## Math 3430, Spring 2019, Systems of equations

We consider systems

$$x' = Ax,\tag{1}$$

where A is a  $2 \times 2$  matrix with constant real coefficients, and  $x : I \to \mathbb{R}^2$ . We look for a solution of the form

$$x(t) = v e^{\lambda t},$$

where v is a constant vector and  $\lambda$  is a scalar. Plugging this into (1) gives

$$\lambda v e^{\lambda t} = e^{\lambda t} A v \quad \Leftrightarrow \quad A v = \lambda v. \tag{2}$$

We are interested in finding a general solution, so we do not want to consider vectors v = 0. Since by definition the eigenvectors are nonzero, finding vectors v and scalars  $\lambda$  satisfying

$$Av = \lambda v, \tag{3}$$

is equivalent to solving an eigenvalue problem for A. The idea is to find n = 2 linearly independent solutions (see Theorem 3 in Section 7.2 and compare to Theorem 2 in Section 2.2). Here is the procedure we follow.

Step 0: Make sure the system is in the matrix form:

$$x' = Ax$$

Step 1: Find the eigenvalues for the matrix A.

**Step 2:** Since the characteristic polynomial equation for a  $2 \times 2$  matrices is quadratic, there are three cases to consider (compare to 2nd order ODE in Chapter 2). We discuss them according to the order of difficulty.

**Case 1: Two real distinct eigenvalues:**  $\lambda_1, \lambda_2 \in \mathbb{R}$ . Recalling a theorem from Linear Algebra we know the corresponding eigenvectors will be linearly independent. We find an eigenvector  $v_1$  corresponding to  $\lambda_1$  and  $v_2$  corresponding to  $\lambda_2$ . The general solution is then given by

$$x(t) = c_1 e^{\lambda_1 t} v_1 + c_2 e^{\lambda_2 t} v_2.$$

**Case 2: Complex Eigenvalues:**  $\alpha \pm i\beta$  ( $\alpha, \beta \in \mathbb{R}, \beta \neq 0$ ). In this case, when we find the eigenvector  $v_1$  corresponding to the eigenvalue  $\lambda_1 = \alpha + i\beta$  and the eigenvector  $v_2$  corresponding to  $\lambda_2 = \alpha - i\beta$ ,

$$e^{\lambda_1 t} v_1$$
 and  $e^{\lambda_2 t} v_2$ 

will be both complex. But since the matrix A is real-valued, it is reasonable to look for real-valued solutions. We follow the same idea as in Chapter 2.

If w(t) is a complex valued solution, we can write it as  $v(t) = u_1(t) + iu_2(t)$ , where  $u_1, u_2$  are real-valued. Next, because (1) is linear, we can show that both  $u_1$  and  $u_2$  are solutions (exercise!). So we take one eigenvalue  $\alpha + i\beta$  and find a corresponding eigenvector v. Next,  $ve^{\lambda t}$  is a solution, so we then extract the real and imaginary parts. They can be shown to be

$$u_1(t) = e^{\alpha t} (B_1 \cos \beta t - B_2 \sin \beta t), \quad u_2(t) = e^{\alpha t} (B_2 \cos \beta t + B_1 \sin \beta t),$$

where

$$B_1 = \mathcal{R}e(v), \quad B_2 = \mathcal{I}m(v),$$

where v is an eigenvector associated to the eigenvalue  $\alpha + i\beta$ . It is another exercise to show  $u_1$  and  $u_2$  are linearly independent. Hence the general solution is given by

$$x(t) = c_1 u_1(t) + c_2 u_2(t).$$

**Case 3: Repeated real eigenvalue.** This is in some sense the most complicated case. Recall from Linear Algebra

- $m_a$  algebraic multiplicity is the number of times  $\lambda$  appears as a root of det $(A \lambda I) = 0$ .
- $m_g$  geometric multiplicity is the number of linearly independent eigenvectors corresponding to  $\lambda$ .

Since we have n = 2, the repeated eigenvalue means, we have  $m_a = 2$ . If it turns out that  $m_g = 2$  also, then we can just take the linear combination as the solution

$$x(t) = c_1 e^{\lambda t} v_1 + c_2 e^{\lambda t} v_2.$$

If  $m_g = 1$ , then we have what is called a *defective eigenvalue*. In this case the general solution can be shown to be given by

$$x(t) = c_1 e^{\lambda t} v + c_2 u_2,$$

where

$$u_2(t) = e^{\lambda t}(tv + w),$$

with v an eigenvector associated to the eigenvalue  $\lambda$ , and w the generalized eigenvector solving  $(A - \lambda I)w = v$ .