

## Summary- Elements of Chapters 7

The goal of this chapter is to solve a linear system of differential equations (we will also be able to solve some special nonlinear systems).

### Special Nonlinear Systems

Given the general nonlinear system,  $\frac{dx}{dt} = f(x, y)$  and  $\frac{dy}{dt} = g(x, y)$ , we can find two kinds of solutions: Equilibrium solutions (more in Chapter 9), and solution curves (also known as integral curves) that are solutions to:

$$\frac{dy}{dx} = \frac{dy/dt}{dx/dt} = \frac{g(x, y)}{f(x, y)}$$

If we're lucky, this will simplify to a form from Chapter 2 (first order equations). Notice that with a linear system,  $x' = ax + by$ ,  $y' = cx + dy$ , then

$$\frac{dy}{dx} = \frac{cx + dy}{ax + by} = \frac{c + d\frac{y}{x}}{a + b\frac{y}{x}}$$

is at least homogeneous (but may also be something else).

### Linear Systems

We're talking about three ways to solve a linear system:

- Using  $dy/dx$ , as in the last section.
- Converting the system to a second order equation (then use Chapter 3 methods)
- Using eigenvalues and eigenvectors. This last form will also be used to do some analysis in Chapter 9.

We started with some basic matrix algebra- Be sure you know how to perform matrix-vector multiplication and matrix-matrix multiplication for  $2 \times 2$  matrices.

### Eigenvalues and Eigenvectors

1. Definition: Given an  $n \times n$  matrix  $A$ , if there is a constant  $\lambda$  and a non-zero vector  $\mathbf{v}$  so that

$$A\mathbf{v} = \lambda\mathbf{v}$$

then  $\lambda$  is an eigenvalue, and  $\mathbf{v}$  is an associated eigenvector.

**NOTE:** Eigenvectors are not unique. That is, if  $\mathbf{v}$  is an eigenvector for  $A$ , so is  $k\mathbf{v}$  (prove it!).

2. If you have not had linear algebra, the main point below is the right-most system of equations. It is important to remember that one!

$$A\mathbf{v} = \lambda\mathbf{v} \Leftrightarrow \begin{matrix} av_1 & +bv_2 & = & \lambda v_1 \\ cv_1 & +dv_2 & = & \lambda v_2 \end{matrix} \Leftrightarrow \begin{matrix} (a - \lambda)v_1 & +bv_2 & = & 0 \\ cv_1 & +(d - \lambda)v_2 & = & 0 \end{matrix} \quad (1)$$

This system has a non-trivial solution for  $v_1, v_2$  only if the determinant of coefficients is 0:

$$\begin{vmatrix} a - \lambda & b \\ c & d - \lambda \end{vmatrix} = 0$$

And this is the **characteristic equation**. We solve this for the eigenvalues:

$$\lambda^2 - (a + d)\lambda + (ad - bc) = 0 \quad \Leftrightarrow \lambda^2 - \text{Tr}(A)\lambda + \det(A) = 0$$

where  $\text{Tr}(A)$  is the trace of  $A$  (which we defined as  $a + d$ ). For each  $\lambda$ , we must go back and solve Equation (1).

3. Notation: Often it is easier to use the notation  $A - \lambda I$  to represent the matrix:

$$A - \lambda I = \begin{bmatrix} a & b \\ c & d \end{bmatrix} - \lambda \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} a - \lambda & b \\ c & d - \lambda \end{bmatrix}$$

- Using this notation, the characteristic equation becomes:  $|A - \lambda I| = 0$ .
- Using this notation, the eigenvector equation is:  $(A - \lambda I)\mathbf{v} = \mathbf{0}$
- The generalized eigenvector  $\mathbf{w}$  solves:  $(A - \lambda I)\mathbf{w} = \mathbf{v}$

## Solve $\mathbf{x}' = A\mathbf{x}$ using Eigenvectors/Eigenvalues

We make the ansatz:

$$\mathbf{x}(t) = e^{\lambda t}\mathbf{v} = e^{\lambda t} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} e^{\lambda t}v_1 \\ e^{\lambda t}v_2 \end{bmatrix}$$

We showed that this implies  $\lambda, \mathbf{v}$  must be an eigenvalue, eigenvector of the matrix  $A$ .

The eigenvalues are found by solving the characteristic equation:

$$\lambda^2 - \text{Tr}(A)\lambda + \det(A) = 0 \quad \lambda = \frac{\text{Tr}(A) \pm \sqrt{\Delta}}{2}$$

The solution is one of three cases, depending on  $\Delta$ :

- Real  $\lambda_1, \lambda_2$  with two eigenvectors,  $\mathbf{v}_1, \mathbf{v}_2$ :

$$\mathbf{x}(t) = C_1 e^{\lambda_1 t} \mathbf{v}_1 + C_2 e^{\lambda_2 t} \mathbf{v}_2$$

- Complex  $\lambda = a + ib$ ,  $\mathbf{v}$  (we only need one):

$$\mathbf{x}(t) = C_1 \text{Re}(e^{\lambda t} \mathbf{v}) + C_2 \text{Im}(e^{\lambda t} \mathbf{v})$$

*Computational Note:* As in Chapter 3, our solutions here are real solutions- That means you should not have an  $i$  in your final answer.

- One eigenvalue, one eigenvector  $\mathbf{v}$ . Get  $\mathbf{w}$  that solves  $(A - \lambda I)\mathbf{w} = \mathbf{v}$ . Then:

$$\mathbf{x}(t) = e^{\lambda t} (C_1 \mathbf{v} + C_2 (t\mathbf{v} + \mathbf{w}))$$

*Computational Note:* You should find that there are an infinite number of possible vectors  $\mathbf{w}$ - Just choose one convenient representative.

You might find this helpful- Below there is a chart comparing the solutions from Chapter 3 to the solutions in Chapter 7:

	Chapter 3	Chapter 7
DE:	$ay'' + by' + cy = 0$	$\mathbf{x}'(t) = A\mathbf{x}(t)$
Ansatz:	$y = e^{rt}$	$\mathbf{x}(t) = e^{\lambda t}\mathbf{v}$
Characteristic Equation:	$ar^2 + br + c = 0$	$\lambda^2 - \text{Tr}(A)\lambda + \det(A) = 0$
Case 1:	$\Delta > 0:$ $y(t) = C_1 e^{r_1 t} + C_2 e^{r_2 t}$	$\Delta > 0:$ $\mathbf{x}(t) = C_1 e^{\lambda_1 t} \mathbf{v}_1 + C_2 e^{\lambda_2 t} \mathbf{v}_2$
Case 2:	$\Delta < 0, r = \alpha + \beta i$ $y(t) = C_1 \text{Re}(e^{rt}) + C_2 \text{Im}(e^{rt})$	$\Delta < 0, \lambda = \alpha + \beta i$ $\mathbf{x}(t) = C_1 \text{Re}(e^{\lambda t} \mathbf{v}) + C_2 \text{Im}(e^{\lambda t} \mathbf{v})$
Case 3:	$\Delta = 0:$ $y(t) = e^{rt}(C_1 + C_2 t)$	$\Delta = 0:$ $\mathbf{v}$ solves $(A - \lambda I)\mathbf{v} = \mathbf{0}$ (as usual) $\mathbf{w}$ solves $(A - \lambda I)\mathbf{w} = \mathbf{v}$ $\mathbf{x}(t) = e^{\lambda t} (C_1 \mathbf{v} + C_2 (t\mathbf{v} + \mathbf{w}))$

### Classification of the Equilibria

The origin is always an equilibrium solution to  $\mathbf{x}' = A\mathbf{x}$ , and we can use the Poincaré Diagram to help us classify the origin. The Poincaré Diagram is based on the discriminant:

$$\Delta = (\text{Tr}(A))^2 - 4\det(A)$$

If  $\Delta = 0$ , we have a parabola in the  $(\text{Tr}(A), \det(A))$  plane. Inside the parabola is where  $\Delta < 0$  and outside the parabola is where  $\Delta > 0$ .