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## HW 1

**1** Let  $V$  be a real finite dimensional vector space with inner product  $\langle \cdot, \cdot \rangle_V$ , and let  $B = \{v_1, v_2, \dots, v_n\}$  be an ordered basis for  $V$ .

a. If  $\Phi$  is the associated coordinate map, show that  $\langle \cdot, \cdot \rangle_{\mathbb{R}^n} := \langle \Phi^{-1}(\cdot), \Phi^{-1}(\cdot) \rangle$  defines an inner product on  $\mathbb{R}^n$ .

b. Show that if  $x, y \in \mathbb{R}^n$ , then  $\langle \cdot, \cdot \rangle_{\mathbb{R}^n} = y^T G x$ , where  $G_{jk} = \langle v_k, v_j \rangle_V$ .

*Proof.* a. Let  $x, y, z \in \mathbb{R}^n$  and  $\alpha, \beta \in \mathbb{R}$ .

Symmetry:  $\langle x, y \rangle = \langle \Phi^{-1}(x), \Phi^{-1}(y) \rangle = \langle \Phi^{-1}(y), \Phi^{-1}(x) \rangle = \langle y, x \rangle$ .

Bilinearity:

$$\begin{aligned} \langle \alpha x + \beta y, z \rangle &= \langle \Phi^{-1}(\alpha x + \beta y), \Phi^{-1}(z) \rangle \\ &= \langle \alpha \Phi^{-1}(x) + \beta \Phi^{-1}(y), \Phi^{-1}(z) \rangle \\ &= \alpha \langle \Phi^{-1}(x), \Phi^{-1}(z) \rangle + \beta \langle \Phi^{-1}(y), \Phi^{-1}(z) \rangle \\ &= \alpha \langle x, z \rangle + \beta \langle y, z \rangle. \end{aligned}$$

Linearity in the second component follows from symmetry.

Positivity:  $\langle x, x \rangle = \langle \Phi^{-1}x, \Phi^{-1}x \rangle \geq 0$  with equality iff  $\Phi^{-1}(x) = 0 \equiv x = 0$  since  $\Phi$  is an isomorphism.

b. Since both sides are linear in each variable, it suffices to check the equation for  $x = e_j$  and  $y = e_k$ .  $\langle e_j, e_k \rangle = \langle \Phi^{-1}(e_j), \Phi^{-1}(e_k) \rangle = \langle v_j, v_k \rangle = e_k^T G_{kj} e_j$ .  $\square$

**2** In the previous problem, suppose that  $B = \{v_1, v_2, \dots, v_n\}$  is simply a subset of vectors in  $V$  and  $U = \text{span}(B)$ . Show that  $B$  is a basis for  $U$  iff  $y^T G x$  is an inner product for  $\mathbb{R}^n$ .

*Proof.* The forward implication follows from (1). For the converse, suppose  $y^T G x$  is an inner product and  $\sum_{i=1}^n a_i v_i = 0$  for  $a_i \in \mathbb{R}$ . Then  $0 = \langle \sum_i a_i v_i, \sum_i a_i v_i \rangle = \sum_{i,j} a_i a_j \langle v_i, v_j \rangle = a^T G a$ , where  $a = (a_1, a_2, \dots, a_n)$ . Since  $G$  is positive definite,  $a = 0$ . Thus  $B$  is linearly independent.  $\square$

**3** Let  $U$  be a subspace of an inner product space  $V$ .

a. Fix  $v \in V$ . Show that  $p \in U$  satisfies  $\min_{u \in U} \|v - u\| = \|v - p\|$  iff  $v - p$  is orthogonal to the subspace  $U$ .

b. Show that  $p$  is unique, given that it exists for  $v$ .

c. Suppose  $p$  exists for every  $v \in V$ . Define  $P : V \rightarrow U$  by  $Pv := p$ . Show that  $P$  is linear and  $P^2 = P$ .

*Proof.* a. Suppose  $\min_{u \in U} \|v - u\| = \|v - p\|$ . If  $v - p$  is not orthogonal to  $U$ , then we can pick  $u \in U$  such that  $\langle v - p, u \rangle \neq 0$ . By multiplying  $u$  by the appropriate phase, WLOG  $\langle v - p, u \rangle > 0$ . Let  $t \in \mathbb{R}$ . Then  $\|v - p - tu\|^2 =$

$\|v - p\|^2 - 2t\langle v - p, u \rangle + t^2\|u\|^2$ , which is minimized when  $t = \frac{\langle v - p, u \rangle}{\|u\|^2}$ . This contradicts the minimality of  $p$ .

Conversely suppose  $v - p$  is orthogonal to  $U$ . Then for any  $u \in U$ , we have  $\|v - u\|^2 = \|v - p + (p - u)\|^2 = \|v - p\|^2 + \|p - u\|^2$ . This is minimized when  $u = p$ .

b. Suppose both  $p$  and  $q$  satisfy the conditions in (a). Note that the orthogonal complement to  $U$ ,  $U^\perp$ , is a subspace. Moreover if  $u \in U \cap U^\perp$ , then  $\langle u, u \rangle = 0$ , so  $u = 0$ . Hence, since  $v - p, v - q \in U^\perp$ , we have  $(v - p) - (v - q) = q - p \in U^\perp$ . Thus,  $q - p \in U \cap U^\perp$ , so  $q = p$ .

c. To see that  $P$  is linear, let  $\alpha, \beta \in \mathbb{C}$  and  $v, w \in V$ . Then for any  $u \in U$ , we have  $\langle \alpha v + \beta w - (\alpha P(v) + \beta P(w)), u \rangle = \alpha \langle v - P(v), u \rangle + \beta \langle w - P(w), u \rangle = 0$ . Hence,  $P(\alpha v + \beta w) = \alpha P(v) + \beta P(w)$ .

To see that  $P^2 = P$ , let  $v \in V$  and  $p = P(v)$ . Then  $P^2(v) = P(p) = \min_{u \in U} \|p - u\| = p = P(v)$ .  $\square$

**4** Let  $V, B, U, G$  be defined as in problem 1, except  $B$  is a basis for  $U$ .

a. Let  $v \in V$  and  $d_k := \langle v, v_k \rangle_V$ . Show that  $p = \sum_j x_j v_j \in U$  is the orthogonal projection of  $v$  onto  $U$  iff the  $x_j$ 's satisfy the *normal equations*,  $d_k = \sum_j G_{kj} x_j$ .

b. Show that the orthogonal projection  $P : V \rightarrow U$  exists.

c. Show that if  $B$  is orthonormal, then  $Pv = \sum_j \langle v, v_j \rangle_V v_j$ .

*Proof.* a. Suppose that  $p$  is the orthogonal projection of  $v$  onto  $U$ . Then  $\langle p - v, v_k \rangle = 0$  for every  $k$ . Hence  $d_k = \langle v, v_k \rangle = \langle p, v_k \rangle = \langle \sum_j x_j v_j, v_k \rangle = \sum_j G_{kj} x_j$ .

Conversely, suppose  $d_k = \sum_j G_{kj} x_j$  for each  $k$ . Then  $\langle v, v_k \rangle = \langle \sum_j x_j v_j, v_k \rangle = \langle p, v_k \rangle$  for each  $k$ . Hence  $\langle p - v, v_k \rangle = 0$  for each  $k$ , so  $\langle p - v, u \rangle = 0$  for all  $u \in U$  since  $(v_k)$  is a basis for  $U$ .

b. Suppose  $z \in \ker(G)$ . Then  $z^T G z = 0$ , so  $z = 0$  since  $G$  is positive definite. Hence  $G$  is invertible, so we can define  $x_j = \sum_k (G^{-1})_{jk} d_k = \sum_k (G^{-1})_{jk} \langle v, v_k \rangle_V$ .

c. If  $B$  is orthonormal,  $G_{jk} = \langle v_k, v_j \rangle_V = \delta_{ij}$ . Hence from (b),  $Pv = \sum_j x_j v_j = \sum_{j,k} (G^{-1})_{jk} \langle v, v_k \rangle_V v_j = \sum_j \langle v, v_j \rangle_V v_j$ .  $\square$

**5** Equality holds in Schwarz's inequality iff  $u$  and  $v$  are linearly dependent.

*Proof.* Suppose  $u, v$  are linearly dependent. If either  $u$  or  $v$  is zero, then equality holds trivially. Otherwise,  $u = kv$  for some scalar  $k$ . Then  $|\langle u, v \rangle| = |\langle u, kv \rangle| = |k| \|u\|^2 = \|u\| \|v\|$ .

Conversely, suppose  $|\langle u, v \rangle| = \|u\| \|v\|$ . If either  $u$  or  $v$  is zero, we are done.

Otherwise, let  $t = \frac{\langle u, v \rangle}{\|v\|^2}$ . Then

$$\begin{aligned}
\|u - tv\|^2 &= \|u\|^2 - 2\Re(t\langle v, u \rangle) + |t|^2\|v\|^2 \\
&= \|u\|^2 - \frac{2}{\|v\|^2}|\langle u, v \rangle|^2 + \frac{|\langle u, v \rangle|^2}{\|v\|^2} \\
&= \|u\|^2 - 2\|u\|^2 + \|u\|^2 \\
&= 0.
\end{aligned}$$

Hence,  $u - tv = 0$ , so  $u, v$  are linearly dependent.  $\square$

**6** Suppose that  $F \in C[0, 1]$ ,  $F(x) \geq 0$ , and  $F(x_0) > 0$  for some  $x_0 \in [0, 1]$ . Show that there is a closed interval  $[a, b] \subset [0, 1]$ ,  $a \neq b$ , that contains  $x_0$  and on which  $F(x) \geq \frac{1}{2}F(x_0)$ .

*Proof.* Since  $F$  is continuous, there exists  $\delta > 0$  such that  $|x - x_0| < \delta$  implies  $|F(x) - F(x_0)| < F(x_0)/2$ . Hence, if  $x \in [x_0 - \delta/2, x_0 + \delta/2]$ , then  $F(x) - F(x_0) > -F(x_0)/2$  which implies  $F(x) > F(x_0)/2$ .  $\square$

**7** Let  $b = \{v_1, \dots, v_n\}$  be a basis for a vector space  $V$ . Define linear functionals  $\phi_k$  for  $1 \leq k \leq n$  via  $\phi_k(v_j) = \delta_{jk}$ .

1. Show that  $B^* := \{\phi_1, \dots, \phi_n\}$  is a basis for  $V^*$ .

2. Let  $V = \mathbb{R}^n$  and suppose that  $B = \{x_1, \dots, x_n\}$  is a basis of column vectors for  $\mathbb{R}^n$ , and let  $X = [x_1 \cdots x_n]$ . Show that  $R^{n*}$  may be identified with the set of  $1 \times n$  row vectors, and that  $B^*$  is then the set of rows of  $X^{-1}$ .

*Proof.* 1) To see that  $B^*$  is linearly independent, suppose  $\sum_{i=1}^n a_i \phi_i = 0$  for some  $a_i \in \mathbb{R}$ . For any  $v_j$ , applying the left-hand side to  $v_j$  yields  $a_j = 0$ . Hence,  $a_i = 0$  for all  $i$ .

To see that  $B^*$  spans  $V^*$ , let  $\psi \in V^*$  and  $x \in V$ . Then  $x = \sum_i x_i v_i$  for some  $x_i \in \mathbb{R}$ . Then  $\psi(x) = \sum_i x_i \psi(v_i) = \sum_i x_i \psi(v_i) \phi_i(v_i) = \sum_i \psi(v_i) \phi_i(x_i v_i) = \sum_i \psi(v_i) \phi_i(x) = (\sum_i \psi(v_i) \phi_i)x$ . Hence,  $\psi = \sum_i \psi(v_i) \phi_i$ .

2) For  $\psi \in \mathbb{R}^{n*}$  let  $T(\psi) = (\psi(e_1), \psi(e_2), \dots, \psi(e_n))$ . Then  $T(\psi)$  acting by matrix multiplication on column vectors in  $\mathbb{R}^n$  is a linear functional. Moreover,  $T(\psi)(e_j) = \psi(e_j)$  for each  $j$ , so  $T(\psi)$  and  $\psi$  must agree everywhere.

For each  $j, k$ ,  $T(\phi_k)(x_j) = \delta_{jk}$ . Hence if  $Y$  is the matrix with rows  $T(\phi_k)$  for  $1 \leq k \leq n$ , then  $YX = I$ . Thus,  $Y = X^{-1}$ .  $\square$