



Analytical Geometry and Linear Algebra II, Lab 12

Similar matrices

Singular value decomposition - SVD

Left and right inverses. Pseudoinverse

Similar Matrices



The primary goal of SVD



To “X-RAY” matrix
(To understand the structure of matrix)

Singular Value Decomposition (SVD)

3 ways of explanation

1. Linear transformation – [Kutz video](#)
2. Algebraic – [MIT video \(Strang\)](#), [Aaron Greiner video](#)
3. As a tool for DS – [Stanford video](#), [Brunton video](#)

According to Kholodov words, SVD was created for *finding an inverse for any matrices*. It is needed in linear transformation related operations. Other properties were found afterwards.



Singular Value Decomposition (SVD)

How to calculate it (2 possible ways)



First approach

1. Find eigenpairs for $A^T A$. Result is Σ and V . ($A^T A = V \Sigma^2 V^T$)
2. Find U , using Σ and V ($AV \Sigma^+ = U$)

Second approach

1. Find eigenpairs for $A^T A$ and AA^T . ($A^T A = V \Sigma^2 V^T$) ($AA^T = U \Sigma^2 U^T$)
2. Put signs correctly

Singular Value Decomposition (SVD)



Obtain SVD for $A = \begin{bmatrix} 3 & 2 & 2 \\ 2 & 3 & -2 \end{bmatrix}$, using second approach (*task was taken*)

1. Eigenpairs of AA^T .

$$AA^T = \begin{bmatrix} 17 & 8 \\ 8 & 17 \end{bmatrix}. \lambda_1 = 25, \lambda_2 = 9. x_{\lambda_1} = \begin{bmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{bmatrix}, x_{\lambda_2} = \begin{bmatrix} \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \end{bmatrix}. U = \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{bmatrix}$$

2. Eigenpairs of $A^T A$.

$$A^T A = \begin{bmatrix} 13 & 12 & 2 \\ 12 & 13 & -2 \\ 2 & -2 & 8 \end{bmatrix}. \lambda_1 = 25, \lambda_2 = 9, \lambda_3 = 0. V = \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{18}} & \frac{2}{3} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{18}} & -\frac{2}{3} \\ 0 & \frac{4}{\sqrt{18}} & -\frac{1}{3} \end{bmatrix}$$

3. Result. $A\Sigma V^T = \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{bmatrix} \begin{bmatrix} 5 & 0 & 0 \\ 0 & 3 & 2 \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 \\ \frac{1}{\sqrt{18}} & -\frac{1}{\sqrt{18}} & \frac{4}{\sqrt{18}} \\ \frac{2}{3} & -\frac{2}{3} & -\frac{1}{3} \end{bmatrix}$

It had to be U

Video

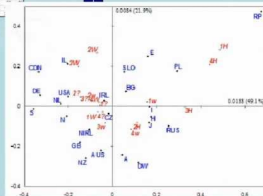


Visualizing high-dimensional data

This allows data matrices of high-dimensionality to be approximated optimally by one of rank 2:



so that the data can be visualized in a two-dimensional space for ease of interpretation



Singular Value Decomposition (SVD)

Properties

TODO



Singular Value Decomposition (SVD)

Where it can be used



- Image compression (slides)
- Pseudo-inverse (slides)
- Dimensionality reduction ([code + comments in pdf](#))
- Least square ([code + comments in pdf](#))
- Principal Component Analysis (PCA) ([video](#))
- Eigenfaces algorithms ([video](#))

SVD Applications

Image compression

Task: We want to compress our image for reducing the size/

Solution: We can represent our picture as a matrix.

Next step is using SVD for reducing matrix rank.



Pseudo-Inverse





Reference material

- Lecture 28: Similar Matrices and Jordan Form.
- Lecture 29: Singular Value Decomposition
- Lecture 33: Left and Right Inverses; Pseudoinverse
- 6. Singular Value Decomposition (SVD)
- *"Introduction to Linear Algebra", pdf pages 375–411*
7 Singular Value Decomposition (SVD)
- *"Linear Algebra and Applications", pdf pages 335–345*
5.6 Similarity Transformations
- *"Linear Algebra and Applications", pdf pages 377–386*
6.3 Singular Value Decomposition

Deserve "A" grade!

– Oleg Bulichev

✉ o.bulichev@innopolis.ru

📍 @Lupasic

🏢 Room 105 (Underground robotics lab)