## Homework 7 Solutions

**5.2 Ex. 8–10.** Clearly  $A_n \begin{bmatrix} 1 & 1 & \dots & 1 \end{bmatrix}^T = n \begin{bmatrix} 1 & 1 & \dots & 1 \end{bmatrix}^T$ , so  $\mathbf{v}_1 = \begin{bmatrix} 1 & 1 & \dots & 1 \end{bmatrix}^T$  is an eigenvector with eigenvalue  $\lambda = n$ . The vectors in Ex. 9 are in the nullspace of  $A_n$ , so they are eigenvectors with eigenvalue zero. They are linearly independent because the matrix whose rows are these vectors is in echelon form. The vector  $\mathbf{v}_1$  is not in the span of the others, since it's not in the nullspace of A, so these n vectors are linearly independent. Hence  $A_n$  is diagonalizable with  $\Lambda$  having diagonal entries n (once) and 0 (n - 1 times).

Now the characteristic polynomial of A is equal to that of  $\Lambda$ , namely  $f(\lambda) = \lambda^{n-1}(\lambda - n)$ . The matrix in the extra part of the problem is  $I_n + A_n$ , so its determinant is  $\det(I_n + A_n) = (-1)^n \det(-I_n - A_n) = (-1)^n f(-1) = (-1)^n (-1)^{n-1} (-1 - n) = n + 1$ .

**5.2 Ex. 20**. Characteristic polynomial is  $(\lambda - 7)(\lambda + 3)^2$ . Eigenvalues are 7, -3, with -3 repeated twice. The vector



spans the  $\lambda = 7$  eigenspace. The vectors

$$\begin{bmatrix} 1\\0\\2 \end{bmatrix}, \begin{bmatrix} 0\\1\\0 \end{bmatrix}$$

span the  $\lambda = 3$  eigenspace. You can verify that 7 - 3 - 3 = 1 = tr(A) and  $7(-3)^2 = 63 = det(A)$ .

**5.2 Ex. 33**. det $(\lambda I_n - A) = det((\lambda I_n - A)^T)$  by Theorem (2.19). Now since  $I_n$  is symmetric, det $((\lambda I_n - A)^T) = det(\lambda I_n - A^T)$ , which is the characteristic polynomial of  $A^T$ .

**5.3 Ex. 9.** The eigenvalues are 2 and -3, and each eigenspace is 1-dimensional. So we can only find two linearly independent eigenvectors.

5.3 Ex. 26. Diagonalizable, with for instance

$$S = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 4 & 1 & 0 & 0 \\ 3 & 0 & 1 & 0 \\ 2 & 0 & 0 & 1 \end{bmatrix}, \quad \Lambda = \begin{bmatrix} 6 & 0 & 0 & 0 \\ 0 & 5 & 0 & 0 \\ 0 & 0 & 5 & 0 \\ 0 & 0 & 0 & 5 \end{bmatrix}$$

**5.3 Ex. 32.** Since A and B have the same eigenvectors, we can diagonalize them with the same matrix S, so  $A = S\Lambda S^{-1}$  and  $B = S\Theta S^{-1}$ , where  $\Lambda$  and  $\Theta$  are diagonal. Diagonal matrices commute, so  $AB = S\Lambda\Theta S^{-1} = S\Theta\Lambda S^{-1} = BA$ .

Problem A. (a)  $\phi(\mathbf{0}) = \mathbf{0}$ , and if  $\phi(\mathbf{v}) = \phi(\mathbf{w}) = \mathbf{0}$ , then  $\phi(\mathbf{v}+\mathbf{w}) = \mathbf{0}$  and  $\phi(r\mathbf{v}) = \mathbf{0}$ . This shows ker( $\phi$ ) contains zero and is closed under addition and scalar multiplication, so ker( $\phi$ ) is a subspace of V. Similarly  $\phi(\mathbf{0}) = \mathbf{0}$  shows that  $\mathbf{0} \in \operatorname{im}(\phi)$ , and  $\phi(\mathbf{v}) + \phi(\mathbf{w}) = \phi(\mathbf{v}+\mathbf{w})$ ,  $r\phi(\mathbf{v}) = \phi(r\mathbf{v})$  shows that  $\operatorname{im}(\phi)$  is closed under addition and scalar multiplication.

(b) We have  $[\phi(\mathbf{v})]_{\mathcal{C}} = A[\mathbf{v}]_{\mathcal{B}}$ . The left-hand side is zero if and only if  $\mathbf{v} \in \ker(\phi)$ , while the right-hand side is zero if and only if  $[\mathbf{v}]_{\mathcal{B}} \in \mathrm{NS}(A)$ .

(c) We have  $\mathbf{w} \in \operatorname{im}(\phi)$  if and only if  $\mathbf{w} = \phi(\mathbf{v})$  for some vector  $\mathbf{v} \in V$ , if and only if  $[\mathbf{w}]_{\mathcal{C}} = A[\mathbf{v}]_{\mathcal{B}}$ , if and only if  $[\mathbf{w}]_{\mathcal{C}} \in \operatorname{CS}(A)$ .

(d) Let  $\dim(V) = n$ ,  $\dim(W) = n$ , so A is an  $m \times n$  matrix. Part (b) shows that  $\mathbf{v} \mapsto [\mathbf{v}]_{\mathcal{B}}$  gives a linear isomorphism from  $\ker(\phi)$  to  $\operatorname{NS}(A)$ , hence  $\dim(\ker(\phi)) = \dim(\operatorname{NS}(A)) = n - \operatorname{rank} A$ . Similarly, part (c) implies that  $\dim(\operatorname{im}(\phi)) = \dim(\operatorname{CS}(A)) = \operatorname{rank} A$ . Hence  $\dim(V) = n = \dim(\ker(\phi)) + \dim(\operatorname{im}(\phi))$ .

(e) T is linear, since the properties of derivatives easily give T(f + g) = T(f) + T(g)and T(rf) = rT(f) for constant r. The kernel of T is the set of solutions of the differential equation f + f' = 0. You can solve it by separation of variables (integrate f'/f = -1), giving the general solution  $f(x) = Ae^{-x}$ . This is only a polynomial if A = 0.

(f) By part (d), the image of  $T: P_{<n} \to P_{<n}$  has dimension  $n = \dim P_{<n}$ , so T is onto, *i.e.*, every  $g(x) \in P_{<n}$  is T(f) for some f. Since this holds for every n, it follows that every polynomial g(x) is equal to f(x) + f'(x) for some polynomial f(x). In fact, f(x) is unique, because T has zero kernel in the space of polynomials.

**Problem B.** (a)  $\Delta(f+g) = f(x+1) + g(x+1) - f(x) - g(x) = \Delta f + \Delta g$  and  $\Delta(rf) = rf(x+1) - rf(x) = r\Delta(f)$ . (b) For k > 0,

$$\Delta C_k(x) = \frac{(x+1)x(x-1)\cdots(x-k+2)}{k!} - \frac{x(x-1)\cdots(x-k+1)}{k!}$$
$$= ((x+1) - (x-k+1))\frac{x(x-1)\cdots(x-k+2)}{k!} = \frac{x(x-1)\cdots(x-k+2)}{(k-1)!} = C_{k-1}(x).$$

By definition,  $\Delta C_0(x) = \Delta 1 = 0$ . Hence the matrix A is  $n \times n$  with entries 1 just above the diagonal and all other entries 0:

$$A = \begin{bmatrix} 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & & \vdots \\ 0 & 0 & 0 & \dots & 1 \\ 0 & 0 & 0 & \dots & 0 \end{bmatrix}.$$

(c)  $\Delta 1 = 0, \ \Delta x = 1, \ \Delta x^2 = 2x + 1$ , so

$$B = \begin{bmatrix} 0 & 1 & 1 \\ 0 & 0 & 2 \\ 0 & 0 & 0 \end{bmatrix}.$$

 $C_0 = 1, C_1 = x, C_2 = (x^2 - x)/2$ , so

$$X = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & -1/2 \\ 0 & 0 & 1/2 \end{bmatrix}.$$

Now compute

$$XAX^{-1} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & -1/2 \\ 0 & 0 & 1/2 \end{bmatrix} \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 2 \end{bmatrix} = \begin{bmatrix} 0 & 1 & 1 \\ 0 & 0 & 2 \\ 0 & 0 & 0 \end{bmatrix}.$$