

# Methods of classification and dimensionality reduction - Report 1

March 19, 2022

## 1. DESCRIPTION OF METHODS

**SVD1.** We get a  $n \times d$  dimensional matrix  $Z$  that we want to approximate by a different matrix  $\tilde{Z}$ . We want somehow  $\tilde{Z}$  to maintain only "the most important" information from  $Z$ . That's what we can get using SVD decomposition of matrix and cutting out the smallest eigenvalues.

Precisely, we want to find matrix  $\tilde{Z}_r$  of rank  $r$  ( $r < \text{rank}(Z)$ ), so that  $\|Z - \tilde{Z}_r\|$  is small. Using SVD decomposition  $Z = U\Lambda^{\frac{1}{2}}V^T$  we construct  $\tilde{Z}$  as

$$\tilde{Z}_r = U_r\Lambda_r^{\frac{1}{2}}V_r^T,$$

where  $\Lambda_r$  contains  $r$  biggest eigenvalues of  $Z$  and  $U_r, V_r$  contains only columns corresponding to those eigenvalues.

**NMF.** We get a real  $n \times d$  dimensional matrix  $Z$ , such that  $n \geq d$ . The aim is to approximate  $Z$  as

$$Z \approx WH,$$

where

- $W$  is  $n \times r$  matrix of non-negative elements,
- $H$  is  $r \times d$  matrix of non-negative elements.

Precisely, we look for

$$\arg \min_{W,H} \|Z - WH\|^2,$$

where  $\|A\|^2 = \sum_{i,j} A_{ij}^2$ .

**SVD2.**

## 2. OUR DATA

**Description.**

**Performing methods.**

**Choosing parameters.**

## 3. RESULTS