MATH 115A HW 5

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1. Prove Theorem 2.10.

(a)
$$T \circ (U_1 + U_2) = T \circ U_1 + T \circ U_2$$

Let
$$v \in V$$

$$T \circ (U_1 + U_2)(v) = T \circ (U_1(v) + U_2(v))$$

$$T(U_1(v) + U_2(v)) = T(U_1(v)) + T(U_2(v)) \text{ (linearity)}$$

$$\therefore T \circ U_1 + T \circ U_2 = T \circ (U_1 + U_2)$$

(b) $T \circ (U_1 \circ U_2) = (T \circ U_1) \circ U_2$

Let
$$v \in V$$

$$T \circ (U_1 \circ U_2)(v) = T((U_1 \circ U_2)(v)) = T(U_1(U_2(v)))$$

$$((T \circ U_1) \circ U_2)(v) = (T \circ U_1)(U_2(v)) = T(U_1(U_2(v)))$$
Since both equal $T(U_1(U_2(v)))$, $T \circ (U_1 \circ U_2) = (T \circ U_1) \circ U_2$

(c) Given $I_V(v) = v$, $I_V \circ T = T \circ I_V = T$

Let
$$v \in V$$

$$(I_v \circ T)(v) = I_v(T(v)) = T(v) \text{ (def. of } I_v)$$

$$(T \circ I_v)(v) = T(I_v(v)) = T(v) \text{ (def. of } I_v)$$
Since both equal $T(v)$, $I_V \circ T = T \circ I_V = T$

(d) $(aU_1) \circ U_2 = a(U_1 \circ U_2)$

Let
$$v \in V$$

$$((aU_1) \circ U_2)(v) = aU_1(U_2(v))$$

$$(a(U_1 \circ U_2))(v) = a((U_1 \circ U_2)(v)) \text{ (linearity)}$$

$$a((U_1 \circ U_2)(v)) = a(U_1(U_2(v))$$
Since both equal $a(U_1(U_2(v))), (aU_1) \circ U_2 = a(U_1 \circ U_2)$

2. $(\lambda A) \cdot B = A \cdot (\lambda B)$ Proof.

$$(AB)_{ij} = \sum_{l=1}^{m} A_{il} B_{lj}, \ 1 \le i \le n, \ 1 \le j \le k$$
(Def. of matrix multiplication)
$$((\lambda A)B)_{ij} = \sum_{l=1}^{m} (\lambda A_{il}) B_{lj}$$

$$= \sum_{l=1}^{m} \lambda (A_{il}) B_{lj} \text{ (Def. scalar multiplication)}$$

$$= \lambda \sum_{l=1}^{m} (A_{il}) B_{lj} \text{ (Distributive mult. of field elements)}$$

$$(A(\lambda B))_{ij} = \sum_{l=1}^{m} A_{il}(\lambda B_{lj})$$
$$= \sum_{l=1}^{m} A_{il}\lambda(B_{lj}) \text{ (Def. scalar multiplication)}$$
$$= \lambda \sum_{l=1}^{m} A_{il}(B_{lj}) \text{ (Distributive mult. of field elements)}$$

Since both $((\lambda A) \cdot B)_{ij}$ and $(A \cdot (\lambda B))_{ij}$ equal $\lambda \sum_{l=1}^{m} A_{il} B_{lj}$, $(\lambda A) \cdot B = A \cdot (\lambda B)$. \square

3. Textbook 2.2 #9 T is linear. Proof.

Let
$$y, z \in V : y = y_1 + y_2 i, z = z_1 + z_2 i$$

$$T(y+z) = T(y_1 + y_2 i + z_1 + z_2 i) = T(y_1 + z_1 + i(y_2 + z_2)) = y_1 + z_1 - i(y_2 + z_2)$$

$$T(y) + T(z) = T(y_1 + y_2 i) + T(z_1 + z_2 i) = y_1 - y_2 i + z_1 - z_2 i = y_1 + z_1 - i(y_2 + z_2)$$

$$\therefore T(y+z) = T(y) + T(z)$$

$$T(\lambda z) = T(\lambda(z_1 + z_2 i)) = T(\lambda z_1 + \lambda z_2 i) = \lambda z_1 - \lambda z_2 i$$

$$\lambda T(z) = \lambda T(z_1 + z_2 i) = \lambda(z_1 - z_2 i) = \lambda z_1 - \lambda z_2 i \text{ (dist. multiplication over a field)}$$

$$\therefore T(\lambda z) = \lambda T(z)$$

Since
$$T(y+z) = T(y) + T(z)$$
 and $T(\lambda z) = \lambda T(z)$, T is linear. \square

Find $[T]_{\beta}$.

$$\beta = \{1, i\}$$

$$T(1) = (1)1 + (-0)i$$

$$T(i) = (0)1 + (-1)i$$

$$[T]_{\beta} = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$$

4. The *i*-th column of A is $A \cdot e_i$ *Proof.*

$$(A \cdot e_i)_{j1} = \sum_{l=1}^n A_{jl} e_{lj} \text{ (Def. matrix multiplication)}$$

$$(A \cdot e_i)_{11} = \sum_{l=1}^n A_{1l} e_{l1}$$

$$= 0(A_{11}) + \dots + 1(A_{1i}) + \dots + 0(A_{1n}) = A_{1i}$$

$$(A \cdot e_i)_{21} = \sum_{l=1}^n A_{2l} e_{l1}$$

$$= 0(A_{21}) + \dots + 1(A_{2i}) + \dots + 0(A_{2n}) = A_{2i}$$

$$\vdots$$

$$(A \cdot e_i)_{k1} = \sum_{l=1}^n A_{kl} e_{l1}$$

$$= 0(A_{k1}) + \dots + 1(A_{ki}) + \dots + 0(A_{kn}) = A_{ki}$$

$$A \cdot e_i = (A_{1i}, A_{2i}, \dots, A_{mi})^T \text{ (this is the } i\text{-th column of } A)$$

$$\therefore A \cdot e_i = \text{ the } i\text{-th column of } A. \quad \Box$$

- 5. $D: \operatorname{Poly}(\mathbb{R}) \to \operatorname{Poly}(\mathbb{R})$ by D(f(x)) = f'(x)
 - (a) D is linear. Proof.

Let
$$f, g \in \operatorname{Poly}(\mathbb{R})$$

 $D(f(x)) = f'(x)$
 $D(g(x)) = g'(x)$
 $D(f(x)) + D(g(x)) = f'(x) + g'(x)$

$$D(f(x) + g(x)) = \frac{d}{dx}(f(x) + g(x))$$

$$\frac{d}{dx}(f(x) + g(x)) = f'(x) + g'(x)$$

$$\therefore D(f(x)) + D(g(x)) = D(f(x) + g(x)) \forall f, g \in \operatorname{Poly}(\mathbb{R})$$
Let $f \in \operatorname{Poly}(\mathbb{R}), \lambda \in \mathbb{R}$

$$\lambda D(f(x)) = \lambda \cdot f'(x)$$

$$D(\lambda f(x)) = \frac{d}{dx}(\lambda f(x))$$
$$\frac{d}{dx}(\lambda f(x)) = \lambda \cdot f'(x)$$

$$\therefore \lambda D(f(x)) = D(\lambda f(x)) \forall f, g \in \text{Poly}(\mathbb{R})$$

Since $D(f(x)) + D(g(x)) = D(f(x) + g(x)) \forall f, g \in \text{Poly}(\mathbb{R})$ and $\lambda D(f(x)) = D(\lambda f(x)) \forall f, g \in \text{Poly}(\mathbb{R})$, D is a linear transformation.

(b) For n = 3, find $[D]_{\beta_2}^{\beta_3}$ w.r.t. bases

$$\beta_3 = \{1, x, x^2, x^3\}, \ \beta_2 = \{1, x, x^2\}$$

$$D(1) = 0 = (0)1 + (0)x + (0)x^2$$

$$D(x) = 1 = (1)1 + (0)x + (0)x^2$$

$$D(x^2) = 2x = (0)1 + (2)x + (0)x^2$$

$$D(x^3) = 3x^2 = (0)1 + (0)x + (3)x^2$$

$$[D]_{\beta_2}^{\beta_3} = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 3 \end{pmatrix}$$

(c) Find $[D]^{\gamma}_{\alpha}$ where

$$\alpha = \{x + 2, x - 2, x^2 + x, x^3 + x^2\}, \ \gamma = \{1, 2x, x^2 + 1\}$$

$$D(x + 2) = 2 = (2)1 + (0)2x + (0)(x^2 + 1)$$

$$D(x - 2) = -2 = (-2)1 + (0)2x + (0)(x^2 + 1)$$

$$D(x^2 + x) = 2x + 1 = (1)1 + (1)2x + (0)(x^2 + 1)$$

$$D(x^3 + x^2) = 3x^2 + 2x = (-3)1 + (1)2x + (3)(x^2 + 1)$$

$$\begin{cases} 2 & -2 & 1 & -3 \end{cases}$$

$$[D]_{\alpha}^{\gamma} = \begin{pmatrix} 2 & -2 & 1 & -3 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 3 \end{pmatrix}$$

- 6. $L_A(\vec{x}) = A \cdot x$
 - (a) Compute $[L_A]_{\beta_3}^{\beta_2}$

$$\begin{pmatrix} 2 & 1 & 0 \\ 1 & 8 & 9 \end{pmatrix}$$

(b) Compute $[L_A]_{\gamma_3}^{\gamma_2}$

$$L_A \begin{pmatrix} 1\\2\\0 \end{pmatrix} = \begin{pmatrix} 4\\17 \end{pmatrix} = \frac{17}{2} \begin{pmatrix} 1\\2 \end{pmatrix} + -\frac{3}{2} \begin{pmatrix} 3\\0 \end{pmatrix}$$

$$L_A \begin{pmatrix} 1\\1\\1 \end{pmatrix} = \begin{pmatrix} 3\\18 \end{pmatrix} = 9 \begin{pmatrix} 1\\2 \end{pmatrix} + -2 \begin{pmatrix} 3\\0 \end{pmatrix}$$

$$L_A \begin{pmatrix} 0\\0\\1 \end{pmatrix} = \begin{pmatrix} 0\\9 \end{pmatrix} = \frac{9}{2} \begin{pmatrix} 1\\2 \end{pmatrix} + -\frac{3}{2} \begin{pmatrix} 3\\0 \end{pmatrix}$$

$$[L_A]_{\gamma_3}^{\gamma_2} = \begin{pmatrix} \frac{17}{2} & 9 & \frac{9}{2}\\ -\frac{3}{2} & -2 & -\frac{3}{2} \end{pmatrix}$$

7. Textbook 2.3 # 3

$$\beta = \{1, x, x^2\}, \ \gamma = \{e_1, e_2, e_3\}$$

(a) Compute $[U]^{\gamma}_{\beta}$, $[T]_{\beta}$, and $[UT]^{\gamma}_{\beta}$

$$U(1+0x+0x^{2}) = (1+0,0,1-0) = (1,0,1) = (1)e_{1} + (0)e_{2} + (1)e_{3}$$

$$U(0+1x+0x^{2}) = (0+1,0,0-1) = (1,0,-1) = (1)e_{1} + (0)e_{2} + (-1)e_{3}$$

$$U(0+0x+1x^{2}) = (0+0,1,0-0) = (0,1,0) = (0)e_{1} + (1)e_{2} + (0)e_{3}$$

$$[U]_{\beta}^{\gamma} = \begin{pmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & -1 & 0 \end{pmatrix}$$

$$T(1+0x+0x^{2}) = 0(3+x) + 2(1) = 2 = (2)1 + (0)x + (0)x^{2}$$

$$T(0+1x+0x^{2}) = 1(3+x) + 2(x) = 3 + 3x = (3)1 + (3)x + (0)x^{2}$$

$$T(0+0x+1x^{2}) = 2x(3+x) + 2(x^{2}) = 6x + 4x^{2} = (0)1 + (6)x + (4)x^{2}$$

$$[T]_{\beta} = \begin{pmatrix} 2 & 3 & 0 \\ 0 & 3 & 6 \\ 0 & 0 & 4 \end{pmatrix}$$

$$U(T(1+0x+0x^{2})) = U(2) = (2,0,2) = (2)e_{1} + (0)e_{2} + (2)e_{3}$$

$$U(T(0+1x+0x^{2})) = U(3+3x) = (6,0,0) = (6)e_{1} + (0)e_{2} + (0)e_{3}$$

$$U(T(0+0x+1x^{2})) = U(6x+4x^{2}) = (6,4,-6) = (6)e_{1} + (4)e_{2} + (-6)e_{3}$$

$$[UT]_{\beta}^{\gamma} = \begin{pmatrix} 2 & 6 & 6 \\ 0 & 0 & 4 \\ 2 & 0 & 6 \end{pmatrix}$$

Verification with Theorem 2.11:

$$[U]_{\beta}^{\gamma} \cdot [T]_{\beta} = \begin{pmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & -1 & 0 \end{pmatrix} \begin{pmatrix} 2 & 6 & 6 \\ 0 & 0 & 4 \\ 2 & 0 & 6 \end{pmatrix} = \begin{pmatrix} 2 & 6 & 6 \\ 0 & 0 & 4 \\ 2 & 0 & 6 \end{pmatrix} \tag{1}$$

(b) Compute $[h(x)]_{\beta}$ and $[U(h(x))]_{\gamma}$.

$$[h(x)]_{\beta} = \begin{pmatrix} 3 \\ -2 \\ 1 \end{pmatrix}$$

$$U(h(x)) = U(3 - 2x + x^{2}) = (1, 1, 5)$$

$$U(h(x)) = U(3 - 2x + x^{2}) = (1, 1, 5)$$
$$[U(h(x))]_{\gamma} = \begin{pmatrix} 1\\1\\5 \end{pmatrix}$$

Verification with Theorem 2.14:

$$\begin{pmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & -1 & 0 \end{pmatrix} \begin{pmatrix} 3 \\ -2 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \\ 5 \end{pmatrix}$$

- 8. Textbook 2.3 #16
 - (a) $\operatorname{Im}(T) \cap \ker(T) = \{0\}$ Proof.

$$\operatorname{rank}(T) = \operatorname{rank}(T^2)$$

$$\operatorname{rank}(T) + \operatorname{nullity}(T) = \dim(V)$$

$$\operatorname{rank}(T^2) + \operatorname{nullity}(T) = \dim(V)$$

$$\operatorname{rank}(T^2) + \operatorname{nullity}(T^2) = \dim(V) \text{ (since } T : V \to V)$$

$$\therefore \operatorname{nullity}(T) = \operatorname{nullity}(T^2) \text{ and } \ker(T) = \ker(T^2)$$

$$\operatorname{Let } v \in \operatorname{Im}(T) \cap \ker(T). \exists w \in V : T(w) = v$$

$$\operatorname{Since } \in \ker(T), T(v) = 0$$

$$T(T(w)) = 0 \therefore T^2(w) = 0$$

$$w \in \ker(T^2) \text{ (def. of nullity)}$$

$$w \in \ker(T) \text{ (since } \ker(T) = \ker(T^2)$$

$$T(w) = 0 \therefore v \text{ must equal } 0 \text{ since } T(w) = v.$$

$$\therefore \operatorname{Im}(T) \cap \ker(T) = \{0\}. \quad \Box$$

Since $\operatorname{rank}(T) + \operatorname{nullity}(T) = \dim(V)$ and $\operatorname{Im}(T) \cap \ker(T) = \{0\}, V = \operatorname{Im}(T) \oplus \ker(T)$