

Principle Component Analysis

Implementing dimensionality reduction

PCA

- Remember that for the matrix X we can perform the Singular Value Decomposition as:

$$\mathbf{X} = \mathbf{USV}^T$$

Where \mathbf{US} gives us the coordinates of a sample in \mathbf{X} in the space of principle components:

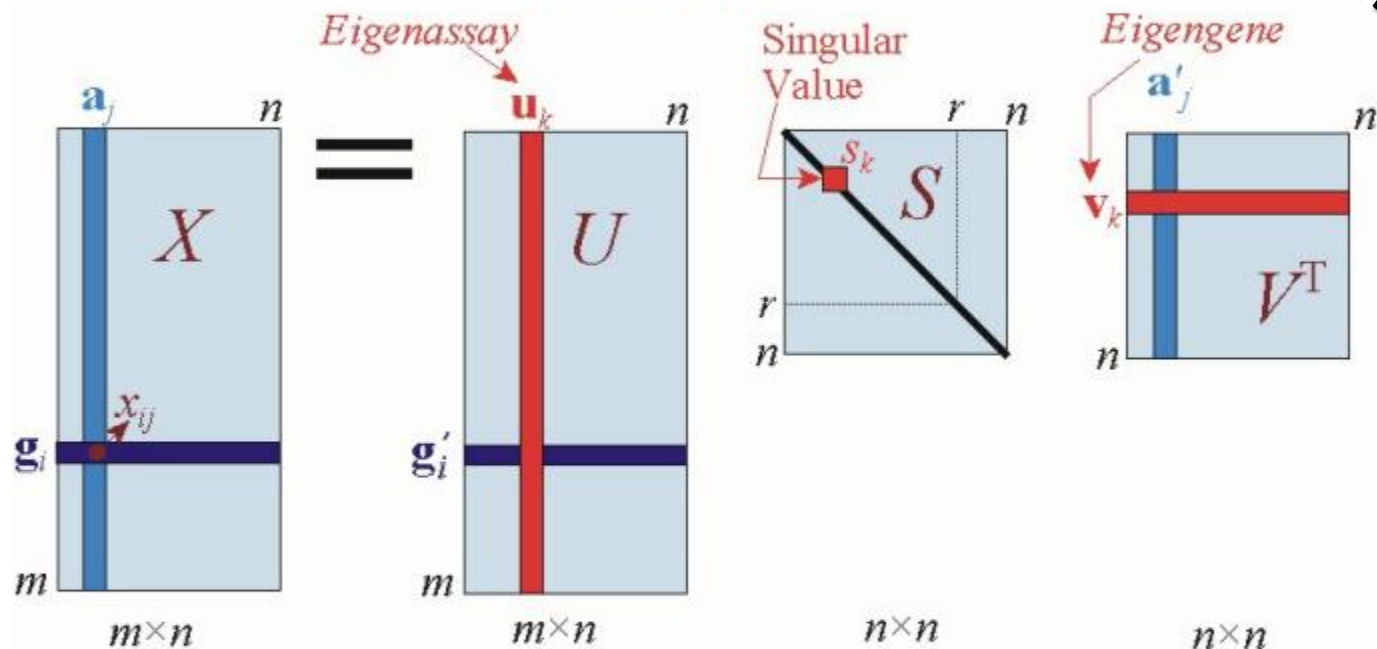
$$\mathbf{x}/\mathbf{v}^T = \mathbf{US}$$

SVD and PCA

- Singular Value Decomposition

- For any matrix X :

$$X = USV^T$$



Data X , one row per data point
Data is zero-centred

US gives coordinates of rows of X in the space of principle components

S is diagonal,
 $S_k > S_{k+1}$,
 S_k is k^{th} largest eigenvalue

Rows of V^T are unit length eigenvectors

Projecting the data into PC space

- Given that US is our original observations projected into Principal Component Space, we will need to solve $X = USV^T$ for US .
- To remove the V^T on the right, we can utilise the fact that a square orthogonal matrix multiplied by its inverse (the transpose) is the Identity Matrix and can be removed.

Projecting our data into PC space

$$X = USV^T$$

$$XV = USV^T V$$

$$XV = USI$$

$$XV = US$$

PCA for dimensionality reduction

- Given some observed data X :
 - Get the mean of the observed data: μ .
 - Perform a Singular Value Decomposition on the mean centred X : obtaining U, S and V .
 - Mean-centre the points in X .
 - Project points onto the new principle component space.

PCA in practice

- Given an N-by-F matrix of **training** data X_{trn} :
 - $\mu = \text{mean}(X_{trn})$;
 - $X_{trn_centred} = X_{trn} - \mu$;
 - $[U, S, V'] = \text{svd}(X_{trn_centred})$;
 - $X_{trn_projected} = X_{trn_centred} * V$;
- Given an M-by-F matrix of **testing** data X_{tst} :
 - $X_{tst_centred} = X_{tst} - \mu$;
 - $X_{tst_projected} = X_{tst_centred} * V$;

How to reduce dimensions

- Select $f < F$ principle component features.
 - $f = 10$;
 - $X_{trn_projected} = X_{trn_projected}(:, 1:f)$;
 - $X_{tst_projected} = X_{tst_projected}(:, 1:f)$;
- Or make the selection during the projection:
 - $X_{trn_projected} = X_{trn_centred} * V(:, 1:f)$;
 - $X_{tst_projected} = X_{tst_centred} * V(:, 1:f)$;