## In the problems below,  $(V, \langle , \rangle)$  is a finite-dimensional inner-product space of positive dimension  $n$ .

**NB 15.1.** (a) Let H be a subspace of V. In class, we defined a map  $\text{proj}_H : V \to H$ called the *orthogonal projection from V to H*. We also showed that  $V = H \oplus H^{\perp}$ . You saw maps that were called *projections* earlier this semester, in the exercises for FIS Section 2.1: given any subspaces  $W_1, W_2$  of V for which  $V = W_1 \oplus W_2$ , a map that the book called the projection (of V) on  $W_1$  along  $W_2$  was defined. Show that, in this terminology, the map  $proj_H$  is precisely the projection of V on H along  $H^{\perp}$ .

(b) Show that, for all  $v \in V$ , the relation

$$
\text{proj}_H(v) + \text{proj}_{H^\perp}(v) = v. \tag{1}
$$

(Equivalently, in the vector space  $\mathcal{L}(V, V)$ , we have

$$
\text{proj}_H + \text{proj}_{H^\perp} = I_V \tag{2}
$$

(c) Do the analogs of equation (1) and (2) hold for any direct-sum decomposition of an arbitrary vector space  $V'$  as  $W_1 \oplus W_2$ , where  $W_1, W_2$  are complementary subspace of  $V'$ ), or is it special to inner-product spaces and orthgonal complements?

**NB 15.2** Consider the case in which  $V = \mathbb{R}^n$ , where  $n \geq 2$ , and  $\langle , \rangle$  is the standard inner product on  $\mathbb{R}^n$ . (For notational ease, in this problem I'll write elements of  $\mathbb{R}^n$ just as ordered *n*-tuples rather than as column vectors.) Let  ${e_i}_{i=1}^n$  be the standard ordered basis of  $\mathbf{R}^n$ .

For 
$$
k \in \{1.2, ..., n-1\}
$$
 let  
\n
$$
H_k = \{(a_1, ..., a_n) \in \mathbb{R}^n : a_i = 0 \text{ for all } i > k\}
$$
\n
$$
= \{(a_1, ..., a_k, 0, 0, ..., 0) : a_1, ..., a_k \in \mathbb{R}\}
$$
\n
$$
= \text{span}\{\mathbf{e}_1, ..., \mathbf{e}_k\}.
$$

(a) Explicitly, what is the subspace  $(H_k)^{\perp}$ ?

(b) Let  $\mathbf{a} = (a_1, \ldots, a_n)$  and let  $k \in \{1, 2, \ldots, n-1\}$ . Compute  $\text{proj}_{H_k}(\mathbf{a})$  and  $\text{proj}_{(H_k)\perp}(\mathbf{a})$  in terms of  $a_1, \ldots a_n$ . (Once you find the first of these, you should be able to write down the the second immediately, without any real computations, from earlier problem(s) on this page.)

**NB 15.3.** Let 
$$
\beta = \{u_1, ..., u_n\}
$$
 be an orthonormal basis of *V*, and let  $v, w \in V$ . Let  $\mathbf{a} = [v]_{\beta} = \begin{pmatrix} a_1 \\ \vdots \\ a_n \end{pmatrix}$  and  $\mathbf{b} = [w]_{\beta} = \begin{pmatrix} b_1 \\ \vdots \\ b_n \end{pmatrix}$ . (Thus  $v = \sum_i a_i u_i$  and  $w = \sum_i b_i u_i$ .)

(a) Show that

$$
\langle v, w \rangle = \sum_{i=1}^{n} a_i b_i = [v]_{\beta} \cdot [w]_{\beta}.
$$
 (3)

In other words, in an orthonormal basis of  $(V, \langle , \rangle)$ , the inner product "looks just like" dot-product. (This is NOT true in an *arbitrary* basis!!)

(b) From equation (3), deduce that if  $a_1, \ldots, a_n$  are the coordinates of a vector  $v \in V$  with respect to an orthonormal basis,

$$
||v||^2 = a_1^2 + a_2^2 + \dots + a_n^2.
$$
 (4)

(Again, this is NOT true for an arbitrary basis!!)