A} yields two eigenvalues:

$$\lambda_1 = 0, \quad \lambda_2 = kx - l$$

and corresponding eigenvectors:

$$s_1 = egin{bmatrix} 1 \ 0 \end{bmatrix}, \;\; s_2 = egin{bmatrix} rac{l}{kx} - 1 \ 1 \end{bmatrix}$$

(b)

Let's elaborate on possible λ 's:

- 1. $x<rac{l}{k}$ then λ_2 is negative and eigenvector s_2 points in the 1^{st} quadrant, the solutions tend to y->0.
- 2. $x=rac{l}{k}$ both eigenvalues are zero, nearby the fixed points line there is no change in the vector $\begin{bmatrix} x \\ y \end{bmatrix}$ at the line $x=rac{l}{k}$.
- 3. $x>\frac{l}{k}$ λ_2 is positive, eigenvector s_2 points towards the 2^{nd} quadrant. The solutions tend towards the eigenvectors.

See the attached image for clarification:

The nullclines here are y=0 ($\frac{dx}{dt}=0$) and $x=\frac{l}{k}$ ($\frac{dy}{dt}=0$).

(e)

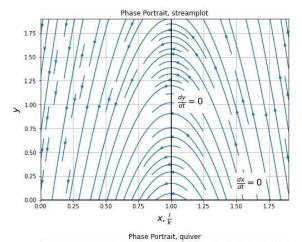
From this we can derive that epidemic occurs only if initial point (x_0,y_0) is to the right of the line $x=\frac{l}{k}$, i.e. $(x_0>\frac{l}{k},y_0\neq 0)$

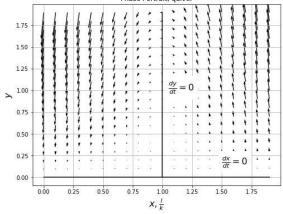
(d)

Now let's plot the phase portrait to confirm our expectations:

```
k = 0.1
1 = 0.1
 def dx(x, y):
           return -k*k*y
           return k*x*y - 1*y
 x = np.arange(0, 2, 0.1)
 y = np.arange(0, 2, 0.1)
 ni, nj = X.shape
U = np.zeros(X.shape)
V = np.zeros(X.shape)
  for i in range(ni):
           for j in range(nj):
    U[i][j] = dx(X[i][j], Y[i][j])
    V[i][j] = dy(X[i][j], Y[i][j])
plt.streamplot(X, Y, U, V)
plt.plot([1/k, 1/k], [0, 1.9], color='black')
plt.plot([0, 1.9], [0, 0], color='black')
plt.xlabel('$x, \\frac{l}{k}\$', fontsize=16)
plt.ylabel('$y\$', fontsize=16)
plt.title('Phase Portrait, streamplot')
plt.title('Phase Portrait, streamplot')
 plt.text(x=1.65, y=1, s='x)-frac{dy}{dt}=0$', backgroundcolor='white', fontsize=16) plt.text(x=1.5, y=0.15, s='x)-frac{dx}{dt}=0$', backgroundcolor='white', fontsize=16) plt.gcf().set_size_inches(8, 6)
 plt.grid()
  plt.quiver(X, Y, U, V)
plt.quiver(X, Y, U, V)
plt.plot([]/k, l/k], [0, 1.9], color='black')
plt.plot([0, 1.9], [0, 0], color='black')
plt.xlabel('$x, \\frac{t}{k}\$', fontsize=16)
plt.ylabel('$y\$', fontsize=16)
plt.title('Phase Portrait, quiver')
plt.text(x=1.05, y=1, s='\\frac{dy}{dt}=0\$', backgroundcolor='white', fontsize=16)
plt.text(x=1.5, y=0.15, s='\\frac{dx}{dt}=0\$', backgroundcolor='white', fontsize=16)
plt.text(\(\text{x}=1.5\), y=0.15, s='\\\frac{dx}{dt}=0\$', backgroundcolor='white', fontsize=16)
 plt.gcf().set_size_inches(8, 6)
 plt.grid()
```

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(c)

$$\frac{dx}{dt} = -kxy,$$

$$\frac{dy}{dt} = kxy - ly,$$

Dividing the second equation by first we obtain:

$$\frac{dy}{dx} = \frac{ly - kxy}{kxy} = \frac{l}{kx} - 1$$

Integrating:

$$dy = \frac{l}{k}\frac{dx}{x} - dx,$$

$$C = \frac{l}{k}\log x - x - y, \ x, y > 0, \ C = Const$$

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