# Solutions to Systems of Linear Differential Equations (DE)

An *n*-dimensional **linear first-order DE system** is one that can be written as a

matrix vector equation -

$$\vec{X}'(t) = A(t)\vec{X}(t) + \vec{f}(t)$$

 $\vec{X}'(t) = A(t)\vec{X}(t) + \vec{f}(t)$  A(t) is an  $n \times n$  matrix  $\vec{X}(t)$  and  $\vec{f}(t)$  are  $n \times 1$  vectors

If  $\vec{f}(t) \equiv \vec{0}$ , the system is **homogenous**, i.e.  $\vec{X}'(t) = A(t)\vec{X}(t)$ 

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Example: 
$$x' = 3x - 2y$$

$$y' = x$$

$$z' = -x + y + 3z$$

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$$\overrightarrow{X'} = \begin{bmatrix} 3 & -2 & 0 \\ 1 & 0 & 0 \\ -1 & 1 & 3 \end{bmatrix} \overrightarrow{X}$$

$$\overrightarrow{X} = \begin{bmatrix} x \\ y \\ z \end{bmatrix}$$

It may be easily verified that 
$$\vec{x}_h = \begin{pmatrix} 2e^{2t} \\ e^{2t} \\ e^{2t} \end{pmatrix}$$
 is a solution to the system  $\begin{aligned} x' &= 3x - 2y \\ y' &= x \\ z' &= -x + y + 3z \end{aligned}$ 

Actually, it can be easily verified that  $\begin{pmatrix} 0 \\ 0 \\ e^t \end{pmatrix}$  and  $\begin{pmatrix} e^t \\ e^t \\ 0 \end{pmatrix}$  are also solutions to the same  $\vec{X}_h' = A\vec{X}_h$ 

Similarly, for the **non**homogenous ODE

$$x' = 3x - 2y + 2 - 2e^{t} 
 y' = x - e^{t} 
 z' = -x + y + 3z + e^{t} - 1$$

$$\overrightarrow{X}' = \begin{bmatrix} 3 & -2 & 0 \\ 1 & 0 & 0 \\ -1 & 1 & 3 \end{bmatrix} \overrightarrow{X} + \begin{bmatrix} 2 - 2e^{t} \\ -e^{t} \\ e^{t} - 1 \end{bmatrix}$$

$$\overrightarrow{X}_{P} = \begin{bmatrix} e^{t} \\ 1 \\ 0 \end{bmatrix}$$

linear combinations of these will also be solutions

$$\vec{X}_P = \begin{pmatrix} e^t \\ 1 \\ 0 \end{pmatrix}$$

**Particular solution** of the system CHECK!

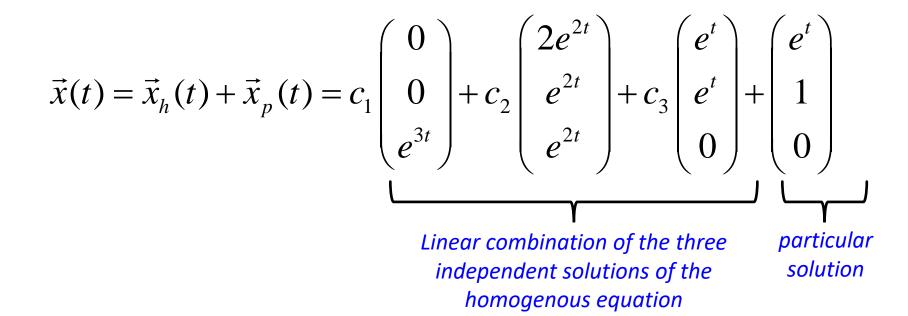
### The Superposition Principle for Homogenous Linear DE Systems

If  $\vec{x}_1(t), \vec{x}_2(t), \dots, \vec{x}_n(t)$  are linearly independent solutions to the homogenous equation  $\vec{X}'(t) = A(t)\vec{X}(t)$  then any linear combinations of these, i.e.

$$c_1\vec{x}_1(t) + c_2\vec{x}_2(t) + \dots + c_n\vec{x}_n(t)$$

is also a solution to that equation for any set of real constants  $c_1, c_2, \ldots, c_n$ 

Using this Superposition Principle and the homogenous and particular solutions obtained earlier -



We need to show that 
$$\vec{x}_1 = \begin{pmatrix} 0 \\ 0 \\ e^{3t} \end{pmatrix}$$
,  $\vec{x}_2 = \begin{pmatrix} 2e^{2t} \\ e^{2t} \\ e^{2t} \end{pmatrix}$ ,  $\vec{x}_3 = \begin{pmatrix} e^t \\ e^t \\ 0 \end{pmatrix}$  are linearly independent on  $(-\infty, \infty)$ 

**Step 1:** Choose a point, say  $t_0 = 0 \in (-\infty, \infty)$ 

**Step 2:** Calculate  $\vec{x}_1(t_0), \vec{x}_2(t_0), \vec{x}_3(t_0)$  and form the column space matrix

$$C = \begin{pmatrix} 0 & 2 & 1 \\ 0 & 1 & 1 \\ 1 & 1 & 0 \end{pmatrix}$$

The columns of  $\boldsymbol{c}$  are obviously independent but we will confirm that in the next slide by computing rref(c)

**Step 3:** Test for linear independence of the columns of C by computing

$$rref(C) = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

Clearly, the column vectors of C must be linearly independent

Alternatively, this could have been shown by calculating and showing that  $\det(C) \neq 0$ 

In general, for a  $n \times n$  linear system, we need n linearly independent solutions  $\vec{X}_1(t), \vec{X}_2(t), \dots, \vec{X}_n(t)$  to form a basis for the solution space with the general solution to the homogenous system given by

$$\vec{X}_h = c_1 \vec{X}_1(t) + c_2 \vec{X}_2(t) + \dots + c_n \vec{X}_n(t)$$
  $c_1, c_1, \dots c_n \in \mathbb{R}$ 

### **Fundamental Matrix:**

Note that  $\overrightarrow{X_h}$  can also be expressed as follows -

$$\vec{x}_h(t) = c_1 \begin{pmatrix} 0 \\ 0 \\ e^{3t} \end{pmatrix} + c_2 \begin{pmatrix} 2e^{2t} \\ e^{2t} \\ e^{2t} \end{pmatrix} + c_3 \begin{pmatrix} e^t \\ e^t \\ 0 \end{pmatrix} \quad \text{OR} \quad \begin{pmatrix} 0 & 2e^{2t} & e^t \\ 0 & e^{2t} & e^t \\ e^{3t} & e^{2t} & 0 \end{pmatrix} \begin{pmatrix} c_1 \\ c_2 \\ c_3 \end{pmatrix}$$

$$\begin{pmatrix} \begin{vmatrix} 1 & 1 \\ \vec{X}_1 & \vec{X}_2 & \vec{X}_3 \\ | & | & | \end{pmatrix}$$

$$\vec{X}(t) \quad \vec{C}$$
Fundamental
Matrix

## Fundamental Matrix X(t) (continued)

(i)  $det(X(t)) \neq 0$ 

One can also show that X'(t) = AX(t)

(ii) The Fundamental Matrix is NOT unique

A different set of linearly independent solutions will produce a different X(t) but that  $\vec{x}_h = X(t)\vec{c}$  would hold

How do we find  $\vec{x}_h$  and  $\vec{x}_p$  for a System of Linear ODEs?

# Consider the Homogenous Solution $\vec{x}_h$ first, i.e. the solution of $\vec{X}' = A\vec{X}$

If we choose solutions of the form  $\vec{x} = e^{\lambda t} \vec{v}$ ,

then substituting in X'(t) = AX(t)

gives  $\lambda e^{\lambda t} \vec{v} = A e^{\lambda t} \vec{v}$ 

Factoring this, we get  $e^{\lambda t}(A - \lambda I)\vec{v} = \vec{0}$ 

Since  $e^{\lambda t}$  can never be zero, we need to find  $\lambda$  and  $\vec{v}$  such that  $(A - \lambda I)\vec{v} = \vec{0}$ 

But a scalar  $\lambda$  and a non-zero vector  $\vec{v}$  satisfying  $(A - \lambda I)\vec{v} = \vec{0}$  are the *eigenvalue* and *eigenvector* of the matrix A

Considering the eigenvalues of A, we will have three main cases –

- (i) Distinct Real Eigenvalues
- (ii) Repeated Real Eigenvalues
- (iii) Complex Eigenvalues

for the eigenvalues of A in X'(t) = AX(t)

Case (i): X'(t) = AX(t) has real eigenvalues  $\lambda_1, \lambda_2, \dots, \lambda_n$   $\lambda_i \neq \lambda_j$  for  $i \neq j$  and the corresponding eigenvectors are  $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n$ 

Note that the eigenvalues are not repeated and, therefore, n independent eigenvectors can be found

For this case, the **General Homogenous Solution** is of the form –

$$\vec{x}(t) = c_1 e^{\lambda_1 t} \vec{v}_1 + c_2 e^{\lambda_2 t} \vec{v}_2 + \dots + c_n e^{\lambda_n t} \vec{v}_n$$

Note that in the case of repeated eigen values, i.e.  $\lambda_i = \lambda_j$   $i \neq j$ , we will need either independent eigenvectors or generalized eigenvectors, as discussed later

Example

Consider the following system of ODEs with initial conditions x(0) = 3, y(0) = 1

$$\frac{dx}{dt} = -2x + y$$

$$\frac{dy}{dt} = x - 2y$$

$$\vec{X}' = \begin{pmatrix} -2 & 1\\ 1 & -2 \end{pmatrix} \vec{X}; \quad \vec{X}(0) = \begin{pmatrix} 3\\ 1 \end{pmatrix}$$

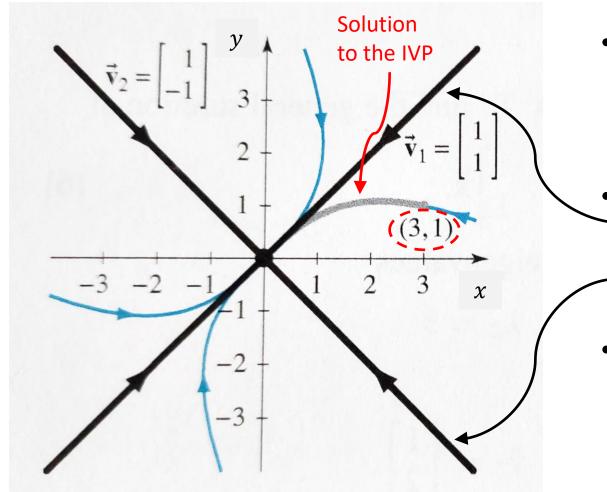
For this, eigenvalues are  $\lambda_1 = -1$ ,  $\lambda_2 = -3$  and eigenvectors  $\vec{v}_1 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ ,  $\vec{v}_2 = \begin{pmatrix} 1 \\ -1 \end{pmatrix}$ 

General Solution: 
$$\vec{x}(t) = c_1 e^{-t} \begin{pmatrix} 1 \\ 1 \end{pmatrix} + c_2 e^{-3t} \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

Using the given initial condition  $\vec{X}(0) = {3 \choose 1} = c_1 {1 \choose 1} + c_2 {1 \choose -1} \Rightarrow c_1 = 2, c_2 = 1$ 

$$\vec{x}(t) = 2e^{-t} \begin{pmatrix} 1 \\ 1 \end{pmatrix} + e^{-3t} \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

Alternatively, 
$$\vec{x}(t) = X(t)\vec{c} = \begin{pmatrix} e^{-t} & e^{-3t} \\ e^{-t} & -e^{-3t} \end{pmatrix} \begin{pmatrix} 2 \\ 1 \end{pmatrix} = \begin{pmatrix} 2e^{-t} + e^{-3t} \\ 2e^{-t} - e^{-3t} \end{pmatrix}$$



**Phase Portrait** 

(Stable Equilibrium at origin, solution from (3,1) in grey)

 Trajectories move towards or away from the equilibrium according to the sign of the eigenvalues (-ive or +ive) associated with the eigenvectors

Along each eigenvector is a unique trajectory called a **SEPRATRIX** that separates the trajectories curving one way from those curving the other way

The equilibrium occurs at the origin and the phase portrait is symmetric about this point

Case (ii): X'(t) = AX(t) with repeated eigenvalues  $\lambda_1, \lambda_2 = \lambda$  and with only one eigenvector  $\vec{v}$ 

Construct an **additional linear independent vector**  $\overrightarrow{u}$  as follows

Step (i): Find  $\vec{v}$  corresponding to  $\lambda$ 

Step (ii) Find a new  $\vec{u} \neq \vec{0}$  such that  $(A - \lambda I)\vec{u} = \vec{v}$ 

Step (iii) With these, try  $\vec{x}(t) = c_1 e^{\lambda t} \vec{v} + c_2 e^{\lambda t} (t \vec{v} + \vec{u})$ 

 $\overrightarrow{u}$  is referred to as the **Generalized Eigenvector** of A

But it is not really an eigenvector as  $A\vec{u} \neq \hat{\lambda} \vec{u}$ 

# Why this approach works?

Let  $\overrightarrow{X}_2(t) = e^{\lambda t}(t\overrightarrow{v} + \overrightarrow{u})$  where we are given that

(a) eigenvalue  $\lambda$  and eigenvector  $\vec{v}$  satisfy  $(A-\lambda I)\vec{v}=\vec{0}$  and (b)  $\vec{X}_1(t)=e^{\lambda t}\vec{v}$  is a solution of  $\vec{X}'=A\vec{X}$ , i.e.  $\vec{X}_1'=A\vec{X}_1$ 

Show that  $\overrightarrow{X}_2' = A\overrightarrow{X}_2$  if we can find  $\overrightarrow{u}$  such that  $(A - \lambda I)\overrightarrow{u} = \overrightarrow{v}$ 

Substituting,  $e^{\lambda t}(\vec{v} + \lambda t I \vec{v} + \lambda I \vec{u}) = e^{\lambda t}(t A \vec{v} + A \vec{u})$  and equating the coefficients of  $t e^{\lambda t}$  and  $e^{\lambda t}$  on the LHS and RHS of this equation, we get –

- 1. Coefficient of  $te^{\lambda t}$ :  $(A \lambda I) \vec{v} = \vec{0}$  This is the original eigenvalue equation that we already had
- 2. Coefficient of  $e^{\lambda t}$ :  $(A \lambda I)\vec{u} = \vec{v}$  We need to solve this to find  $\vec{u}$  and use it to find  $\vec{X}_2(t) = e^{\lambda t}(t\vec{v} + \vec{u})$

Example: Consider 
$$\vec{X}' = A\vec{X} = \begin{pmatrix} 2 & -1 \\ 4 & 6 \end{pmatrix} \vec{X}$$

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 Eigenvalue  $\lambda = 4$  (repeated)

Eigenvector  $\vec{v} = \begin{pmatrix} 1 \\ -2 \end{pmatrix}$ 

One solution  $\vec{x}_1(t) = e^{4t} \begin{pmatrix} 1 \\ -2 \end{pmatrix}$ 

If we follow the earlier approach of Lecture 1 of Module 3 then we should try our second solution as  $\vec{x}_2(t) = te^{4t}\vec{v}$ . However, substituting this  $\vec{x}_2(t)$  in  $\vec{X}' = A\vec{X}$ , we find that this does not work!

> See Example 6, pg. 363 of Farlow textbook

Example: Consider 
$$\vec{X}' = A\vec{X} = \begin{pmatrix} 2 & -1 \\ 4 & 6 \end{pmatrix} \vec{X}$$
 Eigenvalue  $\lambda = 4$  (repeated)

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One solution  $\vec{x}_1(t) = e^{4t} \begin{pmatrix} 1 \\ -2 \end{pmatrix}$ 

Eigenvalue  $\lambda = 4$  (repeated)

Instead, we try a Generalized Eigenvector  $\vec{u}$  such that  $\vec{x}_2(t) = e^{4t}(t\vec{v} + \vec{u})$  is a solution to  $\vec{x}_2' = A\vec{x}_2$ 

This can be simplified to (1)  $(A-4I)\vec{v}=\vec{0}$  and (2)  $(A-4I)\vec{u}=\vec{v}$  by equating the coefficients of  $e^{4t}$  and  $te^{4t}$  on both sides of  $\vec{x}_2' = A\vec{x}_2$ 

Here (1) is the original eigenvalue equation for  $\lambda = 4$  and  $\vec{v} = \begin{pmatrix} 1 \\ -2 \end{pmatrix}$  and will not give us anything new

For (2), 
$$(A - 4I)\vec{u} = \vec{v} \implies \begin{pmatrix} -2 & -1 \\ 4 & 2 \end{pmatrix} \begin{pmatrix} u_1 \\ u_2 \end{pmatrix} = \begin{pmatrix} 1 \\ -2 \end{pmatrix} \implies 2u_1 + u_2 = -1$$

Choosing 
$$u_1 = K(say) \Rightarrow u_2 = -2K-1$$
 or  $\vec{u} = \begin{pmatrix} u_1 \\ u_2 \end{pmatrix} = K\begin{pmatrix} 1 \\ -2 \end{pmatrix} + \begin{pmatrix} 0 \\ -1 \end{pmatrix}$ 

Therefore, 
$$\vec{x}_2(t) = te^{4t} \begin{pmatrix} 1 \\ -2 \end{pmatrix} + Ke^{4t} \begin{pmatrix} 1 \\ -2 \end{pmatrix} + e^{4t} \begin{pmatrix} 0 \\ -1 \end{pmatrix} \Rightarrow \vec{x}_2(t) = te^{4t} \begin{pmatrix} 1 \\ -2 \end{pmatrix} + e^{4t} \begin{pmatrix} 0 \\ -1 \end{pmatrix}$$

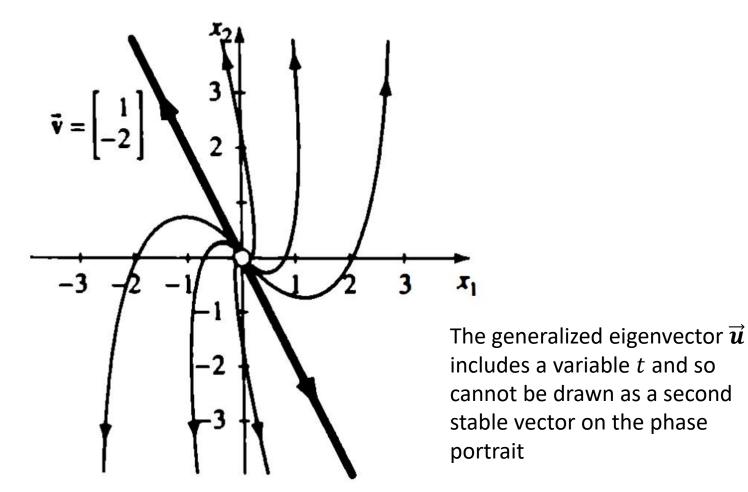
We drop the middle term as that is just a multiple of our first solution

The two solutions are then -

$$\vec{x}_1(t) = e^{4t} \begin{pmatrix} 1 \\ -2 \end{pmatrix}$$
 and 
$$\vec{x}_2(t) = e^{4t} \begin{pmatrix} t \\ -2t - 1 \end{pmatrix}$$

#### **Final Solution**

$$\vec{x}(t) = c_1 e^{4t} \begin{pmatrix} 1 \\ -2 \end{pmatrix} + c_2 e^{4t} \begin{pmatrix} t \\ -2t - 1 \end{pmatrix}$$



Phase Portrait with

- (a) Unstable Equilibrium at the origin
- (b) Double Eigenvalue at  $\lambda_1 = \lambda_2 = 4$
- (c) A single eigenvector

## Subsequent Lectures:

- (i) Complex Eigenvalues
- (ii) Particular solutions  $\vec{X}_p$  for systems of linear ODEs
- (iii) Phase Portraits and Stability Analysis