

Orthogonality (Unit 3)

Defn: Let V be an inner product space.

The vectors $u, v \in V$ are said to be orthogonal and u is said to be orthogonal to v if

$$\langle u, v \rangle = 0.$$

Ex (1) Consider the vectors
 $u = (1, 1, 1)$, $v = (1, 2, -3)$, $w = (1, -4, 3)$ in \mathbb{R}^3 ;

Then $\langle u, v \rangle = 1 + 2 - 3 = 0$

$$\langle u, w \rangle = 1 - 4 + 3 = 0$$

$$\langle v, w \rangle = 1 - 8 - 9 = -16.$$

Thus u is orthogonal to v & w .
 but v & w are not orthogonal.

Ex (2) Consider the functions $\sin t$ & $\cos t$ in the vector space $C[-\pi, \pi]$ of continuous functions on the closed interval $[-\pi, \pi]$. Then

$$\begin{aligned} \langle \sin t, \cos t \rangle &= \int_{-\pi}^{\pi} \sin t \cos t \, dt \\ &= \left(\frac{1}{2} \sin^2 t \right)_{-\pi}^{\pi} = 0. \end{aligned}$$

Thus $\sin t$ & $\cos t$ are orthogonal functions in the vector space $C[-\pi, \pi]$.

Ex In \mathbb{R}^3 , having Euclidean inner product.
For which values of k , u & v are orthogonal?

- (i) $u = (1, 4, 2)$, $v = (3, -2, k)$.
(ii) $u = (k, -2, 4)$, $v = (k, k, -2)$

Soln (i) If u & v are orthogonal
 $\langle u, v \rangle = 0$

$$\Rightarrow \langle u, v \rangle = 3 - 8 + 2k = 0.$$
$$-5 + 2k = 0$$
$$k = \frac{5}{2}$$

(ii) $\langle u, v \rangle = 0$

$$\Rightarrow k^2 - 2k - 8 = 0.$$
$$k = \underline{\underline{-2, 4.}}$$

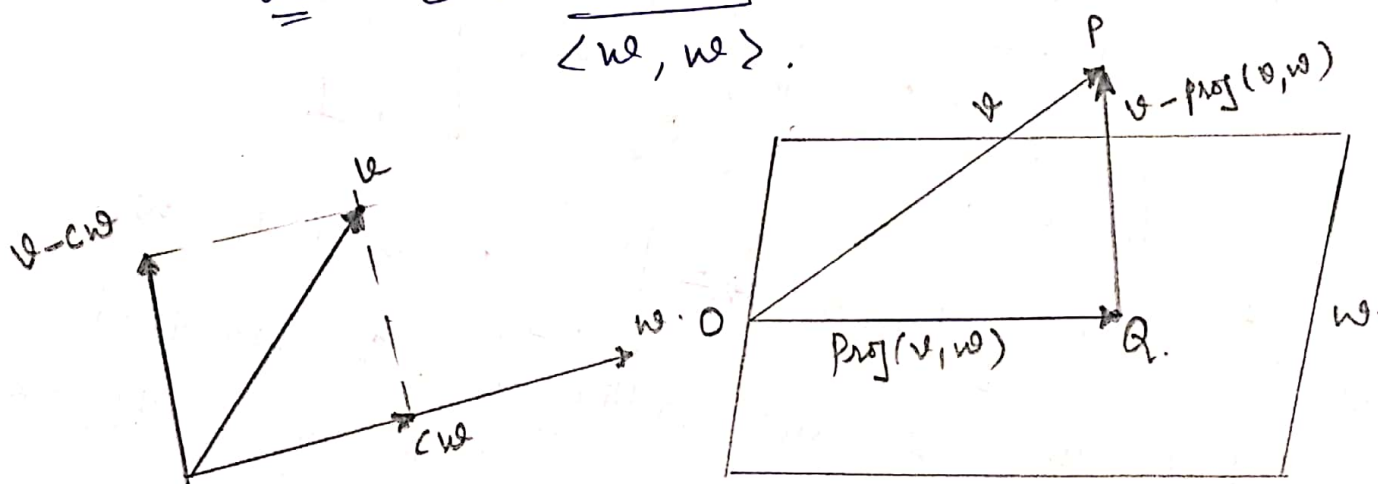
Projections.

Let V be an inner product space. Suppose w is a given non zero vector in V , and suppose v is another vector. The projection of v along w will be the multiple cw of w such that $v' = v - cw$ is orthogonal to w . This means

$$\langle v - cw, w \rangle = 0.$$

$$\Leftrightarrow \langle v, w \rangle - c \langle w, w \rangle = 0.$$

$$\Leftrightarrow c = \frac{\langle v, w \rangle}{\langle w, w \rangle}.$$



The projection of v along w is denoted & defined by -

$$\text{proj}(v, w) = cw = \frac{\langle v, w \rangle}{\langle w, w \rangle} w.$$

Ex Let w be the plane in \mathbb{R}^3 , with equation $x - y + 2z = 0$, and let $v = (3, -1, 2)$. Find the orthogonal projection of v onto w & let the component of v orthogonal to w .

Soln

$$W: x - y + 2z = 0.$$

The subspace w is a plane through the origin in \mathbb{R}^3 . From the equation of the plane, we have $x = y - 2z$, so w consists of vectors of the form

$$\begin{bmatrix} y - 2z \\ y \\ z \end{bmatrix} = y \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} + z \begin{bmatrix} -2 \\ 0 \\ 1 \end{bmatrix}.$$

$$\Rightarrow u_1 = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} \text{ \& } u_2 = \begin{bmatrix} -2 \\ 0 \\ 1 \end{bmatrix} \text{ are basis of } w.$$

but they are not orthogonal, so we find another non-zero vector in w , which is orthogonal to either one of these.

Let $w = \begin{bmatrix} x \\ y \\ z \end{bmatrix}$ is a vector in w , which is orthogonal to u_1 . Then $x - y + 2z = 0$, Since w is in the plane w .

Since $u_1 \cdot w = 0$, we have $x + y = 0$.

$$\text{Solving } \begin{cases} x - y + 2z = 0 \\ x + y = 0 \end{cases}$$

$$\Rightarrow \begin{cases} x = -z, \\ y = z. \end{cases}$$

$$\Rightarrow w = \begin{bmatrix} -z \\ z \\ z \end{bmatrix} = w = \begin{bmatrix} -1 \\ 1 \\ 1 \end{bmatrix}.$$

So $w = \begin{bmatrix} -1 \\ 1 \\ 1 \end{bmatrix}$ is an orthogonal set in w
 & hence an orthogonal basis.

So the orthogonal basis for w are -

$$u_1 = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} \quad \& \quad u_2 = \begin{bmatrix} -1 \\ 1 \\ 1 \end{bmatrix}.$$

$$\text{we have } u_1 \cdot v = 2, \quad u_2 \cdot v = -2$$

$$u_1 \cdot u_1 = 2, \quad u_2 \cdot u_2 = 3$$

$$\text{Proj}_w(v) = \left(\frac{u_1 \cdot v}{u_1 \cdot u_1} \right) u_1 + \left(\frac{u_2 \cdot v}{u_2 \cdot u_2} \right) u_2$$

$$= \frac{2}{2} \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} - \frac{2}{3} \begin{bmatrix} -1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 5/3 \\ 1/3 \\ -2/3 \end{bmatrix}$$

The component of v orthogonal to w is
 the vector.

$$\text{Perp}_w(v) = v - \text{Proj}_w(v)$$

$$= \begin{bmatrix} 2 \\ 1 \\ -1 \\ 2 \end{bmatrix} - \begin{bmatrix} 5/3 \\ 1/3 \\ -2/3 \end{bmatrix} = \begin{bmatrix} 4/3 \\ -4/3 \\ 8/3 \end{bmatrix}$$

It is easy to see that $\text{Proj}_w(v)$ is in w ,
 since it satisfies the equation of the plane.
 It is equally easy to see that $\text{Perp}_w(v)$ is
 orthogonal to w , since it is a scalar
 multiple of the normal vector $\begin{bmatrix} 1 \\ -1 \\ 2 \end{bmatrix}$ to w .

The Gram Schmidt Process

Ex Find an orthogonal basis for the subspace W of \mathbb{R}^3 given by -

$$W = \left\{ \begin{bmatrix} x \\ y \\ z \end{bmatrix} : x - y + 2z = 0 \right\}.$$

Soln

we have $x = y - 2z$, so W consists of vectors of the form

$$\begin{bmatrix} y - 2z \\ y \\ z \end{bmatrix} = y \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} + z \begin{bmatrix} -2 \\ 0 \\ 1 \end{bmatrix}.$$

Let $W = \text{span}(x_1, x_2)$ where

$$x_1 = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} \quad \& \quad x_2 = \begin{bmatrix} -2 \\ 0 \\ 1 \end{bmatrix}.$$

* by Gram Schmidt process: To construct an orthogonal basis for W .

Starting with x_1 , we get a second vector that is orthogonal to it by taking the component of x_2 orthogonal to x_1 .

we set $v_1 = x_1$, so.

$$v_2 = \text{perp}_{x_1}(x_2) = x_2 - \text{proj}_{x_1}(x_2)$$
$$= x_2 - \left(\frac{x_2 \cdot x_1}{x_1 \cdot x_1} \right) x_1$$

$$v_2 = \begin{bmatrix} -2 \\ 0 \\ 1 \end{bmatrix} - \frac{(-2)}{(2)} \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} -1 \\ 1 \\ 1 \end{bmatrix}$$

Thus $\{v_1, v_2\}$ is an orthogonal set of vectors in W .
They are L.I. so form a basis.

Ex Using Gram Schmidt process. find an orthogonal basis for \mathbb{R}^3 that contains the vector $v_1 = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$.

Soln we first find any basis for \mathbb{R}^3 containing v_1 , if we take $x_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$ & $x_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$.

then $\{v_1, x_2, x_3\}$ is clearly a basis for \mathbb{R}^3 .
Using Gram Schmidt process.

$$v_2 = x_2 - \left(\frac{v_1 \cdot x_2}{v_1 \cdot v_1} \right) v_1 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} - \left(\frac{2}{14} \right) \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} = \begin{bmatrix} -1/7 \\ 5/7 \\ -3/7 \end{bmatrix}$$

$$v_2' = \begin{bmatrix} -1 \\ 5 \\ -3 \end{bmatrix}$$

and finally, $v_3 = x_3 - \left(\frac{v_1 \cdot x_3}{v_1 \cdot v_1} \right) v_1 - \left(\frac{v_2' \cdot x_3}{v_2' \cdot v_2'} \right) v_2'$

$$= \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} - \left(\frac{3}{14} \right) \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} - \left(\frac{-3}{35} \right) \begin{bmatrix} -1 \\ 5 \\ -3 \end{bmatrix} = \begin{bmatrix} -3/10 \\ 0 \\ 1/10 \end{bmatrix}$$

$$v_3' = \begin{bmatrix} -3 \\ 0 \\ 1 \end{bmatrix}$$

Then $\{v_1, v_2', v_3'\}$ is an orthogonal basis for \mathbb{R}^3 , that contains v_1 .

Ex

Apply Gram-Schmidt process to construct an orthonormal basis for the subspace

$W = \text{span}(x_1, x_2, x_3)$ of \mathbb{R}^4 where

$$x_1 = \begin{bmatrix} 1 \\ -1 \\ -1 \\ 1 \end{bmatrix}, \quad x_2 = \begin{bmatrix} 2 \\ 1 \\ 0 \\ 1 \end{bmatrix}, \quad x_3 = \begin{bmatrix} 2 \\ 2 \\ 1 \\ 2 \end{bmatrix}$$

Soln

First we note that $\{x_1, x_2, x_3\}$ is a L.I set, so it forms a basis for W .

we set $v_1 = x_1$.

To compute the component of x_2 orthogonal to $W_1 = \text{span}(v_1)$

$$\begin{aligned} v_2 &= \text{perp}_{W_1}(x_2) = x_2 - \left(\frac{v_1 \cdot x_2}{v_1 \cdot v_1} \right) v_1 \\ &= \begin{bmatrix} 2 \\ 1 \\ 0 \\ 1 \end{bmatrix} - \left(\frac{2}{4} \right) \begin{bmatrix} 1 \\ -1 \\ -1 \\ 1 \end{bmatrix} = \begin{bmatrix} 3/2 \\ 3/2 \\ 1/2 \\ 1/2 \end{bmatrix} \end{aligned}$$

(rescaling)

$$\leftarrow v_2' = 2v_2 = \begin{bmatrix} 3 \\ 3 \\ 1 \\ 1 \end{bmatrix}$$

To find component of x_3 orthogonal to $W_2 = \text{span}(x_1, x_2) = \text{span}(v_1, v_2)$

using the orthogonal basis $\{v_1, v_2'\}$

$$\begin{aligned} v_3 &= \text{perp}_{W_2}(x_3) = x_3 - \left(\frac{v_1 \cdot x_3}{v_1 \cdot v_1} \right) v_1 - \left(\frac{v_2' \cdot x_3}{v_2' \cdot v_2'} \right) v_2' \\ &= \begin{bmatrix} 2 \\ 2 \\ 1 \\ 2 \end{bmatrix} - \left(\frac{1}{4} \right) \begin{bmatrix} 1 \\ -1 \\ -1 \\ 1 \end{bmatrix} - \left(\frac{15}{20} \right) \begin{bmatrix} 3 \\ 3 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} -1/2 \\ 0 \\ 1/2 \\ 1 \end{bmatrix} \end{aligned}$$

(rescaling)

$$\leftarrow v_3' = 2v_3 = \begin{bmatrix} -1 \\ 0 \\ 1 \\ 2 \end{bmatrix}$$

$\{v_1, v_2', v_3'\}$ forms an orthogonal basis for W .

To obtain an orthonormal basis, we normalize each vector —

$$q_1 = \frac{v_1}{\|v_1\|} = \frac{1}{2} \begin{bmatrix} 1 \\ -1 \\ -1 \\ 1 \end{bmatrix} = \begin{bmatrix} 1/2 \\ -1/2 \\ -1/2 \\ 1/2 \end{bmatrix}$$

$$q_2 = \frac{v_2'}{\|v_2'\|} = \frac{1}{2\sqrt{5}} \begin{bmatrix} 3 \\ 3 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 3/2\sqrt{5} \\ 3/2\sqrt{5} \\ 1/2\sqrt{5} \\ 1/2\sqrt{5} \end{bmatrix} = \begin{bmatrix} 3\sqrt{5}/10 \\ 3\sqrt{5}/10 \\ \sqrt{5}/10 \\ \sqrt{5}/10 \end{bmatrix}$$

$$q_3 = \frac{v_3'}{\|v_3'\|} = \left(\frac{1}{\sqrt{6}}\right) \begin{bmatrix} -1 \\ 0 \\ 1 \\ 2 \end{bmatrix} = \begin{bmatrix} -1/\sqrt{6} \\ 0 \\ 1/\sqrt{6} \\ 2/\sqrt{6} \end{bmatrix} = \begin{bmatrix} -\sqrt{6}/6 \\ 0 \\ \sqrt{6}/6 \\ \sqrt{6}/3 \end{bmatrix}$$

$\{q_1, q_2, q_3\}$ is an orthonormal basis for W .

QR factorization.

Ex Find a QR factorization of

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

Solⁿ The orthonormal basis for $\text{col}(A)$ produced by the Gram-Schmidt process was - (refer previous example)

$$q_1 = \begin{bmatrix} 1/2 \\ -1/2 \\ -1/2 \\ 1/2 \end{bmatrix}, \quad q_2 = \begin{bmatrix} 3\sqrt{5}/10 \\ 3\sqrt{5}/10 \\ \sqrt{5}/10 \\ \sqrt{5}/10 \end{bmatrix}, \quad q_3 = \begin{bmatrix} -\sqrt{6}/6 \\ 0 \\ \sqrt{6}/6 \\ \sqrt{6}/3 \end{bmatrix}$$

$$Q = [q_1, q_2, q_3] = \begin{bmatrix} 1/2 & 3\sqrt{5}/10 & -\sqrt{6}/6 \\ -1/2 & 3\sqrt{5}/10 & 0 \\ -1/2 & \sqrt{5}/10 & \sqrt{6}/6 \\ 1/2 & \sqrt{5}/10 & \sqrt{6}/3 \end{bmatrix}$$

Since $A = QR$ for some upper triangular matrix R . To find R , we use the fact that Q has orthonormal columns & hence $Q^T Q = I$

Therefore, $Q^T A = Q^T QR = IR = R$.

$$R = Q^T A = \begin{bmatrix} 1/2 & -1/2 & -1/2 & 1/2 \\ 3\sqrt{5}/10 & 3\sqrt{5}/10 & \sqrt{5}/10 & \sqrt{5}/10 \\ -\sqrt{6}/6 & 0 & \sqrt{6}/6 & \sqrt{6}/3 \end{bmatrix} \begin{bmatrix} 1 & 2 & 2 \\ -1 & 1 & 2 \\ -1 & 0 & 1 \\ 1 & 1 & 2 \end{bmatrix}$$

$$= \begin{bmatrix} 2 & 1 & 1/2 \\ 0 & \sqrt{5} & 3\sqrt{5}/2 \\ 0 & 0 & \sqrt{6}/2 \end{bmatrix}$$

Singular Value Decomposition (SVD)

Defn Singular Values: If A is an $m \times n$ matrix, the singular values of A are the square roots of the eigen values of $A^T A$ and are denoted by $\sigma_1, \sigma_2, \dots, \sigma_n$. It is conventional to arrange the singular values, so that $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_n$.

Ex Find the Singular Values of $A = \begin{bmatrix} 1 & 1 \\ 1 & 0 \\ 0 & 1 \end{bmatrix}$

Solⁿ

The matrix

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

has eigen values

$$(2-\lambda)(2-\lambda)-1=0$$

$$\lambda = 3, 1.$$

$$\lambda_1 = 3, \quad \lambda_2 = 1.$$

Singular values are $\sigma_1 = \sqrt{\lambda_1} = \sqrt{3}$ and $\sigma_2 = \sqrt{\lambda_2} = 1.$

Defⁿ: SVD (Singular Value Decomposition)

Let A be an $m \times n$ matrix with singular values $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_r > 0$ and $\sigma_{r+1} = \sigma_{r+2} = \dots = \sigma_n = 0$. Then there exist an $m \times m$ orthogonal matrix U , an $n \times n$ orthogonal matrix V , and an $m \times n$ matrix

Σ of the form

$$\Sigma = \begin{bmatrix} D & 0 \\ 0 & 0 \end{bmatrix}_{m \times n}, \text{ where } D = \begin{bmatrix} \sigma_1 & & 0 \\ & \ddots & \\ & & \sigma_r \end{bmatrix}$$

Such that $A = U \Sigma V^T$, is called.

a Singular Value Decomposition (SVD) of A .

[The columns of U are called left singular vectors of A , columns of V are called right singular vectors of A , the matrices U & V are not uniquely determined by A , but Σ must contain the singular values of A .

Ex Find a Singular Value decomposition of -

(a) $A = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$.

Soln we compute

$$A^T A = \begin{bmatrix} 1 & 1 & 0 \\ 1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Eigen values are $\lambda_1 = 2, \lambda_2 = 1, \lambda_3 = 0$ &
corresponding eigen vectors $\begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}$

These vector are orthogonal, so we normalize them to obtain

$$v_1 = \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \\ 0 \end{bmatrix}$$

Since $\|v_1\| = \sqrt{1+1+0} = \sqrt{2}$
(Dividing each element by $\|v_1\|$.)

$$v_2 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}, \quad v_3 = \begin{bmatrix} -1/\sqrt{2} \\ 1/\sqrt{2} \\ 0 \end{bmatrix}$$

Singular Values of A are

$\sigma_1 = \sqrt{2}, \sigma_2 = \sqrt{1} = 1, \sigma_3 = 0$, thus,

$$V = \begin{bmatrix} 1/\sqrt{2} & 0 & -1/\sqrt{2} \\ 1/\sqrt{2} & 0 & 1/\sqrt{2} \\ 0 & 1 & 0 \end{bmatrix} \quad \& \quad \Sigma = \begin{bmatrix} \sqrt{2} & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

To find U, we compute \longrightarrow

$$u_1 = \frac{Av_1}{\sigma_1} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$u_2 = \frac{Av_2}{\sigma_2} = \frac{1}{1} \begin{bmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$\therefore U = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$, These vectors u_1 & u_2 form an orthonormal basis (std basis) for \mathbb{R}^2 .

$$\therefore \text{SVD} \rightarrow A = U \Sigma V^T$$

$$A = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \sqrt{2} & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} & 0 \\ 0 & 0 & 1 \\ -1/\sqrt{2} & 1/\sqrt{2} & 0 \end{bmatrix}$$

(Which can be easily checked,
Note that V had to be transposed.

Also, note that the singular value σ_3 does not appear in Σ .)

Ex (2) Find SVD for $A = \begin{bmatrix} 1 & 1 \\ 1 & 0 \\ 0 & 1 \end{bmatrix}$.

[Try yourself]

Ans: $A = U \Sigma V^T$.

$$\begin{bmatrix} 1 & 1 \\ 1 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 2\sqrt{6} & 0 & -1/\sqrt{3} \\ 1/\sqrt{6} & -1/\sqrt{2} & 1/\sqrt{3} \\ 1/\sqrt{6} & 1/\sqrt{2} & 1/\sqrt{3} \end{bmatrix} \begin{bmatrix} \sqrt{3} & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ -1/\sqrt{2} & 1/\sqrt{2} \end{bmatrix}$$