

Feb 13  
2012 MAT119A

# 2-D Phase Plane Ch 5

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2x2 Nonlinear Autonomous System -

$$\dot{\underline{x}} = \begin{pmatrix} \dot{x} \\ \dot{y} \end{pmatrix} = f(\underline{x}) = \begin{pmatrix} f_1(x,y) \\ f_2(x,y) \end{pmatrix}$$

• Rest Point:  $f(\underline{x}) = 0 \Rightarrow \underline{x}(t) = \underline{x}$  is a constant soln.

• Assume:  $f$  is Lipschitz & contin  $\underline{x}$  in Domain  $U \subseteq \mathbb{R}^2$  where soln's take values -

$\exists K$  st

$$\|f(\underline{x}_2) - f(\underline{x}_1)\| \leq K \|\underline{x}_2 - \underline{x}_1\|$$

$\exists!$  Thm: There exists a unique soln of ivp

$\forall \underline{x}_0 \in U$ : i.e.  $\exists!$  Soln of  
 $\dot{\underline{x}} = f(\underline{x})$   
 $\underline{x}(t_0) = \underline{x}_0$ .

Soln exists for  $t \in (t_0 - \epsilon, t_0 + \epsilon)$

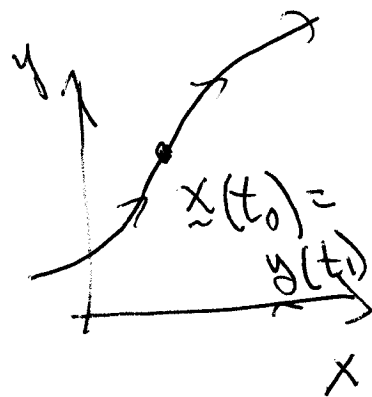
(2)

- Theorem: If  $\underline{x}(t)$  is a solution, so is  $\underline{x}(t+c)$  for any constant  $c$ .

Pf. Assume  $\dot{\underline{x}}(t) = f(\underline{x}(t))$ . Then set

$$\underline{y}(t) = \underline{x}(t+c)$$

(Same values, different time -)



Then

$$\dot{\underline{y}}(t) = \frac{d}{dt} \underline{x}(t+c) = \dot{\underline{x}}(t+c)$$

$$= f(\underline{x}(t+c))$$

$$= f(\underline{y}(t)) \quad \checkmark$$

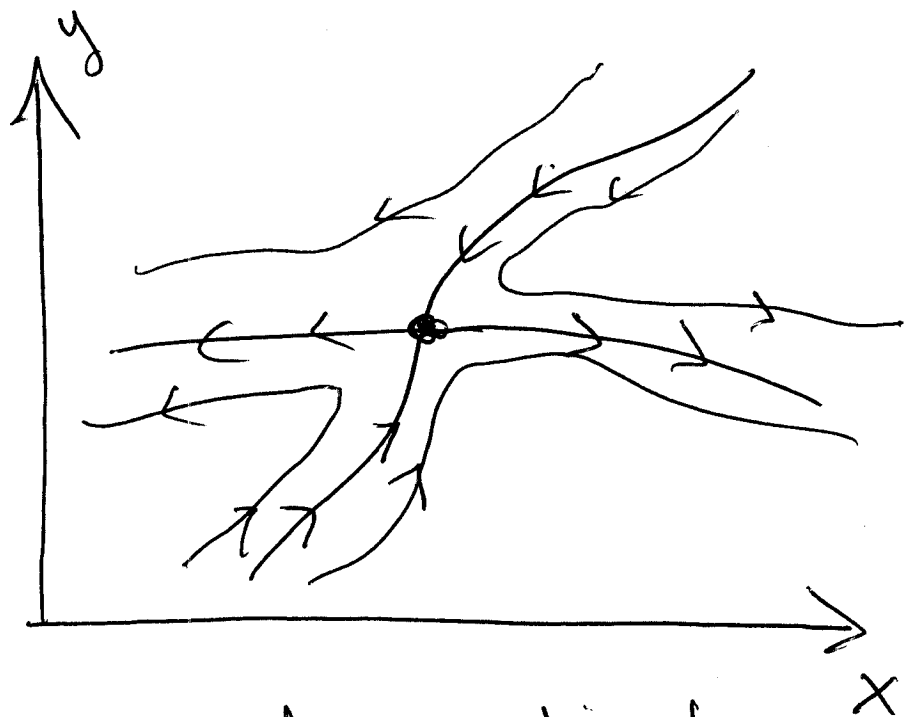
Note: This requires autonomous: Eg if

$$\dot{\underline{x}} = f(\underline{x}, t), \quad \underline{y}(t) = \underline{x}(t+c)$$

$$\dot{\underline{y}} = \dot{\underline{x}}(t+c) = f(\underline{x}(t+c), t+c) = f(\underline{y}(t), t+c) \neq f(\underline{y}(t), t) \quad !$$

• Defn: The phase portrait for soln's of  $\dot{x} = f(x)$  is the graph of solution trajectories in the  $xy$ -plane.

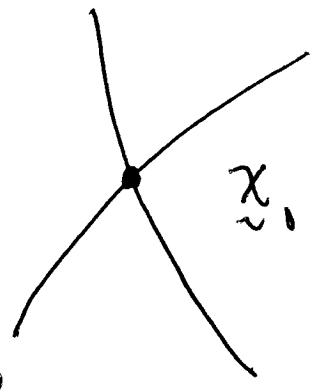
Defn: The trajectory of a solution  $x(t)$  is the image curve in the  $xy$ -plane



"time dependent suppressed - time direction indicated by arrows..."

Cor: Two <sup>different</sup> solution trajectories cannot cross in the  $xy$ -plane:

Pf Assume two solutions  $\underline{x}(t)$  &  $\underline{y}(t)$  cross at point  $(x_0, y_0) = \underline{x}_0$  in the  $xy$ -plane.



Then  $\underline{x}(t_0) = \underline{x}_0 = \underline{y}(t_1)$  at some  $t_0, t_1$ .

Now let  $c = t_1 - t_0$ . Then

$$\underline{z}(t) = \underline{y}(t + c) \Rightarrow \underline{z}(t_0) = \underline{y}(t_0 + t_1 - t_0) = \underline{y}(t_1) = \underline{x}_0$$

Thus: both  $\underline{x}(t)$  &  $\underline{z}(t)$  solve  $\dot{\underline{x}} = f(\underline{x})$

$$\dot{\underline{x}} = f(\underline{x})$$

$$\underline{x}(t_0) = \underline{x}_0$$

$\Rightarrow \exists!$  Thm  $\Rightarrow$  they must be same solution ✓

$\Rightarrow \underline{x}(\cdot)$  &  $\underline{y}(\cdot)$  had same trajectory

Defn: The phase portrait is the graph of the solution trajectories in the xy-plane.

Defn: Soln Trajectory is the curve in xy-plane given by image of  $\vec{x}(t)$ .

Con: trajectories can only "intersect" at rest pts (where solutions never touch.)



Conclude: Rest points are fundamental,

Simplest Autonomous Systems: Linear, Constant Coeff, Homogeneous

$$\dot{\vec{x}} = A\vec{x} = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} \quad (*)$$

Rest Point @  $\vec{x} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$ .

• Note: Solutions of (\*) form a vector space i.e., ~~Sol~~ solution space is closed under addition & scalar multiplication of solutions:

Eg: If  $\underline{x}(t), \underline{y}(t)$  solve  $\dot{\underline{x}} = A\underline{x}, \dot{\underline{y}} = A\underline{y}$

Then set  $\underline{z}(t) = c_1 \underline{x} + c_2 \underline{y}$

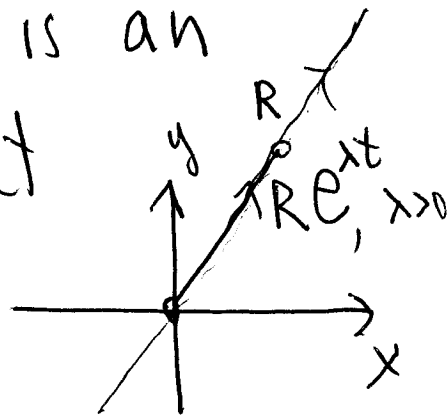
$$\begin{aligned} \dot{\underline{z}}(t) &= c_1 \dot{\underline{x}}(t) + c_2 \dot{\underline{y}}(t) \\ &= c_1 A \underline{x}(t) + c_2 A \underline{y}(t) \\ &= A (c_1 \underline{x}(t) + c_2 \underline{y}(t)) \\ &= A \underline{z}(t) \checkmark \end{aligned}$$

In particular: if  $\dot{\underline{x}} = A\underline{x}$ , the  $-\dot{\underline{x}} = A(-\underline{x})$  is also a solution.

## Classification of Rest Points:

Main Theorem: If  $(\lambda, R)$  is an eigenpair of matrix  $A$  so that

$$AR = \lambda R,$$



Then  $\underline{x}(t) = R e^{\lambda t}$  solves  $\dot{\underline{x}} = A \underline{x}$ .

Proof: Let  $\underline{x}(t) = R e^{\lambda t}$ . Then

$$\dot{\underline{x}}(t) = R \lambda e^{\lambda t}$$

$$A \underline{x}(t) = A R e^{\lambda t} = R \lambda e^{\lambda t} \quad \checkmark$$

Note: The eigen solutions tend to rest point  $\underline{x} = (0, 0)$  in forward time when  $\lambda < 0$ , and in backward time when  $\lambda > 0$ .

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Conclude: To solve  $\dot{\underline{x}} = A\underline{x}$ , find  
two eigensolutions

$$\underline{x}_1 = R_1 e^{\lambda_1 t}$$

$$\underline{x}_2 = R_2 e^{\lambda_2 t}$$

& general solution is:

$$\underline{x}(t) = C_1 R_1 e^{\lambda_1 t} + C_2 R_2 e^{\lambda_2 t}$$

Q: How do we know this is all soln's?

Ans:  $\exists!$  soln to  $\dot{\underline{x}} = A\underline{x}$

$$\underline{x}(0) = \underline{x}_0 = \begin{bmatrix} x_0 \\ y_0 \end{bmatrix}$$

$$\text{So } \underline{x}(t) = C_1 R_1 e^{\lambda_1 t} + C_2 R_2 e^{\lambda_2 t}$$

$$\underline{x}(0) = C_1 R_1 + C_2 R_2 = \begin{bmatrix} R_1 & R_2 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} C_1 \\ C_2 \end{bmatrix} = \begin{bmatrix} x_0 \\ y_0 \end{bmatrix}$$

Can always solve if  $R_1, R_2$  form a basis.



(8)

We have: Theorem: If  $(\lambda_1, R_1)$   $(\lambda_2, R_2)$  are eigenpairs of matrix  $A$ , and  $R_1, R_2$  are linearly independent so that

$$\det \begin{bmatrix} R_1 & R_2 \\ 1 & 1 \end{bmatrix} \neq 0,$$

then the general solution of

$$\dot{\underline{x}} = A \underline{x}$$

is

$$\underline{x}(t) = c_1 R_1 e^{\lambda_1 t} + c_2 R_2 e^{\lambda_2 t}, \quad \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} R_1 & R_2 \\ 1 & 1 \end{bmatrix}^{-1} \begin{bmatrix} x_0 \\ y_0 \end{bmatrix}$$

Proof: The unique soln of the initial value problem is  $\dot{\underline{x}} = A \underline{x}$  is  $\begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} R_1 & R_2 \\ 1 & 1 \end{bmatrix}^{-1} \begin{bmatrix} x_0 \\ y_0 \end{bmatrix}$   
 $\underline{x}(0) = \underline{x}_0$

$\Rightarrow$  we can solve every ivp  $\Rightarrow$  we have every solution!

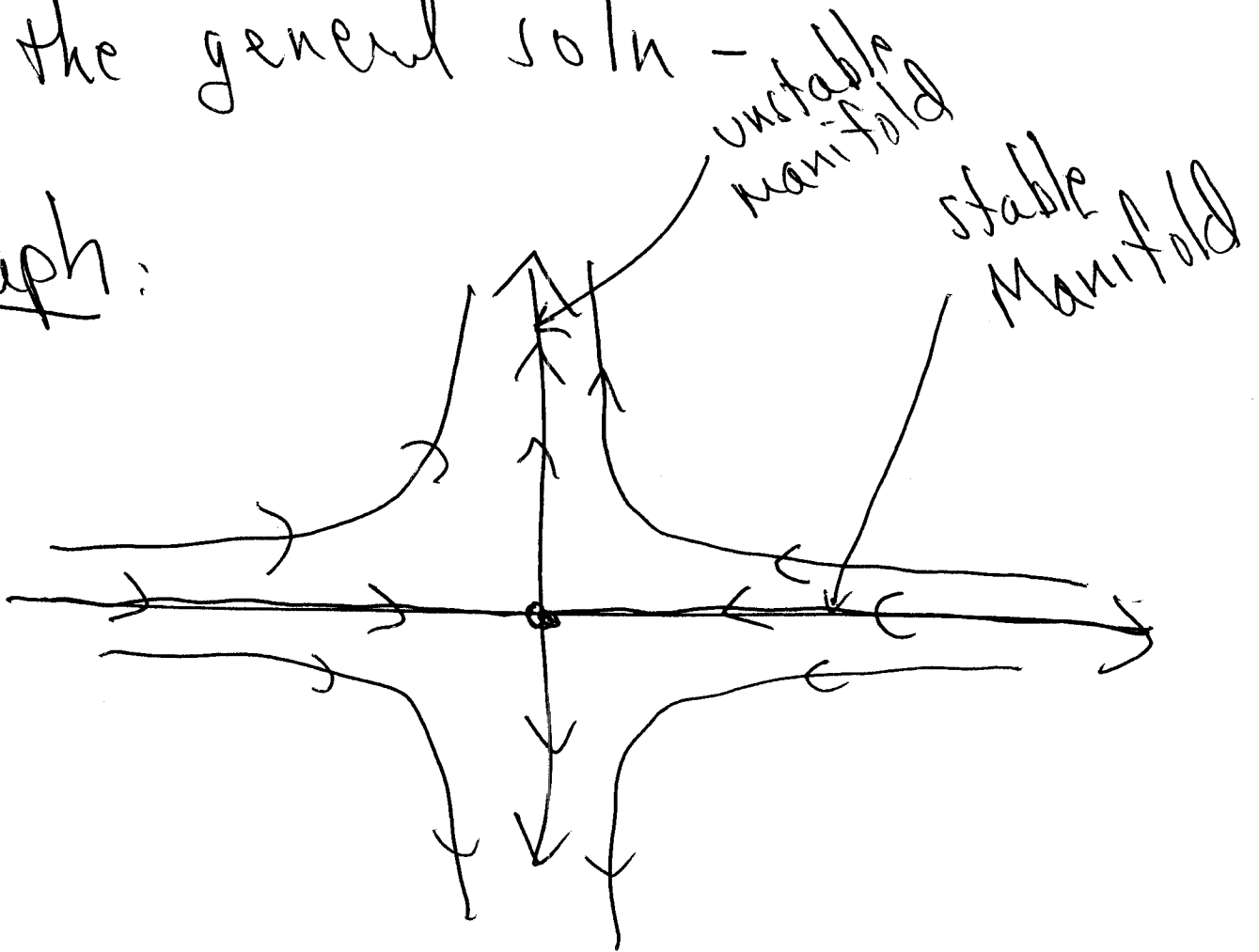
2) Ex:  $\dot{\underline{x}} = \begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$

$(-1, \begin{pmatrix} 1 \\ 0 \end{pmatrix})$  &  $(1, \begin{pmatrix} 0 \\ 1 \end{pmatrix})$  are eigenpairs -

$$\underline{x}(t) = c_1 \begin{pmatrix} 1 \\ 0 \end{pmatrix} e^{-t} + c_2 \begin{pmatrix} 0 \\ 1 \end{pmatrix} e^t$$

is the general soln -

Graph:



# Canonical Cases -

(1)  $A = \begin{bmatrix} \lambda & 0 \\ 0 & \mu \end{bmatrix}$  distinct eigenvalues, same sign  
 $\lambda, \mu > 0, \lambda, \mu < 0$

(2)  $A = \begin{bmatrix} \lambda & 0 \\ 0 & \mu \end{bmatrix}$  distinct eigenvalues, opp sign  
 $\lambda < 0 < \mu$

(3)  $\begin{bmatrix} \alpha & B \\ -B & \alpha \end{bmatrix}$  complex evals  $\lambda = \alpha \pm i\beta$

(4)  $\begin{bmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix}$  coincident evals - basis  
 $(\underline{x}(t) = \vec{v} e^{\lambda t} \text{ is soln})$  of e-vectors

(5)  $\begin{bmatrix} \lambda & 1 \\ 0 & \lambda \end{bmatrix}$  degenerate core - one  
 e-value, one e-vector

Note:  $\lambda = 0 \Rightarrow$  2nd order isn't leady term?  
 $\Rightarrow$  not so interesting ~ degenerate

$$\lambda^2 - \underbrace{\text{tr}A}_{\tau} \lambda + \underbrace{\det A}_{\Delta}$$

$$\lambda = \frac{\tau \pm \sqrt{\tau^2 - 4\Delta}}{2}$$

$\tau^2 - 4\Delta = 0$  where  
eigenvalues go complex

