

# 1 Some examples

1. The functions below are examples of inner products:

(a).  $V = \mathbb{C}([0, 1]) = \{f : [0, 1] \rightarrow \mathbb{C} \text{ continuous}\}.$   
 $\langle f, g \rangle = \int_0^1 f \bar{g}$

(b).  $V = M_n(\mathbb{C}), \langle A, B \rangle = \text{tr}(AB^*),$  where  $B^* = \overline{B}^t.$

Proof: The conditions of the inner products can be established as below:

- $\langle A + B, C \rangle = \text{tr}((A + B)C^*) = \text{tr}(AC^* + BC^*) = \text{tr}(AC^*) + \text{tr}(BC^*) = \langle A + C, B + C \rangle.$
- for any constant  $c, \langle cA, B \rangle = \text{tr}(c(AB^*)) = c \text{tr}(AB^*) = c \langle A, B \rangle.$
- $\langle A, B \rangle = \text{tr}(AB^*) = \text{tr}(\overline{AB^t}) = \sum (\overline{AB^t})_{ii} = \sum A_{ij} \overline{B_{ji}} = \sum A_{ij} \overline{B_{ij}}, \forall 1 \leq i, j \leq n.$  Similarly,  $\langle B, A \rangle = \sum B_{ij} \overline{A_{ij}}.$  Now for  $a, b \in \mathbb{C}$  we have  $\overline{\overline{a} + \overline{b}} = \overline{\overline{a} + \overline{b}}, \overline{ab} = \overline{a} \overline{b}$  and  $\overline{\overline{a}} = a.$  Therefore  $\overline{ab} = \overline{a} \overline{b} = \overline{ab}.$  This gives  $A_{ij} \overline{B_{ij}} = \overline{B_{ij} \overline{A_{ij}}}$  and therefore  $\langle A, B \rangle = A_{ij} \overline{B_{ij}} = \overline{B_{ij} \overline{A_{ij}}} = \overline{\langle B, A \rangle}.$
- From above,  $\langle A, A \rangle = \sum A_{ij} \overline{A_{ij}} = \sum \|A_{ij}\|^2.$  This is obviously nonnegative, and it is zero if and only if all  $\|A_{ij}\|$ 's are zero, meaning that  $A_{ij}$  must be itself a zero (i.e. a zero vector).

2. In assignment 1 problem 1, we have seen that the pairing isn't an inner product because there exists nonzero vector  $\mathbf{x}$  satisfying  $\langle \mathbf{x}, \mathbf{x} \rangle = 0.$  We now show that the pairing  $\mathbf{x}^t \begin{pmatrix} 1 & i \\ -i & 1 \end{pmatrix} \overline{\mathbf{y}}$  satisfies all other properties.

Notice that, if  $\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$  and  $\mathbf{y} = \begin{pmatrix} y_1 \\ y_2 \end{pmatrix}$  then

$$\langle \mathbf{x}, \mathbf{y} \rangle = \mathbf{x}^t A \overline{\mathbf{y}} = \begin{pmatrix} x_1 & x_2 \end{pmatrix} \begin{pmatrix} 1 & i \\ -i & 1 \end{pmatrix} \begin{pmatrix} \overline{y_1} \\ \overline{y_2} \end{pmatrix} = \begin{pmatrix} x_1 & x_2 \end{pmatrix} \begin{pmatrix} \overline{y_1} + i\overline{y_2} \\ -i\overline{y_1} + \overline{y_2} \end{pmatrix} = \begin{pmatrix} x_1(\overline{y_1} + i\overline{y_2}) + x_2(-i\overline{y_1} + \overline{y_2}) \end{pmatrix}.$$

We establish the following:

- $\langle \mathbf{x} + \mathbf{y}, \mathbf{z} \rangle = \mathbf{x} + \mathbf{y}^t A \overline{\mathbf{z}} = (\mathbf{x}^t + \mathbf{y}^t) A \overline{\mathbf{z}} = \mathbf{x}^t A \overline{\mathbf{z}} + \mathbf{y}^t A \overline{\mathbf{z}} = \langle \mathbf{x}, \mathbf{z} \rangle + \langle \mathbf{y}, \mathbf{z} \rangle.$
- For any constant  $c, \langle c\mathbf{x}, \mathbf{y} \rangle = (c\mathbf{x}^t) A \overline{\mathbf{y}} = c(\mathbf{x}^t A \overline{\mathbf{y}}) = c\langle \mathbf{x}, \mathbf{y} \rangle.$
- Before proving  $\langle \mathbf{x}, \mathbf{y} \rangle = \overline{\langle \mathbf{y}, \mathbf{x} \rangle},$  we need the following properties about complex numbers: for any complex numbers  $a$  and  $b,$  we have  $\overline{\overline{a} + \overline{b}} = \overline{\overline{a} + \overline{b}};$  for any complex numbers  $a$  and  $b, \overline{a \cdot b} = \overline{a} \overline{b}.$  Therefore,

$$\langle \mathbf{x}, \mathbf{y} \rangle = \begin{pmatrix} x_1(\overline{y_1} + i\overline{y_2}) + x_2(-i\overline{y_1} + \overline{y_2}) \end{pmatrix}, \langle \mathbf{y}, \mathbf{x} \rangle = \begin{pmatrix} y_1(\overline{x_1} + i\overline{x_2}) + y_2(-i\overline{x_1} + \overline{x_2}) \end{pmatrix}.$$

We have  $x_1 \overline{y_1} = \overline{\overline{x_1} y_1} = \overline{\overline{x_1} y_1},$  and similarly  $x_2 \overline{y_2} = \overline{\overline{x_2} y_2} = \overline{\overline{x_2} y_2}.$  In addition,  $i(x_1 \overline{y_2} - x_2 \overline{y_1}) = i(\overline{\overline{x_1} y_2} - \overline{\overline{x_2} y_1}) = -i(\overline{\overline{x_2} y_1} - \overline{\overline{x_1} y_2}) = \overline{i(\overline{x_2} y_1 - \overline{x_1} y_2)} = \overline{i(\overline{x_2} y_1 - \overline{x_1} y_2)}.$  Therefore,

$$x_1(\overline{y_1} + i\overline{y_2}) + x_2(-i\overline{y_1} + \overline{y_2}) = x_1 \overline{y_1} + x_2 \overline{y_2} + i(x_1 \overline{y_2} - x_2 \overline{y_1}) = \overline{\overline{x_1} y_1} + \overline{\overline{x_2} y_2} + \overline{i(\overline{x_2} y_1 - \overline{x_1} y_2)} = \overline{\overline{x_1} y_1 + \overline{x_2} y_2 + i(\overline{x_2} y_1 - \overline{x_1} y_2)} \\ = \overline{y_1(\overline{x_1} + i\overline{x_2}) + y_2(-i\overline{x_1} + \overline{x_2})}, \text{ establishing the claim.}$$

- Now  $\langle \mathbf{x}, \mathbf{x} \rangle = \begin{pmatrix} x_1(\overline{x_1} + i\overline{x_2}) + x_2(-i\overline{x_1} + \overline{x_2}) \end{pmatrix} = (x_1 \overline{x_1} + x_2 \overline{x_2} + i(x_1 \overline{x_2} - x_2 \overline{x_1}) = |x_1|^2 + |x_2|^2 + i x_1 \overline{x_2} + (-i) \overline{x_1} x_2 = |x_1|^2 + |x_2|^2 + i x_1 \overline{x_2} + \overline{i x_1 \overline{x_2}} = |x_1|^2 + |x_2|^2 + i x_1 \overline{x_2} + \overline{i x_1 \overline{x_2}} = |x_1|^2 + |x_2|^2 + 2\text{Re}(i x_1 \overline{x_2}),$  because  $a + \overline{a} = 2\text{Re}(a).$  Now,  $|2\text{Re}(i x_1 \overline{x_2})| \leq |2(i x_1 \overline{x_2})| \leq 2|x_1 x_2|$  so  $-2|x_1 x_2| \leq |2\text{Re}(i x_1 \overline{x_2})| \leq 2|x_1 x_2|,$  so  $|x_1|^2 + |x_2|^2 + 2\text{Re}(i x_1 \overline{x_2}) \geq |x_1|^2 + |x_2|^2 - 2|x_1||x_2| = (|x_1| - |x_2|)^2,$  so the pairing is always nonnegative. Notice, however, it could happen that this quantity is indeed 0 even with both  $x_1, x_2$  nonzero.

# 2 Proofs of identities

1. Given basis  $\{\vec{w}_1, \dots, \vec{w}_n\}$  of an inner product space, prove that the the set of vectors  $\{\vec{v}_1, \dots, \vec{v}_n\}$  defined as  $\vec{v}_1 = \vec{w}_1$  and

$$\vec{v}_k = \vec{w}_k - \sum_{i=1}^{k-1} \frac{\langle \vec{w}_k, \vec{v}_i \rangle}{\|\vec{v}_i\|^2} \vec{v}_i, \quad \forall k \in [2, n]$$

is an orthogonal basis.

**Proof:** (Credits to textbook and prof). First, we prove that  $\langle \vec{i}, \vec{j} \rangle = 0, \forall i \neq j$ . We also proceed by inducting on  $n$ . Base case where  $n = 1$  is trivial. Suppose the claim holds for  $n = 1, 2, \dots, k-1$  for some  $k$ , we have: for any  $j < k$ ,

$$\begin{aligned} \langle \vec{v}_k, \vec{v}_j \rangle &= \langle \vec{w}_k - \sum_{i=1}^{k-1} \frac{\langle \vec{w}_k, \vec{v}_i \rangle}{\|\vec{v}_i\|^2} \vec{v}_i, \vec{v}_j \rangle = \langle \vec{w}_k, \vec{v}_j \rangle - \left\langle \sum_{i=1}^{k-1} \frac{\langle \vec{w}_k, \vec{v}_i \rangle}{\|\vec{v}_i\|^2} \vec{v}_i, \vec{v}_j \right\rangle = \langle \vec{w}_k, \vec{v}_j \rangle - \sum_{i=1}^{k-1} \frac{\langle \vec{w}_k, \vec{v}_i \rangle}{\|\vec{v}_i\|^2} \langle \vec{v}_i, \vec{v}_j \rangle \\ &= \langle \vec{w}_k, \vec{v}_j \rangle - \frac{\langle \vec{w}_k, \vec{v}_j \rangle}{\|\vec{v}_j\|^2} \langle \vec{v}_j, \vec{v}_j \rangle = \langle \vec{w}_k, \vec{v}_j \rangle - \langle \vec{w}_k, \vec{v}_j \rangle = 0, \end{aligned}$$

justifying the claim. (By induction hypothesis,  $\langle \vec{i}, \vec{j} \rangle = 0$  for any  $i < j < k$ .)

Next, notice that none of the vectors  $\vec{v}_i$  can be zero; each of the vectors  $\vec{v}_k$  can be written as the linear combination of  $\vec{w}_1, \dots, \vec{w}_k$ , with the coefficient of  $\vec{w}_k$  being 1. Since  $\vec{w}_1, \dots, \vec{w}_k$  are linearly independent, the claim follows.

Finally, in class we have seen that a set of nonzero orthogonal vectors must be linearly independent. Since the set of vectors  $\{\vec{v}_1, \dots, \vec{v}_n\}$  has  $n$  elements and are linearly independent, this set is also a basis. The conclusion follows.

2. Given a finite dimensional inner-product space  $V$  and let  $W$  be its subspace with orthonormal basis  $\{\vec{w}_1, \dots, \vec{w}_k\}$ . Then for each  $\vec{v} \in V$  there exists a unique  $\vec{w} \in W$  and  $\vec{w}' \in W^\perp$  satisfying  $\vec{w} + \vec{w}' = \vec{v}$ , given by the following formula:

$$\vec{w} = \sum_{i=1}^k \frac{\langle \vec{v}, \vec{w}_i \rangle}{\|\vec{w}_i\|^2} \vec{w}_i, \vec{w}' = \vec{v} - \vec{w}.$$

**Proof:** since a subspace (or a vector space, in general) is closed under addition,  $\vec{w}$  described above is in  $W$ . To show that  $\vec{w}' \in W^\perp$ , we notice the following for all  $j \in [1, n]$ :

$$\langle \vec{w}', \vec{w}_j \rangle = \langle \vec{v} - \sum_{i=1}^k \frac{\langle \vec{v}, \vec{w}_i \rangle}{\|\vec{w}_i\|^2} \vec{w}_i, \vec{w}_j \rangle = \langle \vec{v}, \vec{w}_j \rangle - \left\langle \sum_{i=1}^k \frac{\langle \vec{v}, \vec{w}_i \rangle}{\|\vec{w}_i\|^2} \vec{w}_i, \vec{w}_j \right\rangle = \langle \vec{v}, \vec{w}_j \rangle - \sum_{i=1}^k \frac{\langle \vec{v}, \vec{w}_i \rangle}{\|\vec{w}_i\|^2} \langle \vec{w}_i, \vec{w}_j \rangle = \langle \vec{v}, \vec{w}_j \rangle - \langle \vec{v}, \vec{w}_j \rangle = 0,$$

because  $\langle \vec{w}_i, \vec{w}_j \rangle$  vanishes whenever  $i \neq j$ , and  $\frac{\langle \vec{v}, \vec{w}_j \rangle}{\|\vec{w}_j\|^2} \langle \vec{w}_j, \vec{w}_j \rangle = \langle \vec{v}, \vec{w}_j \rangle$ .

To show that the numbers  $\vec{w}$  and  $\vec{w}'$  are unique, suppose that there exists  $\vec{w}_1, \vec{w}_2 \in W$  and  $\vec{w}'_1, \vec{w}'_2 \in W^\perp$  satisfying  $\vec{w}_1 + \vec{w}'_1 = \vec{w}_2 + \vec{w}'_2$ . Now,  $\vec{w}_1 - \vec{w}_2 \in W$  and  $\vec{w}'_1 - \vec{w}'_2 = -(\vec{w}_1 - \vec{w}_2) \in W^\perp$ , which means the vector  $\vec{w}_1 - \vec{w}_2$  is in both  $W$  and  $W^\perp$  (the product of any vector in  $W$  and any scalar constant is also in  $W$ ). Notice, however, that this means  $\|\vec{w}_1 - \vec{w}_2\| = 0$  by the definition of  $W$  and  $W^\perp$ , so  $\vec{w}_1 - \vec{w}_2 = 0$  or  $\vec{w}_1 = \vec{w}_2$ , showing that such pair of numbers must be unique.

3. Let  $V$  be a finite dimensional transformation. Then for each transformation  $T : V \rightarrow V$  there is a unique transformation  $T^*$  satisfying  $\langle T(\vec{x}), \vec{y} \rangle = \langle \vec{x}, T^*(\vec{y}) \rangle$  for all  $\vec{x}, \vec{y} \in V$ .

**Proof:** Let  $n$  be the dimension of  $V$ , and denote  $\{\vec{v}_1, \dots, \vec{v}_n\}$  by an orthonormal basis of  $V$ . We use the fact that each linear transformation is uniquely determined by the values of  $T(\vec{v}_1), \dots, T(\vec{v}_n)$ . That is, for each  $n$ -tuples of vectors  $\{\vec{w}_1, \dots, \vec{w}_n\}$  there is a unique linear transformation  $T$  such that  $T(\vec{v}_i) = \vec{w}_i$ . Suppose

that numbers  $a_{ij} \in \mathbb{C}, 1 \leq i, j \leq n$  are such that  $T(\vec{v}_i) = \sum_{j=1}^n a_{ij} \vec{v}_j$ , we have, for each  $i, k$ ,  $\langle T(\vec{v}_i), \vec{v}_k \rangle =$

$\left\langle \sum_{j=1}^n a_{ij} \vec{v}_j, \vec{v}_k \right\rangle = a_{ik}$ . Suppose that there is a linear transformation  $T^*$  such that  $\langle T(\vec{x}), \vec{y} \rangle = \langle \vec{x}, T^*(\vec{y}) \rangle$  for all

$\vec{x}, \vec{y} \in V$ . Let  $b_{ij}$  be numbers such that  $T^*(\vec{v}_i) = \sum_{j=1}^n b_{ij} \vec{v}_j$  then we have  $a_{ik} = \langle T(\vec{v}_i), \vec{v}_k \rangle = \langle (\vec{v}_i), T^*(\vec{v}_k) \rangle =$

$\overline{\langle T^*(\vec{v}_k), \vec{v}_i \rangle} = \overline{\left\langle \sum_{j=1}^n b_{kj} \vec{v}_j, \vec{v}_i \right\rangle} = \overline{a_{ki}}$ , therefore we must have  $T^*(\vec{v}_i) = \sum_{j=1}^n b_{ij} \vec{v}_j = T^*(\vec{v}_i) = \sum_{j=1}^n \overline{b_{ji}} \vec{v}_j$ . This uniquely defines  $T^*$ .

Conversely, let  $T^*$  be as defined, given  $T$ . From above we already have the relation  $\langle T(\vec{v}_i), \vec{v}_k \rangle = \langle (\vec{v}_i), T^*(\vec{v}_k) \rangle$  for each pair of orthonormal basis. Let  $\vec{x} = \sum_{i=1}^n x_i \vec{v}_i$  and  $\vec{y} = \sum_{i=1}^n y_i \vec{v}_i$  then we have

$$\begin{aligned} \langle T(\vec{x}), \vec{y} \rangle &= \langle T(\sum_{i=1}^n x_i \vec{v}_i), \sum_{i=1}^n y_i \vec{v}_i \rangle = \langle \sum_{i=1}^n x_i T(\vec{v}_i), \sum_{i=1}^n y_i \vec{v}_i \rangle = \sum_{i=1}^n \sum_{j=1}^n x_i y_j \langle T(\vec{v}_i), \vec{v}_j \rangle \\ &= \sum_{i=1}^n \sum_{j=1}^n x_i y_j \langle \vec{v}_i, T^*(\vec{v}_j) \rangle = \langle \sum_{i=1}^n x_i \vec{v}_i, \sum_{j=1}^n y_j T^*(\vec{v}_j) \rangle = \langle \sum_{i=1}^n x_i \vec{v}_i, T^*(\sum_{j=1}^n y_j \vec{v}_j) \rangle = \langle \vec{x}, T^*(\vec{y}) \rangle \end{aligned}$$

4. Let  $A = [T]_\beta$  for some orthonormal basis  $\beta$  is a finite dimensional space  $V$ . Then  $[T]_\beta^* = [T^*]_\beta$ .

**Proof:** Let our orthonormal basis be  $\{\vec{v}_1, \dots, \vec{v}_n\}$ . This proof relies on the following fact:  $A_{ij} = \langle T(\vec{v}_j), \vec{v}_i \rangle$ .

This is because for each  $j$ ,  $[T\vec{v}_j]_\beta = [T]_\beta[\vec{v}_j]_\beta = \text{Col}_j(A)$  so  $\langle T(\vec{v}_j), \vec{v}_i \rangle = \sum_{k=1}^n A_{kj} \langle \vec{v}_k, \vec{v}_i \rangle = A_{ij}$ , as desired.

Thus for each  $i, j$  we have  $([T]_\beta^*)_{ij} = (A^*)_{ij} = \overline{A_{ji}^t} = \overline{A_{ji}} = \overline{\langle T(\vec{v}_i), \vec{v}_j \rangle} = \overline{\langle \vec{v}_i, T^*(\vec{v}_j) \rangle} = \langle T^*(\vec{v}_j), \vec{v}_i \rangle = ([T^*]_\beta)_{ij}$ .