

## Recitation 9

Thursday October 6, 2022

### 1 Recap

#### 1.1 Determinant

##### 1.1.1 Algebraic View

The *determinant* of a square matrix  $A \in \mathbb{R}^{n \times n}$  is defined as

$$\det A = \sum_{\sigma} \left( (-1)^{\text{sign}(\sigma)} \cdot \prod_{i=1}^n A_{i,\sigma(i)} \right) \quad (1)$$

where  $\sigma$  iterates over any permutation of  $\{1, \dots, n\}$  and  $\text{sign}(\sigma)$  is the parity of  $\sigma$ . An equivalent definition (more computationally convenient) for the determinant is

$$\det A = (\pm 1) \cdot (\text{product of pivots in } \text{ref}(A)),$$

where the sign depends on the parity of the number of row exchanges in the REF.

##### 1.1.2 Geometric View

The *determinant* of a square matrix  $A \in \mathbb{R}^{n \times n}$  is the scaling factor between  $\text{vol}(S)$  and  $\text{vol}(\phi_A(S))$  taking into account handedness; where  $\phi_A(S) = \{Ax : x \in S\}$  is the region which  $\phi_A$  maps  $S$  into.

##### 1.1.3 Properties

Let  $A, B$  be  $n \times n$  matrices. Let  $C$  be another square matrix and  $D$  be a matrix with proper dimension.

1. Swapping two rows (or two columns) of  $A$  *negates* the determinant.
2. Adding a row to another row *does not* change the determinant.
3. Adding a column to another column *does not* change the determinant.
4. Multiplying a row/column by a scalar  $c$  *changes* the determinant by a factor of  $c$ .
5.  $\det A^T = \det A$
6.  $\det AB = \det A \cdot \det B$
7.  $\det(A + B) \neq \det A + \det B$
8.  $\det \begin{pmatrix} A & D \\ 0 & C \end{pmatrix} = \det A \cdot \det C.$

## 1.2 Square Matrices Revisited

Let  $A$  be an  $n \times n$  square matrix. Then, the following statements are equivalent.

1.  $A$  is invertible, i.e.  $A^{-1}$  exists
2.  $A$  has both a left inverse and a right inverse
3. The columns of  $A$  are linearly independent
4. The rows of  $A$  are linearly independent
5.  $Ax = b$  is uniquely solvable for every  $b \in \mathbb{R}^n$
6.  $N(A) = \{\mathbf{0}\}$
7.  $C(A) = \mathbb{R}^n$
8.  $\text{Rank}(A) = n$
9.  $\det A \neq 0$

## 1.3 Projection

Suppose that we want to find the orthogonal projection of a given vector  $w$  onto a  $k$ -dimensional subspace  $\mathcal{V}$ . In other words, we want to find a vector  $v = \text{proj}_{\mathcal{V}} w$  for which 1)  $v \in \mathcal{V}$ , and 2)  $(w - v) \perp u$  for all  $u \in \mathcal{V}$ .

Suppose that  $\{v_1, \dots, v_k\}$  is an orthonormal basis of  $\mathcal{V}$ . Then,

$$\begin{aligned} v = \text{proj}_{\mathcal{V}} w &= \sum_{i=1}^k (w \cdot v_i) v_i = \sum_{i=1}^k v_i \cdot (v_i \cdot w) = \sum_{i=1}^k v_i \cdot (v_i^\top w) \\ &= \sum_{i=1}^k (v_i v_i^\top) w = \left( \sum_{i=1}^k v_i v_i^\top \right) w = Pw \end{aligned}$$

when  $P = \sum_{i=1}^k v_i v_i^\top$  is an  $n \times n$  matrix. The matrix  $P$  is the *orthogonal projection matrix* onto the subspace  $\mathcal{V}$ , and can also be written as

$$P = VV^\top, \quad \text{where } V = [v_1 \ \dots \ v_k] \in \mathbb{R}^{n \times k}.$$

Note that  $P$  is symmetric (i.e.  $P^\top = P$ ) and  $\text{rank}(P) = \text{rank}(V) = k$ .

Another interesting property of  $V$  is that  $V^\top V = I_k$ . Indeed, this is a necessary and sufficient condition for  $\{v_1, \dots, v_k\}$  to be orthonormal.

## 2 Exercises

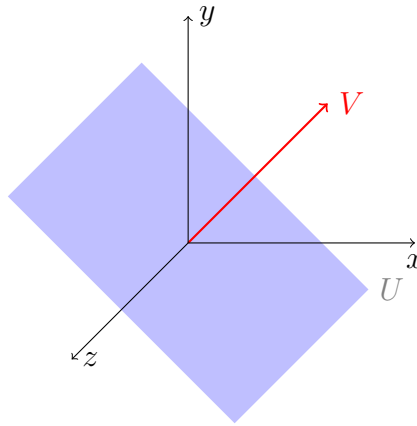
1. Use determinant properties to show that if  $A$  and  $B$  are square matrices such that  $AB$  is invertible, then both  $A$  and  $B$  are invertible.

2. Let  $A = \begin{bmatrix} 1 & -2 & 1 \\ 3 & \alpha & -5 \\ -1 & -1 & 2 \end{bmatrix}$ . What values of  $\alpha$  makes  $A$  *not* invertible?

3. Consider the two subspaces  $U$  and  $W$  as follows.

$$U = \left\{ \begin{bmatrix} u_1 \\ u_2 \\ u_3 \end{bmatrix} : u_1 - u_2 = 0, u_1 - u_3 = 0 \right\} = \text{Span} \left\{ \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \right\}$$

$$W = \left\{ \begin{bmatrix} w_1 \\ w_2 \\ w_3 \end{bmatrix} : w_1 + w_2 + w_3 = 0 \right\} = \text{Span} \left\{ \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix} \right\}$$



The two subspaces are orthogonal complements (check!). It can be easily seen that the given set of generators bases are actually bases (why?).

- Find an orthonormal basis of  $U$ . Use it to derive the projection matrix  $P_U$  which projects vectors onto the subspace  $U$ .
  - Find an orthonormal basis of  $W$ . Use it to derive the projection matrix  $P_W$  which projects vectors onto the subspace  $W$ .
  - Notice that  $P_U + P_W = I_3$ . It turns out that this is not a coincidence. For any orthogonal complement subspaces  $U, W \subseteq \mathbb{R}^n$ , the sum of their corresponding projection matrices is exactly  $I_n$ . Can you explain why it is always the case?  
*Hint: Use orthogonal decomposition. What exactly is each component of the decomposition?*
4. Recall in 2D plane, a mirror matrix  $M$  is the matrix for which given a vector  $x$ , then  $Mx$  is the mirror image of  $x$  across a line  $L$  that passes through the origin.
- What can we tell about  $M^2$  using the following fact: if we reflect a vector  $v$  across the line  $L$  *twice*, we end up with the original vector  $v$ .
  - Using the answer from part (a), find the possible values of  $\det M$ .
  - Take any 2-dimensional region  $\mathcal{S}$  you like. What is the region that is produced from left-multiplying any  $x \in \mathcal{S}$  by  $M$ ? In other words, what is the region  $\phi_M(\mathcal{S}) = \{Mx : x \in \mathcal{S}\}$ ?
  - What is the volume of  $\phi_M(\mathcal{S})$ ? How is it compared to the volume of  $\mathcal{S}$ ? What can we tell about  $\det M$ ?
  - Recall from problem set 2 that if  $L$  makes angle  $\theta$  with the  $x$ -axis, then we have  $M = \begin{bmatrix} \cos 2\theta & \sin 2\theta \\ \sin 2\theta & -\cos 2\theta \end{bmatrix}$ . Is the expression consistent with part a and d?