



Aalto University

Linear algebra

Exercise sheet 10 / Model solutions

1. (a) Let $\mathcal{V} \subset \mathbb{R}^n$ be a subspace and $\{v_1, \dots, v_k\}$ an *orthonormal basis* for \mathcal{V} with respect to the inner product $\langle \cdot, \cdot \rangle : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$. Show that for any $x \in \mathbb{R}^n$ it holds that

$$x^\perp := x - \sum_{i=1}^k \langle x, v_i \rangle v_i \in \mathcal{V}^\perp,$$

where $\mathcal{V}^\perp \subset \mathbb{R}^n$ is the orthogonal complement of \mathcal{V} with respect to the inner product $\langle \cdot, \cdot \rangle$, see Definition 2.3 on page 45 of the lecture notes. (This proves that equation (75) there does define a direct sum $\mathbb{R}^n = \mathcal{V} \oplus \mathcal{V}^\perp$.)

- (b) Prove that any projection matrix $P \in \mathbb{R}^{n \times n}$ is an orthogonal projection in the Euclidian inner product, if $P^T = P$.

Hints: (a) Write an arbitrary vector of the subspace \mathcal{V} in the form $v = \sum_{i=1}^k \alpha_i v_i$ and show by direct calculation that $\langle x^\perp, v \rangle = 0$. (b) With a projection P we may write the direct sum $\mathbb{R}^n = R(P) \oplus R(I - P)$; check equation (72) from the lecture notes. You have to show that $R(I - P) = R(P)^\perp$. Since moreover $\mathbb{R}^n = R(P) \oplus R(P)^\perp$, it's actually enough to prove that $R(I - P) \subseteq R(P)^\perp$.

Solution:

- (a) We may write as $v = \sum_{j=1}^k \alpha_j v_j$ an arbitrary element of the subspace $\mathcal{V} \subset \mathbb{R}^n$. Then

$$\begin{aligned} \langle x^\perp, v \rangle &= \left\langle x - \sum_{i=1}^k \langle x, v_i \rangle v_i, \sum_{j=1}^k \alpha_j v_j \right\rangle \\ &= \left\langle x, \sum_{j=1}^k \alpha_j v_j \right\rangle - \left\langle \sum_{i=1}^k \langle x, v_i \rangle v_i, \sum_{j=1}^k \alpha_j v_j \right\rangle \\ &= \sum_{j=1}^k \alpha_j \langle x, v_j \rangle - \sum_{i=1}^k \sum_{j=1}^k \alpha_j \langle x, v_i \rangle \langle v_i, v_j \rangle \\ &= \sum_{j=1}^k \alpha_j \langle x, v_j \rangle - \sum_{j=1}^k \alpha_j \langle x, v_j \rangle = 0, \end{aligned}$$

where we used for the fourth equality that

$$\langle v_i, v_j \rangle = \begin{cases} 1 & \text{if } i = j \\ 0 & \text{if } i \neq j. \end{cases}$$

Hence x^\perp is orthogonal to every element of \mathcal{V} , that is, $x^\perp \in \mathcal{V}^\perp$.

- (b) As suggested in the hint, let us show the inclusion $R(I - P) \subset R(P)^\perp$. Take $\mathbf{x} \in R(I - P)$, meaning that $\mathbf{x} = (I - P)\mathbf{z}$ for some $\mathbf{z} \in \mathbb{R}^n$. Let us show that \mathbf{x} is orthogonal to any element $\mathbf{y} = P\mathbf{w}$ of $R(P)$, where $\mathbf{w} \in \mathbb{R}^n$:

$$\begin{aligned}\langle \mathbf{x}, \mathbf{y} \rangle_2 &= \mathbf{x}^T \mathbf{y} \\ &= ((I - P)\mathbf{z})^T (P\mathbf{w}) \\ &= \mathbf{z}^T (I - P)^T P\mathbf{w} \\ &= \mathbf{z}^T (I^T - P^T) P\mathbf{w} \\ &= \mathbf{z}^T (I - P) P\mathbf{w} \\ &= \mathbf{z}^T (P - P^2)\mathbf{w} = 0,\end{aligned}$$

where in the fifth step we used that $P = P^T$ by assumption and in the last step that $P^2 = P$ because P is a projection. So we have that $\mathbf{x} \in R(P)^\perp$, that is, $R(I - P) \subseteq R(P)^\perp$.

Why is this enough? Namely, why must it be that $R(I - P) = R(P)^\perp$? Let $\mathbf{v} \in R(P)^\perp$. Since $\mathbb{R}^n = R(P) \oplus R(I - P)$ (see formula (72) on page 45 of the lecture notes) there exist uniquely determined vectors $\mathbf{w}_1 \in R(P)$ and $\mathbf{w}_2 \in R(I - P)$ such that

$$\mathbf{v} = \mathbf{w}_1 + \mathbf{w}_2.$$

Since $\mathbf{v} \in R(P)^\perp$, by taking the inner product with \mathbf{w}_1 we get

$$0 = \langle \mathbf{v}, \mathbf{w}_1 \rangle = \langle \mathbf{w}_1, \mathbf{w}_1 \rangle + \langle \mathbf{w}_2, \mathbf{w}_1 \rangle = \|\mathbf{w}_1\|^2,$$

where $\langle \mathbf{w}_2, \mathbf{w}_1 \rangle = 0$ by the inclusion proved in the exercise. So then $\mathbf{w}_1 = 0$ and $\mathbf{v} = \mathbf{w}_2 \in R(I - P)$.

2. Let

$$\mathcal{V} = \text{span} \left\{ \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \right\}.$$

- (a) Find an orthogonal projection P such that $R(P) = \mathcal{V}$.
 (b) Choose some other basis for the subspace \mathcal{V} . Compute P again in this new basis.

Solution: Let us consider the situation in general. We want to find a matrix $P \in \mathbb{R}^{n \times n}$ such that P is a projection matrix for the orthogonal projection to the subspace $R(V)$, where $V \in \mathbb{R}^{n \times m}$ is the matrix whose columns generate the subspace \mathcal{V} .

Let $\mathbf{x} \in \mathbb{R}^n$ be a vector we want to project. It can be written as

$$\mathbf{x} = V\mathbf{a} + W\mathbf{b}, \tag{1}$$

where $\mathbf{a} \in \mathbb{R}^m$, $\mathbf{b} \in \mathbb{R}^k$, and $W \in \mathbb{R}^{n \times k}$. Furthermore let us choose W so that $R([V \ W]) = \mathbb{R}^n$ (that is, the columns of the matrices V and W together generate \mathbb{R}^n) and $V^T W = 0$,

hence the projection would be orthogonal. (We can make this choice, since for \mathbb{R}^n there exists an orthogonal basis such that for some $k \in \mathbb{N}$ some k basis vectors generate the subspace $R(V)$. Now the rest of the basis vectors go to the matrix W and they are orthogonal to $R(V)$, as they are orthogonal to the columns of V .)

So now the task is to find a P such that $P\mathbf{x} = V\mathbf{a}$. Multiplying equation (1) by V^T on the left, we get

$$V^T\mathbf{x} = V^TV\mathbf{a} + V^TW\beta = V^TV\mathbf{a}.$$

The columns of V are chosen to be linearly independent, so that $N(V) = \{\mathbf{0}\}$, from which follows by some previous homework that we also have $N(V^TV) = \{\mathbf{0}\}$, meaning that V^TV is invertible. We then get that $\mathbf{a} = (V^TV)^{-1}V^T\mathbf{x}$, hence $P\mathbf{a} = V(V^TV)^{-1}V^T\mathbf{x}$. And therefore

$$P = V(V^TV)^{-1}V^T. \quad (2)$$

(a) So now when we pick

$$V = \begin{bmatrix} 1 & 1 \\ 1 & 1 \\ 0 & 1 \end{bmatrix}$$

as given in the assignment, and pluggin it in (2) we get

$$P = \begin{bmatrix} 1/2 & 1/2 & 0 \\ 1/2 & 1/2 & 0 \\ 0 & 0 & 1 \end{bmatrix}.$$

And to conclude we also have $R(P) = \mathcal{V}$, because clearly every point of \mathcal{V} we can express as a linear combination of the columns of P , and conversely P is a projection to the subspace \mathcal{V} .

(b) This time let us choose as basis $\{(1, 1, 0), (0, 0, 1)\}$ for \mathcal{V} . We get, again by using (2),

$$P = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} & 0 \\ \frac{1}{2} & \frac{1}{2} & 0 \\ 0 & 0 & 1 \end{bmatrix}.$$

This was expected, as the projection depends on the space on which we project, not on its basis.

3. Let

$$\mathcal{V} = \text{span} \left\{ \begin{bmatrix} 1 \\ 2 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 2 \\ 1 \end{bmatrix} \right\} \quad \text{and} \quad \mathcal{W} = \text{span} \left\{ \begin{bmatrix} 2 \\ 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} \right\}.$$

Denote

$$V = \begin{bmatrix} 1 & 1 \\ 2 & 1 \\ 1 & 2 \\ 0 & 1 \end{bmatrix} \quad \text{and} \quad W = \begin{bmatrix} 2 & 1 \\ 1 & 1 \\ 1 & 1 \\ 0 & 1 \end{bmatrix}.$$

You may assume known that $\mathbb{R}^4 = \mathcal{V} \oplus \mathcal{W}$.

- (a) Using the matrix V give a formula for the orthogonal projection matrix $Q \in \mathbb{R}^{4 \times 4}$ such that $R(Q) = \mathcal{V}$.
- (b) Using the matrices V and W give a formula for the projection matrix $P \in \mathbb{R}^{4 \times 4}$ such that $R(P) = \mathcal{V}$ and $N(P) = \mathcal{W}$.
- (c) Compute the projection matrices in (a) and (b) using Matlab. Check that $Q^2 = Q$, $P^2 = P$, $Q^T(I - Q) = 0$, $PV = V$ and $PW = 0$ (within floating point accuracy).

Hints: (b) If P is constructed so that it projects the direct sum $\mathbb{R}^4 = \mathcal{V} \oplus \mathcal{W}$ to the first component, what is the null space of P ?

- (a) We can use equation (2), in which now we have $P = Q$ and

$$V = \begin{bmatrix} 1 & 1 \\ 2 & 1 \\ 1 & 2 \\ 0 & 1 \end{bmatrix}.$$

- (b) Let us divide an arbitrary vector \mathbf{x} in components so that the columns of V and W ($V \in \mathbb{R}^{n \times m}$, $W \in \mathbb{R}^{n \times k}$) together form a basis for \mathbb{R}^n (from which follows that $[V \ W]$ is invertible):

$$\mathbf{x} = V\mathbf{a} + W\mathbf{b}.$$

Next we find the matrix P for the projection $\mathbf{x} \mapsto V\mathbf{a}$: from

$$\mathbf{x} = V\mathbf{a} + W\mathbf{b} = \begin{bmatrix} V & W \end{bmatrix} \begin{bmatrix} \mathbf{a} \\ \mathbf{b} \end{bmatrix},$$

it follows that

$$\begin{bmatrix} \mathbf{a} \\ \mathbf{b} \end{bmatrix} = \begin{bmatrix} V & W \end{bmatrix}^{-1} \mathbf{x}$$

which we can use in order to compute

$$P\mathbf{x} = V\mathbf{a} = V \begin{bmatrix} \mathbf{I}_m & \mathbf{0}_{m \times k} \end{bmatrix} \begin{bmatrix} \mathbf{a} \\ \mathbf{b} \end{bmatrix} = V \begin{bmatrix} \mathbf{I}_m & \mathbf{0}_{m \times k} \end{bmatrix} \begin{bmatrix} V & W \end{bmatrix}^{-1} \mathbf{x}.$$

And this means that

$$P = V \begin{bmatrix} \mathbf{I}_m & \mathbf{0}_{m \times k} \end{bmatrix} \begin{bmatrix} V & W \end{bmatrix}^{-1} = \begin{bmatrix} V & \mathbf{0}_{m \times k} \end{bmatrix} \begin{bmatrix} V & W \end{bmatrix}^{-1}. \quad (3)$$

Let us now choose W so that $N(P) = \mathcal{W}$. First of all, we know that $N(P) = R(I - P)$, and here is a proof: (\subseteq) If $\mathbf{x} \in N(P)$, then $P\mathbf{x} = 0$, so that $\mathbf{x} = (I - P)\mathbf{x}$, and so $\mathbf{x} \in R(I - P)$. (\supseteq) If $\mathbf{x} \in R(I - P)$, then $\mathbf{x} = (I - P)\mathbf{y}$ for some $\mathbf{y} \in \mathbb{R}^n$, and since $(I - P)$ is a projection matrix (see the first part of problem 1), we get

$$\mathbf{x} = (I - P)\mathbf{y} = (I - P)(I - P)\mathbf{y} = (I - P)\mathbf{x},$$

hence $P\mathbf{x} = 0$, so that $\mathbf{x} \in N(P)$. So that's why $N(P) = R(I - P)$. Now, since $W\mathbf{b} = (I - P)\mathbf{x}$, as $R(W) = R(I - P)$, we have

$$\mathcal{W} = N(P) = R(I - P) = R(W),$$

that is, we can choose

$$W = \begin{bmatrix} 2 & 1 \\ 1 & 1 \\ 1 & 1 \\ 0 & 1 \end{bmatrix}.$$

Since we want to project to the subspace \mathcal{V} , we choose as in part (a)

$$V = \begin{bmatrix} 1 & 1 \\ 2 & 1 \\ 1 & 2 \\ 0 & 1 \end{bmatrix}.$$

Now we then get the desired projection P in (3).

(c) Matlab code:

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V = [1 1; 2 1; 1 2; 0 1]
W = [2 1; 1 1; 1 1; 0 1]
Q = V*inv(V'*V)*V'
P = V*[eye(size(V)(2)) zeros(size(V)(2))]*inv([V W])
Q^2                                % =Q? i.e. is it a proj. matrix?
P^2                                % =P?
Q'*(eye(size(Q))-Q)                % =0? orthogonality
```

We get the expected results. As projection matrices we get

$$Q = \begin{bmatrix} 0.176471 & 0.294118 & 0.235294 & 0.058824 \\ 0.294118 & 0.823529 & 0.058824 & -0.235294 \\ 0.235294 & 0.058824 & 0.647059 & 0.411765 \\ 0.058824 & -0.235294 & 0.411765 & 0.352941 \end{bmatrix} = \frac{1}{17} \begin{bmatrix} 3 & 5 & 4 & 1 \\ 5 & 14 & 1 & -4 \\ 4 & 1 & 11 & 7 \\ 1 & -4 & 7 & 6 \end{bmatrix},$$

$$P = \begin{bmatrix} -6.6667e-01 & 3.3333e-01 & 1.0000e+00 & -6.6667e-01 \\ -1.0000e+00 & 1.0000e+00 & 1.0000e+00 & -1.0000e+00 \\ -1.0000e+00 & 1.1102e-16 & 2.0000e+00 & -1.0000e+00 \\ -3.3333e-01 & -3.3333e-01 & 1.0000e+00 & -3.3333e-01 \end{bmatrix}$$

$$= \frac{1}{3} \begin{bmatrix} -2 & 1 & 3 & -2 \\ -3 & 3 & 3 & -3 \\ -3 & 0 & 6 & -3 \\ -1 & -1 & 3 & -1 \end{bmatrix}.$$

4. Let $\mathbf{a}, \mathbf{b} \in \mathbb{R}^3 \setminus \{0\}$ be such that $\mathbf{a}^T \mathbf{b} = 0$ and define the matrix $P \in \mathbb{R}^{3 \times 3}$ as

$$P = \frac{\mathbf{a}\mathbf{a}^T}{\|\mathbf{a}\|_2^2} + \frac{\mathbf{b}\mathbf{b}^T}{\|\mathbf{b}\|_2^2}.$$

- (a) Is P a projection matrix? How about $I - P$, where I is the identity matrix?
 (b) Is P an orthogonal projection?
 (c) Let $\mathbf{a} = [1, 1, 1]^T$ and $\mathbf{b} = [1, -2, 1]^T$. Find $\mathbf{c} \in \mathbb{R}^3$ such that

$$I - P = \frac{\mathbf{c}\mathbf{c}^T}{\|\mathbf{c}\|_2^2}.$$

Hints: (a) See Definition 2.1 on page 45 of the lecture notes. (b) Use problem 1(b). (c) The vector \mathbf{c} has to be orthogonal to \mathbf{a} and \mathbf{b} , since $R(I - P) = R(P)^\perp$.

Solution:

- (a) By definition a matrix A is a projection matrix if $A^2 = A$ (that is, once projected a point doesn't move anymore). For P we get

$$\begin{aligned} P^2 &= \frac{(\mathbf{a}\mathbf{a}^T)(\mathbf{a}\mathbf{a}^T)}{\|\mathbf{a}\|_2^4} + \frac{(\mathbf{b}\mathbf{b}^T)(\mathbf{b}\mathbf{b}^T)}{\|\mathbf{b}\|_2^4} + \frac{(\mathbf{a}\mathbf{a}^T)(\mathbf{b}\mathbf{b}^T)}{\|\mathbf{a}\|_2^2\|\mathbf{b}\|_2^2} + \frac{(\mathbf{b}\mathbf{b}^T)(\mathbf{a}\mathbf{a}^T)}{\|\mathbf{b}\|_2^2\|\mathbf{a}\|_2^2} \\ &= \frac{\mathbf{a}(\mathbf{a}^T\mathbf{a})\mathbf{a}^T}{\|\mathbf{a}\|_2^4} + \frac{\mathbf{b}(\mathbf{b}^T\mathbf{b})\mathbf{b}^T}{\|\mathbf{b}\|_2^4} + \frac{\mathbf{a}(\mathbf{a}^T\mathbf{b})\mathbf{b}^T}{\|\mathbf{a}\|_2^2\|\mathbf{b}\|_2^2} + \frac{\mathbf{b}(\mathbf{b}^T\mathbf{a})\mathbf{a}^T}{\|\mathbf{b}\|_2^2\|\mathbf{a}\|_2^2} \\ &= \frac{\mathbf{a}\mathbf{a}^T}{\|\mathbf{a}\|_2^2} + \frac{\mathbf{b}\mathbf{b}^T}{\|\mathbf{b}\|_2^2} = P, \end{aligned}$$

where in the third step we use that $\mathbf{a}^T\mathbf{a} = \|\mathbf{a}\|_2^2$ and $\mathbf{a}^T\mathbf{b} = \mathbf{b}^T\mathbf{a} = 0$. So P is a projection matrix.

For an arbitrary projection matrix, it holds that $I - P$ is also a projection matrix, because

$$(I - P)^2 = I^2 - 2P + P^2 = I - 2P + P = I - P.$$

- (b) The projection P splits any vector \mathbf{x} in two components:

$$\mathbf{x} = P\mathbf{x} + (I - P)\mathbf{x}.$$

The projection P is orthogonal if these components are orthogonal to each other. So P is orthogonal if $(P\mathbf{x})^T((I - P)\mathbf{x}) = 0$. We get

$$\begin{aligned} (P\mathbf{x})^T((I - P)\mathbf{x}) &= (P\mathbf{x})^T\mathbf{x} - (P\mathbf{x})^TP\mathbf{x} \\ &= \mathbf{x}^TP^T\mathbf{x} - \mathbf{x}^TP^TP\mathbf{x} \\ &= \mathbf{x}^TP\mathbf{x} - \mathbf{x}^TP\mathbf{x} \\ &= 0, \end{aligned}$$

where we use that P is symmetric and $P^2 = P$. Then P is an orthogonal projection. The same we could show by checking that $P^T(I - P) = 0$, because then the subspaces $R(P)$ and $R(I - P)$ are orthogonal, from which follows that the associated components are orthogonal to each other.

- (c) The projection matrix of the form $A = \frac{\mathbf{a}\mathbf{a}^T}{\|\mathbf{a}\|^2}$ is an orthogonal projection to the subspace $\text{span}\{\mathbf{a}\}$ (see the lecture notes). Then P is a map to the subspace $\text{span}\{\mathbf{a}, \mathbf{b}\}$, also an orthogonal projection, which we proved in the previous part.

For an orthogonal projection P the components $P\mathbf{x}$ and $(I - P)\mathbf{x}$ are always orthogonal, and $(I - P)\mathbf{x}$ is now the projection to the subspace $\text{span}\{\mathbf{c}\}$, the vector \mathbf{c} must be orthogonal to the vectors \mathbf{a} and \mathbf{b} . So then we get \mathbf{c} by cross product:

$$\mathbf{c} = \mathbf{a} \times \mathbf{b} = \begin{bmatrix} 3 \\ 0 \\ -3 \end{bmatrix}.$$

We can still rescale the vector \mathbf{c} , for instance we can choose $\mathbf{c} = [1 \ 0 \ -1]^T$.