

#### **OPTIMIZATION**

#### MASTER IN FUNDAMENTAL PRINCIPLES OF DATA SCIENCE

# OPTIMIZATION PROBLEM 2

CONCRETE MIXING PROBLEM



#### Author

Vladislav Nikolov Vasilev

FACULTY OF MATHEMATICS AND COMPUTER SCIENCE

ACADEMIC YEAR 2021-2022

## 1 Problem description

Suppose that we are mixing concrete and are using n different gravel sizes  $s_1, \ldots, s_n$ .

The ideal mixture is given by  $c = (c_1, \ldots, c_n)$ , where  $c_i$   $(0 \le c_i \le 1)$  is the fraction of size  $s_i$  in the mix, and  $\sum_{i=1}^n c_i = 1$ .

Gravel mixtures come from m different mines. The gravel composition at each mine  $j=1,\ldots,m$  is given by  $C_j=(c_{1j},\ldots,c_{nj})$ , where  $0 \le c_{ij} \le 1$  for all  $i=1,\ldots,n$  and  $\sum_{i=1}^n c_{ij}=1$ .

Let  $x = (x_1, ..., x_m)$  be the vector that represents the fraction of gravel of the mines in the mixture, where  $0 \le x_j \le 1$  for all j = 1, ..., m and  $\sum_{j=1}^m x_j = 1$ .

Find the best possible approximation  $x = (x_1, ..., x_m)$  of the ideal mixture,  $c = (c_1, ..., c_n)$ , by using the material from the m mines.

### 2 Solution

We have that  $\boldsymbol{x} \in \mathbb{R}^m$ ,  $\boldsymbol{c} \in \mathbb{R}^n$  and  $C \in \mathbb{R}^{n \times m}$ , where the matrix  $C = (C_1, \dots, C_m)$  has  $C_j$  as columns.

Finding the best possible vector  $\boldsymbol{x}$  means finding a vector such that

$$Cx = c \tag{1}$$

This means that the vector result of the matrix-vector product Cx has to be as similar as possible to c. With this in mind, we can rewrite expression (1) as the following optimization problem:

$$\boldsymbol{x}^{\star} = \min_{\boldsymbol{x} \in \mathbb{R}^m} \|C\boldsymbol{x} - \boldsymbol{c}\|^2 \tag{2a}$$

subjet to 
$$\sum_{j=1}^{m} x_j = 1$$
, and  $x_j \ge 0$  (2b)

where  $x^*$  is the best possible solution out of all the feasible ones. Note that minimizing  $\|Cx - c\|^2$  is the same as minimizing  $\|Cx - c\|$ .

Now, knowing that  $\|x\| = \sqrt{x^T x}$ , the distance found in expression (2a) can be rewritten as follows:

$$||C\mathbf{x} - \mathbf{c}||^{2} = (C\mathbf{x} - \mathbf{c})^{T}(C\mathbf{x} - \mathbf{c})$$

$$= \mathbf{x}^{T}C^{T}C\mathbf{x} - \mathbf{x}^{T}C^{T}\mathbf{c} - \mathbf{c}^{T}C\mathbf{x} + \mathbf{c}^{T}\mathbf{c}$$

$$= \mathbf{x}^{T}C^{T}C\mathbf{x} - (\mathbf{x}^{T}C^{T}\mathbf{c})^{T} - \mathbf{c}^{T}C\mathbf{x} + \mathbf{c}^{T}\mathbf{c}$$

$$= \mathbf{x}^{T}C^{T}C\mathbf{x} - \mathbf{c}^{T}C\mathbf{x} - \mathbf{c}^{T}C\mathbf{x} + \mathbf{c}^{T}\mathbf{c}$$

$$= \mathbf{x}^{T}C^{T}C\mathbf{x} - 2\mathbf{c}^{T}C\mathbf{x} + ||\mathbf{c}||^{2}$$
(3)

Thanks to the last expression in (3), we can clearly see that this is a **Quadratic Optimization** problem because the objective function is quadratic and all the restrictions are linear.

Notice that if  $C \in \mathbb{R}^{n \times m}$ , then  $C^T C \in \mathbb{R}^{n \times n}$  and  $\mathbf{c}^T C \in \mathbb{R}^m$ . Let  $z = C\mathbf{x}$ , then:

$$\mathbf{x}^T C^T C \mathbf{x} = \mathbf{z}^T \mathbf{z} = \|\mathbf{z}\|^2 \ge 0, \quad \forall \mathbf{x} \in \mathbb{R}^m$$
 (4)

From the inequality in expression (4) we know that the Euclidean norm is only going to be 0 if  $\mathbf{x} = 0$ . Taking also into account the problem restrictions that can be seen in (2b), we know that  $\sum_{j=1}^{m} x_j = 1$ , which means that  $\mathbf{x} \neq 0$ . This implies that  $C^TC$  is **positive definite** and that an optimal value can be found.

Thus, we can conclude that a non-zero value of  $\boldsymbol{x}$  can be found so that the distance  $\|C\boldsymbol{x} - \boldsymbol{c}\|^2$  is as close to 0 as possible, which means that the found value of  $\boldsymbol{x}$  is going to be the optimal solution to this minimization problem.