

## MATH 63CM HOMEWORK 4 SOLUTIONS

(a). We prove the contrapositive. Suppose that  $\mathbf{x}_1, \dots, \mathbf{x}_n$  are linearly dependent at some time  $t_1$ . Then there are constants  $c_1, \dots, c_n$ , not all zero, such that

$$\sum c_i \mathbf{x}_i(t_1) = 0.$$

Then  $x(t) := \sum c_i \mathbf{x}_i(t)$  is a solution of the initial value problem

$$\begin{aligned} \mathbf{x}'(t) &= A(t)\mathbf{x}(t), \\ \mathbf{x}(t_1) &= 0. \end{aligned}$$

The 0 function is also a solution of this initial value problem. But we know solutions are unique. Thus  $\mathbf{x}(\cdot) \equiv 0$ , so  $\mathbf{x}_1, \dots, \mathbf{x}_n$  are linearly dependent at all times.

(a). (Alternate solution). Let  $X(t)$  be the  $n \times n$  matrix whose columns are  $\mathbf{x}_1(t) \dots \mathbf{x}_n(t)$ . Note that

$$X'(t) = A(t)X(t).$$

(That is because, by definition of matrix multiplication, column  $j$  of  $A(t)X(t)$  is  $A(t)\mathbf{x}_j(t)$ .) Since the columns of  $X(t_0)$  are independent,  $\det X(t_0) \neq 0$ . By Liouville's Theorem (Proposition 3.13 in the text),

$$\det X(t) = \det X(t_0) e^{\int_{t_0}^t \text{tr } A(s) ds}.$$

Thus  $\det X(t) \neq 0$  for all  $t$ , so the columns of  $X(t)$  are independent for each  $t$ .

(b). Trivially, each linear combination of the  $\mathbf{x}_i$  is a solution:

$$\frac{d}{dt} \sum_i c_i \mathbf{x}_i(t) = \sum_i c_i \mathbf{x}'_i(t) = \sum_i c_i A \mathbf{x}_i(t) = A \left( \sum_i c_i \mathbf{x}_i(t) \right).$$

To see that we get all solutions in this way, let  $\mathbf{x}(\cdot)$  be any solution of the equation. Since  $\mathbf{x}_1(t_0), \dots, \mathbf{x}_n(t_0)$  are  $n$  independent vectors in  $\mathbf{R}^n$ , they form a basis for  $\mathbf{R}^n$ . Thus  $\mathbf{x}(t_0)$  is a linear combination of the  $\mathbf{x}_i(t_0)$ :

$$\mathbf{x}(t_0) = \sum_i c_i \mathbf{x}_i(t_0)$$

for suitable constants  $c_1, \dots, c_n$ . Now  $\mathbf{x}(t)$  and  $\sum_i c_i \mathbf{x}_i(t)$  are two solutions of the ODE that are equal at time  $t_0$ . By the uniqueness theorem, they are equal for all  $t$ .

2. We have

$$\frac{dv}{dx} = \frac{1}{x} \frac{dy}{dx} - \frac{y}{x^2} = \frac{\phi(v) - v}{x}.$$

By separation of variables, we therefore have

$$\int \frac{dv}{\phi(v) - v} = \int \frac{dx}{x} = \log |x| + C,$$

so

$$|x| = \tilde{C} \exp \left\{ \int \frac{dv}{\phi(v) - v} \right\}.$$

3. This falls under the framework of the previous problem with

$$\phi(v) = \frac{1+v}{1-v}.$$

Thus we have

$$\begin{aligned} |x| &= \tilde{C} \exp \left\{ \int \frac{dv}{\frac{1+v}{1-v} - v} \right\} \\ &= \tilde{C} \exp \left\{ \int \frac{1-v}{1+v-v(1-v)} dv \right\} \\ &= \tilde{C} \exp \left\{ \int \frac{1-v}{1+v^2} dv \right\} \\ &= \tilde{C} \exp \left\{ \arctan v - \frac{1}{2} \log |1+v^2| \right\} \\ &= \tilde{C} \frac{\exp \{ \arctan(y/x) \}}{\sqrt{1+(y/x)^2}} \\ &= \tilde{C} |x| \frac{\exp \{ \arctan(y/x) \}}{\sqrt{x^2+y^2}}. \end{aligned}$$

Therefore, we have for  $x \neq 0$

$$\tilde{C} \frac{\exp \{ \arctan(y/x) \}}{\sqrt{x^2+y^2}} = 1.$$

4. (a). The equilibrium points are when  $x_2^2 + x_1 x_2 = 2$  and  $x_1^2 + x_1 x_2 = 2$ . Adding and subtracting the two equations, we see that this occurs when  $(x_1 + x_2)^2 = 4$  and  $x_1^2 = x_2^2$ . The solutions to this system are  $(x_1, x_2) = (1, 1)$  and  $(x_1, x_2) = (-1, -1)$ .

Around  $(1, 1)$ , the linearized equation is

$$\begin{aligned} \tilde{x}_1' &= \tilde{x}_1 + 3\tilde{x}_2 \\ \tilde{x}_2' &= 3\tilde{x}_1 + \tilde{x}_2; \end{aligned}$$

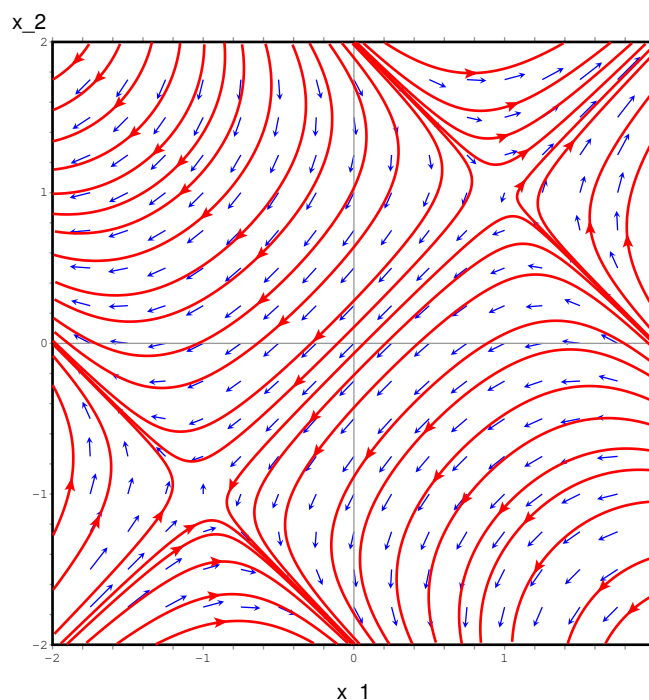
the eigenvalues of this system are  $-2$  and  $4$ , so this is a hyperbolic equilibrium point and we have a saddle.

Around  $(-1, -1)$ , the linearized equation is

$$\begin{aligned} \tilde{x}_1' &= -\tilde{x}_1 - 3\tilde{x}_2 \\ \tilde{x}_2' &= -3\tilde{x}_1 - \tilde{x}_2; \end{aligned}$$

the eigenvalues of this system are  $2$  and  $-4$ , so this is a hyperbolic equilibrium point and we have a saddle.

(b). The phase portrait is here:



(c). We have

$$\begin{aligned}(x_1 + x_2)' &= (x_1 + x_2)^2 - 4 \\ (x_1 - x_2)' &= (x_1 + x_2)(x_1 - x_2)\end{aligned}$$

From this we see that the curve  $x_1 = x_2$  is invariant under the evolution. Thus it is the unstable manifold for the equilibrium point  $(1, 1)$  and the stable manifold for the equilibrium point  $(-1, -1)$ . (The stability can be checked by checking the sign of the derivative along the manifold.) We also see that the curves  $x_1 + x_2 = 2$  and  $x_1 + x_2 = -2$  are invariant under the evolution. Therefore, the curve  $x_1 + x_2 = 2$  is the stable manifold for the equilibrium point  $(1, 1)$ , and  $x_1 + x_2 = -2$  is the unstable manifold for the equilibrium point  $(-1, -1)$ .

5. Note that

$$\begin{aligned}\det(\lambda I - A) &= \begin{vmatrix} \lambda - 3 & 1 \\ -1 & \lambda - 1 \end{vmatrix} (\lambda - 2) \\ &= ((\lambda - 3)(\lambda - 1) + 1)(\lambda - 2) \\ &= (\lambda^2 - 4\lambda + 4)(\lambda - 2) \\ &= (\lambda - 2)^3.\end{aligned}$$

Consequently the matrix  $N = A - 2I = \begin{bmatrix} 1 & -1 & 0 \\ 1 & -1 & 1 \\ 0 & 0 & 0 \end{bmatrix}$  is nilpotent (indeed  $N^3 = 0$ )

and commutes with  $A$ . Thus

$$\begin{aligned} e^{At} &= e^{t(2I+N)} \\ &= e^{2tI} e^{tN} \\ &= e^{2t} \left( I + tN + \frac{t^2}{2!} N^2 \right) \\ &= e^{2t} \left( I + t \begin{bmatrix} 1 & -1 & 0 \\ 1 & -1 & 1 \\ 0 & 0 & 0 \end{bmatrix} + \frac{t^2}{2} \begin{bmatrix} 0 & 0 & -1 \\ 0 & 0 & -1 \\ 0 & 0 & 0 \end{bmatrix} \right) \\ &= e^{2t} \begin{bmatrix} (1+t) & -t & -t^2/2 \\ t & (1-t) & t \\ 0 & 0 & 1 \end{bmatrix} \end{aligned}$$

6. Let  $k$  be a positive integer such that  $N^k = 0$ . Then

$$(I + N)(I - N + N^2 - \dots - N^{k-1}) = I - N^k = I.$$

7. Let  $\mathbf{x} \in \mathbf{C}^n$ . Then (by  $(*)$  in the statement of the problem), there exist  $\mathbf{x}_i \in \ker(\lambda_i I - A)^{\nu_i}$  such that

$$\mathbf{x} = \mathbf{x}_1 + \dots + \mathbf{x}_k.$$

Note that if  $q(z)$  and  $\hat{q}(z)$  are two polynomials, then  $q(A)\hat{q}(A) = \hat{q}(A)q(A)$ . Thus

$$\begin{aligned} p(A)\mathbf{x}_j &= \prod_{i=1}^k (\lambda_i I - A)^{\nu_i} \mathbf{x}_j \\ &= \left( \prod_{i \neq j} (\lambda_i I - A)^{\nu_i} \right) (\lambda_j I - A)^{\nu_j} \mathbf{x}_j \\ &= 0. \end{aligned}$$

Thus

$$p(A)\mathbf{x} = p(A) \left( \sum_{j=1}^k \mathbf{x}_j \right) = \sum_{j=1}^k p(A)\mathbf{x}_j = 0.$$

We have shown that  $p(A)\mathbf{x} = 0$  for every vector  $\mathbf{x}$ . Thus  $p(A) = 0$ .