TPajdla. Elements of Geometry for Computer Vision and Robotics 2020-10-4 (pajdla@cvut.cz)

§ 1 $\lambda_1 = \lambda_2 = \lambda_3 = 1$: Let $\lambda_1 = \lambda_2 = \lambda_3 = 1$. Then $p(\lambda) = (\lambda - 1)^3 = 1$

 $\lambda^3 - 3\lambda^2 + 3\lambda - 1$. It means that $r_{11} + r_{22} + r_{33} = 3$ and since $r_{11} \le 1$,

 $r_{22} \le 1$, $r_{33} \le 1$, it leads to $r_{11} = r_{22} = r_{33} = 1$, which implies R = I. Then

I - R = 0 and all non-zero vectors of \mathbb{R}^3 are eigenvectors of R. Notice that

• Next, consider $\lambda_1 = 1$ and $\lambda_2 = \lambda_3 = -1$. The eigenvectors \vec{v} corre-

There is always at least one one-dimensional space of such vectors. We

which means that there is a one-dimensional space of real eigenvectors

corresponding to 1 and a two-dimensional space of real eigenvectors cor-

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responding to -1. Notice that rank of R - I is two here.

2.1.3 Eigenvectors of \mathbb{R} . $\mathbb{R} \in \mathbb{R}^{3\times 3}$ $\mathbb{R}^{T}\mathbb{R} = \mathbb{I}$ def $\mathbb{R} = 1$

2.1.3 Eigenvectors of R.
$$\emptyset \in \mathbb{R}^{3 \times 5}$$
 $\mathbb{R}^{7} \mathbb{R} = \mathbb{I}$ $\mathbb{R}^{3 \times 5} \mathbb{R}^{7} \mathbb{R}^{7}$

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

sponding to $\lambda_2 = \lambda_3 = -1$ are solutions to

also see that there is a rotation matrix

rank of R - I is zero in this case.

(2.17)

 $= \lambda^3 - (r_{11} + r_{22} + r_{33}) \lambda^2$

 $= \lambda^3 - \operatorname{trace} R(\lambda^2 - \lambda) - 1$

 $= (\lambda - 1)(\lambda^2 + (1 - \operatorname{trace} R)\lambda + 1)$

 $+(r_{11}\,r_{22}-r_{21}\,r_{12}+r_{11}\,r_{33}-r_{31}\,r_{13}+r_{22}\,r_{33}-r_{23}\,r_{32})\,\lambda$

 $= \lambda^3 - (r_{11} + r_{22} + r_{33}) \lambda^2 + (r_{33} + r_{22} + r_{11}) \lambda - |\mathbf{R}|$

 $+r_{11}(r_{23}r_{32}-r_{22}r_{33})-r_{21}(r_{32}r_{13}-r_{12}r_{33})+r_{31}(r_{13}r_{22}-r_{12}r_{23})$

N ∈ R3 \ {0}}

Of 1R3 from the ingentees.

(2.12)

(2.13)

(2.14)

(2.15)

(2.16)

T Pajdla. Elements of Geometry for Computer Vision and Robotics 2020-10-4 (pajdla@cvut.cz) $\S 2 \mid \lambda_1 = 1, \lambda_2 = \lambda_3 = -1$: How does the situation look for a general R

with eigenvalues 1, -1, -1? Consider an eigenvector \vec{v}_1 corresponding to 1 and an eigenvector \vec{v}_2 corresponding to -1. They are linearly independent. Otherwise there has to be $s \in \mathbb{R}$ such that $\vec{v}_2 = s \vec{v}_1 \neq 0$ and then

ney are linearly independent.
$$= s \vec{v}_1 \neq 0 \text{ and then}$$
(2.20)
(2.21)
(2.22)

$$\vec{v}_2 = s \vec{v}_1 \tag{2.20}$$

$$R \vec{v}_2 = s R \vec{v}_1 \tag{2.21}$$

$$-\vec{v}_2 = s \vec{v}_1 \tag{2.22}$$
 leading to $s = -s$ and therefore $s = 0$ which contradicts $\vec{v}_2 \neq 0$. Now, let us look at vectors $\vec{v}_3 \in \mathbb{R}^3$ defined by

leading to
$$\underline{s} = -\underline{s}$$
 and therefore $\underline{s} = 0$ which contradicts $\vec{v}_2 \neq 0$. Now, let us look at vectors $\vec{v}_3 \in \mathbb{R}^3$ defined by
$$\begin{bmatrix} \vec{v}_1^\top \\ \vec{v}_2^\top \end{bmatrix} \vec{v}_3 = 0$$
 (2.23)
The above linear system has a one-dimensional space of solutions since

Then write: $\begin{bmatrix} \vec{v}_1^{\top} \\ \vec{v}_2^{\dagger} \end{bmatrix} \mathbf{R}^{\top} \vec{v}_3 = \begin{bmatrix} \vec{v}_1^{\top} \mathbf{R}^{\top} \\ \vec{v}_2^{\dagger} \mathbf{R}^{\top} \end{bmatrix} \vec{v}_3 = \begin{bmatrix} \vec{v}_1^{\top} \\ -\vec{v}_2^{\dagger} \end{bmatrix} \underline{\vec{v}_3} = 0$

We see that
$$R^{\top}\vec{v}_3$$
 and \vec{v}_3 are in the same one-dimensional space, i.e. they are linearly dependent and we can write

the rows of its matrix are independent. Chose a fixed solution $\vec{v}_3 \neq 0$.

 $\mathbf{R}^{\top}\vec{v}_3 = s\,\vec{v}_3 \quad \neg$ (2.25)

for some non-zero
$$s \in \mathbb{C}$$
. Multiplying equation 2.25 by R from the left and dividing both sides by s gives

 $\frac{1}{s}\vec{v}_3 = R\vec{v}_3$ $= R\vec{v}_3$ $= \text{liquid} \quad of (2.26)$ Clearly, \vec{v}_3 is an eigenvector of R. Since it is not a multiple of \vec{v}_1 , it must correspond to eigenvalue -1. Moreover, $\vec{v}_2^{\top} \vec{v}_3 = 0$ and hence they are

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 $\frac{\text{1 dent.}}{0} \quad \overrightarrow{0} \neq \overrightarrow{N}_{1} \sim 1$ $(2.20) \quad \overrightarrow{0} \neq \overrightarrow{N}_{2} \sim -1$ $\downarrow \underline{1}$ $R_{N_{1}} = N_{1} \Rightarrow N_{1}^{T} R_{2}^{T} = N_{1}^{T}$ $R\vec{N}_{1} = -\vec{N}_{2} \rightarrow N_{2}^{T}R^{T} = -N_{2}^{T}$

$$\frac{\vec{v}_1 \vec{v}_2}{\vec{v}_1 \vec{v}_2} LI$$

$$\frac{\vec{v}_1 \vec{v}_2}{\vec{v}_1 \vec{v}_2} \frac{\vec{v}_2 \vec{v}_3}{\vec{v}_1 \vec{v}_2}$$

$$\frac{\vec{v}_1 \vec{v}_2}{\vec{v}_1 \vec{v}_2} = \left(\frac{\vec{v}_1 \vec{v}_1}{\vec{v}_2 \vec{v}_2} \right)$$

 $\left\langle \begin{bmatrix} \vec{N}_1^T \\ \vec{N}_1^T \end{bmatrix} \right\rangle = \left\langle \begin{bmatrix} \vec{N}_1^T \\ -\vec{N}_2^T \end{bmatrix} \right\rangle$

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linearly independent. We have shown that if -1 is an eigenvalue of R, then there are always at least two linearly independent vectors corresponding to the eigenvalue -1, and therefore there is a two-dimensional space of eigenvectors corresponding to -1. Notice that the rank of R-I is two in this case since the two-dimensional subspace corresponding to -1 can be complemented only by a one-dimensional subspace corresponding to 1 to

avoid intersecting the subspaces in a non-zero vector.
§3 General
$$\lambda_1, \lambda_2, \lambda_3$$
: Finally, let us look at arbitrary (even non-real) eigenvalues. Assume $\lambda = x + yi$ for real x, y . Then we have
$$\frac{R\vec{v} = (x + yi)\vec{v}}{R\vec{v}} = (x + yi)\vec{v}$$
 (2.27) If $y \neq 0$, vector \vec{v} must be non-real since otherwise we would have a real vector on the left and a non-real vector on the right. Further-

more, the eigenvalues are pairwise distinct and hence there are three onedimensional subspaces of eigenvectors (we now understand the space as \mathbb{C}^3 over \mathbb{C}). In particular, there is exactly one one-dimensional subspace

corresponding to eigenvalue 1. The rank of R-I is two.

I at
$$\frac{1}{2}$$
 be an eigenvector of a retation matrix D. Then

Let
$$\vec{v}$$
 be an eigenvector of a rotation matrix R. Then
$$\mathbf{R} \, \vec{v} = (x + yi) \, \vec{v}$$

$$\mathbf{R} \, \vec{v} = (x + yi) \, \vec{v}
\mathbf{R}^{\top} \mathbf{R} \, \vec{v} = (x + yi) \, \mathbf{R}^{\top} \vec{v}$$

$$\begin{array}{rcl}
\widetilde{\mathbf{I}} \cdot \overrightarrow{v} &=& (x+yi) \, \underline{\mathbf{R}}^{\top} \overrightarrow{v} \\
& & \\
& & \\
\end{array} \qquad \begin{array}{rcl}
(2.30) \\
(2.31)
\end{array}$$

(2.28)(2.29)

$$\Rightarrow (x - yi)\vec{v} = \mathbf{R}^{\top}\vec{v}$$
eigenvector \vec{v} of \mathbf{R} corresponding to eigenvalue $x + yi$ is the

We see that the eigenvector \vec{v} of R corresponding to eigenvalue x + yi is the eigenvector of \mathbb{R}^{\top} corresponding to eigenvalue x-yi and vice versa. Thus, there is the following interesting correspondence between eigenvalues and eigenvectors of R and R^{T} . Considering eigenvalue-eigenvector pairs

$$R^{T}R = I$$

 $\lambda_{2} \equiv \vec{N}$, $y \neq 0$

121= | x - yi | = 1

3,=1, 2= x+yi, xy ∈ R

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 $(1, \vec{v}_1)$, $(x + yi, \vec{v}_2)$, $(x - yi, \vec{v}_3)$ of R we have $(1, \vec{v}_1)$, $(x - yi, \vec{v}_2)$, $(x + yi, \vec{v}_3)$ pairs of \mathbb{R}^{\top} , respectively.

§ **4 Orthogonality of eigenvectors** The next question to ask is what are the angles between eignevectors of R? We will considers pairs $(\lambda_1 = 1, \vec{v}_1)$, $(\lambda_2 = x + yi, \vec{v}_2)$, $(\lambda_3 = x - yi, \vec{v}_3)$ of eigenvectors associated with their respective eigenvalues. For instance, vector \vec{v}_1 denotes an eigenvector associated with egenvalue 1.

If all eigenvalues are equal to 1, i.e. R = I, then all non-zero vectors of \mathbb{R}^3 are eigenvectors of R and hence we can alway find two eignevectors

containing a given angle. In particular, we can choose three mutually orthogonal eignevectors. If $\lambda_1 = 1$ and $\lambda_2 = \lambda_3 = -1$, then we have seen that every \vec{v}_1 is perpendicular to \vec{v}_2 and \vec{v}_3 and that \vec{v}_2 and \vec{v}_3 can be any two non-zero vectors in a two-dimensional subspace of \mathbb{R}^3 , which is orthogonal to \vec{v}_1 . Therefore, for every angle, there are \vec{v}_2 and \vec{v}_3 which contain it. In particular, it is possible to choose \vec{v}_2 to be orthogonal to \vec{v}_3 and hence there are three

mutually orthogonal eigenvectors. Finally, if λ_2 , λ_3 are non-real, i.e. $y \neq 0$, we have three mutually distinct eigenvalues and hence there are exactly three one-dimensional subspaces (each without the zero vector) of eigenvectors. If two eigenvectors are from the same subspace, then they are linearly dependent and hence they contain the zero angle.

Let us now evaluate $\vec{v}_1^{\dagger} \vec{v}_2$

 $0 = \underline{\vec{v}_1^{\dagger} \vec{v}_2} = \underline{\vec{v}_1^{\dagger} \vec{v}_2} = \overline{\vec{v}_1^{\dagger}} \underline{R}^{\dagger} \underline{R} \underline{\vec{v}_2} = \overline{\vec{v}_1^{\dagger}} (x + yi) \underline{\vec{v}_2} = (x + yi) \underline{\vec{v}_1^{\dagger} \vec{v}_2}$ (2.33) We conclude that either $(x + yi) \neq 1$ or $\underline{\vec{v}_1^{\dagger} \vec{v}_2} = 0$. Since the latter can't be the case as $\underline{y} \neq 0$, the former must hold true. We see that $\underline{\vec{v}_1}$ is orthogonal to $\underline{\vec{v}_2}$. We can show that $\underline{\vec{v}_1}$ is orthogonal to $\underline{\vec{v}_3}$ exactly in the same way.

 $R^{T} = (\lambda_{1} \vec{N}_{1})_{1} (x-y_{1} \vec{N}_{2})_{1} (x+y_{1} \vec{N}_{3})$ $\lambda_{1} = \lambda_{2} = \lambda_{3} = 1$ $\lambda_{1} = \lambda_{2} = \lambda_{3} = 1$ $\lambda_{1} = \lambda_{3} = -1$ $\lambda_{1} = \lambda_{3} = 1$ $\lambda_{1} = \lambda_{3} = 1$ $\lambda_{1} = \lambda_{3} = 1$ $\lambda_{2} = \lambda_{3} = 1$ $\lambda_{3} = \lambda_{3} = 1$ $\lambda_{4} = \lambda_{3} = 1$ $\lambda_{4} = \lambda_{4} = \lambda_{4}$ $\lambda_{5} = \lambda_{5} = \lambda_{5} = 1$ $\lambda_{6} = \lambda_{5} = \lambda_{5} = 1$ $\lambda_{7} = \lambda_{7} = \lambda_{7} = \lambda_{7} = 1$ $\lambda_{1} = \lambda_{2} = \lambda_{3} = 1$ $\lambda_{1} = \lambda_{2} = \lambda_{3} = 1$ $\lambda_{2} = \lambda_{3} = 1$ $\lambda_{3} = \lambda_{3} = 1$ $\lambda_{1} = \lambda_{2} = \lambda_{3} = 1$ $\lambda_{2} = \lambda_{3} = 1$ $\lambda_{3} = \lambda_{3} = 1$ $\lambda_{1} = \lambda_{2} = \lambda_{3} = 1$ $\lambda_{2} = \lambda_{3} = 1$ $\lambda_{3} = \lambda_{3} = 1$ $\lambda_{1} = \lambda_{2} = \lambda_{3} = 1$ $\lambda_{2} = \lambda_{3} = 1$ $\lambda_{3} = \lambda_{3} = 1$ $\lambda_{4} = \lambda_{3} = 1$ $\lambda_{5} = \lambda_{5} = \lambda_{5} = 1$ $\lambda_{7} = \lambda_{7} = \lambda_$

Real: $\vec{x} \cdot \vec{y} = (|\vec{x}|| ||\vec{y}|| \cos \vec{x}(\vec{x}|\vec{y}))$

 $\mathbb{R} \qquad (1 \overline{N}_1) (x + 3 i \overline{N}_2) (x - 3 i \overline{N}_3)$

TPajdla. Elements of Geometry for Computer Vision and Robotics 2020-10-4 (pajdla@cvut.cz) ~ 上小 Let us next consider the angle between eigenvectors \vec{v}_2 and \vec{v}_3

> (2.34)(2.35)

(2.36)

(2.41)

$$\vec{v}_3^\dagger \vec{v}_2 = \vec{v}_3^\dagger \mathbf{R}^\top \mathbf{R} \vec{v}_2 = (\mathbf{R} \vec{v}_3)^\dagger \mathbf{R} \vec{v}_2 = ((x - yi) \vec{v}_3)^\dagger (x + yi) \vec{v}_2 \qquad (2.34)$$

$$= \vec{v}_3^\dagger (x + yi) (x + yi) \vec{v}_2 \qquad (2.35)$$

$$\vec{v}_3^\dagger \vec{v}_2 = (x^2 + 2xyi - y^2) \vec{v}_3^\dagger \vec{v}_2 \qquad (2.36)$$
We conclude that either $(x^2 + (2xyi - y^2)) = 1$ or $\vec{v}_3^\dagger \vec{v}_2 = 0$. The former implies $xy = 0$ and threfore $x = 0$ since $y \neq 0$ but then $-y^2 = 1$, which is, for a real y , impossible. We see that $\vec{v}_3^\dagger \vec{v}_2 = 0$, i.e. vectors \vec{v}_2 are orthogonal

dent. Therefore

which implies

implies xy = 0 and threfore x = 0 since $y \neq 0$ but then $-y^2 = 1$, which is, for a real y, impossible. We see that $\vec{v}_{3}^{\dagger}\vec{v}_{2}=0$, i.e. vectors \vec{v}_{2} are orthogonal to vectors \vec{v}_3 . Clearly, it is always possible to choose three mutually orhogonal eigenvectors. We can further normalize them to unit legth and thus obtain an orthonormal basis as non-zero orthogonal vectors are linearly indepen-

 $\frac{\mathbf{R}\,\vec{v}_2}{2} = \mathbf{R}\,(\vec{u} + \vec{w}i) = \mathbf{R}\,\vec{u} + \mathbf{R}\,\vec{w}\,i \qquad = \\
(x + yi)\,\vec{v}_2 = (x + yi)\,(\vec{u} + \vec{w}i) = x\,\vec{u} - y\,\vec{w} + (x\vec{w} + y\vec{u})i$ (2.39)(2.40)

 $\underline{\mathbf{R}} \, \vec{u} = x \, \vec{u} - y \, \vec{w}$ and $\underline{\mathbf{R}} \, \vec{w} = x \, \vec{w} + y \, \vec{u}$

y ER In general, not always possible $\begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \overrightarrow{v} = \chi \overrightarrow{v}$ $R\begin{bmatrix}\vec{v}_1 & \vec{v}_2 & \vec{v}_3\end{bmatrix} = \begin{bmatrix}\vec{v}_1 & \vec{v}_2 & \vec{v}_3\end{bmatrix} \begin{bmatrix}\lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3\end{bmatrix} (2.37) = \begin{bmatrix}\lambda_1 & 0 & 1 \\ \lambda_1 & \lambda_2 & 1 \end{bmatrix}$ $\begin{bmatrix}\lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3\end{bmatrix} (2.37) = \begin{bmatrix}\lambda_1 & 0 & 1 \\ \lambda_1 & \lambda_2 & 1 \end{bmatrix}$ Not of (by normalization) $\overrightarrow{V_2}, \overrightarrow{V_3} \in C^3 \setminus \{\overrightarrow{0}\} \longrightarrow \overrightarrow{W_1}$ $\overrightarrow{V_2}, \overrightarrow{V_3} \in C^3 \setminus \{\overrightarrow{0}\} \longrightarrow \overrightarrow{W_1}$ $\overrightarrow{V_1} = \overrightarrow{W_1} + \overrightarrow{W_1}$

T Pajdla. Elements of Geometry for Computer Vision and Robotics 2020-10-4 (pajdla@cvut.cz) Now, let us compare two expressions: $R(\vec{u} - \vec{w}i)$ and $(x - yi)(\vec{u} - \vec{w}i)$

 $\Im \left(\begin{pmatrix} R(\vec{u} - \vec{w}i) &= R\vec{u} - R\vec{w}i = x\vec{u} - y\vec{w} - (x\vec{w} + y\vec{u})i & (2.42) \\ (x - yi)(\vec{u} - \vec{w}i) &= x\vec{u} - y\vec{w} - (x\vec{w} + y\vec{u})i & (2.43) \end{pmatrix} \right)$

We see that
$$R(\vec{u} - \vec{w}i) = (x - yi)(\vec{u} - \vec{w}i)$$
 (2.44)

$$R(\vec{u} - \vec{w}i) = (x - yi)(\vec{u} - \vec{w}i)$$
which means that $(x - yi, \vec{u} - \vec{w}i)$ are an eigenvalue-eigenvector pair of

R. It is important to understand what has been shown. We have shown that if $\vec{u} + \vec{w}i$ is an eigenvector of R corresponding to an eigenvalue λ ,

then the conjugated vector $\vec{u} - \vec{w}i$ is an eignevector of R corresponding to eigenvalue, which is conjugated to λ (This does not mean that every two

eigenvectors corresponding to x + yi and x - yi must be conjugated). The conclusion from the previous analysis is that the both non-real

eigenvectors of R are generated by the same two real vectors \vec{u} and \vec{w} . Let

us look at the angle between
$$\vec{u}$$
 and \vec{w} . Consider that
$$0 = \vec{v}_3^{\dagger} \vec{v}_2 = (\vec{u} - \vec{w}i)^{\dagger} (\vec{u} + \vec{w}i) = (\vec{u}^{\top} + \vec{w}^{\top}i) (\vec{u} + \vec{w}i) \qquad (2.45)$$

$$= (\vec{u}^{\top}\vec{u} - \vec{w}^{\top}\vec{w}) + (\vec{u}^{\top}\vec{w} + \vec{w}^{\top}\vec{u}) i \qquad (2.46)$$

$$\vec{u}^{\top}\vec{u} = \vec{w}^{\top}\vec{w} \quad \text{and} \quad \vec{w}^{\top}\vec{u} = 0$$

Finally, let us consider

and hence

$$u \cdot u = w \cdot w \quad \text{and} \quad w \cdot u = 0$$

which means that vectors
$$\vec{y}$$
 and \vec{w} are orthogonal

which means that \vec{u} and \vec{w} are also orthogonal to \vec{v}_1 .

which means that vectors \vec{u} and \vec{w} are orthogonal.

$$u = 0$$
 gonal.

(2.49)

(2.50)

でして

$$\vec{N}_2 = (\vec{N} + \vec{W}i) \quad \vec{U}_1 \vec{W} \in \mathbb{R}^3$$

$$\overrightarrow{N}_1 \sim 1$$
 $\overrightarrow{M}_1 \stackrel{\rightarrow}{M}$

 $0 = \vec{v}_1^{\top} \vec{v}_2 = \vec{v}_1^{\top} \vec{u} + \vec{v}_1^{\top} \vec{w} i$

 $\vec{v}_1^{\mathsf{T}} \vec{u} = 0$ and $\vec{v}_1^{\mathsf{T}} \vec{w} = 0$

T Pajdla. Elements of Geometry for Computer Vision and Robotics 2020-10-4 (pajdla@cvut.cz) $A^3 \sim \mathbb{R}^3$

2.1.4 Rotation axis

A one-dimensional subspace generated by an eigenvector \vec{v}_1 of R corresponding to $\lambda = 1$, is called the *rotation axis* (or axis of rotation) of R. If R = I, then there is an infinite number of rotation axes, otherwise there is exactly one. Vectors \vec{v} , which are in a rotation axis of rotation R, remain unchanged by R, i.e. $R\vec{v} = \vec{v}$.

Consider that the eigenvector of R corresponding to 1 is also an eigenvector of R^{\top} since

$$\begin{array}{cccc}
\overrightarrow{\Gamma} \cdot & & & & & & & & & \\
& \overrightarrow{\Gamma} \cdot \overrightarrow{v_1} & = & & & & & & \\
& & \overrightarrow{\Gamma} \cdot \overrightarrow{v_1} & = & & & & & \\
\end{array}$$

It implies

$$\underbrace{\mathbf{T}} \cdot \vec{v}_1 = \underline{\mathbf{R}}^{\mathsf{T}} \vec{v}_1$$

$$\vec{v}_1 = (R - R^{\top}) \vec{v}_1 = \frac{r_{21}}{r_{23} - r_{32}} \vec{v}_1 = \frac{r_{23} - r_{32}}{r_{23} - r_{32}} \vec{v}_1 = \frac{r_{23} - r_{23}}{r_{23} - r_{23}} \vec{v}_2 = \frac{r_{23} - r_{23}}{r_{23} - r_{23}} \vec{v}_1 = \frac{r_{23} - r_{23}}{r_{23} - r_{23}} \vec{v}_2 = \frac{r_{23} - r_{23}}{r_{23} - r_{23}} \vec{v}_3 = \frac{r_{23} - r_{23}}{r_{23} - r_{23}} \vec{v}_3$$

$$-r_{21}$$
 $r_{13} - r_{31}$ r_{33} r_{33}

Clearly, we have a nice formula for an eigenvector corresponding to
$$\lambda_1 = 1$$
, when vector $\begin{bmatrix} r_{32} - r_{23} & r_{13} - r_{31} & r_{21} - r_{12} \end{bmatrix}^{\mathsf{T}}$ is non-zero. That is when

 $R - R^{T}$ is a non-zero matrix, which is exactly when R is not symmetric. Let us now investigate the situation when R is symmetric. Then, $R = \mathbb{R}' = \mathbb{R}$ 10 in eigens poce is mapped to itself element - wise

Rototion axis

R(R+I)=(R+I)R (111) = 1. (111)

L- 70 = 11 N11

(2.53)

(2.54)

(2.57)

 $R^{\top} = R^{-1}$ and therefore R(R+I) = RR + R = I + R = R+I

T Pajdla. Elements of Geometry for Computer Vision and Robotics 2020-10-4 (pajdla@cvut.cz) which shows that the non-zero columns of the matrix R+I are eigenvectors

corresponding to the unit eigenvalue. Clearly, at least one of the columns must be non-zero since otherwise, R = -I and |R| would be minus one, which is impossible for a rotation.

2.1.5 Rotation angle

Rotation angle θ of rotation R is the angle between a non-zero real vector \vec{x} which is orthogonal to \vec{v}_1 and its image R \vec{x} . There holds

non-zero \vec{u} we get the following contradiction

 $= \underbrace{\frac{x(\vec{u}^{\top}\vec{u} + \vec{w}^{\top}\vec{w}) + y(\vec{u}^{\top}\vec{u} - \vec{w}^{\top}\vec{w})}_{\vec{u}^{\top}\vec{u} + \vec{w}^{\top}\vec{w}}$

which is orthogonal to
$$\vec{v}_1$$
 and its image $R\vec{x}$. There holds
$$\cos \theta = \frac{\vec{x}^T R \vec{x}}{2\pi i}$$

 $\vec{x} = \vec{u} + \vec{w}$

their sum can be zero only if they both are zero since otherwise for, e.g., a

21

 $0 = \vec{u}^{\top} \vec{0} = \vec{u}^{\top} (\vec{u} + \vec{v}) = \vec{u}^{\top} \vec{u} + \vec{u}^{\top} \vec{v} = \vec{u}^{\top} \vec{u} \neq 0$

Let us set

$$\cos\theta = \frac{\vec{x}^{\top} \vec{R} \vec{x}}{\vec{x}^{\top} \vec{x}}$$

 $\cos \theta = \frac{\vec{x}^{\mathsf{T}} \mathbf{R} \vec{x}}{\vec{z}^{\mathsf{T}} \vec{z}}$

(2.58)

(2.59)

Clearly, \vec{x} is a real vector which is orthogonal to \vec{v}_1 since both \vec{u} and \vec{w} are.

Let's see that it is non-zero. Vector \vec{v}_2 is an eigenvector and thus

 $\longrightarrow 0 \neq \vec{v}_2^{\top} \vec{v}_2 = \vec{u}^{\top} \vec{u} + \vec{w}^{\top} \vec{w}$

and therefore $\vec{u} \neq \vec{0}$ or $\vec{w} \neq \vec{0}$. Vectors $\vec{u}_{\underline{i}} \cdot \vec{w}$ are orthogonal and therefore

(2.60)

(2.62)

(2.63)

マ、マ、マニューが、マュニューが、

So votation angle

~ * (\vec{x} , R \vec{x})

で、上が、 が、上が

Let us now evaluate $\frac{\vec{x} + \vec{y} \cdot \vec{x}}{\cos \theta} = \frac{\vec{x}^{\top} \mathbf{R} \cdot \vec{x}}{\vec{x}^{\top} \vec{x}} = \frac{(\vec{u} + \vec{w})^{\top} \mathbf{R} \cdot (\vec{u} + \vec{w})}{(\vec{u} + \vec{w})^{\top} (\vec{u} + \vec{w})} = \frac{(\vec{u} + \vec{w})^{\top} (\vec{x} \cdot \vec{u} - y \cdot \vec{w} + x \cdot \vec{w} + y \cdot \vec{u})}{\vec{u}^{\top} \vec{u} + \vec{w}^{\top} \vec{w}} = \frac{\vec{w} \cdot \vec{w}}{\vec{w}}$

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We have used equation 2.41 and equation 2.48 We see that the rotation angle is given by the real part of λ_2 (or λ_3). Consider the characteristic equation of R, Equation 2.13

 $\cos \theta = \frac{1}{2} (\operatorname{trace} R - 1)$

We have seen that rank (R - I) = 0 for R = I and rank (R - I) = 2 for all

Let us next investigate the relationship between the range and the null space of (R - I). The null space of (R - I) is generated by eigenvectors

Now assume that vector \vec{v} is also in the range of (R - 1). Then, there is

a vector $\vec{a} \in \mathbb{R}^3$ such that $\vec{v} = (\mathbf{R} - \mathbf{I})\vec{a}$. Let us evaluate/the square of the

 $\vec{v}^{\top}\vec{v} = \vec{v}^{\top}(\mathbf{R} - \mathbf{I})\vec{a} = (\vec{v}^{\top}\mathbf{R} - \vec{v}^{\top})\vec{a} = (\vec{v}^{\top} - \vec{v}^{\top})\vec{a} = \underline{0}$

which implies $\vec{v} = \vec{0}$. We have used result 2.32 with x = 1 and y = 0.

Hence, the range of R - I intersects the null space of R - I in the zero

 $= (\lambda - 1)(\lambda - x - yi)(\lambda - x + yi)$

We see that trace R = 2x + 1 and thus

2.1.6 Matrix (R - I).

rotation matrices $R \neq I$.

length of \vec{v}

vector.

quation of R, Equation 2.13
$$0 = \lambda^3 - \operatorname{trace} (R \lambda^2 + (R_{11} + R_{22} + R_{33}) \lambda - |R|)$$
 (2.64)

$$= (\lambda - 1)(\lambda - x - yi)(\lambda - x + yi)$$

$$= \lambda^3 - (2x + 1)\lambda^2 + (x^2 + 2x + y^2)\lambda - (x^2 + y^2)$$
(2.

$$(2.65)$$

$$x^2 + y^2) \qquad (2.66)$$



Simple formule for D

range = ({A x , xar})

mull = ({x ep3 Ax=})

For motion oxis

molerstonding

 $R\vec{N}_1 = \vec{N}_1 = I \cdot \vec{N}_1$

RN, - RIN, = (R-I) N, = 0

vand, R-I < 3 | N, \$ 0

corresponding to 1 since $(\mathbf{R} - \mathbf{I}) \vec{v} = 0$ implies $\mathbf{R} \vec{v} = \vec{v}$.

2.1.7 Tangent space to rotations

The set of rotation matrices

The set of rotation matrices
$$\mathcal{Q} = \int \mathbf{p}$$

 $\mathcal{R} = \{ \mathbf{R} \in \mathbb{R}^{3 \times 3} \mid \mathbf{R}^{\mathsf{T}} \mathbf{R} = \mathbf{I}, \mid \mathbf{R} \mid = 1 \}$

can be understood as a subset of \mathbb{R}^9 with

$$\mathbf{r} = \begin{bmatrix} r_{11} & r_{21} & r_{31} & r_{12} & r_{22} & r_{32} & r_{12} & r_{23} & r_{3} \end{bmatrix}^{\mathsf{T}} \text{ representing } \mathbf{R} = \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix}$$
(2.70)

Rotation constraints in definition 2.69 are algebraic and thus \mathcal{R} is a *an affine* variety 3. Let us investigate how does look the tangent space to \mathcal{R} .

To get the tangent space to \mathcal{R} , we will first find the normal N_R to \mathcal{R} at rotation R and then take its orthogonal complement T_R , which is tangent to \mathcal{R} at R. In the end, we will write it all down in a convenient matrix form.

The space N_R , normal to \mathcal{R} , is generated by columns of the *Jacobian matrix* [?] of constraints in 2.69 written in a matrix form as

$$C = \begin{bmatrix} r_{11} r_{12} + r_{21} r_{22} + r_{31} r_{32} \\ r_{11} r_{13} + r_{21} r_{23} + r_{31} r_{33} \\ r_{12} r_{13} + r_{22} r_{23} + r_{32} r_{33} \\ r_{11}^{2} + r_{21}^{2} + r_{31}^{2} - 1 \\ r_{12}^{2} + r_{22}^{2} + r_{32}^{2} - 1 \\ r_{13}^{2} + r_{23}^{2} + r_{33}^{2} - 1 \\ r_{11} r_{22} r_{33} - r_{11} r_{23} r_{32} - r_{12} r_{21} r_{33} + r_{12} r_{23} r_{31} + r_{13} r_{21} r_{32} - r_{13} r_{22} r_{31} - 1 \end{bmatrix}$$

$$(2.71)$$

³Affine variety is a subset of a linear space defined by algebraic constraints

The Jacobian matrix of C is obtained as

$$\mathbf{J}_{ij} = \frac{\partial \mathsf{C}_i}{\partial \mathbf{r}_j}, \qquad \mathbf{J} = \begin{bmatrix} r_{12} & r_{22} & r_{32} & r_{11} & r_{21} & r_{31} & 0 & 0 & 0 \\ r_{13} & r_{23} & r_{33} & 0 & 0 & 0 & r_{11} & r_{21} & r_{31} \\ 0 & 0 & 0 & r_{13} & r_{23} & r_{33} & r_{12} & r_{22} & r_{32} \\ 2r_{11} & 2r_{21} & 2r_{31} & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 2r_{12} & 2r_{22} & 2r_{32} & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 2r_{13} & 2r_{23} & 2r_{33} \\ J_{71} & J_{72} & J_{73} & J_{74} & J_{75} & J_{76} & J_{77} & J_{78} & J_{79} \end{bmatrix}$$
 with

$$J_{71} = r_{22} r_{33} - r_{23} r_{32}$$

$$J_{72} = -r_{12} r_{33} + r_{13} r_{32}$$

$$J_{73} = r_{12} r_{23} - r_{13} r_{22}$$

$$J_{74} = -r_{21} r_{33} + r_{23} r_{31}$$

$$J_{75} = r_{11} r_{33} - r_{13} r_{31}$$

$$J_{76} = -r_{11} r_{23} + r_{13} r_{21}$$

$$J_{77} = r_{21} r_{32} - r_{22} r_{31}$$

$$J_{78} = -r_{11} r_{32} + r_{12} r_{31}$$

$$J_{79} = r_{11} r_{22} - r_{12} r_{21}$$

Jacobian matrix J is a 7×9 matrix. The first three rows of J contain the elements of two columns of R. The next three rows contain one column of R. It suggests to construct a basis T of the tangent space T_R to R from

columns of R. We can check that

$$JT = 0 \quad \text{with} \quad T = \begin{bmatrix}
0 & -r_{13} & r_{12} \\
0 & -r_{23} & r_{22} \\
0 & -r_{33} & r_{32} \\
r_{13} & 0 & -r_{11} \\
r_{23} & 0 & -r_{21} \\
r_{33} & 0 & -r_{31} \\
-r_{12} & r_{11} & 0 \\
-r_{22} & r_{21} & 0 \\
-r_{32} & r_{31} & 0
\end{bmatrix}.$$
(2.72)

Next, we can see that each column of T contains two different columns of R and hence T x = 0 for a non-zero x implies that every two columns of R are linearly dependent, which is impossible. Therefore, T has rank equal to three at least.

Finally, the first six rows of J contain columns of R. We see that $\begin{bmatrix} \mathbf{x}^\top & 0 \end{bmatrix} \mathbf{J} = 0$ for a non-zero \mathbf{x} implies that columns of R are linearly dependent, which is impossible. Therefore, the rank of N_R is not smaller than six. Hence, the dimension of the tangent space T_R is exactly three at every $R \in \mathcal{R}$ and T is indeed a basis of T_R .

Let us now rewrite the above back into a matrix form by inverting the matrix vectorization used in [2.70]. We rewrite columns of T into three matrices

$$T_{1} = \begin{bmatrix} 0 & r_{13} & -r_{12} \\ 0 & r_{23} & -r_{22} \\ 0 & r_{33} & -r_{32} \end{bmatrix}, T_{2} = \begin{bmatrix} -r_{13} & 0 & r_{11} \\ -r_{23} & 0 & r_{21} \\ -r_{33} & 0 & r_{31} \end{bmatrix}, T_{3} = \begin{bmatrix} r_{12} & -r_{11} & 0 \\ r_{22} & -r_{21} & 0 \\ r_{32} & -r_{31} & 0 \end{bmatrix}$$
(2.73)

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and then can write the reformated tangent space of rotations at R for some real vector $\mathbf{s} = \begin{bmatrix} s_1 & s_2 & s_3 \end{bmatrix}$ as

$$T_{R}(s) = T_{1} s_{1} + T_{2} s_{2} + T_{3} s_{3}$$

$$= \begin{bmatrix} -s_{2} \begin{bmatrix} r_{13} \\ r_{23} \\ r_{33} \end{bmatrix} + s_{3} \begin{bmatrix} r_{12} \\ r_{22} \\ r_{32} \end{bmatrix}, s_{1} \begin{bmatrix} r_{13} \\ r_{23} \\ r_{33} \end{bmatrix} - s_{3} \begin{bmatrix} r_{11} \\ r_{21} \\ r_{31} \end{bmatrix}, -s_{1} \begin{bmatrix} r_{12} \\ r_{22} \\ r_{32} \end{bmatrix} + s_{2} \begin{bmatrix} r_{11} \\ r_{21} \\ r_{31} \end{bmatrix} \end{bmatrix}$$

$$= \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix} \begin{bmatrix} 0 & -s_{3} & s_{2} \\ s_{3} & 0 & -s_{1} \\ -s_{2} & s_{1} & 0 \end{bmatrix}$$

$$= R[s]$$

$$(2.76)$$

The first order approximation of rotations around R is then obtained as

$$R + T_{R}(s) = R + R[s]_{\times} = R(I + [s]_{\times})$$
(2.77)

In particular, vectors in the tangent spaces to the space of rotations at the identity, which are called *infinitesimal rotations*, are

$$T_{I}(s) = [s]_{\times} \tag{2.78}$$

and the first order approximation of rotations at identity is

$$I + T_{I}(s) = I + [s]_{\times}$$
 (2.79)