# NUMERICAL OPTIMIZATION

Sheet 1: Linear Algebra refresh

In this sheet, each task will contain a theory refresher and an actual exercise.

### EXERCISE ONE Give the definition of:

- 1. Linear independence of a set of vectors.
- 2. Vector subspace (sometimes called linear subspace) of  $\mathbb{R}^d$ .
- 3. Rank of a set of vectors or of a matrix.

Solve the following system of equalities using Gaussian elimination.

$$3x + 2y + 4z + w = 0$$
$$2x + 4y + 2z + 2w = 5$$
$$-x + 2y + 2z + w = 7$$
$$4z = 2$$

What is the geometric shape of the space of all solutions? Also, what is the necessary and sufficient condition on rank (rank of what?) for the system to have a unique solution?

## EXERCISE TWO

- 1. Give the definition of a determinant, what an invertible matrix is. What is the connection between a determinant of a matrix and its invertibility?
- 2. Explain how to compute determinants of  $2 \times 2$  and  $3 \times 3$  matrices quickly (the Sarrus rule).
- 3. Explain the full formula for a determinant that uses permutations of columns and the sgn function.

#### EXERCISE THREE

If you have not heard them in Linear Algebra, familiarize yourselves with the following definitions:

**D:** A set  $A \subseteq \mathbb{R}^d$  is an affine subspace, if A is of the form L+v for some linear subspace L and a shift vector  $v \in \mathbb{R}^d$ . By "A is of the form L+v" we mean a bijection between vectors of L and vectors of A given as b(u) = u + v. Each affine subspace has a dimension, defined as the dimension of its associated linear subspace L.

**D:** A vector x is an affine combination of a finite set of vectors  $a_1, a_2, \ldots, a_n$  if  $x = \sum_{i=1}^n \alpha_i a_i$ , where  $\alpha_i$  are real numbers satisfying  $\sum_{i=1}^n \alpha_i = 1$ .

A set of vectors  $V \subseteq \mathbb{R}^d$  is affinely independent if it holds that no vector  $v \in V$  is an affine combination of the rest.

**D:** Given a set of vectors  $V \subseteq \mathbb{R}^d$ , we can think of its *affine span*, which is a set of vectors A that are all possible affine combinations of any finite subset of V.

Similar to the linear spaces, affine spaces have a finite basis, so we do not need to consider all finite subsets of V, but we can generate the affine span as affine combinations of the basis.

**D:** A hyperplane is any affine space in  $\mathbb{R}^d$  of dimension d-1. Thus, on a 2D plane, any line is a hyperplane. In the 3D space, any plane is a hyperplane, and so on.

A hyperplane splits the space  $\mathbb{R}^d$  into two halfspaces. We count the hyperplane itself as a part of both halfspaces.

First, solve the following two proving exercises:

1. Prove that every hyperplane can be expressed as the set of solutions of

$$\{x \in \mathbb{R}^d \,|\, c^T x = b\},\,$$

where  $c \in \mathbb{R}^d$  and  $b \in \mathbb{R}$ .

2. Prove that each *proper* affine subspace (proper means any subspace which is not the full  $\mathbb{R}^d$ ) can be expressed as an intersection of finitely many affine hyperplanes.

Next, two exercises on your understanding of higher dimensions, which can be solved using the tools above.

- 1. Can two 2D planes (affine subspaces of dimension 2) intersect in exactly one point, if we are working in  $\mathbb{R}^4$ ?
- 2. Can two affine subspaces of dimension 3 within  $\mathbb{R}^5$  intersect in exactly one point?

#### Exercise four

- 1. Give the definition of a *positive definite* matrix. Include at least two more equivalent descriptions of this property (i.e., a matrix is positive definite if and only if ...).
- 2. One tool to detect if a matrix is positive definite is using the Cholesky decomposition. Give a definition of this decomposition and explain the Cholesky algorithm (or one of its variants).
- 3. Finally, use the Cholesky algorithm (by hand) to show that this matrix is positive definite:

$$A = \begin{pmatrix} 2 & 2 & 0 \\ 2 & 5 & 3 \\ 0 & 3 & 8 \end{pmatrix}$$

## Exercise five

- 1. Give the definition of eigenvalues and eigenvectors.
- 2. Can you comment on the algorithmic complexity of computing the set of all eigenvalues?
- 3. One way to compute eigenvalues of a small matrix A uses a formula that ties eigenvalues to roots of a specific polynomial  $p_A(t)$  that arises from the determinant. Define this polynomial and use this technique to find the eigenvalues of the following matrix A:

$$A = \begin{pmatrix} 1 & 2 & 0 \\ 3 & -1 & 3 \\ 1 & -2 & 2 \end{pmatrix}$$

EXERCISE SIX One final algorithm that we should learn about is Gram-Schmidt orthogonalization. This is a process to find an orthogonal basis from any given vector set (usually the set is independent, so we are converting from a set of linearly independent vectors to a set of orthogonal vectors, preserving the spanned subspace).

Explain the algorithm and use it (by hand) to convert the following set of vectors to their orthogonal representation:

$$V = \{v_1, v_2, v_3\}, \text{ where } v_1 = (1, 0, 1, 1), v_2 = (2, 1, -1, 0), \text{ and } v_3 = (1, 1, 1, 1).$$