	P <i>n</i> P			
	• We can solve PnP by minimizing $\varepsilon_{PnP,GEO} = \sum_{k=1}^{n} d_{PP}(\mathbf{y}_k, \mathbf{y}'_k)^2$ , where $\mathbf{y}'_k = \mathbf{R}\bar{\mathbf{x}}_k + \bar{\mathbf{t}}$ ,			
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Spring 2014 Lecture 7A Representations of 3D rotations	over $\mathbf{R} \in SO(3)$ and $\mathbf{t} \in \mathbb{R}^3$ • Initial solution from P3P • We need to parameterize $\mathbf{R} \in SO(3)$			
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Stivi	Parameterization of SO(3)			
<ul> <li>A similar case appears in SfM, where we minimize</li> </ul>	• Each R $\in$ SO(3) is a 3 $ imes$ 3 matrix that satisfies			
$\boldsymbol{\varepsilon}_{BA} = \sum_{k=1}^{m} \sum_{j=1}^{p} w_{kj} d_{\mathrm{PP}}(\mathbf{y}_{kj}, \mathbf{C}_k \mathbf{x}_j)^2,$	$\mathbf{R}^{T}\mathbf{R} = \mathbf{I}$ and $\det(\mathbf{R}) = +1$			
over the camera poses: $\mathbf{C}_{\mathbf{k}} \sim (\mathbf{R}_{\mathbf{k}}  \mathbf{t}_{\mathbf{k}})$	<ul> <li>How can we change R to R' such that these constraints are maintained?</li> </ul>			
<ul> <li>Each rotation <b>R</b><sub>k</sub> ∈ SO(3) needs to be parameterized</li> </ul>				
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### Axis-angle representation

- Any rotation **R** is characterized by
  - a rotation axis **n** (normalized) (2 dof)
  - a rotation angle  $\alpha$  (1 dof)

such that  ${\bf R}$  rotates the angle  $\alpha$  about  ${\bf n}$  according to the "right-hand rule"

 $(\mathbf{n}, \alpha)$ same R as  $(-\mathbf{n}, -\alpha)$ 



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# Rodrigues' rotation formula

- Given **R**, how do we determine **n** and α?
- Based on Rodrigues' formula:

$$\cos \alpha = \frac{\operatorname{trace}(\mathbf{R}) - 1}{2}, \qquad \qquad \frac{\mathbf{R} - \mathbf{R}^{\top}}{2} = \sin \alpha \, [\, \mathbf{\hat{n}} \,]_{\times},$$

• Notice: ambiguity at  $\alpha$  =  $\pi$ 

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# Rodrigues' rotation formula

- Given (**n**, *α*), how do we determine **R**?
- Use Rodrigues' rotation formula:

```
\mathbf{R}(\hat{\mathbf{n}}, \alpha) = \hat{\mathbf{n}} \, \hat{\mathbf{n}}^\top + \cos \alpha \, (\mathbf{I} - \hat{\mathbf{n}} \, \hat{\mathbf{n}}^\top) + \sin \alpha \, [\, \hat{\mathbf{n}}\,]_{\times}.
```

 $\mathbf{R}(\hat{\mathbf{n}},\alpha) = \mathbf{I} + (1 - \cos \alpha) (\hat{\mathbf{n}} \, \hat{\mathbf{n}}^{\top} - \mathbf{I}) + \sin \alpha [\hat{\mathbf{n}}]_{\times}.$ 

$$\mathbf{R}(\hat{\mathbf{n}},\alpha) = \mathbf{I} + (1 - \cos \alpha) \left[ \hat{\mathbf{n}} \right]_{\times}^2 + \sin \alpha \left[ \hat{\mathbf{n}} \right]_{\times}.$$

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### Using so(3) so(3) • $[\mathbf{m}]_{\times} \in$ so(3) has eigensystem: • so(3) is the set of $3 \times 3$ anti-symmetric matrices 0, +i |m|, -i |m|• Can be parameterized by $\mathbf{m} \in \mathbb{R}^3$ : $[\mathbf{m}]_{\sim}$ eigenvalues • 2 alternative mappings $so(3) \rightarrow SO(3)$ $\mathbf{m}$ , $\mathbf{p}$ -i $\mathbf{q}$ , $\mathbf{p}$ +i $\mathbf{q}$ eigenvectors - Matrix exponential Cayley transformation where (m, p, q) is a right-handed orthogonal basis in $\mathbb{R}^3$ Geometry in Computer Vision Geometry in Computer Vision 10 13 June 2014 9 13 June 2014 Klas Nordberg Klas Nordberg Matrix exponential Matrix exponential • The matrix exponential function is defined for • If **M** is diagonalized by unitary **E**: a square matrix **M** as $\mathbf{E}^*\mathbf{E} = \mathbf{I}$ $\mathbf{M} = \mathbf{E} \mathbf{D} \mathbf{E}^*$ , **D** diagonal, eigenvalues eigenvectors $e^{\mathbf{M}} = \mathbf{I} + \mathbf{M} + \frac{1}{2}\mathbf{M}^2 + \frac{1}{6}\mathbf{M}^3 + \ldots = \sum_{k=0}^{\infty} \frac{1}{k!}\mathbf{M}^k,$ its exponential can be expressed as $\mathbf{e}^{\mathbf{M}} = \mathbf{E} \ \mathbf{e}^{\mathbf{D}} \ \mathbf{E}^{*} \qquad \exp \begin{pmatrix} d_{1} & 0 & \dots & 0 \\ 0 & d_{2} & \dots & 0 \\ \vdots & \vdots & \ddots & 0 \\ 0 & 0 & \dots & d \end{pmatrix} = \begin{pmatrix} e^{d_{1}} & 0 & \dots & 0 \\ 0 & e^{d_{2}} & \dots & 0 \\ \vdots & \vdots & \ddots & 0 \\ 0 & 0 & \dots & d \end{pmatrix}$ Geometry in Computer Vision Geometry in Computer Vision 13 June 2014 11 13 June 2014 12 Klas Nordberg Klas Nordberg



### Cayley transformation Parameterization of SO(3) • If $\mathbf{M} = [a \mathbf{n}]_{\vee}$ for normalized $\mathbf{n}$ and $a \in \mathbb{R}$ : $C(\mathbf{M})$ $\mathbf{R} \in SO(3)$ $M \in so(3)$ $C(\mathbf{M}) = \mathbf{R}(\mathbf{n}, \alpha), \quad a = \tan(\alpha/2)$ $C(\mathbf{R})$ • Inverse transformation: The *C* mapping is easy to implement. $M = (I - R)(I + R)^{-1} = C(R)$ It can be differentiated w.r.t. $\mathbf{M} = [\mathbf{m}]_{\vee}$ But less trivial than (n, $\alpha$ ) $\rightarrow$ SO(3) Geometry in Computer Vision Geometry in Computer Vision 17 13 June 2014 13 June 2014 18 Klas Nordberg Klas Nordberg Quaternions Quaternions

 Quaternions can be seen as a generalization of complex numbers to the case where we have three distinct imaginary units:

$$q = a + i b + j c + k d$$

 $\mathbb{H}$ 



- Alternatively, we can see ℍ as an algebra on ℝ<sup>4</sup>, allowing us to multiply vectors in ℝ<sup>4</sup> to produce vectors in ℝ<sup>4</sup>
- Alternatively, we can see  $\mathbb{H}$  as an algebra on  $\mathbb{R} \times \mathbb{R}^3$ , consisting of ordered pairs of a real number and a vector in  $\mathbb{R}^3$
- q = (s, v)  $\in \mathbb{H}$
- $|q|^2 = s^2 + |\mathbf{v}|^2$

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### From $\mathbb{H}$ to SO(3) Quaternions and SO(3) • Sandwich product: • Given unit guaternion $q = (q_1, q_2, q_3, q_4) = (\cos(\alpha/2), \sin(\alpha/2) n)$ : $\mathbf{q} \circ \mathbf{p} \circ \mathbf{q}^{-1} = \dots = (\mathbf{0}, \mathbf{R}(\mathbf{n}, \alpha) \mathbf{u})$ $\mathbf{R} = \begin{pmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{pmatrix} = \begin{pmatrix} q_1^2 + q_2 - q_3^2 - q_4^2 & 2(q_2q_3 - q_1q_4) & 2(q_1q_3 + q_2q_4) \\ 2(q_2q_3 + q_1q_4) & q_1^2 - q_2 + q_3^2 - q_4^2 & 2(q_3q_4 - q_1q_2) \\ 2(q_2q_4 - q_1q_3) & 2(q_1q_2 + q_3q_4) & q_1^2 - q_2 - q_3^2 + q_4^2 \end{pmatrix}$ • Each rotation can be represented by a (A) quaternion q = (cos( $\alpha/2$ ), sin( $\alpha/2$ ) **n**) • Double embedding: both g and -g works • Each element in **R** is a quadratic function in q Geometry in Computer Vision Geometry in Computer Vision 13 June 2014 25 13 June 2014 26 Klas Nordberg Klas Nordberg From SO(3) to $\mathbb{H}$ Parameterization of SO(3) • From the previous mapping: (A) $q_1^2 = \frac{1 + r_{11} + r_{22} + r_{33}}{4}, \qquad q_2^2 = \frac{1 + r_{11} - r_{22} - r_{33}}{4},$ $q \in \mathbb{H}$ $\mathbf{R} \in SO(3)$ |q|=1 $q_3^2 = \frac{1 - r_{11} + r_{22} - r_{33}}{4}, \qquad q_4^2 = \frac{1 - r_{11} - r_{22} + r_{33}}{4}.$ (B) and $r_{12} + r_{21} = 4 q_2 q_3,$ $r_{13} + r_{31} = 4 q_2 q_4,$ $r_{23} + r_{32} = 4 q_3 q_4,$ Both (A) and (B) are easy to implement. $r_{21} - r_{12} = 4 q_1 q_4,$ $r_{13} - r_{31} = 4 q_1 q_3,$ $r_{32} - r_{23} = 4 q_1 q_2.$ (A) can be differentiated w.r.t. unit quaternion $q \in \mathbb{R}^4$ (B) Geometry in Computer Vision Geometry in Computer Vision 13 June 2014 27 13 June 2014 28 Klas Nordberg Klas Nordberg

## Euler angles

- We can decompose any **R** ∈ SO(3) into a product of 3 rotations around *fixed axes*
- For example:
  - $\mathbf{R} = \operatorname{Rot}_{z}(\alpha_{1}) \operatorname{Rot}_{x}(\alpha_{2}) \operatorname{Rot}_{z}(\alpha_{3})$
- $(\alpha_1, \alpha_2, \alpha_3)$  are the *Euler angles* of **R**

### **Euler angles**

• There are straight-forward mappings

 $(\alpha_1, \alpha_2, \alpha_3) \leftrightarrow \mathbf{R} \in SO(3)$ 

• Notice: rotations about the z-axis always have an ambiguous representation:

**R**(
$$\alpha_1$$
, 0,  $\alpha_3$ ) = **R**( $\alpha_1$ + $\Delta$ , 0,  $\alpha_3$ - $\Delta$ )

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## Euler angles

- This ambiguity implies that **D**, the derivatives of **R** with respect to (α<sub>1</sub>, α<sub>2</sub>, α<sub>3</sub>) is rank deficient when α<sub>2</sub> = 0
- If Euler angles are use as a parameterization of **R** in a non-linear optimization, there will be a stationary point for all points ( $\alpha_1$ , 0,  $\alpha_3$ ) where the optimization can get stuck