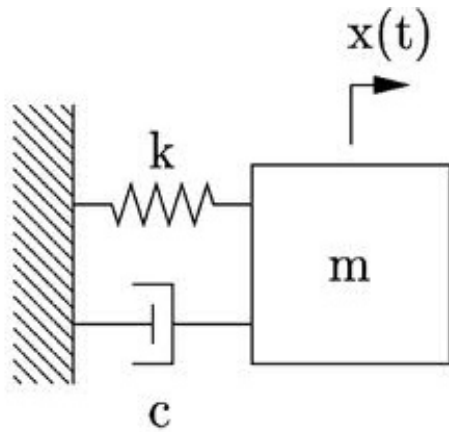


Phase Portraits for 2nd order linear systems

Dr. Mitch Pryor

Phase Portraits

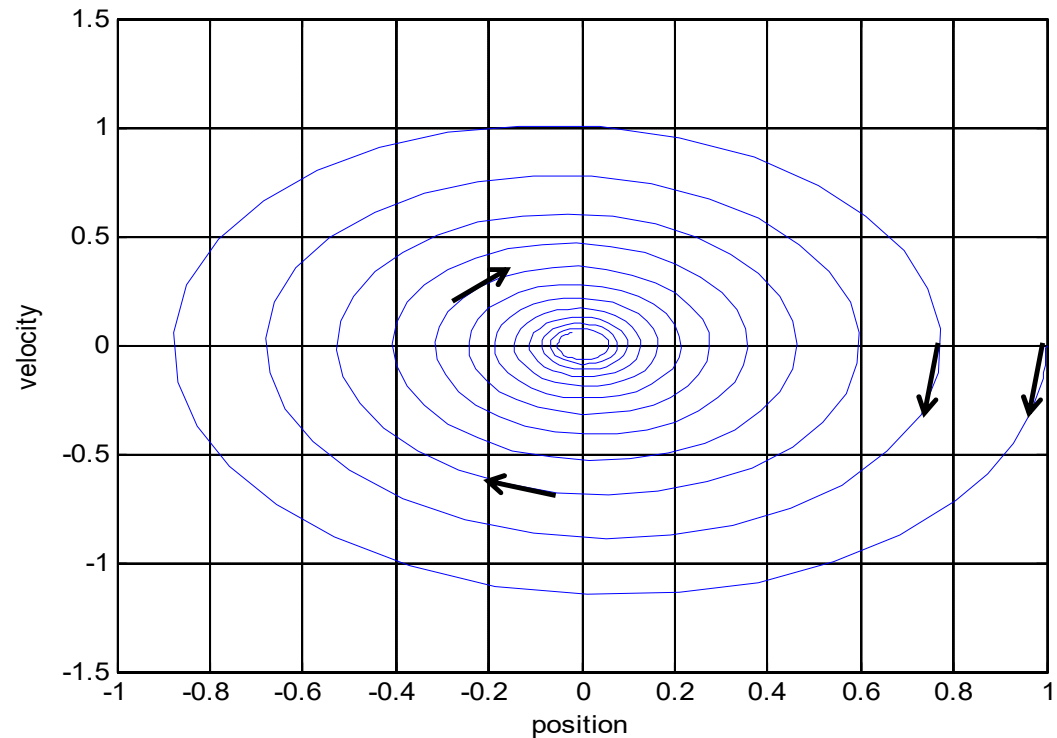


Let...

$$k = 3, b = 0.2, m = 1, F = 0$$

Some initial conditions...

$$z(0) = \begin{Bmatrix} 1 \\ 0 \end{Bmatrix}$$



```
%m-s-d solution
[t, z] = ode45( @test, [0 60 ], [ 1 0 ] );
plot( z(:,1), z(:,2));
grid on;
```

```
function zprime = test( t, z )
m = 2; b = .2; k = 3; F = 0;
zprime = [
    z(2);
    -(b/m)*z(2) - (k/m)*z(1) + F; ];
```

A general Linear 2nd order system

Linear systems have only one equilibrium point where $\frac{d\mathbf{z}_e}{dt} = 0$

$$\dot{z}_1 = f_1(z_1, z_2) = az_1 + bz_2 = 0$$

Only solution at (0,0), the Origin!

$$\dot{z}_2 = f_2(z_1, z_2) = cz_1 + dz_2 = 0$$

True for all linear systems.

There are 4 possible solutions to a second order system.

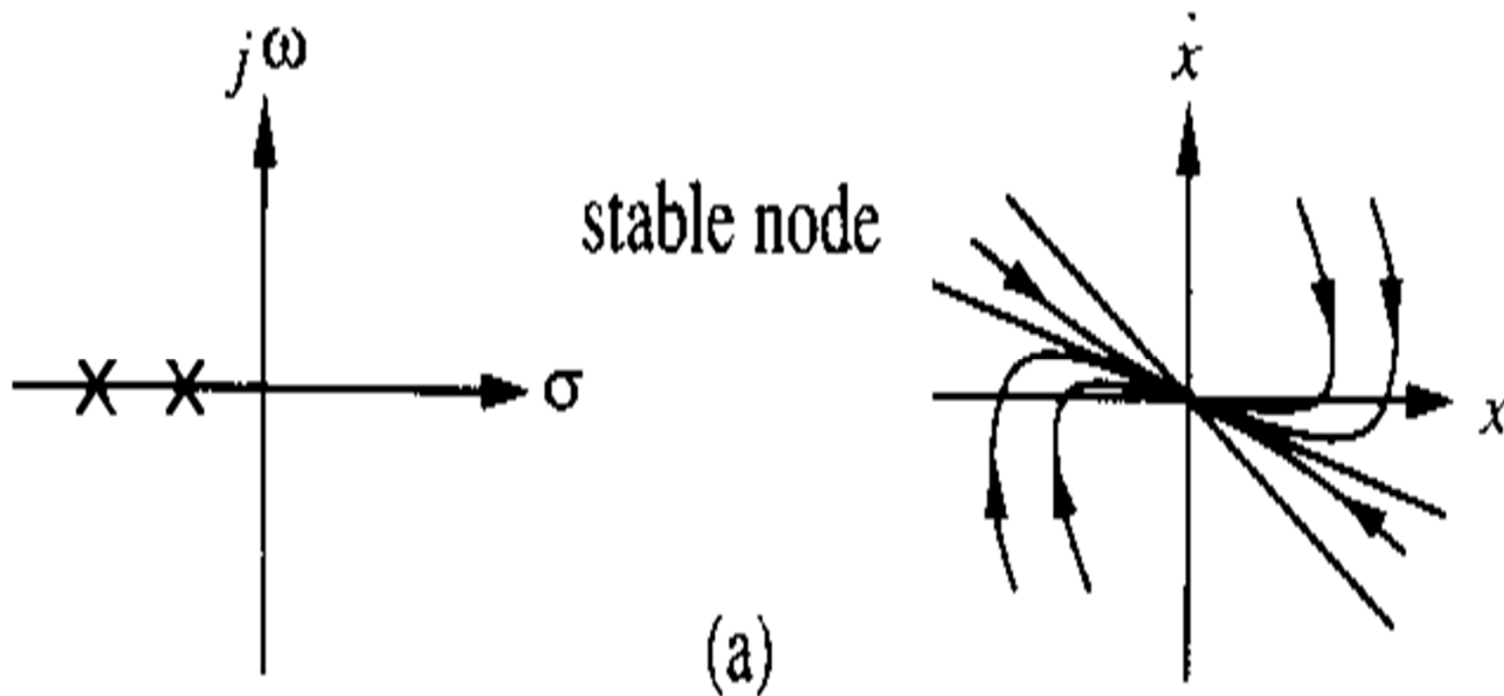
$$\mathbf{z}(t) = k_1 e^{\lambda_1 t} \boldsymbol{\eta}_1 + k_2 e^{\lambda_2 t} \boldsymbol{\eta}_2$$

1. The eigenvalues are both real and have the same sign.
2. The eigenvalues are both real and have opposite signs.
3. The eigenvalues are complex conjugates with nonzero real parts.
4. The eigenvalues are complex conjugates with real parts equal to zero

“What do the phase portraits look like for each scenario?”

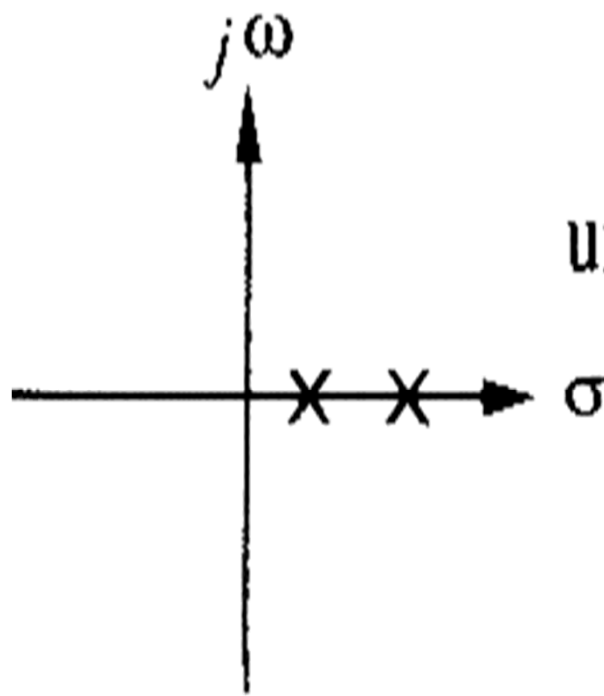
Both eigenvalues are real and negative

$$\ddot{x} + a\dot{x} + bx = 0 \quad \Rightarrow \quad \mathbf{z}(t) = k_1 e^{\lambda_1 t} \boldsymbol{\eta}_1 + k_2 e^{\lambda_2 t} \boldsymbol{\eta}_2$$



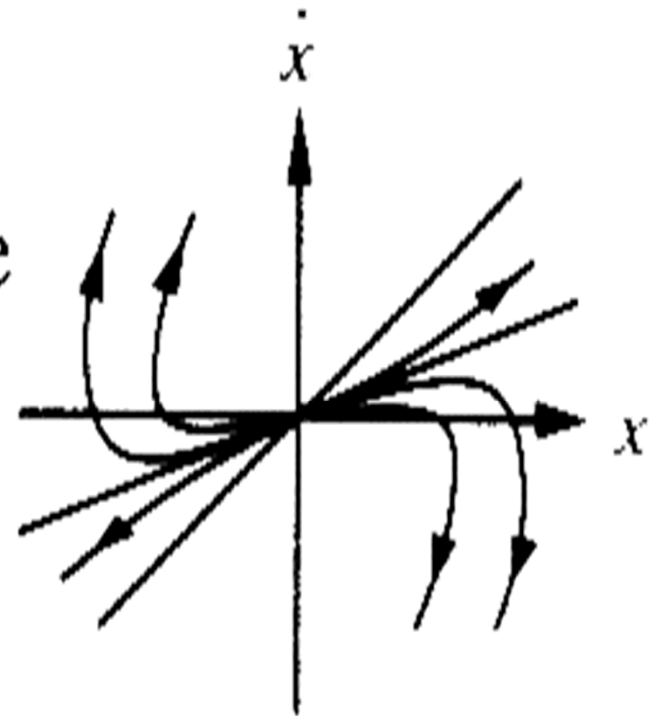
Both eigenvalue are real and positive

$$\ddot{x} + a\dot{x} + bx = 0 \quad \Rightarrow \quad \mathbf{z}(t) = k_1 e^{\lambda_1 t} \boldsymbol{\eta}_1 + k_2 e^{\lambda_2 t} \boldsymbol{\eta}_2$$



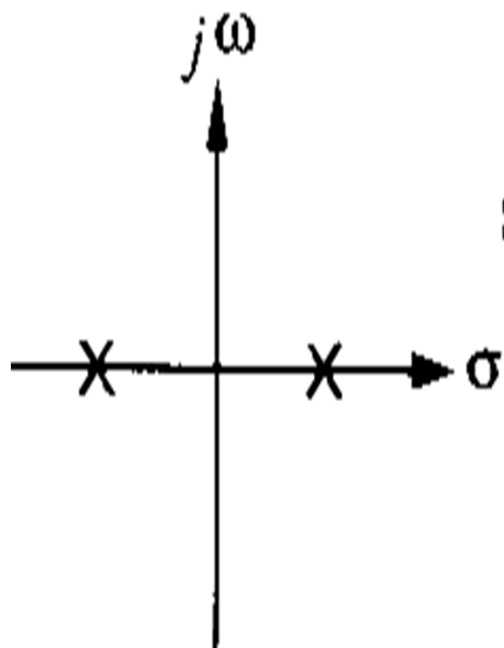
unstable node

(b)



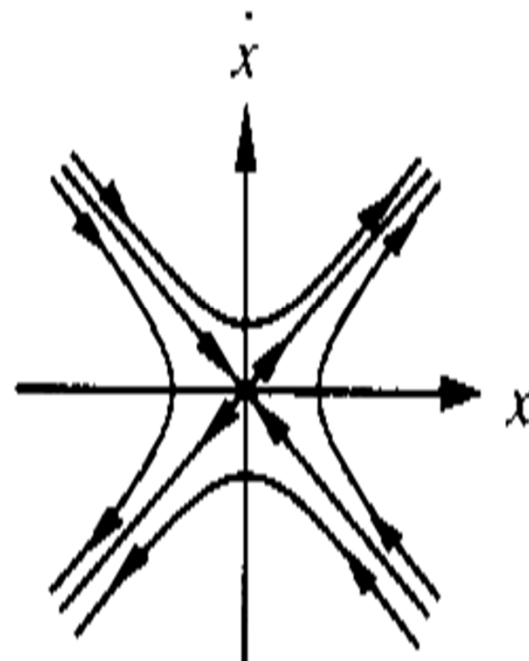
Both eigenvalues are real with different signs

$$\ddot{x} + a\dot{x} + bx = 0 \quad \Rightarrow \quad \mathbf{z}(t) = k_1 e^{\lambda_1 t} \boldsymbol{\eta}_1 + k_2 e^{\lambda_2 t} \boldsymbol{\eta}_2$$



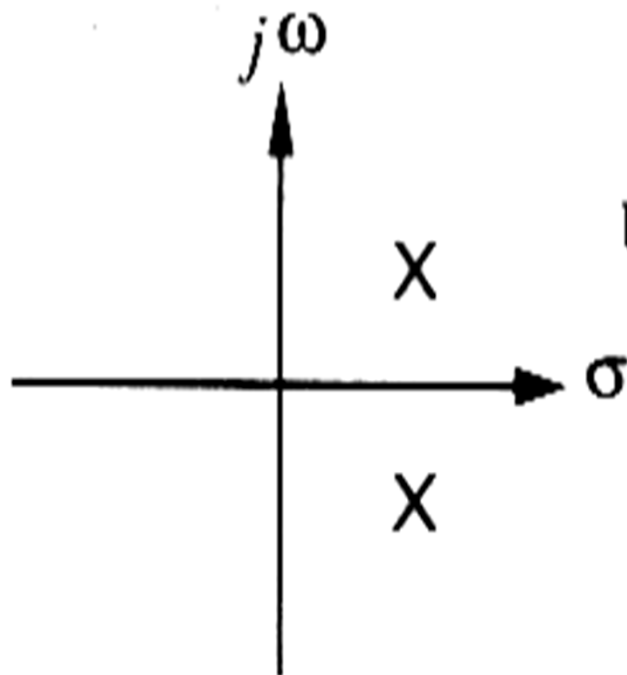
saddle point

(c)

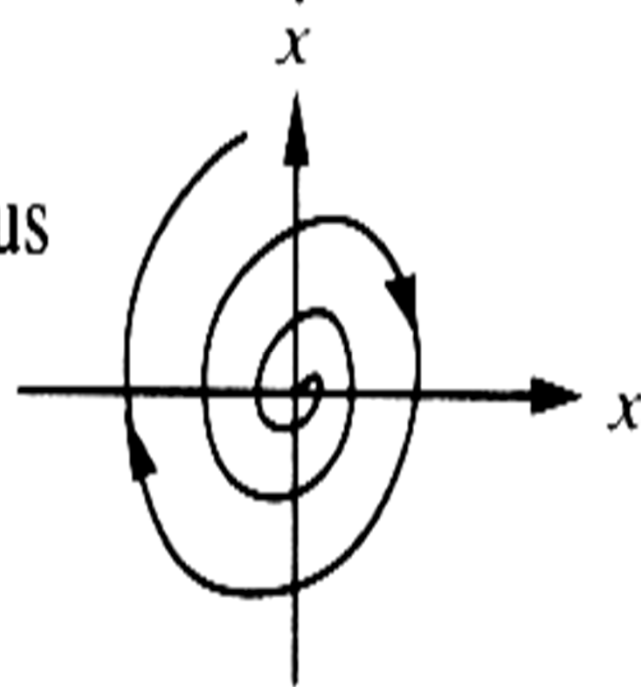


Both eigenvalue positive complex conjugates

$$\ddot{x} + a\dot{x} + bx = 0 \quad \Rightarrow \quad \mathbf{z}(t) = k_1 e^{\lambda_1 t} \boldsymbol{\eta}_1 + k_2 e^{\lambda_2 t} \boldsymbol{\eta}_2$$



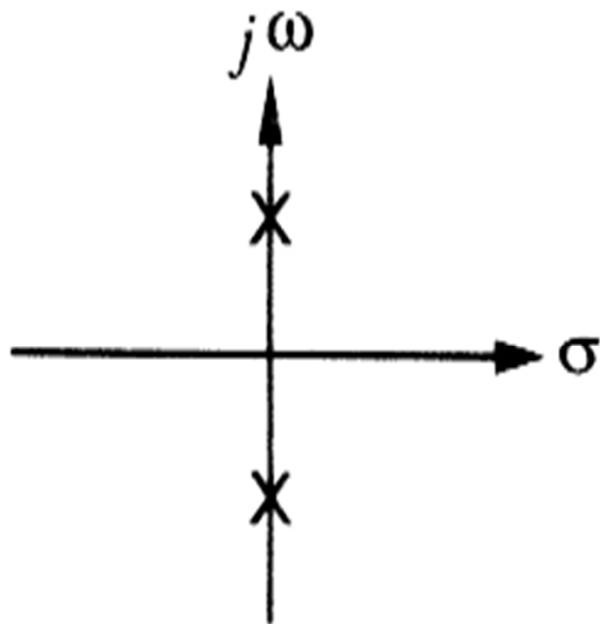
unstable focus



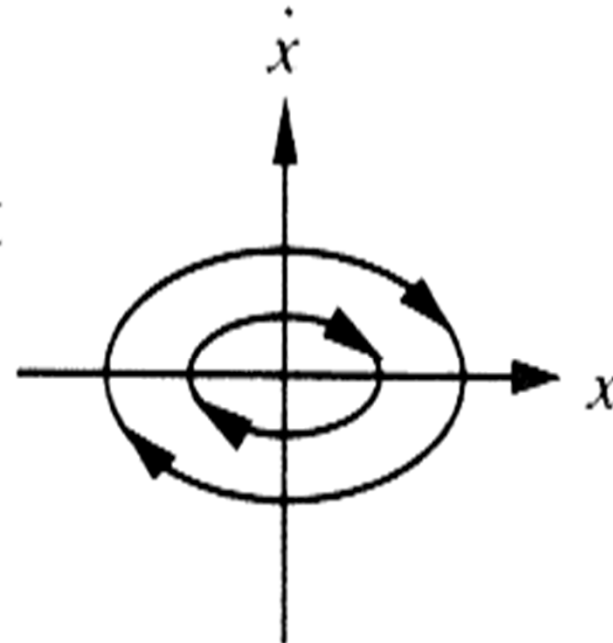
(e)

Both eigenvalues have no real component

$$\ddot{x} + a\dot{x} + bx = 0 \quad \Rightarrow \quad \mathbf{z}(t) = k_1 e^{\lambda_1 t} \boldsymbol{\eta}_1 + k_2 e^{\lambda_2 t} \boldsymbol{\eta}_2$$



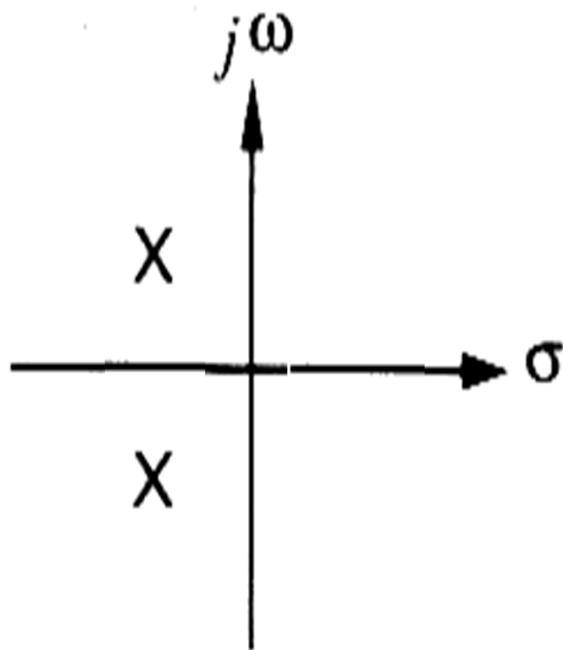
center point



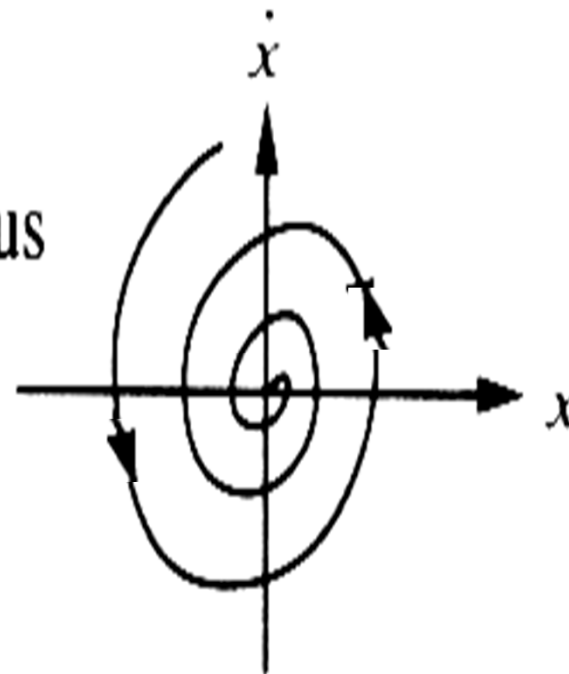
(f)

Both eigenvalues are complex conjugates with
negative real components

$$\ddot{x} + a\dot{x} + bx = 0 \quad \Rightarrow \quad \mathbf{z}(t) = k_1 e^{\lambda_1 t} \boldsymbol{\eta}_1 + k_2 e^{\lambda_2 t} \boldsymbol{\eta}_2$$



stable focus



(e)

Back to the nonlinear example...

In our example...

$$\ddot{x} + 0.6\dot{x} + 3x + x^2 = 0$$

$$z_1 = x \Rightarrow \dot{z}_1 = z_2$$

$$z_2 = \dot{x} \Rightarrow \dot{z}_2 = -0.6z_2 - 3z_1 - z_1^2$$

$$f_1(z_1, z_2) = z_2 = 0$$

$$f_2(z_1, z_2) = -0.6(0) - 3z_1 - z_1^2 = 0$$

$$0 = z_1(-3 - z_1)$$

$(0,0)$ and $(-3,0)$ are equilibrium points;

We can linearize about the equilibrium points and examine stability in near z_e

Find the Jacobian for the system....

$$\mathbf{J} = \begin{bmatrix} \frac{\partial f_1}{\partial z_1} & \frac{\partial f_1}{\partial z_2} \\ \frac{\partial f_2}{\partial z_1} & \frac{\partial f_2}{\partial z_2} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -3-2z_1 & -0.6 \end{bmatrix}$$

Like a gradient, but for multiple variables

At $(0,0)$

$$\mathbf{J} = \begin{bmatrix} 0 & 1 \\ -3 & -0.6 \end{bmatrix} \Rightarrow \lambda = \begin{bmatrix} -0.3000 + 1.7059i \\ -0.3000 - 1.7059i \end{bmatrix}$$

so stable near $(0,0)$

At $(-3,0)$

$$\mathbf{J} = \begin{bmatrix} 0 & 1 \\ 3 & -0.6 \end{bmatrix} \Rightarrow \lambda = \begin{bmatrix} 1.4578 \\ -2.0578 \end{bmatrix}$$

so unstable saddle point at $(-3,0)$

Our solution...

At $(0,0)$

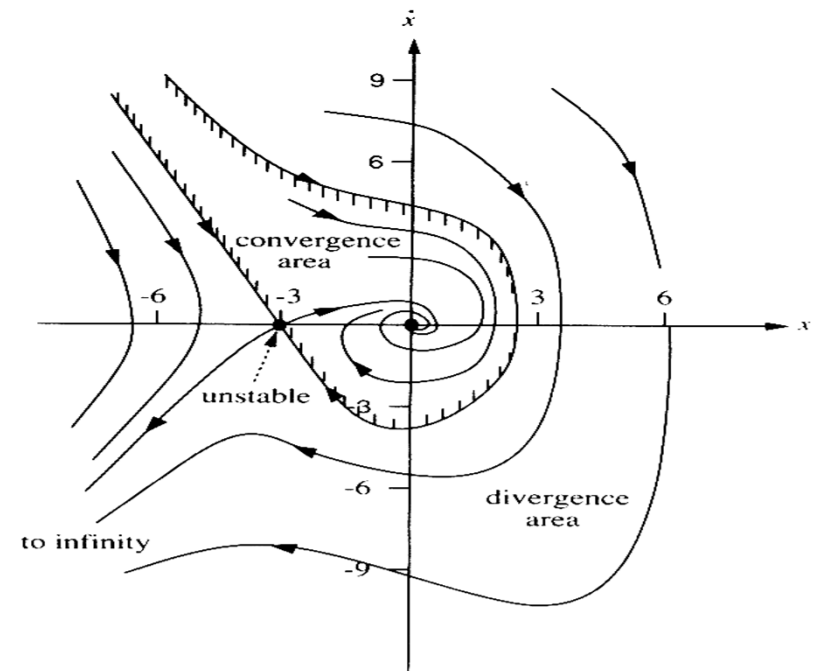
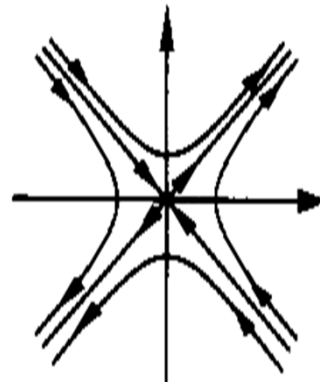
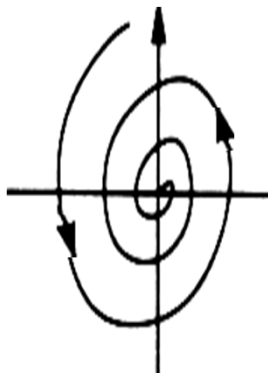
$$\mathbf{J} = \begin{bmatrix} 0 & 1 \\ -3 & -0.6 \end{bmatrix} \Rightarrow \lambda = \begin{bmatrix} -0.3000 + 1.7059i \\ -0.3000 - 1.7059i \end{bmatrix}$$

so stable near $(0,0)$

At $(-3,0)$

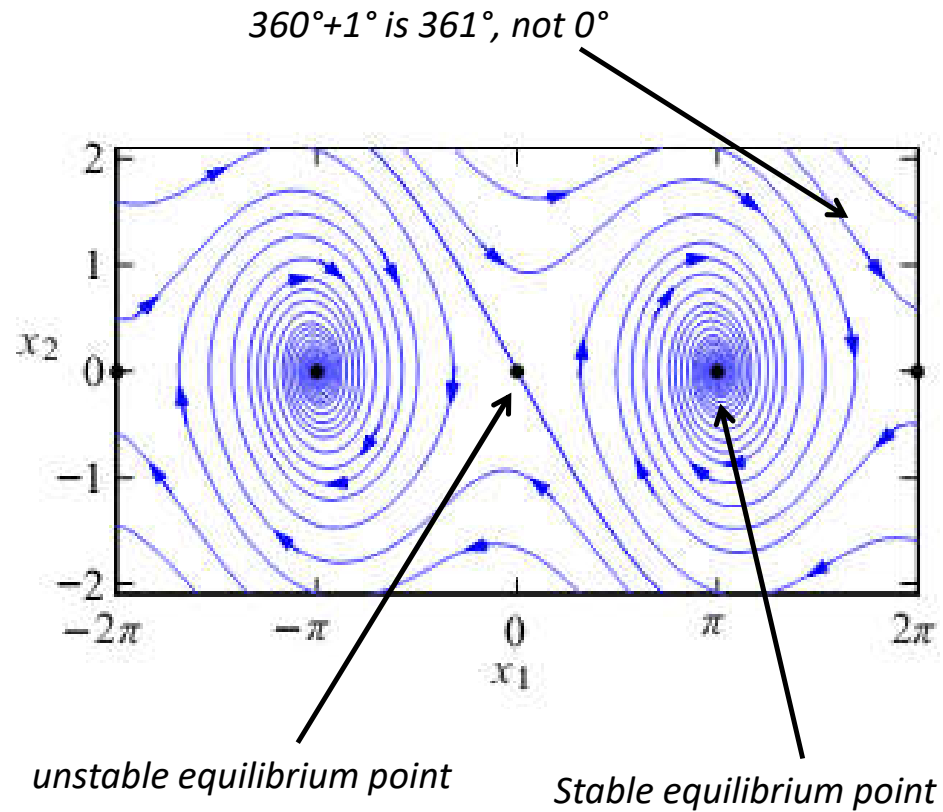
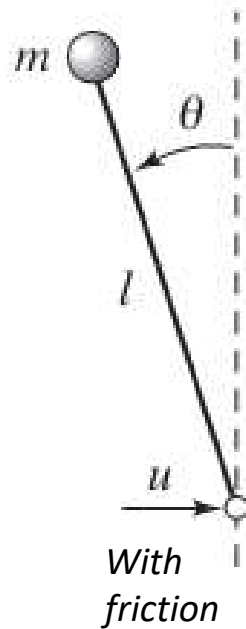
$$\mathbf{J} = \begin{bmatrix} 0 & 1 \\ 3 & -0.6 \end{bmatrix} \Rightarrow \lambda = \begin{bmatrix} 1.4578 \\ -2.0578 \end{bmatrix}$$

so unstable saddle point at $(-3,0)$



Tada!

Final Example, Pendulum



Summary

- Summarized the relationship between eigensystems and phase portraits for second order systems
 - From linear algebra, we see the clear relationship between eigenvalues and stability.
 - We see that there is always 1 (and only 1) equilibrium point at the origin.
 - We also see how eigenvectors impact a systems behavior.
- We can find the Jacobian (i.e. linearize) nonlinear systems at identified equilibrium points to determine system behavior near them.
- From this we can infer, that a controller that stabilizes a system or changes its dynamic behavior must modify the system's underlying eigenvalues and eigenvectors.