

$[\hat{\mathbf{k}}]_{\times}$: cross product matrix for unit vector $\hat{\mathbf{k}}$

Matrix $[\hat{\mathbf{k}}]_{\times}$ as a composition of the following transformations:

① removes the projection of $\vec{\mathbf{v}}$ onto $\hat{\mathbf{k}}$, computed as $\hat{\mathbf{k}} \hat{\mathbf{k}}^T \vec{\mathbf{v}}$

\leftrightarrow

projects $\vec{\mathbf{v}}$ onto plane orthogonal to $\hat{\mathbf{k}}$

$$(I - \hat{\mathbf{k}} \hat{\mathbf{k}}^T) \vec{\mathbf{v}}$$

equal to the projection matrix onto the plane spanned by orthonormal columns of

Q :

$$I - \hat{\mathbf{k}} \hat{\mathbf{k}}^T = Q Q^T$$

Image below shows

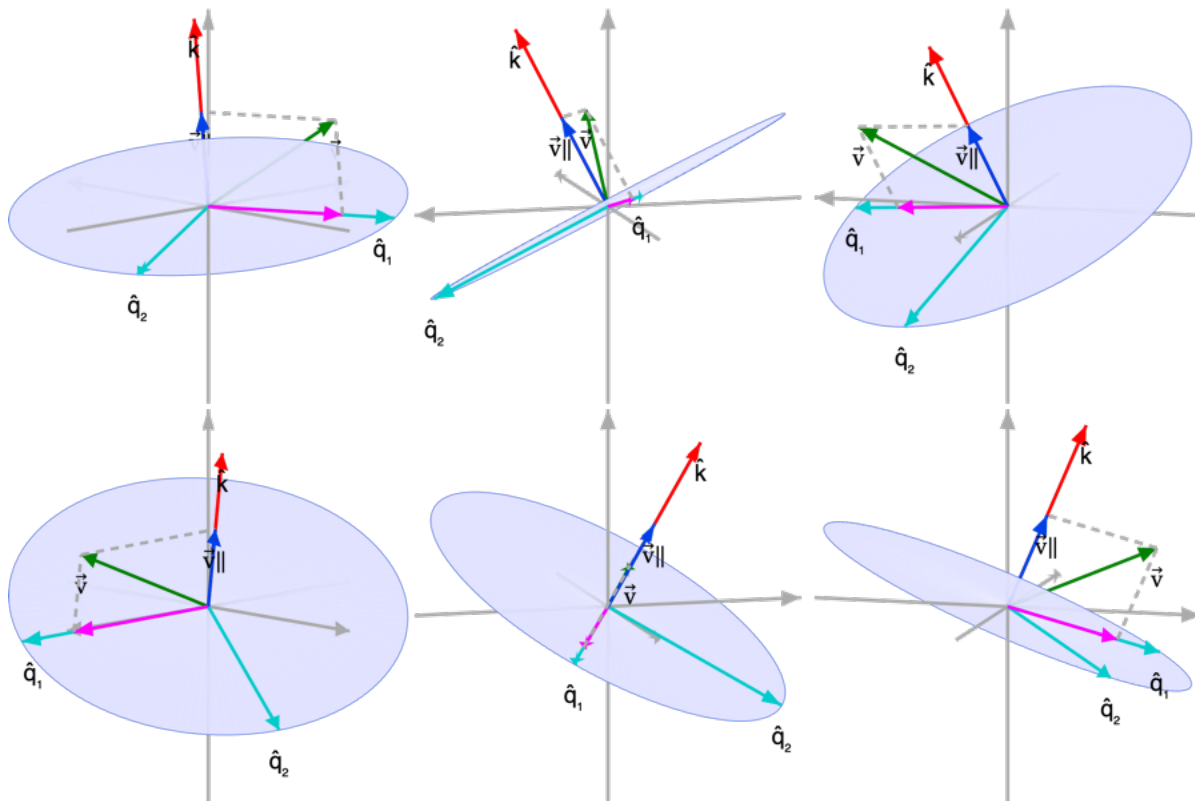
- $\hat{\mathbf{k}}$ (red)
- $\vec{\mathbf{v}}$ (green)
- $\vec{\mathbf{v}}_{\parallel} = \hat{\mathbf{k}} \hat{\mathbf{k}}^T \vec{\mathbf{v}}$ (blue)
- $\vec{\mathbf{v}}_{\perp} = (I - \hat{\mathbf{k}} \hat{\mathbf{k}}^T) \vec{\mathbf{v}}$ (magenta)

- $\hat{\mathbf{q}}_1$ (cyan), can be computed as $\frac{\vec{\mathbf{v}}_{\perp}}{|\vec{\mathbf{v}}_{\perp}|}$

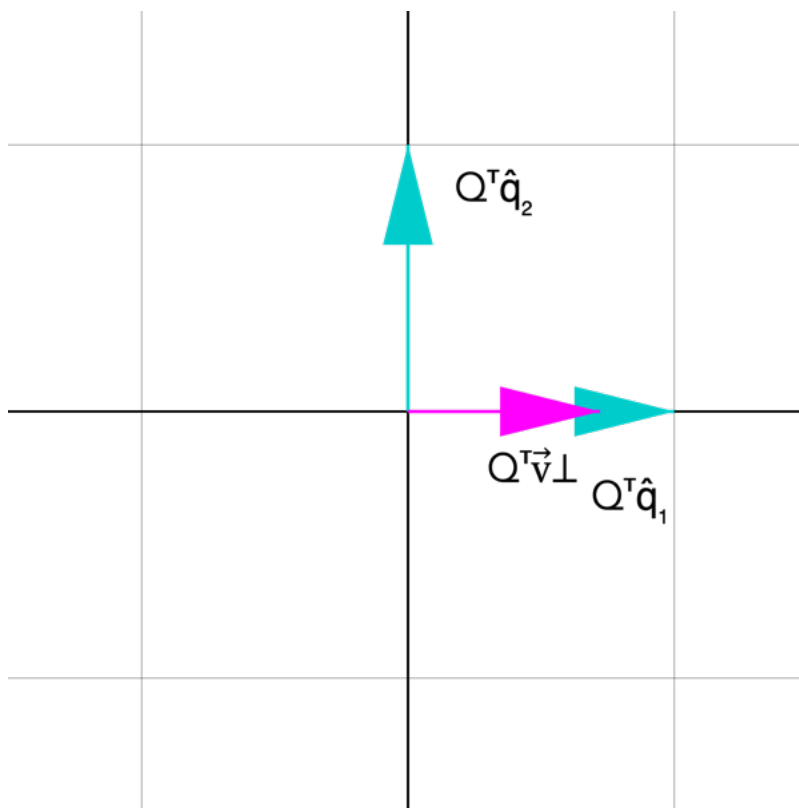
- $\hat{\mathbf{q}}_2$ (cyan), computed from system of equations $\begin{bmatrix} \hat{\mathbf{q}}_1 \\ \hat{\mathbf{k}} \end{bmatrix} \vec{\mathbf{q}} = \vec{\mathbf{0}}$

$\vec{\mathbf{q}}$ sign chosen so that $(\hat{\mathbf{q}}_1, \vec{\mathbf{q}}, \hat{\mathbf{k}})$ is positively oriented

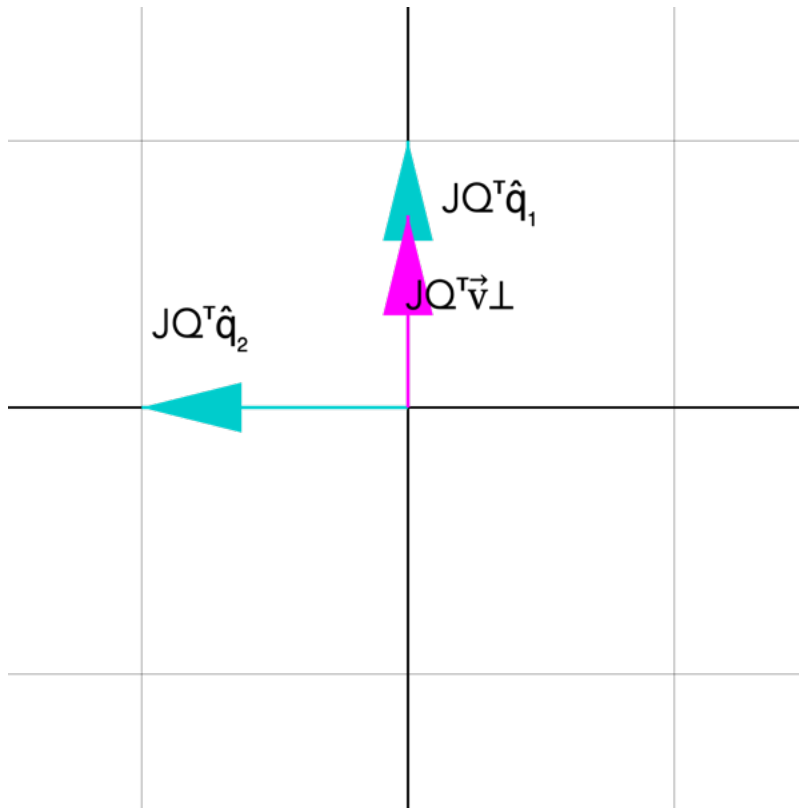
and normalized as $\hat{\mathbf{q}}_2 = \frac{\vec{\mathbf{q}}}{|\vec{\mathbf{q}}|}$



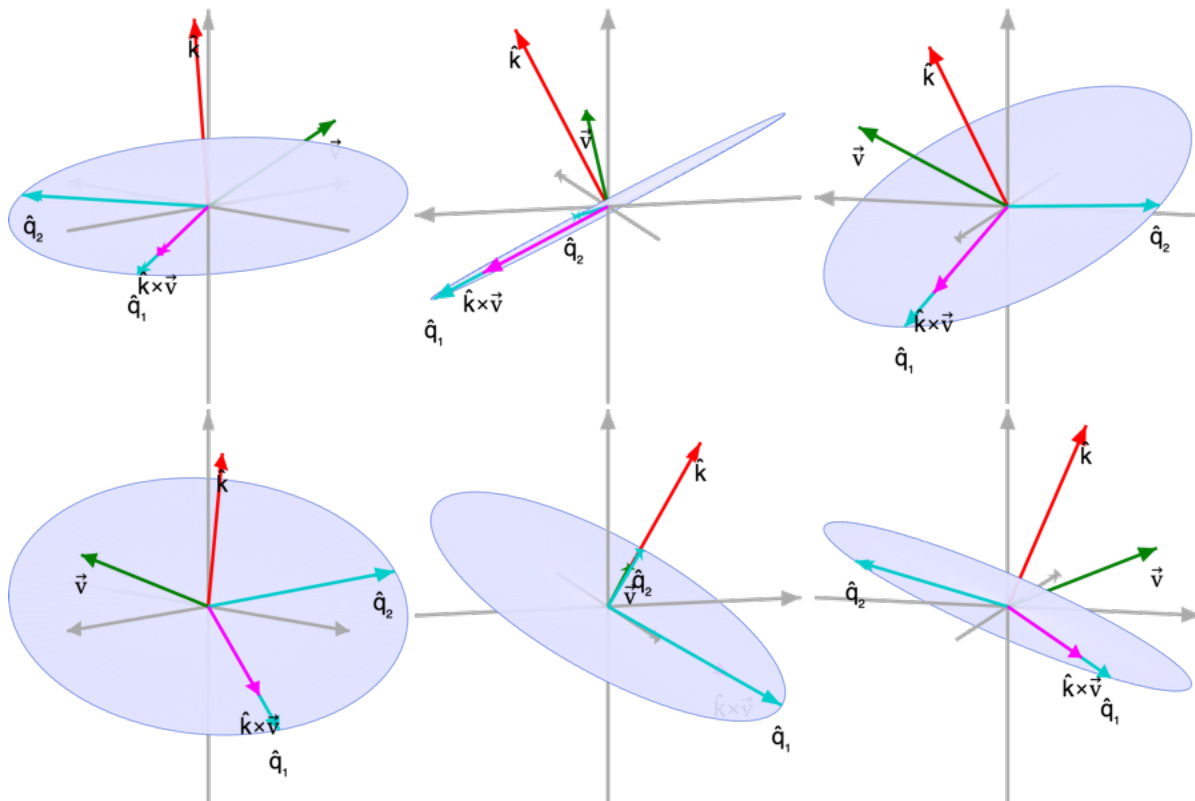
② expresses vectors in the orthogonal plane in \mathbb{R}^2 coordinates: $Q^T = \begin{bmatrix} \hat{q}_1^T \\ \hat{q}_2^T \end{bmatrix}$



③ applies the 90° rotation: $J = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$



④ lifts back to \mathbb{R}^3 : $Q = \begin{bmatrix} \hat{q}_1 & \hat{q}_2 \end{bmatrix}$



$$(I - \hat{k} \hat{k}^T) = Q Q^T$$

↓

$$[\hat{k}]_{\times} = Q J Q^T (I - \hat{k} \hat{k}^T) = Q J (Q^T Q) Q^T = Q J Q^T$$

$$Q J = \left[\hat{q}_1 \mid \hat{q}_2 \right] \left[\begin{array}{c|c} 0 & -1 \\ \hline 1 & 0 \end{array} \right] = \left[\hat{q}_2 \mid -\hat{q}_1 \right]$$

$$[\hat{k}]_{\times} = (Q J) Q^T = \left[\hat{q}_2 \mid -\hat{q}_1 \right] \begin{bmatrix} \hat{q}_1^T \\ \hat{q}_2^T \end{bmatrix} = \hat{q}_2 \hat{q}_1^T - \hat{q}_1 \hat{q}_2^T$$

The algebraic manipulations shown below confirm expected geometric actions of $[\hat{k}]_{\times}$

$$\bullet [\hat{k}]_{\times} \hat{k} = (\hat{q}_2 \hat{q}_1^T - \hat{q}_1 \hat{q}_2^T) \hat{k} =$$

$$\hat{q}_2 (\hat{q}_1^T \hat{k}) - \hat{q}_1 (\hat{q}_2^T \hat{k}) = \vec{0}$$

\leftrightarrow

maps \hat{k} to $\vec{0}$

$$\begin{aligned} \bullet [\hat{k}] \times \hat{q}_1 &= (\hat{q}_2 \hat{q}_1^T - \hat{q}_1 \hat{q}_2^T) \hat{q}_1 = \\ &\hat{q}_2 (\hat{q}_1^T \hat{q}_1) - \hat{q}_1 (\hat{q}_2^T \hat{q}_1) = \\ &\hat{q}_2 - 0 \hat{q}_1 = \hat{q}_2 \end{aligned}$$

\leftrightarrow

rotates $\hat{q}_1 \rightarrow \hat{q}_2$

$$\begin{aligned} \bullet [\hat{k}] \times \hat{q}_2 &= (\hat{q}_2 \hat{q}_1^T - \hat{q}_1 \hat{q}_2^T) \hat{q}_2 = \\ &\hat{q}_2 (\hat{q}_1^T \hat{q}_2) - \hat{q}_1 (\hat{q}_2^T \hat{q}_2) = \\ &\hat{q}_2 0 - \hat{q}_1 = -\hat{q}_1 \end{aligned}$$

\leftrightarrow

rotates $\hat{q}_2 \rightarrow -\hat{q}_1$

The above identities uniquely describe
how $[\hat{k}] \times$ transforms the 3 spanning vectors

Note the particular form of the matrix, called skew-symmetric

$$\hat{q}_2 \hat{q}_1^T - \hat{q}_1 \hat{q}_2^T = \begin{bmatrix} 0 & q_{12} \times q_{21} - q_{11} \times q_{22} & q_{12} \times q_{31} - q_{11} \times q_{32} \\ q_{22} \times q_{11} - q_{21} \times q_{12} & 0 & q_{22} \times q_{31} - q_{21} \times q_{32} \\ q_{32} \times q_{11} - q_{31} \times q_{12} & q_{32} \times q_{21} - q_{31} \times q_{22} & 0 \end{bmatrix}$$

On the next page we will derive
the unique matrix that satisfies all the geometric properties
independent of choice of \hat{q}_1 & \hat{q}_2

$$[\hat{\mathbf{k}}]_{\times} = \mathbf{Q} \mathbf{J} \mathbf{Q}^T = \begin{bmatrix} 0 & -k_3 & k_2 \\ k_3 & 0 & -k_1 \\ -k_2 & k_1 & 0 \end{bmatrix}$$



Computing entries of $[\hat{\mathbf{k}}]_{\times}$

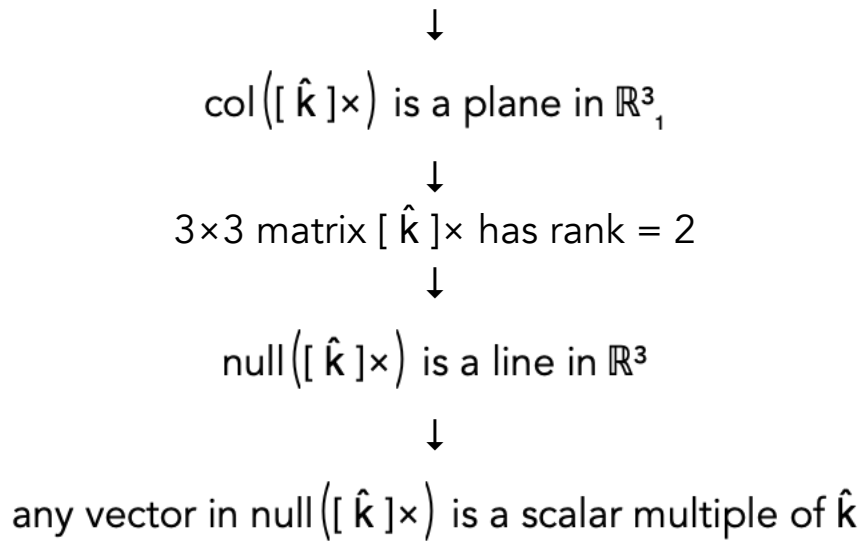
As was shown above, $[\hat{\mathbf{k}}]_{\times}$ has a skew-symmetric form:

$$\hat{\mathbf{q}}_2 \hat{\mathbf{q}}_1^T - \hat{\mathbf{q}}_1 \hat{\mathbf{q}}_2^T = \begin{bmatrix} 0 & \mathbf{q}_{12} \times \mathbf{q}_{21} - \mathbf{q}_{11} \times \mathbf{q}_{22} & \mathbf{q}_{12} \times \mathbf{q}_{31} - \mathbf{q}_{11} \times \mathbf{q}_{32} \\ \mathbf{q}_{22} \times \mathbf{q}_{11} - \mathbf{q}_{21} \times \mathbf{q}_{12} & 0 & \mathbf{q}_{22} \times \mathbf{q}_{31} - \mathbf{q}_{21} \times \mathbf{q}_{32} \\ \mathbf{q}_{32} \times \mathbf{q}_{11} - \mathbf{q}_{31} \times \mathbf{q}_{12} & \mathbf{q}_{32} \times \mathbf{q}_{21} - \mathbf{q}_{31} \times \mathbf{q}_{22} & 0 \end{bmatrix}$$

The above matrix can be re-written as $[\hat{\mathbf{k}}]_{\times} = \begin{bmatrix} 0 & -c & b \\ c & 0 & -a \\ -b & a & 0 \end{bmatrix}$

Recall geometric actions of $[\hat{\mathbf{k}}]_{\times}$:

- maps $\hat{\mathbf{k}}$ to $\vec{0}$
- rotates $\hat{\mathbf{q}}_1 \rightarrow \hat{\mathbf{q}}_2$
- rotates $\hat{\mathbf{q}}_2 \rightarrow -\hat{\mathbf{q}}_1$



From algebraic properties of skew-symmetric $([\hat{\mathbf{k}}]_{\times})$, we can infer that

$$\text{null}([\hat{\mathbf{k}}]_{\times}) \text{ is spanned by a vector } \vec{\mathbf{a}} = \begin{bmatrix} a \\ b \\ c \end{bmatrix}$$

$$\begin{array}{c} \downarrow \\ \text{same 1-dimensional space is spanned by } \hat{\mathbf{k}} \text{ \& } \vec{\mathbf{a}} \\ \downarrow \end{array}$$

$$\begin{bmatrix} a \\ b \\ c \end{bmatrix} = (d \hat{\mathbf{k}}) \text{ where } d \text{ is a scalar}$$

Next, we need to find the unique d that satisfies

- $([\hat{\mathbf{k}}]_{\times}) \hat{\mathbf{q}}_1 = \hat{\mathbf{q}}_2$
- $([\hat{\mathbf{k}}]_{\times}) \hat{\mathbf{q}}_2 = -\hat{\mathbf{q}}_1$

We will consider a simple case of \mathbb{R}^3 where

- $\vec{e}_3 = \hat{k}$
- $\vec{e}_1 = \hat{q}_1$
- $\vec{e}_2 = \hat{q}_2$

In this case,

- $d \left([\hat{k}]_{\times} \right) \vec{e}_1 = \vec{e}_2$

- $d \left([\hat{k}]_{\times} \right) \vec{e}_1 = d \begin{bmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$

$d = 1$ is the only scalar that satisfies the above

↓

$$[\hat{k}]_{\times} = Q J Q^T = \begin{bmatrix} 0 & -k_3 & k_2 \\ k_3 & 0 & -k_1 \\ -k_2 & k_1 & 0 \end{bmatrix}$$

Summary:

① geometry provided us the axis is the null space
 algebra indicated that the null space is generated by vector $\vec{a} = [a \ b \ c]^T$
 both null spaces are 1-dimensional

↓

they are the same object:

$$\text{null}([\hat{k}]_{\times}) = \text{span}(\hat{k}) = \text{span}(\vec{a})$$



$$\vec{a} = d \hat{k} \text{ for some scalar } d$$

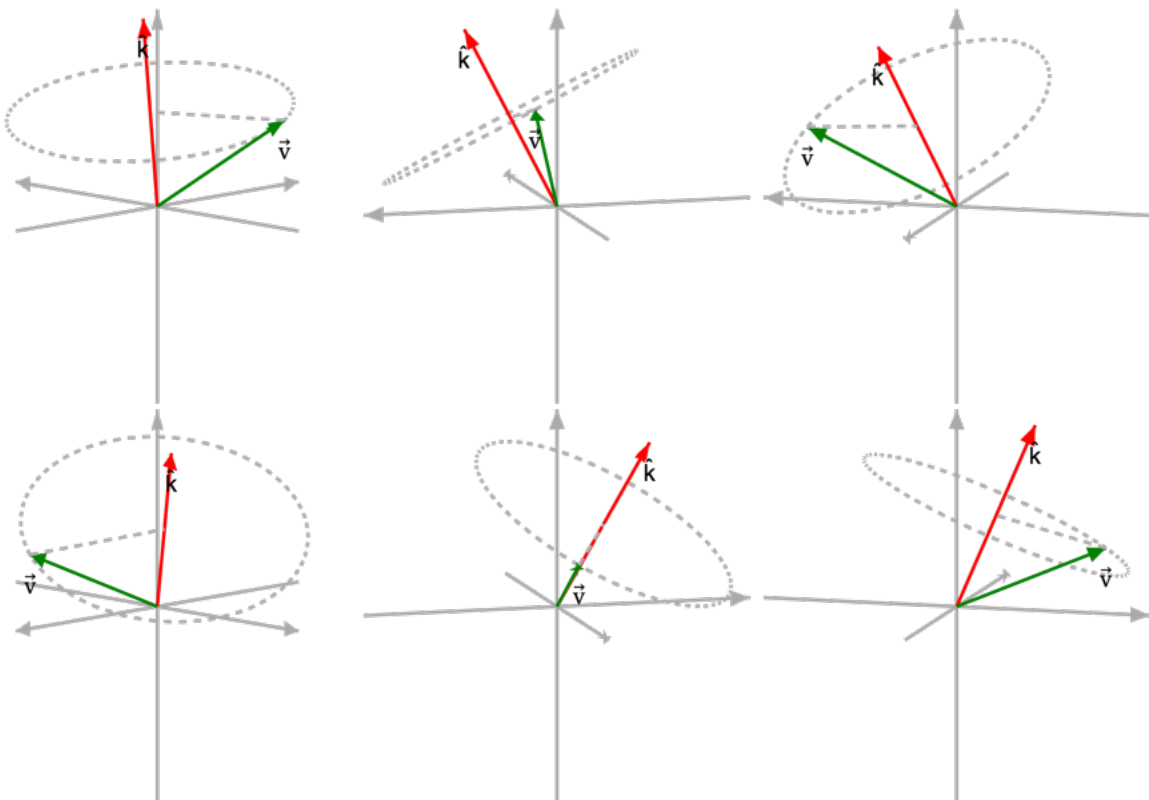
- ② we used another geometric property, considered the simplest case and obtained $d = 1$
- ③ even though the derivation employed non-unique matrix $Q = [\hat{q}_1 \mid \hat{q}_2]$, $[\hat{k}]^\times = Q J Q^T$ is independent of Q



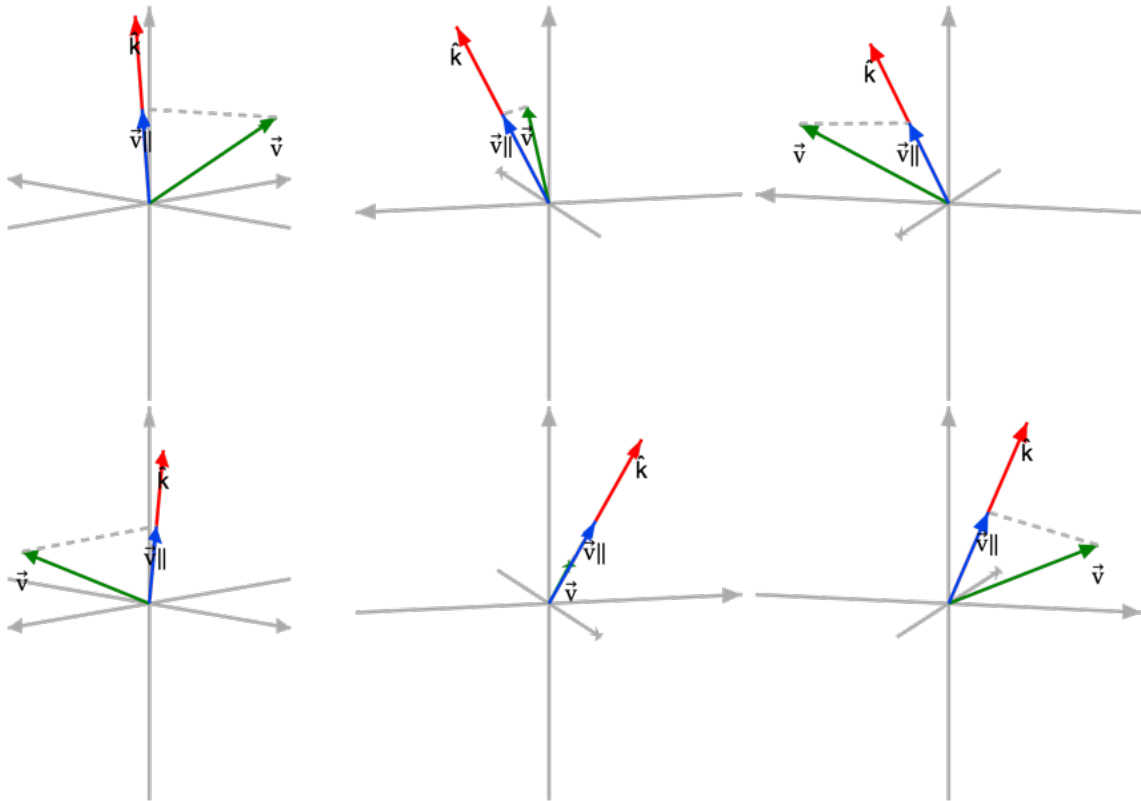
Rotation around a vector (Rodrigues algorithm)

Problem stated:

rotate vector \vec{v} (green) around unit vector \hat{k} (red) by angle θ where \vec{v} & \hat{k} are non-collinear



- ① Calculate projection of \vec{v} onto \hat{k} as
 $\vec{v}_{||} = \hat{k} \hat{k}^T \vec{v}$ (blue)



- ② Define a unit vector \hat{q}_1 (cyan) so that $\hat{q}_1 \cdot \hat{k} = 0$

while there is not a unique choice for \hat{q}_1 ,
 one particular instance can be computed by normalizing $\vec{v}_{\perp} = \vec{v} - \vec{v}_{||}$

$$\hat{q}_1 = \frac{\vec{v}_{\perp}}{|\vec{v}_{\perp}|}$$

- ③ Define another unit vector \hat{q}_2 (cyan) such that

- $\hat{q}_2 \cdot \hat{q}_1 = 0$
- $\hat{q}_2 \cdot \hat{k} = 0$

• equivalently $\begin{bmatrix} \hat{k}^T \\ \hat{q}_1^T \end{bmatrix} \vec{x} = \vec{0}$

\hat{k} and \hat{q}_1 are orthogonal (linearly independent) vectors in \mathbb{R}^3



solution space is one-dimensional

- choose any non-zero \vec{x} so that $(\hat{q}_1, \vec{x}, \hat{k})$ is positively oriented

- normalize as $\hat{q}_2 = \frac{\vec{x}}{|\vec{x}|}$



once \hat{q}_1 is chosen, \hat{q}_2 is uniquely determined up to the sign

(alternatively, \hat{q}_2 can be computed as cross product: $\hat{q}_2 = \hat{k} \times \hat{q}_1$)

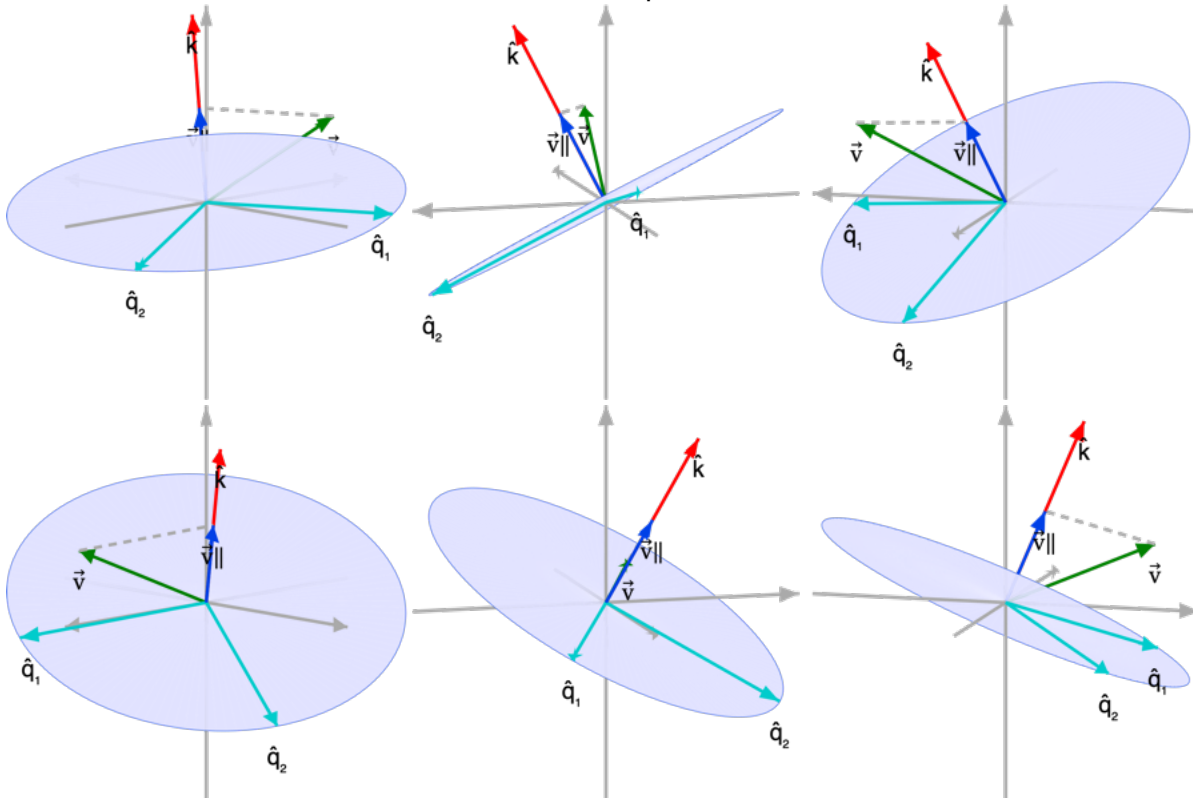
④ \hat{q}_1 & \hat{q}_2 now span the plane orthogonal to \hat{k}

and we have $Q = [\hat{q}_1 \mid \hat{q}_2]$

Q is orthonormal



$(Q Q^T)$ is the matrix of projection onto Q



⑤ \vec{v}_\perp (magenta) is the projection of \vec{v} onto the plane

$$\bullet \vec{v}_\perp = \vec{v} - \vec{v}_\parallel$$

↓

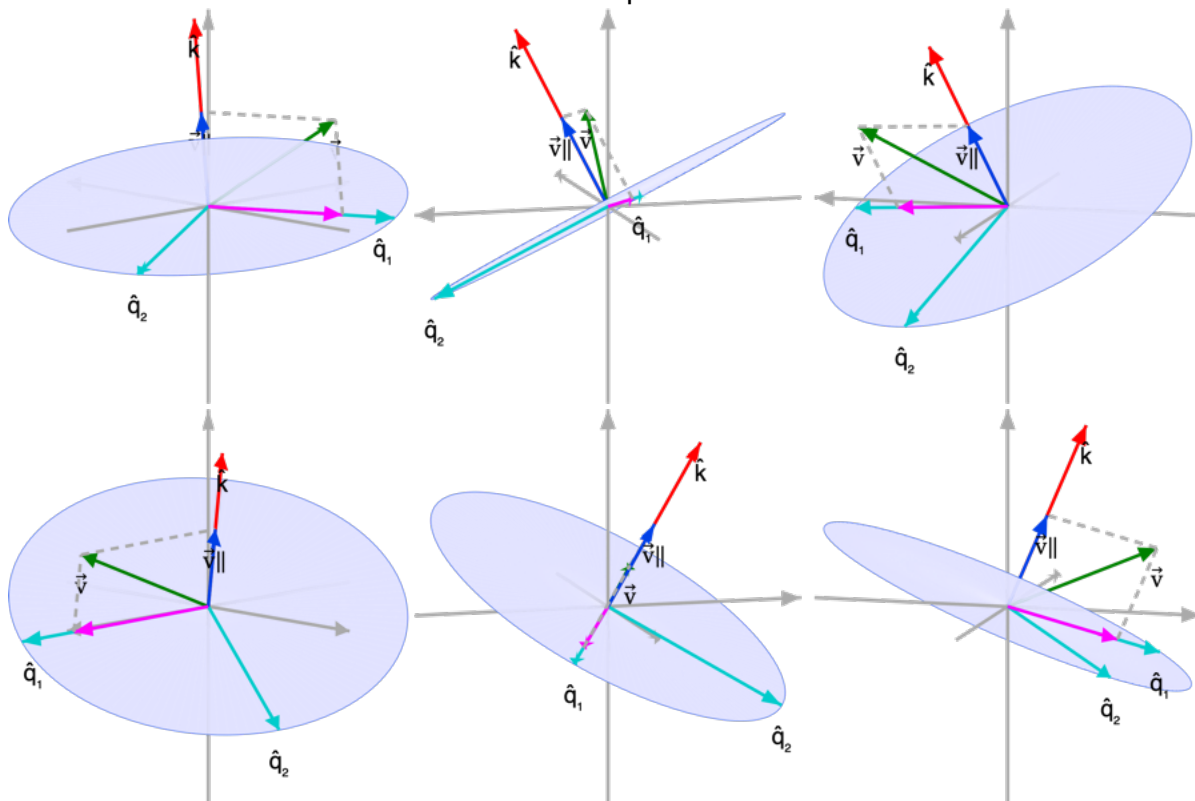
$$\vec{v}_\perp = (I - \hat{k} \hat{k}^T) \vec{v}$$

$$\text{Also, } \vec{v}_\perp = Q Q^T \vec{v}$$

As \vec{v} rotates around \hat{k} ,

• \vec{v}_\parallel stays the same

• \vec{v}_\perp rotates inside the plane to become \vec{w}_\perp



⑥ Represent the plane via 2D coordinates as

multiplication by Q^T ($\mathbb{R}^3 \rightarrow \mathbb{R}^2$):

$$\bullet Q^T \hat{q}_1 = \begin{bmatrix} \hat{q}_1^T \\ \hat{q}_2^T \end{bmatrix} \hat{q}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \hat{e}_1(2D)$$

$$\bullet Q^T \hat{q}_2 = \begin{bmatrix} \hat{q}_1^T \\ \hat{q}_2^T \end{bmatrix} \hat{q}_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \hat{e}_2(2D)$$

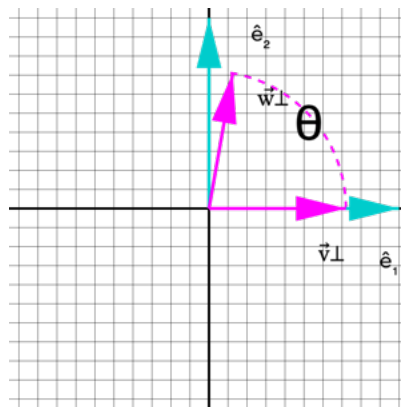
$$\bullet \vec{v}_\perp(3D) = s \hat{q}_1 \text{ (s is a scalar)}$$



$$\bullet \vec{v}_\perp(2D) = s \hat{e}_1$$

⑦ $\vec{v}_\perp(2D)$ rotated inside the plane by θ becomes $\vec{w}_\perp(2D)$

$$\vec{w}_\perp(2D) = R(\theta) (\vec{v}_\perp(2D)) = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix} (\vec{v}_\perp(2D))$$



⑧ Represent \vec{w}_\perp in 3D coordinates

$$Q (\mathbb{R}^2 \rightarrow \mathbb{R}^3)$$

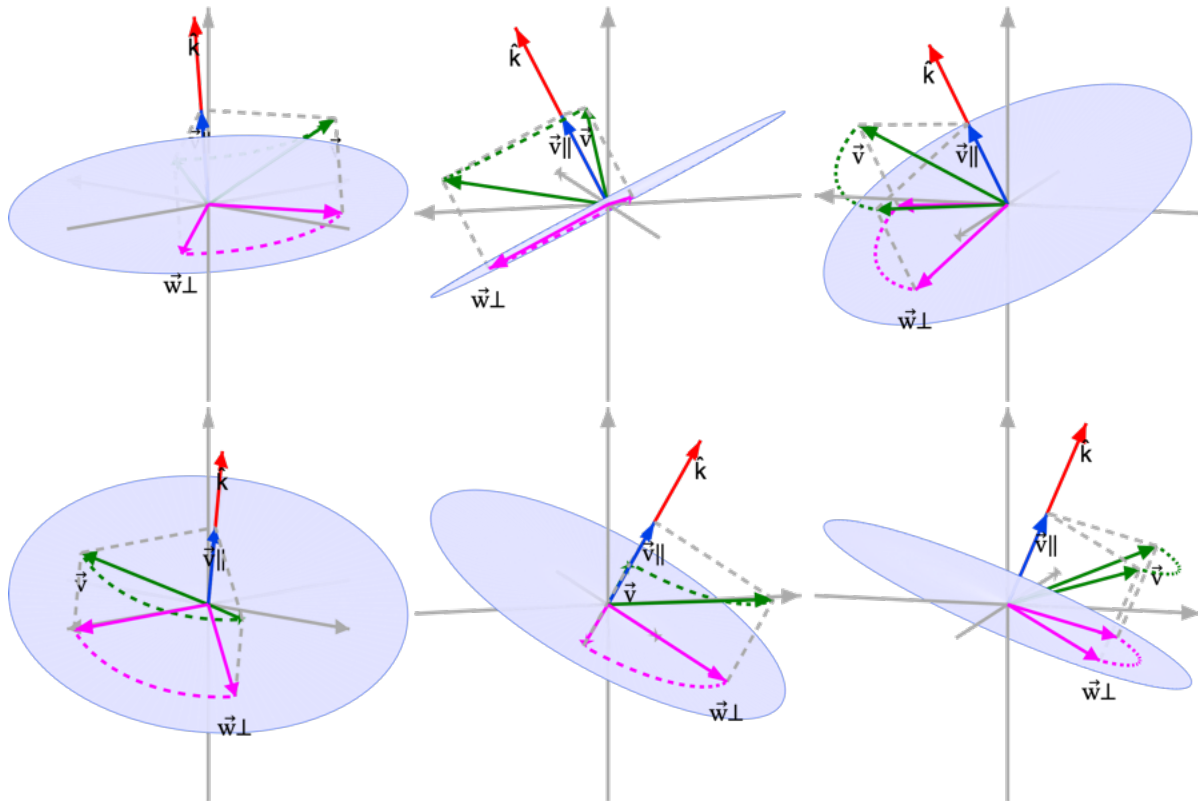
(from plane to 3D coordinates)

$$\bullet Q \hat{e}_1 = \hat{q}_1$$

$$\bullet Q \hat{e}_2 = \hat{q}_2$$

$$\bullet Q (\vec{w}_\perp \in \mathbb{R}^2) \rightarrow (\vec{w}_\perp \in \mathbb{R}^3)$$

⑨ Reconstruct $\vec{w} = \vec{w}_\perp + \vec{v}_\parallel$ (green)



⑩ Combining these steps gives the following

$$\vec{w} = \hat{k} \hat{k}^T \vec{v} + Q R(\theta) Q^T (I - \hat{k} \hat{k}^T) \vec{v}$$

- $\hat{k} \hat{k}^T \vec{v}$ is the unchanged projection onto \hat{k}
- $(I - \hat{k} \hat{k}^T) \vec{v}$ is the component of \vec{v} inside the plane or \vec{v}_\perp
 - $Q R(\theta) Q^T (I - \hat{k} \hat{k}^T) =$
 - $Q R(\theta) (Q^T Q) Q^T = Q R(\theta) Q^T$

is the sequence of transformations applied to \vec{v}_\perp

While this geometric construction is sufficient to compute the rotated vector \hat{w} , the standard closed-form expression

$$\hat{w} = A \vec{v} = [\cos\theta I + (1 - \cos\theta) \hat{k} \hat{k}^T + \sin\theta [\hat{k}]_\times] \vec{v}$$

will be derived on the next page

as a compact representation of the same operations, useful for algebraic manipulation and computational efficiency



Matrix A derived from above algorithm

On the previous page, we showed that

$$\vec{w} = \vec{v}_{\parallel} + Q R(\theta) Q^T \vec{v}_{\perp} = \hat{k} \hat{k}^T \vec{v} + Q R(\theta) Q^T (I - \hat{k} \hat{k}^T) \vec{v}$$

- axis of rotation: unit vector $\hat{k} = [k_1, k_2, k_3]^T$
- \vec{v} is a vector we want to rotate around \hat{k}
- $\hat{k} \hat{k}^T \vec{v}$ is the projection of \vec{v} onto $\text{span}(\hat{k})$
- $Q = [\hat{q}_1 | \hat{q}_2]$ is a 3×2 matrix with orthonormal columns such that

$$\hat{q}_1 \cdot \hat{k} = 0, \hat{q}_2 \cdot \hat{k} = 0$$

- $Q Q^T$ is the matrix of projection onto $\text{col}(Q)$
- $(I - \hat{k} \hat{k}^T) \vec{v}$ is also projection of \vec{v} onto $\text{col}(Q)$

↓

$$Q R(\theta) Q^T (I - \hat{k} \hat{k}^T) =$$

$$Q R(\theta) (Q^T Q) Q^T =$$

$$Q R(\theta) Q^T$$

↓

$$\vec{w} = \hat{k} \hat{k}^T \vec{v} + Q R(\theta) Q^T \vec{v}$$

$$\bullet R(\theta) = \left[\begin{array}{c|c} \cos\theta & -\sin\theta \\ \hline \sin\theta & \cos\theta \end{array} \right]$$

$$Q R(\theta) Q^T = \left[\hat{q}_1 | \hat{q}_2 \right] \left[\begin{array}{c|c} \cos\theta & -\sin\theta \\ \hline \sin\theta & \cos\theta \end{array} \right] \left[\begin{array}{c} \hat{q}_1 \\ \hat{q}_2 \end{array} \right] =$$

$$\left[\hat{\mathbf{q}}_1 \mid \hat{\mathbf{q}}_2 \right] \begin{bmatrix} \cos\theta \hat{\mathbf{q}}_1 - \sin\theta \hat{\mathbf{q}}_2 \\ \sin\theta \hat{\mathbf{q}}_1 + \cos\theta \hat{\mathbf{q}}_2 \end{bmatrix} =$$

$$\hat{\mathbf{q}}_1 (\cos\theta \hat{\mathbf{q}}_1 - \sin\theta \hat{\mathbf{q}}_2) + \hat{\mathbf{q}}_2 (\sin\theta \hat{\mathbf{q}}_1 + \cos\theta \hat{\mathbf{q}}_2) = \\ \cos\theta (\hat{\mathbf{q}}_1 \hat{\mathbf{q}}_1^T + \hat{\mathbf{q}}_2 \hat{\mathbf{q}}_2^T) + \sin\theta (\hat{\mathbf{q}}_2 \hat{\mathbf{q}}_1^T - \hat{\mathbf{q}}_1 \hat{\mathbf{q}}_2^T)$$

Sum of outer products

$$\hat{\mathbf{q}}_1 \hat{\mathbf{q}}_1^T + \hat{\mathbf{q}}_2 \hat{\mathbf{q}}_2^T = \mathbf{Q} \mathbf{Q}^T$$

↓

$$\mathbf{Q} \mathbf{R}(\theta) \mathbf{Q}^T = \cos\theta \mathbf{Q} \mathbf{Q}^T + \sin\theta (\hat{\mathbf{q}}_2 \hat{\mathbf{q}}_1^T - \hat{\mathbf{q}}_1 \hat{\mathbf{q}}_2^T)$$

↓

$$\vec{\mathbf{w}} = \hat{\mathbf{k}} \hat{\mathbf{k}}^T \vec{\mathbf{v}} + \cos\theta (\mathbf{I} - \hat{\mathbf{k}} \hat{\mathbf{k}}^T) \vec{\mathbf{v}} + \sin\theta (\hat{\mathbf{q}}_2 \hat{\mathbf{q}}_1^T - \hat{\mathbf{q}}_1 \hat{\mathbf{q}}_2^T) \vec{\mathbf{v}}$$

So the full 3×3 rotation matrix is

$$\mathbf{A} = \hat{\mathbf{k}} \hat{\mathbf{k}}^T + \cos\theta (\mathbf{I} - \hat{\mathbf{k}} \hat{\mathbf{k}}^T) + \sin\theta (\hat{\mathbf{q}}_2 \hat{\mathbf{q}}_1^T - \hat{\mathbf{q}}_1 \hat{\mathbf{q}}_2^T) = \\ \cos\theta \mathbf{I} + (1 - \cos\theta) \hat{\mathbf{k}} \hat{\mathbf{k}}^T + \sin\theta (\hat{\mathbf{q}}_2 \hat{\mathbf{q}}_1^T - \hat{\mathbf{q}}_1 \hat{\mathbf{q}}_2^T)$$

In the previous page, we showed that

$\hat{\mathbf{q}}_2 \hat{\mathbf{q}}_1^T - \hat{\mathbf{q}}_1 \hat{\mathbf{q}}_2^T$ is the cross product matrix of $\hat{\mathbf{k}}$

$$[\hat{\mathbf{k}}]_{\times} = \begin{bmatrix} 0 & -k_3 & k_2 \\ k_3 & 0 & -k_1 \\ -k_2 & k_1 & 0 \end{bmatrix}$$

and

$$A = \cos\theta I + (1 - \cos\theta) \hat{k} \hat{k}^T + \sin\theta \begin{bmatrix} 0 & -k_3 & k_2 \\ k_3 & 0 & -k_1 \\ -k_2 & k_1 & 0 \end{bmatrix}$$

On the previous page, we demonstrated an algorithm that uses normalized \vec{v}_\perp to compute one instance of \hat{q} . However, the derived matrix A is independent of choice of \vec{v} .

Our tutorial used the following sequence of derivations:

- First, the cross product matrix $[\hat{k}]_\times$ was identified from the action of the particular 90° rotation in \hat{k}_\perp and the special structure of $J = \text{Rot}(90^\circ)$ made that identification possible
- Next, the rotation algorithm around \hat{k} was built geometrically from projection onto \hat{k} , rotation in \hat{k}_\perp and lifting back through the orthonormal basis $Q = [\hat{q}_1 \mid \hat{q}_2]$
 - Finally, expanding that algorithm produced Rodrigues' formula, with $[\hat{k}]_\times$ appearing as the skew-symmetric term in the final matrix expression



Eigenvectors of A : matrix of rotation around \hat{k}

① Geometrically, \hat{k} is not affected by A

We confirm with algebra:

$$A \hat{k} =$$

$$\left(\cos\theta I + (1 - \cos\theta) \hat{k} \hat{k}^T + \sin\theta (\hat{q}_2 \hat{q}_1^T - \hat{q}_1 \hat{q}_2^T) \right) \hat{k} =$$

$$\begin{aligned} \cos\theta \hat{k} + (1 - \cos\theta) \hat{k} (\hat{k}^T \hat{k}) + \sin\theta (\hat{q}_2 \hat{q}_1^T - \hat{q}_1 \hat{q}_2^T) \hat{k} = \\ \cos\theta \hat{k} + (1 - \cos\theta) \hat{k} + \sin\theta (\hat{q}_2 (\hat{q}_1^T \hat{k}) - \hat{q}_1 (\hat{q}_2^T \hat{k})) = \hat{k} \end{aligned}$$

$$A \hat{k} = \hat{k}$$

↓

\hat{k} is an eigenvector of A with $\lambda = 1$

② The remaining action of A is a rotation inside the plane orthogonal to \hat{k}

Eigenvalue-eigenvector pairs of 2D rotation matrix $R = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix}$

λ_{2D}	\vec{x}
$e^{i\theta}$	$[1, -i]^T$
$e^{-i\theta}$	$[1, i]^T$

Previously, we showed that

$$A = \hat{k} \hat{k}^T + Q R Q^T$$

columns of Q are orthonormal and orthogonal to \hat{k}

↓

$$\hat{k}^T Q = \vec{0} \text{ and } Q^T Q = I$$

Next, we show that

if \vec{x} is an eigenvector of R , then $Q \vec{x}$ is an eigenvector of A :

$$\begin{aligned} A (Q \vec{x}) &= \hat{k} \hat{k}^T Q \vec{x} + Q R Q^T Q \vec{x} = \\ &= \hat{k} (\hat{k}^T Q) \vec{x} + Q R (Q^T Q) \vec{x} = \end{aligned}$$

$$\vec{0} + Q R \vec{x}$$

$$\text{If } R \vec{x} = \lambda \vec{x}$$

$$A(Q \vec{x}) = Q(R \vec{x}) = Q(\lambda \vec{x}) = \lambda(Q \vec{x})$$

↓

- $Q \vec{x}$ is an eigenvector of A
- $\lambda(3D) = \lambda(2D)$

$$\bullet [\hat{q}_1 \mid \hat{q}_2] \begin{bmatrix} 1 \\ -i \end{bmatrix} = \hat{q}_1 - i \hat{q}_2$$

$$\bullet [\hat{q}_1 \mid \hat{q}_2] \begin{bmatrix} 1 \\ i \end{bmatrix} = \hat{q}_1 + i \hat{q}_2$$

Eigenvalue-eigenvector pairs of A

λ_{3D}	$\vec{y} = Q \vec{x}$
1	\hat{k}
$e^{i\theta}$	$\hat{q}_1 - i \hat{q}_2$ (infinitely many depending on choice of \hat{q}_1 & \hat{q}_2)
$e^{-i\theta}$	$\hat{q}_1 + i \hat{q}_2$ (infinitely many depending on choice of \hat{q}_1 & \hat{q}_2)



$$\vec{v}_{\parallel} = \hat{k} \hat{k}^T \vec{v}$$

$$\vec{v}_{\perp} = \vec{v} - \vec{v}_{\parallel}$$

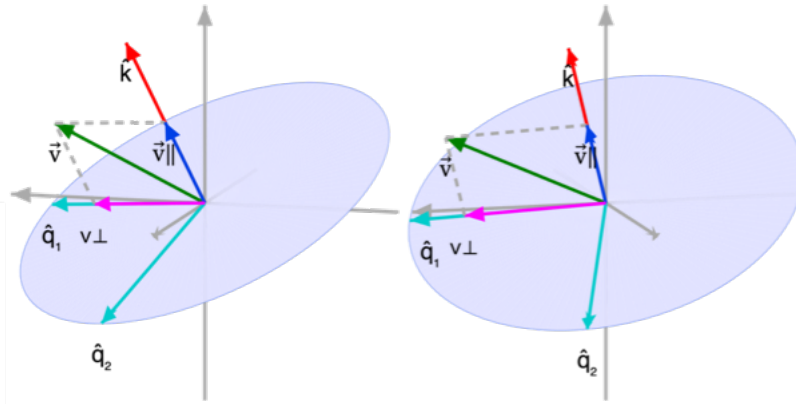
$$\hat{q}_1 = \frac{\vec{v}_{\perp}}{|\vec{v}_{\perp}|}$$

$$\hat{q}_2 = \hat{k} \times \hat{q}_1 \text{ (cross product of unit vectors)}$$

$$Q = [\hat{q}_1 \mid \hat{q}_2]$$

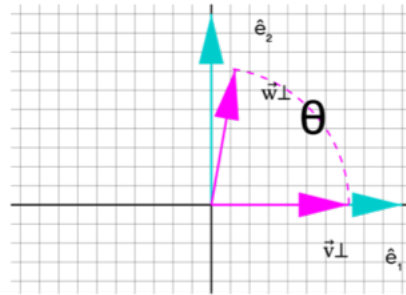
$$\hat{\vec{w}} = \hat{\vec{v}}_{\parallel} + \cos\theta \hat{\vec{v}}_{\perp} + \sin\theta (\hat{\vec{k}} \times \hat{\vec{v}}_{\perp})$$

$$\hat{\vec{w}} = \hat{\vec{v}}_{\parallel} + Q R(\theta) Q^T \hat{\vec{v}}$$



$Q^T (\mathbb{R}^3 \rightarrow \mathbb{R}^2)$
 (plane coordinates)
 $Q^T (\vec{v}_{\perp} \in \mathbb{R}^3) \rightarrow (\vec{v}_{\perp} \in \mathbb{R}^2)$
 $Q^T \hat{q}_1 = \hat{e}_1$
 $Q^T \hat{q}_2 = \hat{e}_2$

$Q^T = [\hat{q}_1^T \mid \hat{q}_2^T]$
 $\hat{q}_1 \cdot \hat{q}_2 = 0$
 \downarrow
 Q^T preserves length & angles



$$(\vec{w}_{\perp} \in \mathbb{R}^2) = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix} (\vec{v}_{\perp} \in \mathbb{R}^2)$$

$Q (\mathbb{R}^2 \rightarrow \mathbb{R}^3)$
 (from plane coordinates)
 $Q (\vec{w}_{\perp} \in \mathbb{R}^2) \rightarrow (\vec{w}_{\perp} \in \mathbb{R}^3)$
 $Q \hat{e}_1 = \hat{q}_1$
 $Q \hat{e}_2 = \hat{q}_2$

$Q = [\hat{q}_1 \mid \hat{q}_2]$
 $\hat{q}_1 \cdot \hat{q}_2 = 0$
 \downarrow
 Q preserves length & angles

