

Singular Value Decomposition

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Singular Value Decomposition

The Singular-Value Decomposition (SVD) is a matrix decomposition method for reducing a matrix to its constituent parts in order to make certain subsequent matrix calculations simpler.

Any real **$m \times n$** matrix A can be decomposed uniquely:

$$A = UDV^T$$

U is **$m \times n$** and column orthogonal ($U^T U = I$)

$$D = \text{diag}(\sigma_1, \sigma_2, \dots, \sigma_n)$$

D is **$n \times n$** and diagonal

σ_i are called *singular* values of A

It is assumed that $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_n \geq 0$

V is **$n \times n$** and orthogonal ($VV^T = V^T V = I$)

SVD (cont'd)

If **$m=n$** , then:

$$A = UDV^T$$

U is **$n \times n$** and orthogonal ($U^T U = U U^T = I$)

D is **$n \times n$** and diagonal

$$D = \text{diag}(\sigma_1, \sigma_2, \dots, \sigma_n)$$

V is **$n \times n$** and orthogonal ($V V^T = V^T V = I$)

SVD (cont'd)

- The columns of U are eigenvectors of AA^T

$$AA^T = UDV^T VDU^T = \tilde{U}D^2U^T$$

- The columns of V are eigenvectors of $A^T A$

$$A^T A = VDU^T UDV^T = \tilde{V}D^2V^T$$

If λ_i is an eigenvalue of $A^T A$ (or AA^T), then $\lambda_i = \sigma_i^2$

Numerical Problem on SVD

$$A = \begin{bmatrix} 4 & 0 \\ 3 & -5 \end{bmatrix}$$

STEP 1: Find $A \cdot A'$ in order to find U

$$A \cdot A' = \begin{bmatrix} 16 & 12 \\ 12 & 34 \end{bmatrix}$$

STEP 2: Find Eigen Vector for A.A'

Find Eigen vector for $A \cdot A'$

$$|A \cdot A' - \lambda I| = 0 \longrightarrow (1)$$

$$- \begin{vmatrix} (16 - \lambda) & 12 \\ 12 & (34 - \lambda) \end{vmatrix} = 0$$

$$\therefore (16 - \lambda) \times (34 - \lambda) - 12 \times 12 = 0$$

$$\therefore (544 - 50\lambda + \lambda^2) - 144 = 0$$

$$\therefore (\lambda^2 - 50\lambda + 400) = 0$$

$$\therefore (\lambda - 10)(\lambda - 40) = 0$$

$$\therefore (\lambda - 10) = 0 \text{ or } (\lambda - 40) = 0$$

\therefore The eigenvalues of the matrix A are given by $\lambda = 10, 40,$

STEP 3: Substitute the value of λ in (1)

1. Eigenvectors for $\lambda = 40$

$$A - \lambda I = \begin{bmatrix} 16 & 12 \\ 12 & 34 \end{bmatrix} - 40 \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} 16 & 12 \\ 12 & 34 \end{bmatrix} - \begin{bmatrix} 40 & 0 \\ 0 & 40 \end{bmatrix}$$

$$= \begin{bmatrix} -24 & 12 \\ 12 & -6 \end{bmatrix}$$

Now, reduce this matrix

$$R_1 \leftarrow R_1 \div -24$$

$$= \begin{bmatrix} 1 & -\frac{1}{2} \\ 12 & -6 \end{bmatrix}$$

$$R_2 \leftarrow R_2 - 12 \times R_1$$

$$= \begin{bmatrix} 1 & -\frac{1}{2} \\ 0 & 0 \end{bmatrix}$$

The system associated with the eigenvalue $\lambda = 40$

$$(A - \lambda I) \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 & -\frac{1}{2} \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\Rightarrow x_1 - \frac{1}{2}x_2 = 0$$

$$\Rightarrow x_1 = \frac{1}{2}x_2$$

\therefore eigenvectors corresponding to the eigenvalue $\lambda = 40$ is

$$v = \begin{bmatrix} \frac{1}{2}x_2 \\ x_2 \end{bmatrix}$$

Let $x_2 = 1$

$$v_1 = \begin{bmatrix} \frac{1}{2} \\ 1 \end{bmatrix}$$

2. Eigenvectors for $\lambda = 10$

$$A - A' - \lambda I = \begin{bmatrix} 16 & 12 \\ 12 & 34 \end{bmatrix} - 10 \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} 16 & 12 \\ 12 & 34 \end{bmatrix} - \begin{bmatrix} 10 & 0 \\ 0 & 10 \end{bmatrix}$$

$$= \begin{bmatrix} 6 & 12 \\ 12 & 24 \end{bmatrix}$$

Now, reduce this matrix
interchanging rows $R_1 \leftrightarrow R_2$

$$= \begin{bmatrix} 12 & 24 \\ 6 & 12 \end{bmatrix}$$

$$R_1 \leftarrow R_1 \div 12$$

$$= \begin{bmatrix} 1 & 2 \\ 6 & 12 \end{bmatrix}$$

$$R_2 \leftarrow R_2 - 6 \times R_1$$

$$= \begin{bmatrix} 1 & 2 \\ 0 & 0 \end{bmatrix}$$

The system associated with the eigenvalue $\lambda = 10$

$$(A - A' - 10I) \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 & 2 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\Rightarrow x_1 + 2x_2 = 0$$

$$\Rightarrow x_1 = -2x_2$$

\therefore eigenvectors corresponding to the eigenvalue $\lambda = 10$ is

$$v = \begin{bmatrix} -2x_2 \\ x_2 \end{bmatrix}$$

Let $x_2 = 1$

$$v_2 = \begin{bmatrix} -2 \\ 1 \end{bmatrix}$$

For Eigenvector-1 $\left(\frac{1}{2}, 1\right)$, Length $L = \sqrt{0.5^2 + 1^2} = 1.11803$

So, normalizing gives $u_1 = \left(\frac{0.5}{1.11803}, \frac{1}{1.11803}\right) = (0.4472, 0.8944)$

For Eigenvector-2 $(-2, 1)$, Length $L = \sqrt{(-2)^2 + 1^2} = 2.23607$

So, normalizing gives $u_2 = \left(\frac{-2}{2.23607}, \frac{1}{2.23607}\right) = (-0.8944, 0.4472)$

$$\therefore \Sigma = \begin{bmatrix} \sqrt{40} & 0 \\ 0 & \sqrt{10} \end{bmatrix} = \begin{bmatrix} 6.32456 & 0 \\ 0 & 3.16228 \end{bmatrix}$$

$$\therefore U = [u_1, u_2] = \begin{bmatrix} 0.44721 & -0.89443 \\ 0.89443 & 0.44721 \end{bmatrix}$$

→ (2)

V is found using formula $v_i = \frac{1}{\sigma_i} A^T \cdot u_i$

$$\therefore V = \begin{bmatrix} 0.70711 & -0.70711 \\ -0.70711 & -0.7071 \end{bmatrix}$$

Firstly take, u_1 (column wise) and sigma 1, to find v_1

Then take u_2 and sigma2 , to find v_2

Another Way to find V and U

$$\therefore A' \cdot A = \begin{bmatrix} 25 & -15 \\ -15 & 25 \end{bmatrix}$$

To Find V using A.A'

Find Eigen vector for $A' \cdot A$

$$|A' \cdot A - \lambda I| = 0$$

$$\begin{vmatrix} (25 - \lambda) & -15 \\ -15 & (25 - \lambda) \end{vmatrix} = 0$$

$$\therefore (25 - \lambda) \times (25 - \lambda) - (-15) \times (-15) = 0$$

$$\therefore (625 - 50\lambda + \lambda^2) - 225 = 0$$

$$\therefore (\lambda^2 - 50\lambda + 400) = 0$$

$$\therefore (\lambda - 10)(\lambda - 40) = 0$$

$$\therefore (\lambda - 10) = 0 \text{ or } (\lambda - 40) = 0$$

\therefore The eigenvalues of the matrix A are given by $\lambda = 10, 40,$

1. Eigenvectors for $\lambda = 40$

$$A' \cdot A - \lambda I = \begin{bmatrix} 25 & -15 \\ -15 & 25 \end{bmatrix} - 40 \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} 25 & -15 \\ -15 & 25 \end{bmatrix} - \begin{bmatrix} 40 & 0 \\ 0 & 40 \end{bmatrix}$$

$$= \begin{bmatrix} -15 & -15 \\ -15 & -15 \end{bmatrix}$$

Now, reduce this matrix

$$R_1 \leftarrow R_1 \div -15$$

$$= \begin{bmatrix} 1 & 1 \\ -15 & -15 \end{bmatrix}$$

$$R_2 \leftarrow R_2 + 15 \times R_1$$

$$= \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$$

The system associated with the eigenvalue $\lambda = 40$

$$(A' \cdot A - 40I) \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\Rightarrow x_1 + x_2 = 0$$

$$\Rightarrow x_1 = -x_2$$

\therefore eigenvectors corresponding to the eigenvalue $\lambda = 40$ is

$$v = \begin{bmatrix} -x_2 \\ x_2 \end{bmatrix}$$

Let $x_2 = 1$

$$v_1 = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

2. Eigenvectors for $\lambda = 10$

$$A' \cdot A - \lambda I = \begin{bmatrix} 25 & -15 \\ -15 & 25 \end{bmatrix} - 10 \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} 25 & -15 \\ -15 & 25 \end{bmatrix} - \begin{bmatrix} 10 & 0 \\ 0 & 10 \end{bmatrix}$$

$$= \begin{bmatrix} 15 & -15 \\ -15 & 15 \end{bmatrix}$$

Now, reduce this matrix

$$R_1 \leftarrow R_1 \div 15$$

$$= \begin{bmatrix} 1 & -1 \\ -15 & 15 \end{bmatrix}$$

$$R_2 \leftarrow R_2 + 15 \times R_1$$

$$= \begin{bmatrix} 1 & -1 \\ 0 & 0 \end{bmatrix}$$

The system associated with the eigenvalue $\lambda = 10$

$$(A' \cdot A - 10I) \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 & -1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\Rightarrow x_1 - x_2 = 0$$

$$\Rightarrow x_1 = x_2$$

\therefore eigenvectors corresponding to the eigenvalue $\lambda = 10$ is

$$v = \begin{bmatrix} x_2 \\ x_2 \end{bmatrix}$$

Let $x_2 = 1$

$$v_2 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

For Eigenvector-1 (- 1, 1), Length $L = \sqrt{(-1)^2 + 1^2} = 1.41421$

So, normalizing gives $v_1 = \left(\frac{-1}{1.41421}, \frac{1}{1.41421} \right) = (-0.7071, 0.7071)$

For Eigenvector-2 (1, 1), Length $L = \sqrt{1^2 + 1^2} = 1.41421$

So, normalizing gives $v_2 = \left(\frac{1}{1.41421}, \frac{1}{1.41421} \right) = (0.7071, 0.7071)$

$$\therefore \Sigma = \begin{bmatrix} \sqrt{40} & 0 \\ 0 & \sqrt{10} \end{bmatrix} = \begin{bmatrix} 6.32456 & 0 \\ 0 & 3.16228 \end{bmatrix}$$

$$\therefore V = [v_1, v_2] = \begin{bmatrix} -0.70711 & 0.70711 \\ 0.70711 & 0.70711 \end{bmatrix}$$

U is found using formula $u_i = \frac{1}{\sigma_i} A \cdot v_i$

$$\therefore U = \begin{bmatrix} -0.44722 & 0.89443 \\ -0.89443 & -0.44722 \end{bmatrix}$$

Verification by using first solution

$$U \times \Sigma = \begin{bmatrix} 0.4472 & -0.8944 \\ 0.8944 & 0.4472 \end{bmatrix} \times \begin{bmatrix} 6.3246 & 0 \\ 0 & 3.1623 \end{bmatrix}$$

$$= \begin{bmatrix} 0.4472 \times 6.3246 - 0.8944 \times 0 & 0.4472 \times 0 - 0.8944 \times 3.1623 \\ 0.8944 \times 6.3246 + 0.4472 \times 0 & 0.8944 \times 0 + 0.4472 \times 3.1623 \end{bmatrix}$$

$$= \begin{bmatrix} 2.8284 + 0 & 0 - 2.8284 \\ 5.6569 + 0 & 0 + 1.4142 \end{bmatrix}$$

$$= \begin{bmatrix} 2.8284 & -2.8284 \\ 5.6569 & 1.4142 \end{bmatrix}$$

$$(U \times \Sigma) \times (V^T) = \begin{bmatrix} 2.8284 & -2.8284 \\ 5.6569 & 1.4142 \end{bmatrix} \times \begin{bmatrix} 0.7071 & -0.7071 \\ -0.7071 & -0.7071 \end{bmatrix}$$

$$= \begin{bmatrix} 2.8284 \times 0.7071 - 2.8284 \times -0.7071 & 2.8284 \times -0.7071 - 2.8284 \times -0.7071 \\ 5.6569 \times 0.7071 + 1.4142 \times -0.7071 & 5.6569 \times -0.7071 + 1.4142 \times -0.7071 \end{bmatrix}$$

$$= \begin{bmatrix} 2 + 2 & -2 + 2 \\ 4 - 1 & -4 - 1 \end{bmatrix}$$

$$= \begin{bmatrix} 4 & 0 \\ 3 & -5 \end{bmatrix}$$

Verification by using second solution

$$U \times \Sigma = \begin{bmatrix} -0.4472 & 0.8944 \\ -0.8944 & -0.4472 \end{bmatrix} \times \begin{bmatrix} 6.3246 & 0 \\ 0 & 3.1623 \end{bmatrix}$$

$$= \begin{bmatrix} -0.4472 \times 6.3246 + 0.8944 \times 0 & -0.4472 \times 0 + 0.8944 \times 3.1623 \\ -0.8944 \times 6.3246 - 0.4472 \times 0 & -0.8944 \times 0 - 0.4472 \times 3.1623 \end{bmatrix}$$

$$= \begin{bmatrix} -2.8284 + 0 & 0 + 2.8284 \\ -5.6569 + 0 & 0 - 1.4142 \end{bmatrix}$$

$$= \begin{bmatrix} -2.8284 & 2.8284 \\ -5.6569 & -1.4142 \end{bmatrix}$$

$$(U \times \Sigma) \times (V^T) = \begin{bmatrix} -2.8284 & 2.8284 \\ -5.6569 & -1.4142 \end{bmatrix} \times \begin{bmatrix} -0.7071 & 0.7071 \\ 0.7071 & 0.7071 \end{bmatrix}$$

$$= \begin{bmatrix} -2.8284 \times -0.7071 + 2.8284 \times 0.7071 & -2.8284 \times 0.7071 + 2.8284 \times 0.7071 \\ -5.6569 \times -0.7071 - 1.4142 \times 0.7071 & -5.6569 \times 0.7071 - 1.4142 \times 0.7071 \end{bmatrix}$$

$$= \begin{bmatrix} 2 + 2 & -2 + 2 \\ 4 - 1 & -4 - 1 \end{bmatrix}$$

$$= \begin{bmatrix} 4 & 0 \\ 3 & -5 \end{bmatrix}$$

Dimensionality Reduction:

Consider first singular value only and U matrix (i.e. Equation (2))

$$\begin{array}{l} \text{Projected} \\ \text{Data} = \end{array} \begin{pmatrix} 0.4472136 & -0.89442719 \\ 0.89442719 & 0.4472136 \end{pmatrix} \times \begin{pmatrix} 6.32456 \\ 0 \end{pmatrix}$$
$$= \begin{pmatrix} 2.8284 \\ 5.6568 \end{pmatrix}$$

SVD properties

A square ($n \times n$) matrix A is singular iff at least one of its singular values $\sigma_1, \dots, \sigma_n$ is zero.

The rank of matrix A is equal to the number of nonzero singular values σ_i

