

Singular Value Decomposition

let A be given matrix

$$[A]_{m \times n} = U_{(m \times r)} \Sigma_{(r \times r)} (V_{(n \times r)})^T$$

A : Input data matrix

where U : is left singular vector
($m \times r$ matrix)

Σ : Singular values

$r \times r$ diagonal matrix

Singular values are in decreasing order

$$\begin{pmatrix} \tau_1 & 0 \\ 0 & \tau_2 \end{pmatrix} \quad \tau_1 > \tau_2$$

V : Right singular vectors
 $n \times r$ matrix where r is 'rank'

SVD properties:

it is always possible to decompose a real matrix

A into $A = U \Sigma V^T$ where

U, Σ, V : unique

U, V : column orthonormal vectors (matrices)

$$U U^T = V V^T = I$$

Columns are orthogonal unit vectors

Σ : diagonal ; $\Sigma = \begin{bmatrix} \sigma_1 & 0 \\ 0 & \sigma_2 \end{bmatrix}$

Entries of principal diagonal positions

(Singular values) are positive.

and they are stored in decreasing order

$$(\sigma_1 \geq \sigma_2 \geq \dots \geq 0)$$

Find singular value decomposition (SVD) of a matrix

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

$$A = U \Sigma V^T$$

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

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

$$(AA^T - I)\mathbf{x} = 0$$

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

Characteristic equation $|A \cdot A^T - I\lambda| = 0$

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

$$(16-\lambda)(34-\lambda) - 144 = 0$$

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

$$\lambda^2 - 50\lambda + 400 = 0$$

$$\lambda^2 - 40\lambda - 10\lambda + 400 = 0$$

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

$$(\lambda - 10)(\lambda - 40)$$

The Eigen values of AA^T are 10 and 40

The Eigen vector for $\lambda = 40$ is

$$(AA^T - I\lambda)\mathbf{x} = 0 \quad \begin{pmatrix} 16-40 & 12 \\ 12 & 34-40 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$= \begin{pmatrix} -24 & 12 \\ 12 & -6 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$R_2 \rightarrow 2R_2 + 1R_1$$

$$= \begin{pmatrix} -24 & 12 \\ -24+24 & -12+12 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$= \begin{pmatrix} -24 & 12 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$= -24x_1 + 12x_2 = 0$$

$$= -12(2x_1 - x_2) = 0$$

$$2x_1 - x_2 = 0 \quad \text{let } x_1 = K \text{ then}$$

$$x_2 = 2K$$

$$\therefore \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K \\ 2K \end{pmatrix} = K \begin{pmatrix} 1 \\ 2 \end{pmatrix}$$

The eigen vector for $\lambda = 10$ is

$$(AA^T - I\lambda)x = 0$$

$$\begin{pmatrix} 16 & -10 & 12 \\ 12 & 34 & -10 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$\begin{pmatrix} 6 & 12 \\ 12 & 24 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \quad R_2 \rightarrow R_2 - 2R_1$$

$$\begin{pmatrix} 6 & 12 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \quad \begin{array}{l} 6x_1 + 12x_2 = 0 \quad 6(x_1 + 2x_2) = 0 \\ \text{let } x_2 = k \quad \text{then} \end{array}$$

$$x_1 = -2x_2 = -2k$$

$$\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} -2k \\ k \end{pmatrix} = k \begin{pmatrix} -2 \\ 1 \end{pmatrix}$$

\therefore For eigen vector for $\lambda = 40$, is $\begin{pmatrix} 1 \\ 2 \end{pmatrix}$

$$u_1 = \begin{pmatrix} 1 \\ 2 \end{pmatrix} \quad \|u_1\| = \sqrt{1+4} = \sqrt{5}$$

$$e_1 = \frac{u_1}{\|u_1\|} = \frac{1}{\sqrt{5}} \begin{pmatrix} 1 \\ 2 \end{pmatrix} = \begin{pmatrix} 1/\sqrt{5} \\ 2/\sqrt{5} \end{pmatrix}$$

for eigen vector for $\lambda = 10$ is $\begin{pmatrix} -2 \\ 1 \end{pmatrix}$

$$u_2 = \begin{pmatrix} -2 \\ 1 \end{pmatrix}$$

$$\|u_2\| = \sqrt{(-2)^2 + 1^2} = \sqrt{4+1} = \sqrt{5}$$

$$e_2 = \frac{u_2}{\|u_2\|} = \frac{\begin{pmatrix} -2 \\ 1 \end{pmatrix}}{\sqrt{5}} = \begin{pmatrix} -2/\sqrt{5} \\ 1/\sqrt{5} \end{pmatrix}$$

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

$$\begin{aligned} v_1 &= \frac{1}{\sigma_1} A^T e_1 \\ &= \frac{1}{\sqrt{40}} \begin{bmatrix} 4 & 3 \\ 0 & -5 \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{5}} \\ \frac{2}{\sqrt{5}} \end{bmatrix} = \frac{1}{\sqrt{40}} \begin{bmatrix} \frac{10}{\sqrt{5}} \\ -\frac{10}{\sqrt{5}} \end{bmatrix} = \begin{bmatrix} \frac{10}{\sqrt{200}} \\ -\frac{10}{\sqrt{200}} \end{bmatrix} \\ &= \begin{bmatrix} \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \end{bmatrix} \end{aligned}$$

$$\begin{aligned} v_2 &= \frac{1}{\sigma_2} A^T e_2 \\ &= \frac{1}{\sqrt{10}} \begin{bmatrix} 4 & 3 \\ 0 & -5 \end{bmatrix} \begin{bmatrix} -\frac{2}{\sqrt{5}} \\ \frac{1}{\sqrt{5}} \end{bmatrix} = \frac{1}{\sqrt{10}} \begin{bmatrix} -\frac{8+3}{\sqrt{5}} \\ -\frac{5}{\sqrt{5}} \end{bmatrix} \\ &= \frac{1}{\sqrt{10}} \begin{bmatrix} -\frac{11}{\sqrt{5}} \\ -\frac{5}{\sqrt{5}} \end{bmatrix} \\ &= \begin{bmatrix} -\frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \end{bmatrix} \end{aligned}$$

$$\therefore V = (v_1, v_2) = \begin{pmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{pmatrix}$$

Verification

$$A = U \Sigma V^T$$

$$\begin{aligned} &= \begin{bmatrix} \frac{1}{\sqrt{5}} & -\frac{2}{\sqrt{5}} \\ \frac{2}{\sqrt{5}} & \frac{1}{\sqrt{5}} \end{bmatrix} \begin{bmatrix} \sqrt{40} & 0 \\ 0 & \sqrt{10} \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{bmatrix} \\ &= \begin{bmatrix} \sqrt{8} & -2\sqrt{8} \\ 2\sqrt{8} & \sqrt{2} \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{bmatrix} \\ &= \begin{bmatrix} \sqrt{4} + \sqrt{4} & 0 \\ \sqrt{16} - 1 & -\sqrt{16} - 1 \end{bmatrix} = \begin{bmatrix} 4 & 0 \\ 3 & -5 \end{bmatrix} \end{aligned}$$

Singular value Decomposition

Let $\{v_1, v_2, v_3 \dots v_n\}$ be the eigen vectors of $(A^T A)_{n \times n}$ and $\{u_1, u_2, \dots u_m\}$ be the eigen vectors corresponding to $(A A^T)_{m \times m}$

Let $\lambda_i = \sigma_i^2$ be the eigen values of $A^T A$

Then σ_i 's are called as the Singular values of A . and $\{v_1, v_2, \dots v_n\}$ and $\{u_1, u_2, \dots u_m\}$ are the singular vectors of A

$$A = U \Sigma V^T = \begin{pmatrix} u_1 & u_2 & \dots & u_m \end{pmatrix}_{m \times m} \begin{pmatrix} \sigma_1 & 0 & 0 & \dots & 0 & 0 & 0 \\ 0 & \sigma_2 & 0 & \dots & 0 & \dots & \\ \vdots & & & & & & \\ 0 & \dots & \dots & \dots & \sigma_n & 0 & 0 \end{pmatrix} \begin{pmatrix} v_1 & v_2 & \dots & v_n \end{pmatrix}_{n \times n}^T$$

Find Singular value decomposition

$$A = \begin{bmatrix} 3 & 2 \\ 3 & -2 \end{bmatrix}$$

$$A \cdot A^T = \begin{bmatrix} 3 & 2 \\ 3 & -2 \end{bmatrix} \begin{bmatrix} 3 & 3 \\ 2 & -2 \end{bmatrix} = \begin{bmatrix} 13 & 5 \\ 5 & 13 \end{bmatrix}$$

Eigen values of $A A^T = |A A^T - \lambda I| = 0$

$$\begin{vmatrix} 13-\lambda & 5 \\ 5 & 13-\lambda \end{vmatrix} = 0$$

$$(13-\lambda)^2 - 25 = 0$$

$$169 + \lambda^2 - 26\lambda - 25 = 0$$

$$\lambda^2 - 26\lambda + 144 = 0$$

$$\lambda^2 - 18\lambda - 8\lambda + 144 = 0$$

$$(\lambda - 8)(\lambda - 18) = 0$$

$$\lambda = 8, \lambda = 18$$

Now consider $A \cdot A^T$ for finding eigen

vectors for $\boxed{\lambda_1 = 18}$ $|A \cdot A^T - \lambda I| = 0$

$$\left[\begin{pmatrix} 13 & 5 \\ 5 & 13 \end{pmatrix} - \begin{pmatrix} 18 & 0 \\ 0 & 18 \end{pmatrix} \right] \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$\begin{pmatrix} -5 & 5 \\ 5 & -5 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$R_2 \rightarrow R_2 + R_1 \sim \begin{pmatrix} -5 & 5 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$-5x_1 + 5x_2 = 0 \Rightarrow 5(-x_1 + x_2) = 0 \Rightarrow -x_1 + x_2 = 0$$

$$\text{let } x_2 = K, \text{ then } x_1 = K \quad \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K \\ K \end{pmatrix} = K \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$
$$u_1 = \begin{pmatrix} 1 \\ 1 \end{pmatrix} \quad K \in \mathbb{R} - \{0\}$$

$$\therefore e_1 = \frac{u_1}{\|u_1\|} = \frac{\begin{pmatrix} 1 \\ 1 \end{pmatrix}}{\sqrt{1+1}} = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{pmatrix}$$

The eigen vectors for the eigen value $\boxed{\lambda_2 = 8}$

$$|A \cdot A^T - \lambda I| = 0$$

$$\left[\begin{pmatrix} 13 & 5 \\ 5 & 13 \end{pmatrix} - \begin{pmatrix} 8 & 0 \\ 0 & 8 \end{pmatrix} \right] \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$\begin{pmatrix} 5 & 5 \\ 5 & 5 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$R_2 \rightarrow R_2 - R_1 \sim \begin{pmatrix} 5 & 5 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$5(x_1 + x_2) = 0 \quad x_1 + x_2 = 0$$

One equation with 2 unknowns we get infinite number of solutions let $x_2 = K$ then $\boxed{x_1 = -K}$

$$\therefore \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} -K \\ K \end{pmatrix} = K \begin{pmatrix} -1 \\ 1 \end{pmatrix} \quad K \in \mathbb{R} - \{0\}$$

$$u_2 = \begin{pmatrix} -1 \\ 1 \end{pmatrix}$$

$$\therefore e_2 = \frac{u_2}{\|u_2\|} = \frac{\begin{pmatrix} -1 \\ 1 \end{pmatrix}}{\sqrt{(-1)^2 + (1)^2}} = \frac{1}{\sqrt{2}} \begin{pmatrix} -1 \\ 1 \end{pmatrix}$$
$$e_2 = \begin{pmatrix} -1/\sqrt{2} \\ 1/\sqrt{2} \end{pmatrix}$$

V is found using formula

$$V_i = \frac{1}{\sigma_i} A^T U_i$$

$$\boxed{\sigma_1 > \sigma_2}$$

$$V_1 = \frac{1}{\sigma_1} A^T U_1$$

$$= \frac{1}{\sqrt{18}} \begin{pmatrix} 3 & 3 \\ 2 & -2 \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{pmatrix}$$

$$= \frac{1}{\sqrt{18}} \begin{pmatrix} \frac{6}{\sqrt{2}} \\ 0 \end{pmatrix} = \begin{pmatrix} \frac{6}{\sqrt{36}} \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

$$V_2 = \frac{1}{\sigma_2} A^T U_2$$

$$= \frac{1}{\sqrt{8}} \begin{pmatrix} 3 & 3 \\ 2 & -2 \end{pmatrix} \begin{pmatrix} -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{pmatrix}$$

$$= \frac{1}{\sqrt{8}} \begin{pmatrix} 0 \\ -\frac{4}{\sqrt{2}} \end{pmatrix} = \begin{pmatrix} 0 \\ -\frac{4}{\sqrt{16}} \end{pmatrix} = \begin{pmatrix} 0 \\ -1 \end{pmatrix}$$

$$\therefore U = \begin{bmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix} \quad \Sigma = \begin{bmatrix} \sqrt{18} & 0 \\ 0 & \sqrt{8} \end{bmatrix}$$

$$V = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$$

$$V^T = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$$

Verification: $U \Sigma V^T$

$$= \begin{bmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix} \begin{bmatrix} \sqrt{18} & 0 \\ 0 & \sqrt{8} \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$$

$$= \begin{pmatrix} \frac{\sqrt{18}}{\sqrt{2}} & -\frac{\sqrt{18}}{\sqrt{2}} \\ \frac{\sqrt{18}}{\sqrt{2}} & \frac{\sqrt{18}}{\sqrt{2}} \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$$

$$= \begin{pmatrix} \frac{\sqrt{9}\sqrt{2}}{\sqrt{2}} & -\frac{\sqrt{9}\sqrt{2}}{\sqrt{2}} \\ \frac{\sqrt{9}\sqrt{2}}{\sqrt{2}} & \frac{\sqrt{9}\sqrt{2}}{\sqrt{2}} \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} = \begin{pmatrix} 3 & -2 \\ 3 & 2 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$$

$$= \begin{pmatrix} 3 & 2 \\ 3 & -2 \end{pmatrix}$$

Find Singular value decomposition (SVD) of a matrix A

$$A = \begin{bmatrix} 1 & 0 & -1 \\ 1 & 1 & 1 \end{bmatrix}_{2 \times 3}$$

$$A^T = \begin{bmatrix} 1 & 1 \\ 0 & 1 \\ -1 & 1 \end{bmatrix}_{3 \times 2}$$

$$A \cdot A^T = \begin{bmatrix} 2 & 0 & -1 \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 0 & 1 \\ -1 & 1 \end{bmatrix} = \begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix}$$

$$A = U \Sigma V^T$$

$\Sigma \rightarrow$ Singular values (Eigen values)
 $\tau_1 > \tau_2$

$U =$ Eigen vectors of $A A^T$. (left singular values)
let them be (e_1, e_2)

$V =$ (Right singular values)

$$V_i = \frac{1}{\sigma_i} A^T \cdot e_i$$

$$\begin{cases} V_1 = \frac{1}{\sigma_1} A^T e_1 \\ V_2 = \frac{1}{\sigma_2} A^T e_2 \end{cases}$$

Step 1: Singular values: characteristic equation is

$$|A A^T - I \tau| = 0$$

$$\begin{vmatrix} 2-\tau & 0 \\ 0 & 3-\tau \end{vmatrix} = 0 \quad (2-\tau)(3-\tau) = 0$$

$$\tau_1 = 3; \tau_2 = 2$$

$$\therefore \Sigma = \begin{bmatrix} \sqrt{\tau_1} & 0 \\ 0 & \sqrt{\tau_2} \end{bmatrix} \\ = \begin{bmatrix} \sqrt{3} & 0 \\ 0 & \sqrt{2} \end{bmatrix}$$

$$\begin{cases} \sigma_1 = \sqrt{3} \\ \sigma_2 = \sqrt{2} \end{cases}$$

Step 2: Computation of left Singular Values

$$|A \cdot A^T - \lambda I| = 0 \quad \text{Eigen vector for } \lambda = 3$$

$$\begin{pmatrix} 2-3 & 0 \\ 0 & 3-3 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$\begin{pmatrix} -1 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$-x_1 = 0 \Rightarrow \boxed{x_1 = 0} \quad \text{Here } x_2 \text{ is missing}$$

$$\text{Variable } \therefore x_2 = k$$

$$\therefore u_1 = \begin{pmatrix} 0 \\ k \end{pmatrix} = k \begin{pmatrix} 0 \\ 1 \end{pmatrix} \quad k \in \mathbb{R} - \{0\}$$

$$e_1 = \frac{u_1}{\|u_1\|} = \frac{\begin{pmatrix} 0 \\ 1 \end{pmatrix}}{\sqrt{1}} = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

Eigen vector $\lambda = 2$

$$\begin{pmatrix} 2-2 & 0 \\ 0 & 3-2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$\begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$x_2 = 0 \quad x_1 \text{ is missing variable } \therefore \text{let } x_1 = k$$

$$\therefore u_2 = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} k \\ 0 \end{pmatrix} = k \begin{pmatrix} 1 \\ 0 \end{pmatrix} \quad u_2 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

$$\therefore \|u_2\| = \sqrt{1+0} = \sqrt{1} = 1$$

$$\therefore U = (e_1, e_2) = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$$

Step 3: $v_i = \frac{1}{\sigma_i} A^T e_i$

$$v_1 = \frac{1}{\sigma_1} A^T e_1 = \frac{1}{\sqrt{3}} \begin{pmatrix} 1 & 1 \\ 0 & 1 \\ -1 & 1 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

$3 \times 2 \quad 2 \times 1$

$$= \frac{1}{\sqrt{3}} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$$

$$= \begin{pmatrix} 1/\sqrt{3} \\ 1/\sqrt{3} \\ 1/\sqrt{3} \end{pmatrix}$$

$$V_2 = \frac{1}{\sigma_2} A^T \cdot e_2$$

$$= \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ 0 & 1 \\ -1 & 1 \end{pmatrix}_{3 \times 2} \begin{pmatrix} 1 \\ 0 \end{pmatrix}_{2 \times 1}$$

$$= \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} = \begin{pmatrix} 1/\sqrt{2} \\ 0 \\ -1/\sqrt{2} \end{pmatrix}$$

$$\therefore V = (V_1, V_2) = \begin{pmatrix} \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{3}} & 0 \\ \frac{1}{\sqrt{3}} & -\frac{1}{\sqrt{2}} \end{pmatrix}$$

$$\begin{aligned} \therefore A &= U \Sigma V^T \\ &= \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} \sqrt{3} & 0 \\ 0 & \sqrt{2} \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{3}} & 0 \\ \frac{1}{\sqrt{3}} & -\frac{1}{\sqrt{2}} \end{pmatrix}^T \\ &= \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} \sqrt{3} & 0 \\ 0 & \sqrt{2} \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{2}} & 0 & -\frac{1}{\sqrt{2}} \end{pmatrix} \\ &= \begin{pmatrix} 0 & \sqrt{2} \\ \sqrt{3} & 0 \end{pmatrix} \begin{pmatrix} 1/\sqrt{3} & 1/\sqrt{3} & 1/\sqrt{3} \\ 1/\sqrt{2} & 0 & -1/\sqrt{2} \end{pmatrix} \\ &= \begin{pmatrix} 1 & 0 & -1 \\ 1 & 1 & 1 \end{pmatrix} \end{aligned}$$