Math 2280-001 Week 11 Mar 27-29 (Exam 2 on Mar 31)

Mon Mar 27: Use last Friday's notes to discuss matrix exponentials

Wed Mar 29: 5.6-5.7 Matrix exponentials, linear systems, and variation of parameters for inhomogeneous systems.

Recall: For the first order system

$$\underline{x}'(t) = A \underline{x}$$

- $\Phi(t)$  is a fundamental matrix (FM) if its *n* columns are a basis for the solution space to the first order system above (i.e.  $\Phi(t)$  is the Wronskian matrix for a basis to the solution space).
- $\Phi(t)$  is an FM if and only if  $\Phi'(t) = A \Phi$  and  $\Phi(0)$  is invertible.
- $e^{tA}$  is the unique matrix solution to

$$X'(t) = AX$$
$$X(0) = I$$

and may be computed either of two ways:

$$e^{tA} = \Phi(t)\Phi(0)^{-1}$$

where  $\Phi(t)$  is any other FM, or via the infinite series

$$e^{tA} = I + tA + \frac{t^2}{2!}A^2 + \frac{t^3}{3!}A^3 + \dots + \frac{t^k}{k!}A^k + \dots$$

Example 1: We showed that if

$$\Lambda = \begin{bmatrix} \lambda_1 & 0 & 0 & \dots & 0 \\ 0 & \lambda_2 & 0 & \dots & 0 \\ \vdots & \vdots & & & \vdots \\ 0 & 0 & \dots & \lambda_n \end{bmatrix}$$

then

$$\mathbf{e}^{t\,\Lambda} = \begin{bmatrix} 1 + t\,\lambda_1 + \frac{t^2}{2!}\,\lambda_1^2 + \dots & 0 & 0 & \dots & 0 \\ 0 & 1 + t\,\lambda_2 + \frac{t^2}{2!}\,\lambda_2^2 + \dots & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & & \vdots & & \vdots \\ 0 & 0 & \dots & 1 + t\,\lambda_n + \frac{t^2}{2!}\,\lambda_n^2 + \dots \end{bmatrix}$$

=

$$\begin{bmatrix} e^{t\lambda_1} & 0 & \dots & 0 \\ 0 & e^{t\lambda_2} & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & e^{t\lambda_n} \end{bmatrix}$$

Example 2) If A is diagonalizable we showed that we got the same answer for  $e^{tA}$  using  $A = SAS^{-1}$ 

and the Taylor series method, as we did we did using the  $\Phi(t)\Phi(0)^{-1}$  method. In fact,  $\Phi(t) = S e^{t\Lambda}$  and  $\Phi(0)^{-1} = S^{-1}$ , so this makes sense.

How to compute e<sup>At</sup> when A is not diagonalizable. This method depends on the fundamental fact about how the generalized eigenspaces of a matrix fit together.

"Recall" (this is really linear algebra material, but most of you haven't seen it.)

For  $A_{n \times n}$  let the characteristic polynomial  $p(\lambda) = det(A - \lambda I)$  factor as

$$p(\lambda) = (-1)^n (\lambda - \lambda_1)^{k_1} (\lambda - \lambda_2)^{k_2} ... (\lambda - \lambda_m)^{k_m}$$

 $p(\lambda) = (-1)^n \left(\lambda - \lambda_1\right)^{k_1} \left(\lambda - \lambda_2\right)^{k_2} ... \left(\lambda - \lambda_m\right)^{k_m}$  Any eigenspace of A for which  $dim\left(E_{\lambda_j}\right) < k_j$  is called <u>defective</u>. If A has any defective eigenspaces then

it is not diagonalizable. (If none of the eigenspaces are defective, then my amalgamating bases for each eigenspace one obtains a basis for  $\mathbb{R}^n$  (or  $\mathbb{C}^n$ , in the case of complex eigendata). However, the larger generalized eigenspaces  $G_{\lambda_i}$  defined by

$$E_{\lambda_{j}} \subseteq G_{\lambda_{j}} := nullspace \left( \left( A - \lambda_{j} I \right)^{k_{j}} \right)$$

do always have dimension  $k_j$ . If bases for each  $G_{\lambda_j}$  are amalgamated they will form a basis for  $\mathbb{R}^n$  or  $\mathbb{C}^n$ .

For each basis vector  $\underline{\mathbf{v}}$  of  $G_{\lambda_i}$  one can construct a basis solution x(t) to x' = Ax as follows:

$$\underline{\boldsymbol{x}}(t) = e^{At}\underline{\boldsymbol{y}} = e^{\lambda_{j}^{T}t} e^{\left(A - \lambda_{j}^{T}\right)t}\underline{\boldsymbol{y}}$$

$$= e^{\lambda_{j}^{T}t} \left(I + t\left(A - \lambda_{j}^{T}I\right) + \frac{t^{2}}{2!} \left(A - \lambda_{j}^{T}I\right)^{2} + \dots\right)\underline{\boldsymbol{y}}$$

$$= e^{\lambda_{j}^{T}t} \left(\underline{\boldsymbol{y}} + t\left(A - \lambda_{j}^{T}I\right)\underline{\boldsymbol{y}} + \frac{t^{2}}{2!} \left(A - \lambda_{j}^{T}I\right)^{2}\underline{\boldsymbol{y}} + \dots\right)$$

$$= e^{\lambda_{j}^{T}t} \left(\underline{\boldsymbol{y}} + t\left(A - \lambda_{j}^{T}I\right)\underline{\boldsymbol{y}} + \frac{t^{2}}{2!} \left(A - \lambda_{j}^{T}I\right)^{2}\underline{\boldsymbol{y}} + \frac{t^{2}}{\left(k_{j}^{T} - 1\right)!} \left(A - \lambda_{j}^{T}I\right)^{k_{j}^{T} - 1}}\underline{\boldsymbol{y}} + 0 + 0 + \dots\right)$$

$$\underline{\boldsymbol{x}}(t) = e^{\lambda_{j}^{T}t} \left(\underline{\boldsymbol{y}} + t\left(A - \lambda_{j}^{T}I\right)\underline{\boldsymbol{y}} + \frac{t^{2}}{2!} \left(A - \lambda_{j}^{T}I\right)^{2}\underline{\boldsymbol{y}} + \frac{t^{2}}{\left(k_{j}^{T} - 1\right)!} \left(A - \lambda_{j}^{T}I\right)^{k_{j}^{T} - 1}}\underline{\boldsymbol{y}}\right).$$

Notes: The final sum is a finite sum because  $\underline{\mathbf{v}} \in nullspace\left(\left(A - \lambda_j I\right)^{\kappa_j}\right)$ ! If  $\underline{\mathbf{v}}$  was an eigenvector, you've just reconstructed

$$\underline{\boldsymbol{x}}(t) = \mathbf{e}^{\lambda_j t} \underline{\boldsymbol{y}}$$

Use the *n* independent solutions found this way to construct a  $\Phi(t)$ , and compute  $e^{At} = \Phi(t)\Phi(0)^{-1}$ .

Exercise 1 Find  $e^{At}$  for the matrix in the system:

$$\begin{bmatrix} x_1' \\ x_2' \end{bmatrix} = \begin{bmatrix} 3 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

tech check:

> with(LinearAlgebra):

$$A := \left[ \begin{array}{cc} 3 & -1 \\ 1 & 1 \end{array} \right];$$

 $factor(Determinant(A - \lambda \cdot IdentityMatrix(2)));$ 

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

$$(\lambda - 2)^2$$
(1)

So the only eigenvalue is  $\lambda = 2$ .

>  $B := A - 2 \cdot IdentityMatrix(2);$ NullSpace(B);

$$B := \begin{bmatrix} 1 & -1 \\ 1 & -1 \end{bmatrix}$$

$$\left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}$$
(2)

but  $E_{\lambda=2}$  is only one-dimensional. However, the generalized eigenspace  $G_{\lambda=2} = nullspace(A-2I)^2$  will be two dimensional:

> NullSpace  $(B^2)$ ;

$$\left\{ \left[\begin{array}{c} 0\\1 \end{array}\right], \left[\begin{array}{c} 1\\0 \end{array}\right] \right\} \tag{3}$$

Use this generalized nullspace basis to construct a basis of solutions to x' = Ax and use the resulting  $\Phi(t)$  to construct the matrix exponential...

>  $xI := t \rightarrow e^{2 \cdot t} \cdot (IdentityMatrix(2) + t \cdot B) \cdot \begin{bmatrix} 1 \\ 0 \end{bmatrix}$ :

xl(t);

 $x2 := t \rightarrow e^{2 \cdot t} \cdot \left( IdentityMatrix(2) + t \cdot B + \frac{t^2}{2} \cdot B^2 \right) \cdot \begin{bmatrix} 0 \\ 1 \end{bmatrix} :$ 

x2(t);

 $\Phi := t \rightarrow \langle x I(t) | x 2(t) \rangle$ :

 $\Phi(t)$ 

 $\Phi(t).\Phi(0)^{-1}$ ;

MatrixExponential( $t \cdot A$ ); #these last two should be the same!! In fact, the last three in this case

$$\begin{bmatrix} e^{2t} (1+t) \\ t e^{2t} \end{bmatrix}$$

$$\begin{bmatrix} -te^{2t} \\ (1-t)e^{2t} \end{bmatrix}$$

$$\begin{bmatrix} e^{2t} (1+t) & -te^{2t} \\ te^{2t} & (1-t)e^{2t} \end{bmatrix}$$

$$\begin{bmatrix} e^{2t} (1+t) & -te^{2t} \\ te^{2t} & (1-t)e^{2t} \end{bmatrix}$$

$$\begin{bmatrix} e^{2t} (1+t) & -te^{2t} \\ te^{2t} & (1-t)e^{2t} \end{bmatrix}$$

$$\begin{bmatrix} e^{2t} (1+t) & -te^{2t} \\ te^{2t} & -e^{2t} (-1+t) \end{bmatrix}$$
(4)

<u>Variation of parameters:</u> This is what fundamental matrices and matrix exponentials are especially good for....they let you solve non-homogeneous systems without guessing. Consider the non-homogeneous first order system

$$\underline{\boldsymbol{x}}'(t) = P(t)\underline{\boldsymbol{x}} + \boldsymbol{f}(t)$$
 \*

Let  $\Phi(t)$  be an FM for the homogeneous system

$$\underline{\boldsymbol{x}}'(t) = P(t)\underline{\boldsymbol{x}}.$$

Since  $\Phi(t)$  is invertible for all t we may do a change of functions for the non-homogeneous system:

$$\underline{\boldsymbol{x}}(t) = \Phi(t)\underline{\boldsymbol{u}}(t)$$

plug into the non-homogeneous system (\*):

$$\Phi'(t)\underline{\boldsymbol{u}}(t) + \Phi(t)\underline{\boldsymbol{u}}'(t) = P(t)\Phi(t)\underline{\boldsymbol{u}}(t) + \boldsymbol{f}(t).$$

Since  $\Phi' = P \Phi$  the first terms on each side cancel each other and we are left with

$$\Phi(t)\underline{\boldsymbol{u}}'(t) = \boldsymbol{f}(t)$$
$$\underline{\boldsymbol{u}}' = \Phi^{-1}\boldsymbol{f}$$

which we can integrate to find a  $\underline{\boldsymbol{u}}(t)$ , hence an  $\underline{\boldsymbol{x}}(t) = \Phi(t)\underline{\boldsymbol{u}}(t)$ .

Remark: This is where the (mysterious at the time) formula for variation of parameters in  $n^{th}$  order linear DE's came from....

"Recall" (February 24 notes):

<u>Variation of Parameters:</u> The advantage of this method is that is always provides a particular solution, even for non-homogeneous problems in which the right-hand side doesn't fit into a nice finite dimensional subspace preserved by L, and even if the linear operator L is not constant-coefficient. The formula for the particular solutions can be somewhat messy to work with, however, once you start computing.

Here's the formula: Let  $y_1(x), y_2(x), ..., y_n(x)$  be a basis of solutions to the homogeneous DE

$$L(y) := y^{(n)} + p_{n-1}(x)y^{(n-1)} + \dots + p_1(x)y' + p_0(x)y = 0.$$

$$E(y, (x) + u_n(x)y_n(x) + \dots + u_n(x)y_n(x) \text{ is a particular solution to}$$

 $L(y) := y^{(n)} + p_{n-1}(x)y^{(n-1)} + \dots + p_1(x)y' + p_0(x)y = 0.$  Then  $y_p(x) = u_1(x)y_1(x) + u_2(x)y_2(x) + \dots + u_n(x)y_n(x)$  is a particular solution to

provided the coefficient functions (aka "varying parameters")  $u_1(x), u_2(x), ... u_n(x)$  have derivatives satisfying the Wronskian matrix equation

$$\begin{bmatrix} y_1 & y_2 & \dots & y_n \\ y_1' & y_2' & \dots & y_n' \\ \dots & \dots & \dots & \dots \\ y_1^{(n-1)} & y_2^{(n-1)} & \dots & y_n^{(n-1)} \end{bmatrix} \begin{bmatrix} u_1' \\ u_2' \\ \vdots \\ u_n' \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ f \end{bmatrix}$$

But if we convert the  $n^{th}$  order DE into a first order system for  $x_1 = y$ ,  $x_2 = y'$  etc. we have

$$\begin{split} x_1 & (=y) \\ x_1' = x_2 & (=y') \\ x_2' = x_3 & (=y'') \\ x_{n-1}' = x_n & (=y^{(n-1)}) \\ x_n' = & (=y^{(n)}) = -p_0(x)y_1 - p_1(x)y_2 - \dots - p_{n-1}(x)y_{n-1} + f. \end{split}$$

And each basis solution y(t) for L(y) = 0 gives a solution  $[y, y', y'', ...y^{(n-1)}]^T$  to the homogeneous system

$$\begin{bmatrix} x_1' \\ x_2' \\ x_3' \\ \vdots \\ x_n' \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 \\ 0 & 0 & 0 & 1 \dots & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ -p_0 & -p_1 & -p_2 & \dots & -p_{n-1} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_n \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 0 \\ \vdots \\ f \end{bmatrix}.$$

So the original Wronskian matrix for the  $n^{th}$  order linear homogeneous DE is a FM for the system above, so the formula we learned in Chapter 3 is a special case of the easier to understand one for first order systems that we just derived, namely

$$\Phi(t)\underline{\boldsymbol{\mu}}'(t) = \boldsymbol{f}(t)$$
$$\underline{\boldsymbol{\mu}}' = \Phi^{-1}\boldsymbol{f}$$

Returning to first order systems, if we want to solve an IVP for a first order system rather than find the complete general solution, then the following two ways are appropriate:

## 1) If you want to solve the IVP

$$\underline{\boldsymbol{x}}'(t) = P(t)\underline{\boldsymbol{x}} + \boldsymbol{f}(t)$$
$$\underline{\boldsymbol{x}}(0) = \underline{\boldsymbol{x}}_0$$

The the solution will be of the form  $\underline{x} = \Phi \underline{u}$  (where  $\underline{u}' = \Phi^{-1} f$  as above). Thus  $\underline{x}_0 = \Phi(0)\underline{u}_0$ 

so

$$\underline{\boldsymbol{u}}_0 = \Phi(0)^{-1}\underline{\boldsymbol{x}}_0.$$

Thus

$$\underline{\boldsymbol{u}}(t) = \underline{\boldsymbol{u}}_0 + \int_0^t \underline{\boldsymbol{u}}'(s) \, \mathrm{d}s$$

$$\underline{\boldsymbol{u}}(t) = \underline{\boldsymbol{u}}_0 + \int_0^t \underline{\boldsymbol{\Phi}}^{-1}(s) \boldsymbol{f}(s) \, \mathrm{d}s.$$

Then

$$\underline{\boldsymbol{x}}(t) = \boldsymbol{\Phi}(t)\underline{\boldsymbol{u}}(t)$$

$$\underline{\boldsymbol{x}}(t) = \boldsymbol{\Phi}(t) \left(\underline{\boldsymbol{u}}_0 + \int_0^t \underline{\boldsymbol{\Phi}}^{-1}(s)\boldsymbol{f}(s) \, \mathrm{d}s\right).$$

## 2) If you want to solve the special case IVP

$$\underline{\boldsymbol{x}}'(t) = A\,\underline{\boldsymbol{x}} + \boldsymbol{f}(t)$$
$$\underline{\boldsymbol{x}}(0) = \underline{\boldsymbol{x}}_0$$

where *A* is a constant matrix, you may derive a special case of the solution formula above just as we did in Chapter 1. This is sort of amazing!

$$\underline{\mathbf{x}}'(t) = A \, \underline{\mathbf{x}} + \mathbf{f}(t)$$

$$\underline{\mathbf{x}}'(t) - A \, \underline{\mathbf{x}} = \mathbf{f}(t)$$

$$e^{-tA}(\underline{\mathbf{x}}'(t) - A \, \underline{\mathbf{x}}) = e^{-tA}\mathbf{f}(t)$$

$$\frac{d}{dt} \left( e^{-tA}\underline{\mathbf{x}}(t) \right) = e^{-tA}\mathbf{f}(t) .$$

Integrate from 0 to *t*:

$$e^{-tA}\underline{x}(t) - \underline{x}_0 = \int_0^t e^{-sA}f(s) ds$$

Move the  $\underline{x}_0$  over and multiply both sides by  $e^{tA}$ :

$$\underline{\boldsymbol{x}}(t) = e^{tA} \left(\underline{\boldsymbol{x}}_0 + \int_0^t e^{-sA} \boldsymbol{f}(s) ds\right).$$

Exercise 2 Consider the non-homogeneous problem related to the homogeneous system in Exercise 1:

$$\begin{bmatrix} x_1' \\ x_2' \end{bmatrix} = \begin{bmatrix} 3 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} t \\ 0 \end{bmatrix}$$
$$\begin{bmatrix} x_1(0) \\ x_2(0) \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}.$$

Solve this system using the formula

$$\underline{\boldsymbol{x}}(t) = e^{tA} \left( \underline{\boldsymbol{x}}_0 + \int_0^t e^{-sA} \boldsymbol{f}(s) \, ds \right)$$

(One could also try undetermined coefficients, but variation of parameters requires no "guessing.")

Tech check: (The commands are sort of strange, but might help in your homework.)

> with(LinearAlgebra):

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

with (Einear Aigeora):
$$A := \begin{bmatrix} 3 & -1 \\ 1 & 1 \end{bmatrix} :$$

$$MatrixExponential(t \cdot A);$$

$$f := t \rightarrow \begin{bmatrix} t \\ 0 \end{bmatrix} :$$

$$x0 := \begin{bmatrix} 0 \\ 0 \end{bmatrix} :$$

$$\begin{bmatrix} e^{2t} (1+t) & -t e^{2t} \\ t e^{2t} & -e^{2t} (-1+t) \end{bmatrix}$$
 (5)

integrand :=  $s \rightarrow simplify(MatrixExponential(-s \cdot A).f(s))$ : #integrand in formula above integrand(t); #checking

$$\begin{bmatrix} -e^{-2t} (-1+t) t \\ -t^2 e^{-2t} \end{bmatrix}$$
 (6)

integrated := unapply(map(int, integrand(s), s = 0..t), t) : #"map" applies a function to each entry of an array... # "unapply" makes a function out of output

> integrated(t); #checking

$$\begin{bmatrix} \frac{1}{2} t^2 e^{-2t} \\ -\frac{1}{4} + \frac{1}{4} e^{-2t} + \frac{1}{2} t e^{-2t} + \frac{1}{2} t^2 e^{-2t} \end{bmatrix}$$
 (7)

>  $x := unapply(simplify(MatrixExponential(t \cdot A).(x0 + integrated(t))), t)$ : x(t); #checking answer

$$\frac{1}{4} t (e^{2t} - 1)$$

$$\frac{1}{4} e^{2t} (-1 + t) + \frac{1}{4} t + \frac{1}{4}$$
(8)

$$\frac{dsolve(\{xl'(t) = 3 \cdot xl(t) - x2(t) + t, x2'(t) = xl(t) + x2(t), xl(0) = 0, x2(0) = 0\});}{\{xl(t) = \frac{1}{4} t e^{2t} - \frac{1}{4} t, x2(t) = -\frac{1}{4} e^{2t} + \frac{1}{4} t + \frac{1}{4} t + \frac{1}{4} t e^{2t}\}}$$
(9)