

HW 6 Dual Stuff

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Exercise 1.6.1

1. We know

$$L([1, \dots, x^{n-1}]) = \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & \ddots & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & a_1 & \cdots & a_1^{n-1} \\ 1 & a_2 & \cdots & a_2^{n-1} \\ \vdots & \vdots & \ddots & \vdots \\ 1 & a_n & \cdots & a_n^{n-1} \end{bmatrix}$$

Then the matrix of L is

$$\begin{bmatrix} 1 & a_1 & \cdots & a_1^{n-1} \\ 1 & a_2 & \cdots & a_2^{n-1} \\ \vdots & \vdots & \ddots & \vdots \\ 1 & a_n & \cdots & a_n^{n-1} \end{bmatrix}$$

It is a Vandermonde matrix.

2. L is invertible \iff the matrix of L is invertible. Note that for Vandermonde matrix L ,

$$\det(L) = \prod_{i>j, i,j \in \{1, \dots, n\}} (a_i - a_j)$$

Then we have a_1, \dots, a_n are distinct $\iff \det(L) \neq 0 \iff L$ is invertible.

3. First prove that $\{ev_{a_1}, \dots, ev_{a_n}\}$ form a basis implies a_1, \dots, a_n are distinct. Suppose $\{ev_{a_1}, \dots, ev_{a_n}\}$ forms a basis and $\exists m, n \in \{1, \dots, n\}$ s.t. $a_m = a_n (m \neq n)$. We know $ev_m = ev_n$, then $\{ev_{a_1}, \dots, ev_{a_n}\}$ are linearly dependent. It is contradictory to the assumption that $\{ev_{a_1}, \dots, ev_{a_n}\}$ is a basis of an n dimensional vector space V^* .

Then prove that a_1, \dots, a_n are distinct implies $\{ev_{a_1}, \dots, ev_{a_n}\}$ form a basis of V^* . From the previous subproblem, we know that there is a unique solution to the interpolation problem with n distinct points. So let $l_i = \prod_{j \neq i} \frac{x - a_j}{a_i - a_j}$, $\{l_i\}$ is the Lagrange basis for V . For $\forall v \in V$, we have

$$v(x) = \sum_{i=1}^n ev_{a_i}(v) l_i$$

For $\forall \alpha \in V^*$, $\forall v \in V$, we get

$$\alpha(v) = \alpha\left(\sum_{i=1}^n ev_{a_i}(v) l_i\right) = \sum_{i=1}^n \alpha(l_i) ev_{a_i}(v)$$

So for any given distinct a_1, \dots, a_n and given $\alpha \in V^*$, α is a linear combination of $\{ev_{a_1}, \dots, ev_{a_n}\}$. I.e. $\{ev_{a_1}, \dots, ev_{a_n}\}$ forms a basis of V^* .

4. Consider polynomials in V . Let

$$[p_{-1}, p_0, p_1] = [1, x, x^2]W$$

for some $W \in M_3(\mathbb{R})$. The dual vectors of $[p_{-1}, p_0, p_1]$ are $[ev_{-1}, ev_0, ev_1]$. Then we know

$$IW^{-1} = \begin{bmatrix} ev_{-1}(1) & ev_{-1}(x) & ev_{-1}(x^2) \\ ev_0(1) & ev_0(x) & ev_0(x^2) \\ ev_1(1) & ev_1(x) & ev_1(x^2) \end{bmatrix} = \begin{bmatrix} 1 & -1 & 1 \\ 1 & 0 & 0 \\ 1 & 1 & 1 \end{bmatrix}$$

So $W = \begin{bmatrix} 0 & 1 & 0 \\ -\frac{1}{2} & 0 & \frac{1}{2} \\ \frac{1}{2} & -1 & \frac{1}{2} \end{bmatrix}$, we have $p_{-1} = \frac{1}{2}x^2 - \frac{1}{2}x$, $p_0 = 1 - x^2$, $p_1 = \frac{1}{2}x^2 + \frac{1}{2}x$.

5. Since $-1, 0, 1, 2$ are distinct, from subproblem 3 we know $\{ev_{-1}, ev_0, ev_1, ev_2\}$ forms a basis of V^* . For $i \in -1, 0, 1, 2$, let $k_i = \prod_{j \neq i} \frac{x-j}{i-j}$, then we have

$$ev_{-2} = \sum_{i=-1}^2 ev_{-2}(k_i) \cdot ev_i$$

Specifically, we have

$$-ev_{-2} + 4ev_{-1} - 6ev_0 + 4ev_1 - ev_2 = 0$$

□

Exercise 1.6.2 Let \mathbf{v} and \mathbf{w} be two polynomials in V . To say α is a dual vector of V is equivalent to say that α is a linear map from V to \mathbb{R} . I.e. we only need to show that for $\forall \lambda, \mu \in \mathbb{R}$, $\alpha(\lambda \mathbf{v} + \mu \mathbf{w}) = \lambda \alpha(\mathbf{v}) + \mu \alpha(\mathbf{w})$. We write the 5 maps mentioned in this questions as $\alpha_1, \dots, \alpha_5$.

1. α_1 is a dual vector. $\alpha_1(\lambda \mathbf{v} + \mu \mathbf{w}) = 6(\lambda \mathbf{v} + \mu \mathbf{w})(5) = \lambda \cdot 6\mathbf{v}(5) + \mu \cdot 6\mathbf{w}(5) = \lambda \alpha_1(\mathbf{v}) + \mu \alpha_1(\mathbf{w})$. So $\alpha_1 \in V^*$.

2. α_2 is not a dual vector. Consider $\mathbf{v} = x^2$. $\alpha_2(\mathbf{v}) = +\infty \notin \mathbb{R}$.

3. α_3 is a dual vector. This map takes the coefficient of term x^2 for a polynomial in V , so it is linear. Let $\mathbf{v} = a_0 + a_1x + a_2x^2$, $\mathbf{w} = b_0 + b_1x + b_2x^2$. We have $\alpha_3(\lambda \mathbf{v} + \mu \mathbf{w}) = \lim_{x \rightarrow +\infty} \frac{\lambda(a_0+a_1x)+\mu(b_0+b_1x)}{x^2} + \lim_{x \rightarrow +\infty} \frac{\lambda a_2 x^2 + \mu b_2 x^2}{x^2} = \lambda a_2 + \mu b_2 = \lambda \alpha_3(\mathbf{v}) + \mu \alpha_3(\mathbf{w})$.

4. α_4 is not a dual vector. For $\forall \mathbf{v} \in V$, let $\mathbf{v} = a_0 + a_1x + a_2x^2$, we have $\alpha_4(\mathbf{v}) = (8a_2 + a_1)(9a_2 + 3a_1 + a_0)$. It is not linear. Specifically, we show a conter example. Let $\mathbf{v} = 1$, $\mathbf{w} = x$, then $\alpha_4(\mathbf{v}) = 0, \alpha_4(\mathbf{w}) = 3, \alpha_4(\mathbf{v} + \mathbf{w}) = 4$, we get $\alpha_4(\mathbf{v}) + \alpha_4(\mathbf{w}) \neq \alpha_4(\mathbf{v} + \mathbf{w})$.

5. α_5 is not a dual vector, since the degree of a polynomial is not a linear function of the polynomial. Specifically, let $\mathbf{v} = \mathbf{w} = x$, then $\mathbf{v} + \mathbf{w} = 2x$, $\deg(\mathbf{v}) = \deg(\mathbf{w}) = 1$, $\deg(\mathbf{v} + \mathbf{w}) = 1 \neq \deg(\mathbf{v}) + \deg(\mathbf{w})$. □

Exercise 1.6.3 Let $\mathbf{v} = \begin{bmatrix} v_1 \\ v_2 \end{bmatrix}$. From the differentiability of f , we have $f(\mathbf{p} + t\mathbf{v}) - f(\mathbf{p}) = tv_1 \frac{\partial f}{\partial x_1}(\mathbf{p}) + tv_2 \frac{\partial f}{\partial x_2}(\mathbf{p}) + o(\|v\|t)$, then $\nabla_{\mathbf{v}} f = \lim_{t \rightarrow 0} (v_1 \frac{\partial f}{\partial x_1}(\mathbf{p}) + v_2 \frac{\partial f}{\partial x_2}(\mathbf{p}) + \frac{o(\|v\|t)}{t}) = \begin{bmatrix} \frac{\partial f}{\partial x_1}(\mathbf{p}) & \frac{\partial f}{\partial x_2}(\mathbf{p}) \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix}$. So $\nabla_{\mathbf{v}} f$ is a linear map from \mathbb{R}^2 to \mathbb{R} , and is thus a dual vector in $(\mathbb{R}^2)^*$. Its “coordinates” under standard dual basis are $\begin{bmatrix} \frac{\partial f}{\partial x_1}(\mathbf{p}) & \frac{\partial f}{\partial x_2}(\mathbf{p}) \end{bmatrix}$. (Of course, the dual basis is $\{\begin{bmatrix} 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \end{bmatrix}\}$) □

Exercise 1.6.4

1. For $\forall \alpha \in \text{Ker}(L^*)$, $L^*(\alpha) = \mathbf{0}$. Then we know for $\forall \mathbf{v} \in V$, $L^*(\alpha)(v) = \alpha \circ L(v) = \mathbf{0}$. For $\forall \mathbf{w} \in \text{Ran}(L)$, we have $\mathbf{w} = L(\mathbf{v})$ for some $\mathbf{v} \in V$. So $\alpha(\mathbf{w}) = \alpha \circ L(v) = \mathbf{0}$.

Conversely, suppose $\forall \alpha \in W^*$ s.t. for $\forall \mathbf{w} \in \text{Ran}(L)$, $\alpha(\mathbf{w}) = \mathbf{0}$. Since for $\forall \mathbf{v} \in V$, $L(\mathbf{v}) \in \text{Ran}(L)$, we have $\alpha \circ L(v) = \alpha(L(v)) = \mathbf{0}$. Note that \mathbf{v} is arbitrary, so $L^*(\alpha) = \alpha \circ L = \mathbf{0}$. I.e. $\alpha \in \text{Ker}(L^*)$.

2. For $\forall \beta \in \text{Ran}(L^*)$, we have $\beta = L^*(\gamma) = \gamma \circ L$ for some $\gamma \in W^*$. Then for $\forall \mathbf{v} \in \text{Ker}(L)$, $\beta(\mathbf{v}) = \gamma(L(\mathbf{v})) = \mathbf{0}$. Conversely, we try to prove if for $\forall \alpha \in V^*$ s.t. $\forall z \in \text{Ker}(L)$, $\alpha(z) = \mathbf{0}$, then $\alpha \in \text{Ran}(L^*)$.

We first show that for $\forall \mathbf{v}, \mathbf{w} \in V$ s.t. $L(\mathbf{v}) = L(\mathbf{w})$, we have $\alpha(\mathbf{v}) = \alpha(\mathbf{w})$. **Proof.** If $L(\mathbf{v}) = L(\mathbf{w})$, then $L(\mathbf{v} - \mathbf{w}) = \mathbf{0}$. I.e. $(\mathbf{v} - \mathbf{w}) \in \text{Ker}(L)$, so $\alpha(\mathbf{v} - \mathbf{w}) = \mathbf{0}$, i.e. $\alpha(\mathbf{v}) = \alpha(\mathbf{w})$. Thus, we know that the value of $\alpha(\mathbf{v})$ is only dependent on $L(\mathbf{v})$.

Suppose $\dim W = m$. We pick a basis for $\text{Ran}(L)$, i.e. $\text{Ran}(L) = \text{span}(\{w_1, \dots, w_k\})$. Then expand this basis such that $W = \text{span}(\{w_1, \dots, w_m\})$. Let $W' = \text{span}(\{w_{k+1}, \dots, w_m\})$, we have $W = \text{Ran}(L) \oplus W'$, i.e. $\forall \mathbf{x} \in W$, there is a unique decomposition $\mathbf{x} = \mathbf{x}_1 + \mathbf{x}_2$ s.t. $\mathbf{x}_1 \in \text{Ran}(L)$, $\mathbf{x}_2 \in W'$. Then we define a map $\beta : W \rightarrow \mathbb{R}$, s.t. $\beta(\mathbf{x}) = \alpha(\mathbf{v})$, where \mathbf{v} is any preimage of \mathbf{x}_1 (We define \mathbf{x}_1 as such: $\mathbf{x} = \mathbf{x}_1 + \mathbf{x}_2$, $\mathbf{x}_1 \in \text{Ran}(L)$, $\mathbf{x}_2 \in W'$). We know β is well defined, since the value of $\alpha(\mathbf{v})$ depends only on $\mathbf{x}_1 = L(\mathbf{v})$. We immediately know that $\beta(\mathbf{x}) = \beta(\mathbf{x}_1)$.

Then we try to show that β is a linear map. Suppose $\forall \mathbf{x}, \mathbf{y} \in W$, we decompose them into $\mathbf{x} = \mathbf{x}_1 + \mathbf{x}_2$, $\mathbf{x}_1 \in \text{Ran}(L)$, $\mathbf{x}_2 \in W'$ and $\mathbf{y} = \mathbf{y}_1 + \mathbf{y}_2$, $\mathbf{y}_1 \in \text{Ran}(L)$, $\mathbf{y}_2 \in W'$. Let $\mathbf{v}, \mathbf{w} \in V$ be any vector s.t. $\mathbf{x}_1 = L(\mathbf{v})$, $\mathbf{y}_1 = L(\mathbf{w})$. Note that $\mathbf{x}_1 + \mathbf{y}_1 = L(\mathbf{v} + \mathbf{w})$, $k\mathbf{x}_1 = L(k\mathbf{v})$, we have $\beta(\mathbf{x} + \mathbf{y}) = \alpha(\mathbf{v} + \mathbf{w}) = \alpha(\mathbf{v}) + \alpha(\mathbf{w}) = \beta(\mathbf{x}_1) + \beta(\mathbf{y}_1) = \beta(\mathbf{x}) + \beta(\mathbf{y})$, $\beta(k\mathbf{x}) = \alpha(k\mathbf{v}) = k\alpha(\mathbf{v}) = k\beta(\mathbf{x}_1) = k\beta(\mathbf{x})$. So β is a linear map, i.e. $\beta \in W^*$.

Then $\forall \mathbf{v} \in V$, $\alpha(\mathbf{v}) = \beta(L(\mathbf{v})) = L^*(\beta)(\mathbf{v})$, i.e. $\alpha \in \text{Ran}(L^*)$. □

Exercise 1.6.5

1. Bilinear:

$$\begin{aligned} (\lambda \mathbf{a} + \mu \mathbf{b}, \mathbf{w}) &= (\lambda \mathbf{a} + \mu \mathbf{b})^T A \mathbf{w} \\ &= \lambda \mathbf{a}^T A \mathbf{w} + \mu \mathbf{b}^T A \mathbf{w} \\ &= \lambda(\mathbf{a}, \mathbf{w}) + \mu(\mathbf{b}, \mathbf{w}) \\ (\mathbf{v}, \lambda \mathbf{a} + \mu \mathbf{b}) &= \mathbf{v}^T A(\lambda \mathbf{a} + \mu \mathbf{b}) \\ &= \lambda \mathbf{v}^T A \mathbf{a} + \mu \mathbf{v}^T A \mathbf{b} \\ &= \lambda(\mathbf{v}, \mathbf{a}) + \mu(\mathbf{v}, \mathbf{b}) \end{aligned}$$

Symetric: Note that A is symetric, we have $(\mathbf{v}, \mathbf{w}) = (\mathbf{v}, \mathbf{w})^T = \mathbf{w}^T A^T \mathbf{v} = \mathbf{w}^T A \mathbf{v} = (\mathbf{w}, \mathbf{v})$.

Positive definite: Note that A is positive definite, we know

$$\begin{cases} \mathbf{v}^T A \mathbf{v} > 0 & \iff \mathbf{v} \neq \mathbf{0} \\ \mathbf{v}^T A \mathbf{v} = 0 & \iff \mathbf{v} = \mathbf{0} \end{cases}$$

So we have $(\mathbf{v}, \mathbf{v}) \geq 0$ and $(\mathbf{v}, \mathbf{v}) = 0 \iff \mathbf{v} = \mathbf{0}$.

2. Note that $(A^{-1}\mathbf{v}, \mathbf{w}) = \mathbf{v}^T (A^{-1})^T A^T \mathbf{w} = \mathbf{v}^T (A^T)^{-1} A \mathbf{w} = \mathbf{v}^T \mathbf{w}$, we have $\text{Riesz}(\mathbf{v}^T) = A^{-1}\mathbf{v}$.

3. We know $\text{Riesz}^{-1}(\mathbf{v}) = (\mathbf{v}|)$, so $\text{Riesz}^{-1}(\mathbf{v}) = \mathbf{v}^T A$.

4. First, we prove

Lemma For any linear bijection $A : V \rightarrow W$, $(A^*)^{-1} = (A^{-1})^*$.

Proof. Suppose arbitrary $\mathbf{v} \in V, \mathbf{w} \in W, \alpha \in V^*, \beta \in W^*$ s.t. $\mathbf{w} = A(\mathbf{v}), \alpha = A^*(\beta)$. We have $(A^{-1})^*(\alpha)(\mathbf{w}) = \alpha \circ A^{-1}(\mathbf{w}) = \alpha(\mathbf{v}) = A^*(\beta)(\mathbf{v}) = \beta \circ A(\mathbf{v}) = \beta(\mathbf{w}) = (A^*)^{-1}(\alpha)(\mathbf{w})$. Thus, $(A^*)^{-1} = (A^{-1})^*$.

Then, let L be the bra map $L : \mathbb{R}^n \rightarrow (\mathbb{R}^n)^*$, $\mathbf{v}, \mathbf{w} \in \mathbb{R}^n$. Consider $L^* : (\mathbb{R}^n)^{**} \rightarrow (\mathbb{R}^n)^*$, we have $L^*(ev_{\mathbf{v}})(\mathbf{w}) = ev_{\mathbf{v}} \circ L(\mathbf{w}) = ev_{\mathbf{v}}(\mathbf{w}|) = (\mathbf{w}, \mathbf{v}) = (\mathbf{v}, \mathbf{w}) = L(\mathbf{v})(\mathbf{w})$. With the canonical isomorphism between $(\mathbb{R}^n)^{**}$ and \mathbb{R}^n , we can say $L = L^*$. Note that $\text{Riesz} = L^{-1}$, from **Lemma**, we know $\text{Riesz}^* = \text{Riesz}$. So $\text{Riesz}^*(\mathbf{v}^T) = \text{Riesz}(\mathbf{v}^T) = A^{-1}\mathbf{v}$.

5. From the previous subproblem, we know $\text{Riesz}^* = \text{Riesz}$, then $\text{Riesz}^{ad} = \text{Riesz}^{-1} \circ \text{Riesz}^* \circ \text{Riesz}^{-1} = \text{Riesz}^{-1} \circ \text{Riesz} \circ \text{Riesz}^{-1} = \text{Riesz}^{-1}$. We have $\text{Riesz}^{ad}(\mathbf{v}) = \text{Riesz}^{-1}(\mathbf{v}) = \mathbf{v}^T A$. □