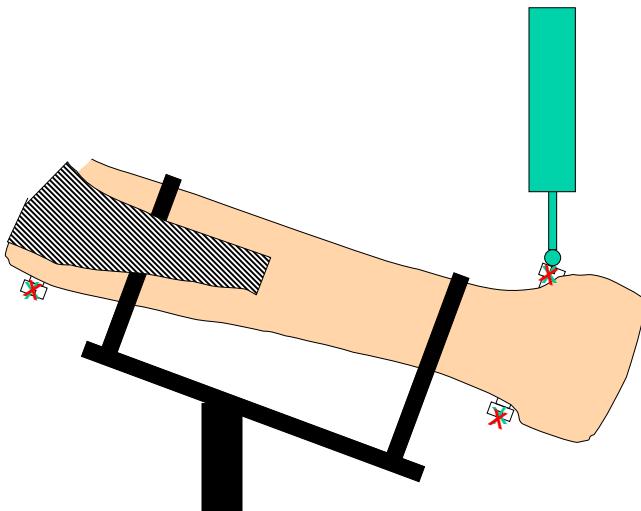


Cartesian Coordinates, Points, and Transformations

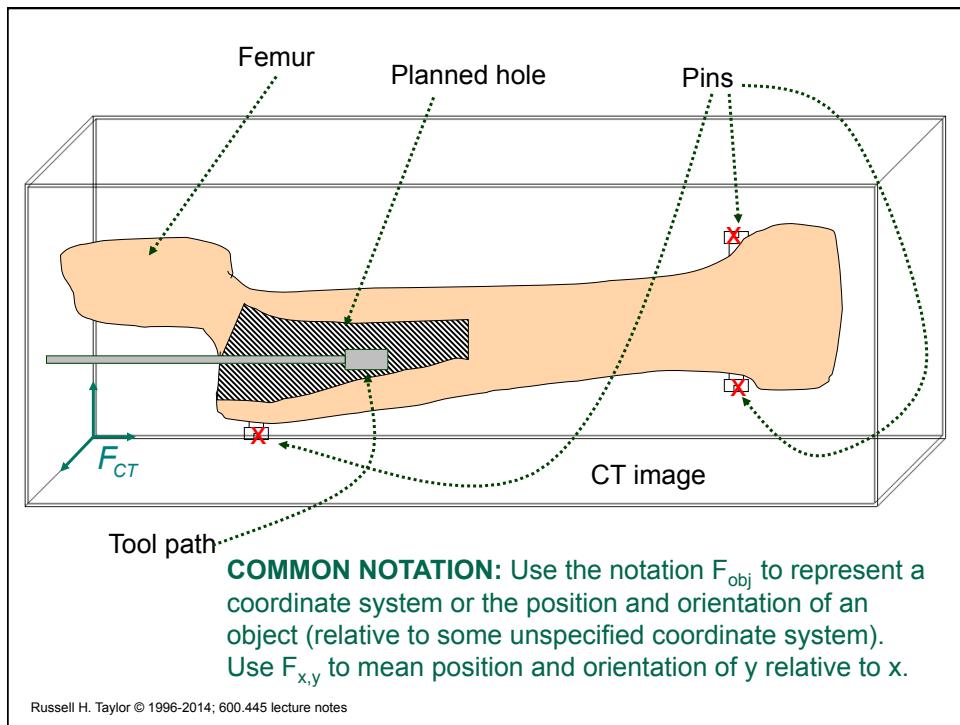
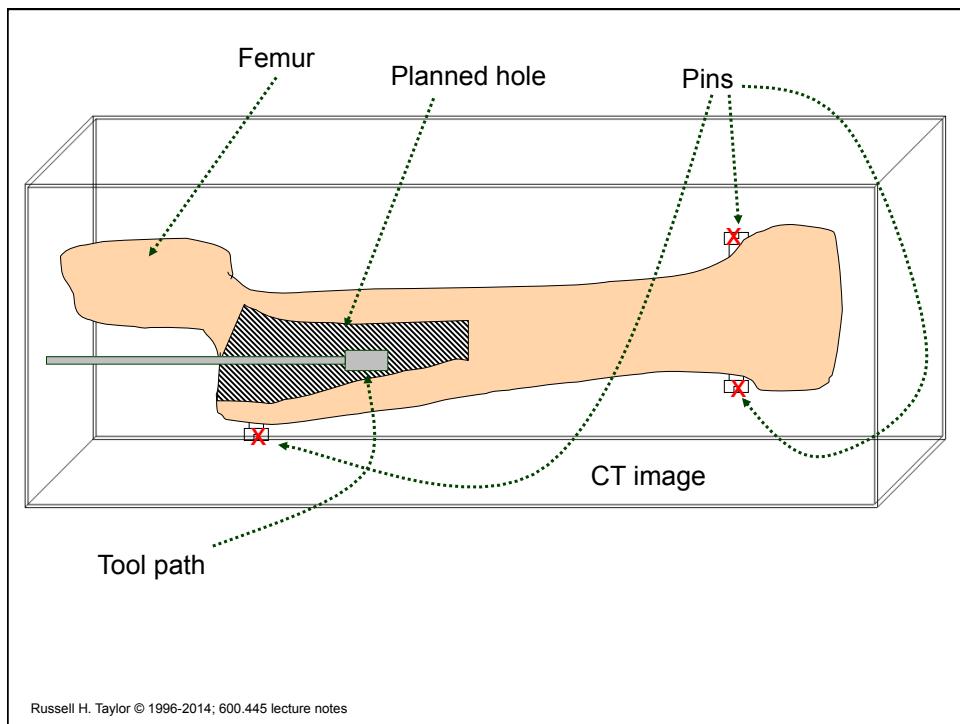
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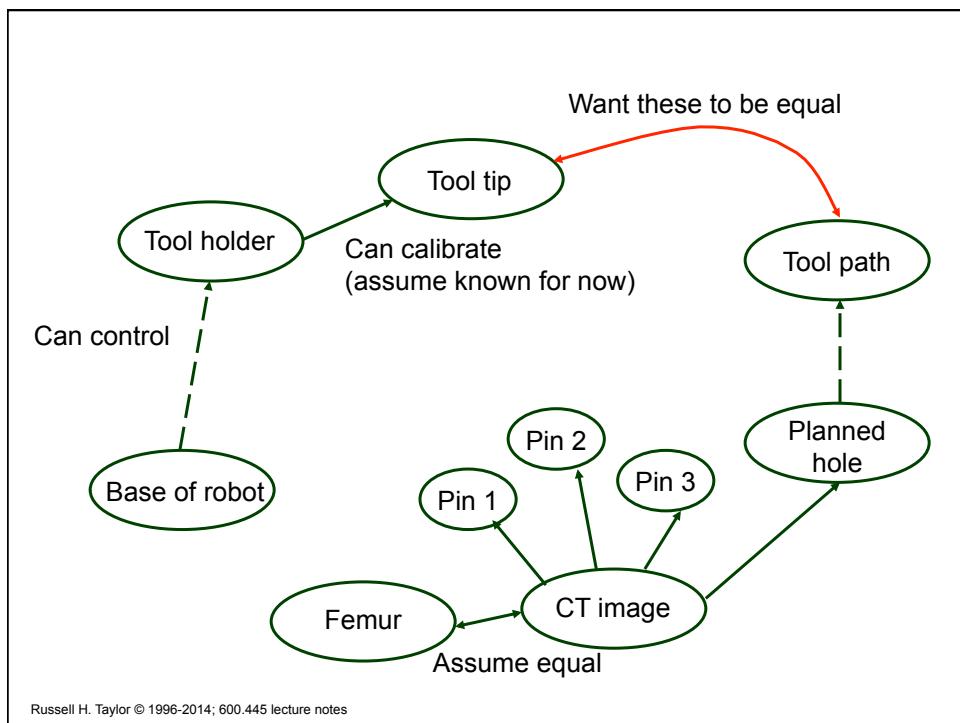
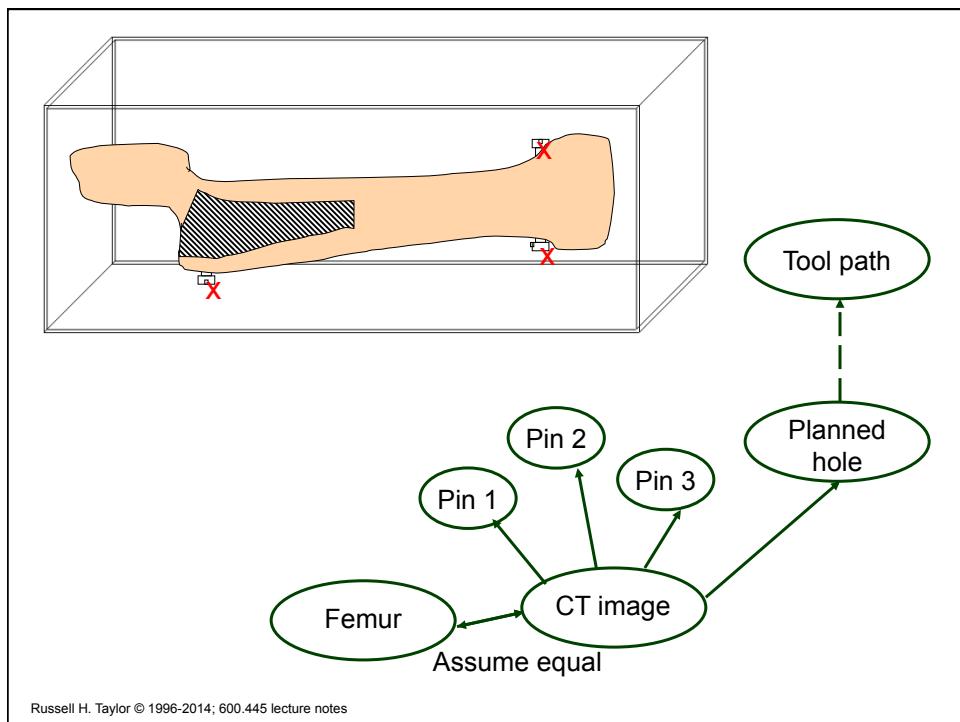
Russell Taylor

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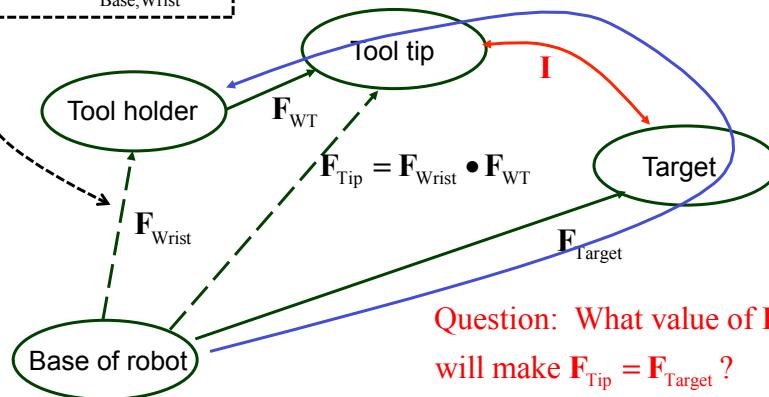


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More correctly, this would be $\mathbf{F}_{\text{Base}, \text{Wrist}}$



Question: What value of $\mathbf{F}_{\text{Wrist}}$ will make $\mathbf{F}_{\text{Tip}} = \mathbf{F}_{\text{Target}}$?

Answer:

$$\begin{aligned}\mathbf{F}_{\text{Wrist}} &= \mathbf{F}_{\text{Target}} \bullet \mathbf{I} \bullet \mathbf{F}_{\text{WT}}^{-1} \\ &= \mathbf{F}_{\text{Target}} \bullet \mathbf{F}_{\text{WT}}^{-1}\end{aligned}$$

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Notational Note

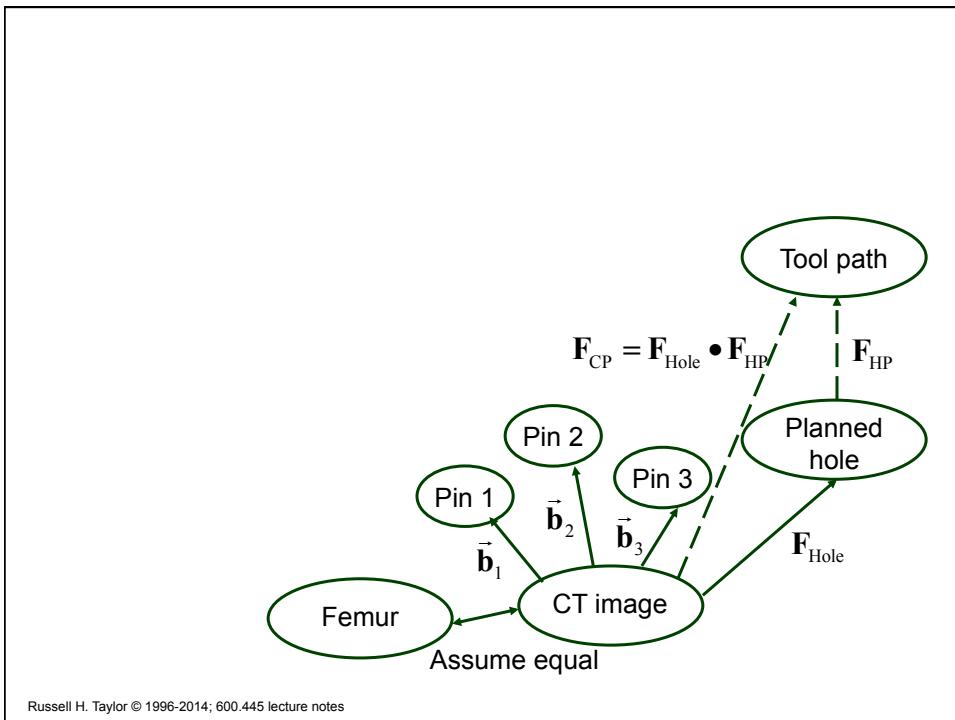
We use the notation $\mathbf{A} \bullet \mathbf{B}$ to represent composition or transformation. Where the context is clear, we may also use \mathbf{AB} for the same thing.

Question: What value of $\mathbf{F}_{\text{Wrist}}$ will make $\mathbf{F}_{\text{Tip}} = \mathbf{F}_{\text{Target}}$?

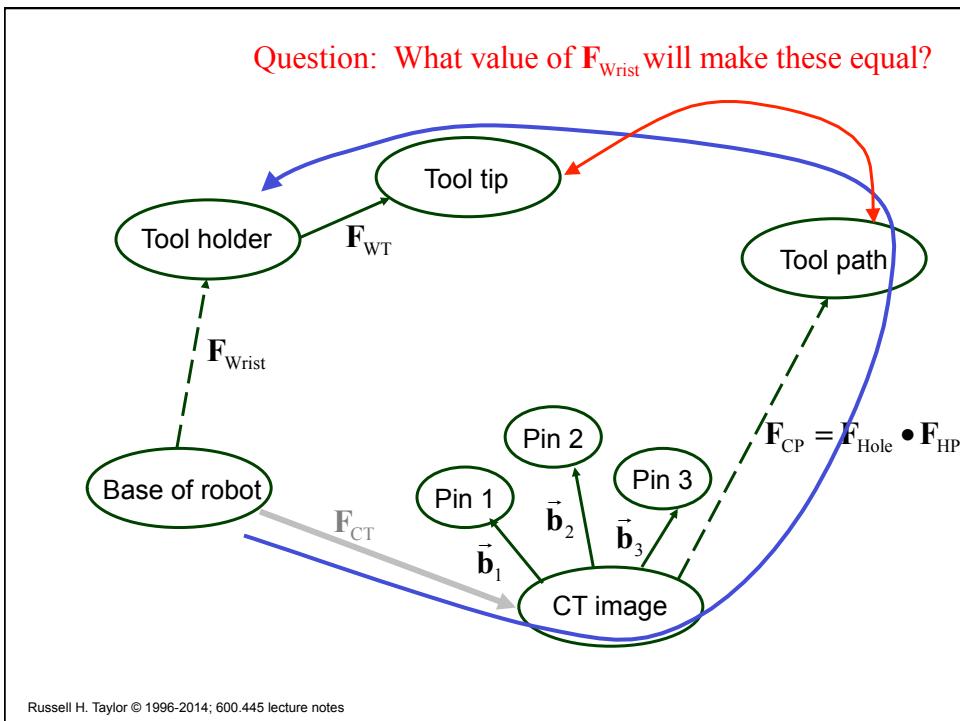
Answer:

$$\begin{aligned}\mathbf{F}_{\text{Wrist}} &= \mathbf{F}_{\text{Target}} \bullet \mathbf{I} \bullet \mathbf{F}_{\text{WT}}^{-1} \\ &= \mathbf{F}_{\text{Target}} \bullet \mathbf{F}_{\text{WT}}^{-1}\end{aligned}$$

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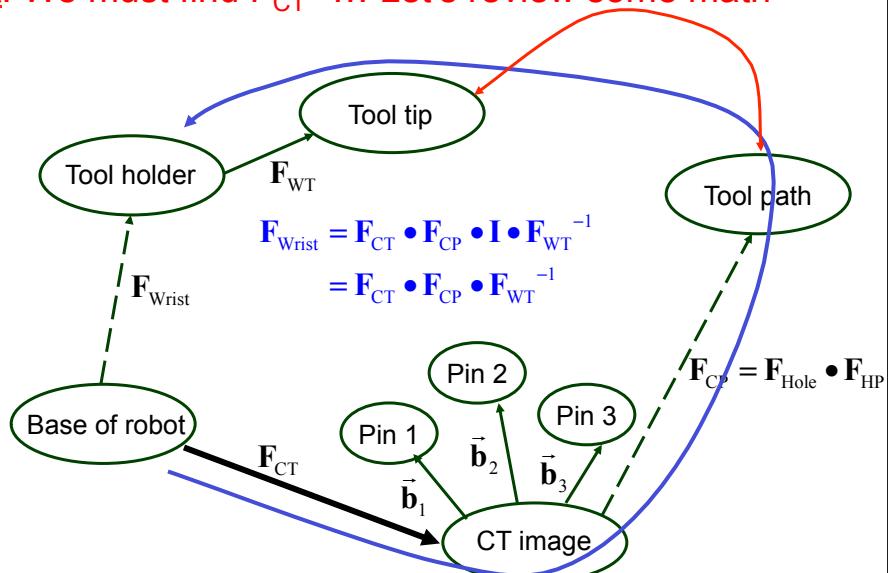


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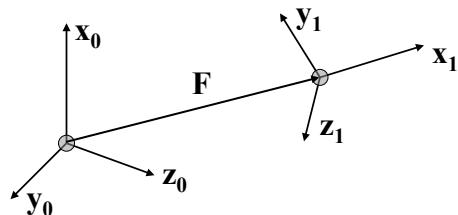
But: We must find F_{CT} ... Let's review some math



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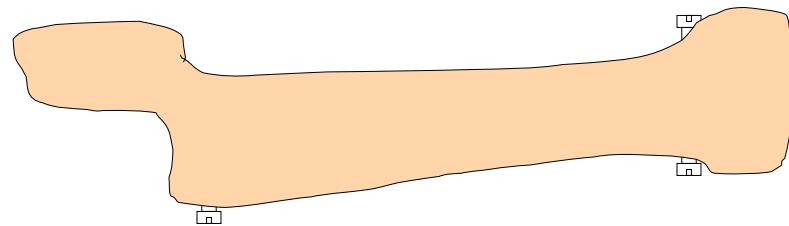
Coordinate Frame Transformation

$$F = [R, p]$$

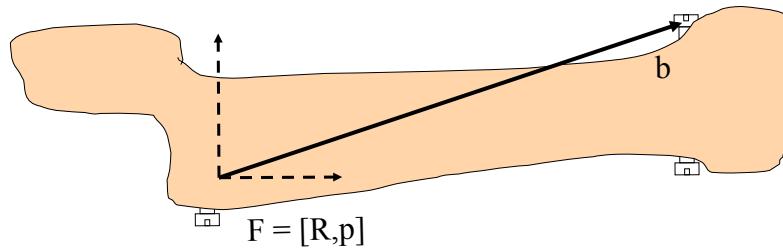


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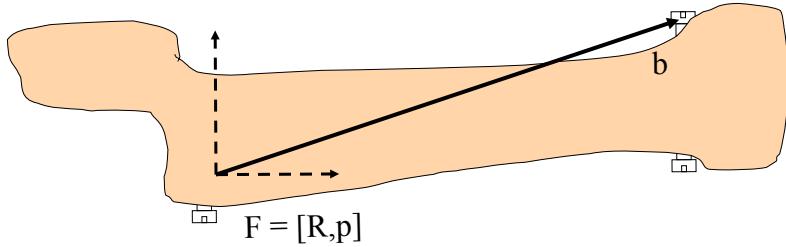
Slide acknowledgment: Sarah Graham and Andy Bzostek



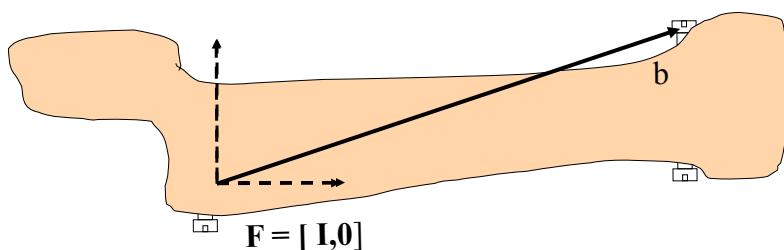
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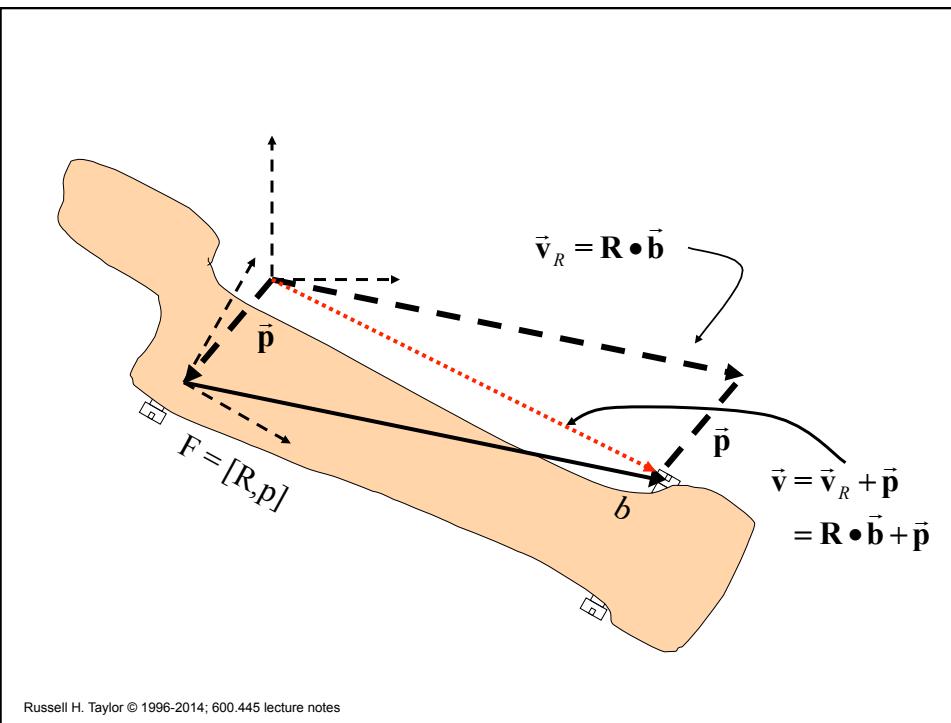
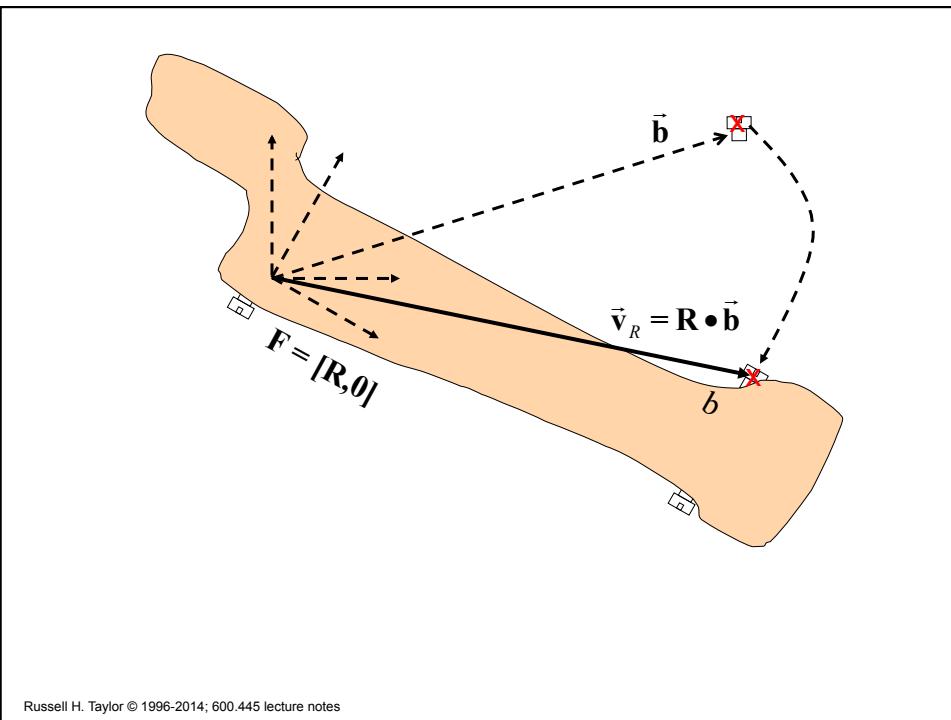
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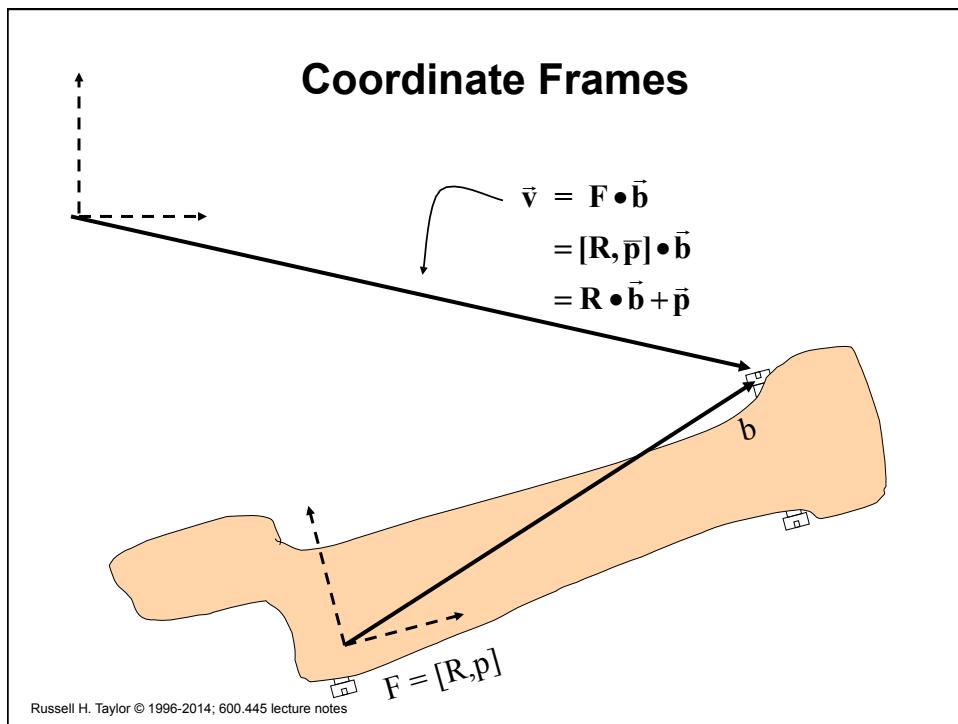


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Forward and Inverse Frame Transformations

Forward

$$F = [R, p]$$

$$v = F \bullet b$$

$$\begin{aligned} &= [R, p] \bullet b \\ &= R \bullet b + p \end{aligned}$$

Inverse

$$F^{-1} v = b$$

$$\begin{aligned} b &= R^{-1} \bullet (v - p) \\ &= R^{-1} \bullet v - R^{-1} \bullet p \end{aligned}$$

$$F^{-1} = [R^{-1}, -R^{-1} \bullet p]$$

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Composition

Assume $\mathbf{F}_1 = [\mathbf{R}_1, \vec{\mathbf{p}}_1]$, $\mathbf{F}_2 = [\mathbf{R}_2, \vec{\mathbf{p}}_2]$

Then

$$\begin{aligned}\mathbf{F}_1 \bullet \mathbf{F}_2 \bullet \vec{\mathbf{b}} &= \mathbf{F}_1 \bullet (\mathbf{F}_2 \bullet \vec{\mathbf{b}}) \\ &= \mathbf{F}_1 \bullet (\mathbf{R}_2 \bullet \vec{\mathbf{b}} + \vec{\mathbf{p}}_2) \\ &= [\mathbf{R}_1, \vec{\mathbf{p}}_1] \bullet (\mathbf{R}_2 \bullet \vec{\mathbf{b}} + \vec{\mathbf{p}}_2) \\ &= \mathbf{R}_1 \bullet (\mathbf{R}_2 \bullet \vec{\mathbf{b}} + \vec{\mathbf{p}}_2) + \vec{\mathbf{p}}_1 \\ &= \mathbf{R}_1 \bullet \mathbf{R}_2 \bullet \vec{\mathbf{b}} + \mathbf{R}_1 \bullet \vec{\mathbf{p}}_2 + \vec{\mathbf{p}}_1 \\ &= [\mathbf{R}_1 \bullet \mathbf{R}_2, \mathbf{R}_1 \bullet \vec{\mathbf{p}}_2 + \vec{\mathbf{p}}_1] \bullet \vec{\mathbf{b}}\end{aligned}$$

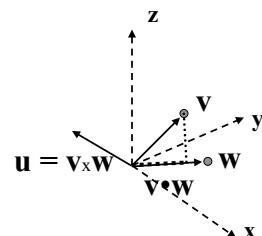
So

$$\begin{aligned}\mathbf{F}_1 \bullet \mathbf{F}_2 &= [\mathbf{R}_1, \vec{\mathbf{p}}_1] \bullet [\mathbf{R}_2, \vec{\mathbf{p}}_2] \\ &= [\mathbf{R}_1 \bullet \mathbf{R}_2, \mathbf{R}_1 \vec{\mathbf{p}}_2 + \vec{\mathbf{p}}_1]\end{aligned}$$

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Vectors

$$\begin{aligned}\mathbf{v}_{col} &= \begin{bmatrix} \mathbf{v}_x \\ \mathbf{v}_y \\ \mathbf{v}_z \end{bmatrix} \\ \mathbf{v}_{row} &= \begin{bmatrix} \mathbf{v}_x & \mathbf{v}_y & \mathbf{v}_z \end{bmatrix}\end{aligned}$$



$$\text{length : } \|\mathbf{v}\| = \sqrt{\mathbf{v}_x^2 + \mathbf{v}_y^2 + \mathbf{v}_z^2}$$

$$\text{dot product : } a = \mathbf{v} \cdot \mathbf{w} = (\mathbf{v}_x \mathbf{w}_x + \mathbf{v}_y \mathbf{w}_y + \mathbf{v}_z \mathbf{w}_z) = \|\mathbf{v}\| \|\mathbf{w}\| \cos \theta$$

$$\text{cross product : } \mathbf{u} = \mathbf{v} \times \mathbf{w} = \begin{bmatrix} \mathbf{v}_y \mathbf{w}_z - \mathbf{v}_z \mathbf{w}_y \\ \mathbf{v}_z \mathbf{w}_x - \mathbf{v}_x \mathbf{w}_z \\ \mathbf{v}_x \mathbf{w}_y - \mathbf{v}_y \mathbf{w}_x \end{bmatrix}, \|\mathbf{u}\| = \|\mathbf{v}\| \|\mathbf{w}\| \sin \theta$$

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Slide acknowledgment: Sarah Graham and Andy Bzostek

Matrix representation of cross product operator

Define

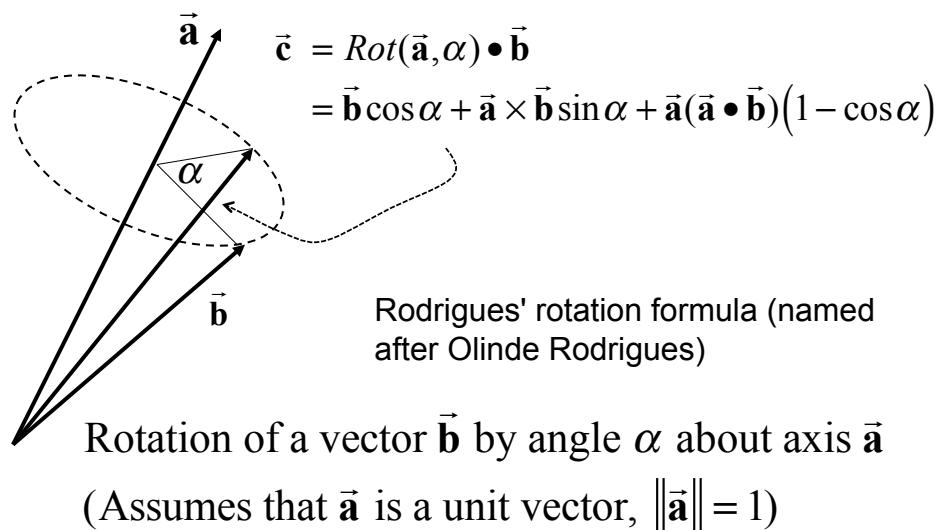
$$\hat{\vec{a}} \stackrel{\Delta}{=} skew(\vec{a}) \stackrel{\Delta}{=} \begin{bmatrix} 0 & -a_z & a_y \\ a_z & 0 & -a_x \\ -a_y & a_x & 0 \end{bmatrix}$$

Then

$$\vec{a} \times \vec{v} = skew(\vec{a}) \bullet \vec{v}$$

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Axis-angle Representations of Rotations



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Exponential representation

Consider a rotation about axis \vec{n} by angle θ . Then

$$e^{skew(\vec{n})\theta} = \mathbf{I} + \theta skew(\vec{n}) + \frac{\theta^2}{2!} skew(\vec{n})^2 + \dots$$

By doing some manipulation, you can show

$$\begin{aligned} Rot(\vec{n}, \theta) &= e^{skew(\vec{n})\theta} \\ &= \mathbf{I} + skew(\vec{n})\sin\theta + skew(\vec{n})^2(1 - \cos\theta) \\ &= \mathbf{I} + skew(\vec{n})\sin\theta + (\vec{n} \bullet \vec{n}^T - \mathbf{I})(1 - \cos\theta) \\ &= \mathbf{I}\cos\theta + skew(\vec{n})\sin\theta + \vec{n} \bullet \vec{n}^T(1 - \cos\theta) \end{aligned}$$

Note that for small θ , this reduces to

$$Rot(\vec{n}, \theta) \approx \mathbf{I} + skew(\theta\vec{n})$$

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Rotations: Some Notation

$Rot(\vec{a}, \alpha)$ = Rotation by angle α about axis \vec{a}

$\mathbf{R}_{\vec{a}}(\alpha)$ = Rotation by angle α about axis \vec{a}

$\mathbf{R}(\vec{a}) = Rot(\vec{a}, \|\vec{a}\|)$

$\mathbf{R}_{xyz}(\alpha, \beta, \gamma) = \mathbf{R}(\vec{x}, \alpha) \bullet \mathbf{R}(\vec{y}, \beta) \bullet \mathbf{R}(\vec{z}, \gamma)$

$\mathbf{R}_{zyx}(\alpha, \beta, \gamma) = \mathbf{R}(\vec{z}, \alpha) \bullet \mathbf{R}(\vec{y}, \beta) \bullet \mathbf{R}(\vec{x}, \gamma)$

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Rotations: A few useful facts

$$Rot(s\vec{a}, \alpha) \bullet \vec{a} = \vec{a} \quad \text{and} \quad \|Rot(\vec{a}, \alpha) \bullet \vec{b}\| = \|\vec{b}\|$$

$$Rot(\vec{a}, \alpha) = Rot(\hat{\mathbf{a}}, \alpha) \quad \text{where } \hat{\mathbf{a}} = \frac{\vec{a}}{\|\vec{a}\|}$$

$$Rot(\hat{\mathbf{a}}, \alpha) \bullet Rot(\hat{\mathbf{a}}, \beta) = Rot(\hat{\mathbf{a}}, \alpha + \beta)$$

$$Rot(\hat{\mathbf{a}}, \alpha)^{-1} = Rot(\hat{\mathbf{a}}, -\alpha)$$

$$Rot(\vec{a}, 0) \bullet \vec{b} = \vec{b} \quad \text{i.e., } Rot(\vec{a}, 0) = \mathbf{I}_{Rot} = \text{the identity rotation}$$

$$Rot(\hat{\mathbf{a}}, \alpha) \bullet \vec{b} = (\hat{\mathbf{a}} \bullet \vec{b}) \hat{\mathbf{a}} + Rot(\hat{\mathbf{a}}, \alpha) \bullet (\vec{b} - (\hat{\mathbf{a}} \bullet \vec{b}) \hat{\mathbf{a}})$$

$$Rot(\hat{\mathbf{a}}, \alpha) \bullet Rot(\hat{\mathbf{b}}, \beta) = Rot(\hat{\mathbf{b}}, \beta) \bullet Rot(Rot(\hat{\mathbf{b}}, -\beta) \bullet \hat{\mathbf{a}}, \alpha)$$

$$Rot(\hat{\mathbf{a}}, \alpha) \bullet \mathbf{R}_\beta = \mathbf{R}_\beta \bullet Rot(\mathbf{R}_\beta^{-1} \bullet \hat{\mathbf{a}}, \alpha)$$

$$\mathbf{R}_\alpha \bullet Rot(\hat{\mathbf{b}}, \beta) = Rot(\mathbf{R}_\alpha \bullet \hat{\mathbf{b}}, \beta) \bullet \mathbf{R}_\alpha$$

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Rotations: more facts

If $\vec{v} = [v_x, v_y, v_z]^T$ then a rotation $\mathbf{R} \bullet \vec{v}$ may be described in

terms of the effects of \mathbf{R} on orthogonal unit vectors, $\vec{\mathbf{e}}_x = [1, 0, 0]^T$,

$$\vec{\mathbf{e}}_y = [0, 1, 0]^T, \vec{\mathbf{e}}_z = [0, 0, 1]^T$$

$$\mathbf{R} \bullet \vec{v} = v_x \vec{\mathbf{r}}_x + v_y \vec{\mathbf{r}}_y + v_z \vec{\mathbf{r}}_z$$

where

$$\vec{\mathbf{r}}_x = \mathbf{R} \bullet \vec{\mathbf{e}}_x$$

$$\vec{\mathbf{r}}_y = \mathbf{R} \bullet \vec{\mathbf{e}}_y$$

$$\vec{\mathbf{r}}_z = \mathbf{R} \bullet \vec{\mathbf{e}}_z$$

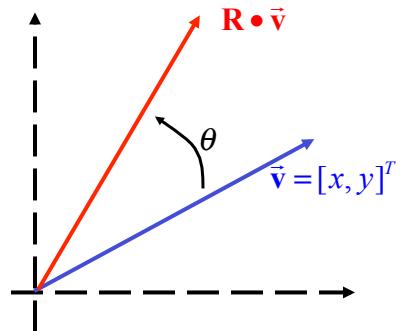
Note that rotation doesn't affect inner products

$$(\mathbf{R} \bullet \vec{b}) \bullet (\mathbf{R} \bullet \vec{c}) = \vec{b} \bullet \vec{c}$$

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Rotations in the plane

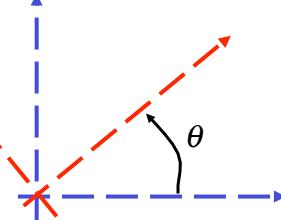
$$\begin{aligned}\mathbf{R} \bullet \begin{bmatrix} x \\ y \end{bmatrix} &= \begin{bmatrix} x \cos \theta - y \sin \theta \\ x \sin \theta + y \cos \theta \end{bmatrix} \\ &= \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \bullet \begin{bmatrix} x \\ y \end{bmatrix}\end{aligned}$$



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Rotations in the plane

$$\begin{aligned}\mathbf{R} \bullet \begin{bmatrix} \vec{\mathbf{e}}_x & \vec{\mathbf{e}}_y \end{bmatrix} &= \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \bullet \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \\ &= \begin{bmatrix} \mathbf{R} \bullet \vec{\mathbf{e}}_x & \mathbf{R} \bullet \vec{\mathbf{e}}_y \end{bmatrix} \\ &= \begin{bmatrix} \vec{\mathbf{r}}_x & \vec{\mathbf{r}}_y \end{bmatrix}\end{aligned}$$



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3D Rotation Matrices

$$\mathbf{R} \bullet \begin{bmatrix} \vec{\mathbf{e}}_x & \vec{\mathbf{e}}_y & \vec{\mathbf{e}}_z \end{bmatrix} = \begin{bmatrix} \mathbf{R} \bullet \vec{\mathbf{e}}_x & \mathbf{R} \bullet \vec{\mathbf{e}}_y & \mathbf{R} \bullet \vec{\mathbf{e}}_z \end{bmatrix}$$

$$= \begin{bmatrix} \vec{\mathbf{r}}_x & \vec{\mathbf{r}}_y & \vec{\mathbf{r}}_z \end{bmatrix}$$

$$\mathbf{R}^T \bullet \mathbf{R} = \begin{bmatrix} \hat{\mathbf{r}}_x^T \\ \hat{\mathbf{r}}_y^T \\ \hat{\mathbf{r}}_z^T \end{bmatrix} \bullet \begin{bmatrix} \vec{\mathbf{r}}_x & \vec{\mathbf{r}}_y & \vec{\mathbf{r}}_z \end{bmatrix}$$

$$= \begin{bmatrix} \vec{\mathbf{r}}_x^T \bullet \vec{\mathbf{r}}_x & \vec{\mathbf{r}}_x^T \bullet \vec{\mathbf{r}}_y & \vec{\mathbf{r}}_x^T \bullet \vec{\mathbf{r}}_z \\ \vec{\mathbf{r}}_y^T \bullet \vec{\mathbf{r}}_x & \vec{\mathbf{r}}_y^T \bullet \vec{\mathbf{r}}_y & \vec{\mathbf{r}}_y^T \bullet \vec{\mathbf{r}}_z \\ \vec{\mathbf{r}}_z^T \bullet \vec{\mathbf{r}}_x & \vec{\mathbf{r}}_z^T \bullet \vec{\mathbf{r}}_y & \vec{\mathbf{r}}_z^T \bullet \vec{\mathbf{r}}_z \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

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Properties of Rotation Matrices

Inverse of a Rotation Matrix equals its transpose:

$$\mathbf{R}^{-1} = \mathbf{R}^T$$

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

The Determinant of a Rotation matrix is equal to +1:

$$\det(\mathbf{R}) = +1$$

Any Rotation can be described by consecutive rotations about the three primary axes, x, y, and z:

$$\mathbf{R} = \mathbf{R}_{z,\theta} \mathbf{R}_{y,\phi} \mathbf{R}_{x,\psi}$$

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Canonical 3D Rotation Matrices

Note: Right-Handed Coordinate System

$$\mathbf{R}_{\bar{x}}(\theta) = \text{Rot}(\bar{x}, \theta) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos(\theta) & -\sin(\theta) \\ 0 & \sin(\theta) & \cos(\theta) \end{bmatrix}$$

$$\mathbf{R}_{\bar{y}}(\theta) = \text{Rot}(\bar{y}, \theta) = \begin{bmatrix} \cos(\theta) & 0 & \sin(\theta) \\ 0 & 1 & 0 \\ -\sin(\theta) & 0 & \cos(\theta) \end{bmatrix}$$

$$\mathbf{R}_{\bar{z}}(\theta) = \text{Rot}(\bar{z}, \theta) = \begin{bmatrix} \cos(\theta) & -\sin(\theta) & 0 \\ \sin(\theta) & \cos(\theta) & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

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Homogeneous Coordinates

- Widely used in graphics, geometric calculations
- Represent 3D vector as 4D quantity
- For our current purposes, we will keep the “scale” $s = 1$

$$\vec{\mathbf{v}} \equiv \begin{bmatrix} xs \\ ys \\ zs \\ s \end{bmatrix} \cong \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix}$$

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Representing Frame Transformations as Matrices

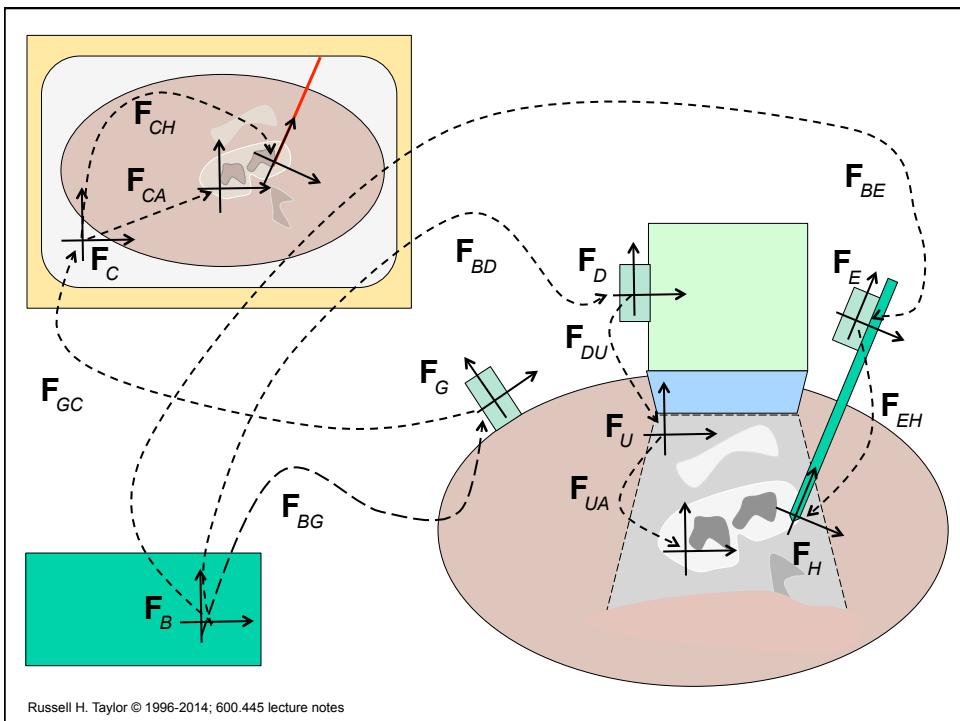
$$\mathbf{v} + \mathbf{p} \rightarrow \begin{bmatrix} 1 & 0 & 0 & \mathbf{p}_x \\ 0 & 1 & 0 & \mathbf{p}_y \\ 0 & 0 & 1 & \mathbf{p}_z \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \mathbf{v}_x \\ \mathbf{v}_y \\ \mathbf{v}_z \\ 1 \end{bmatrix} = \mathbf{P} \bullet \mathbf{v}$$

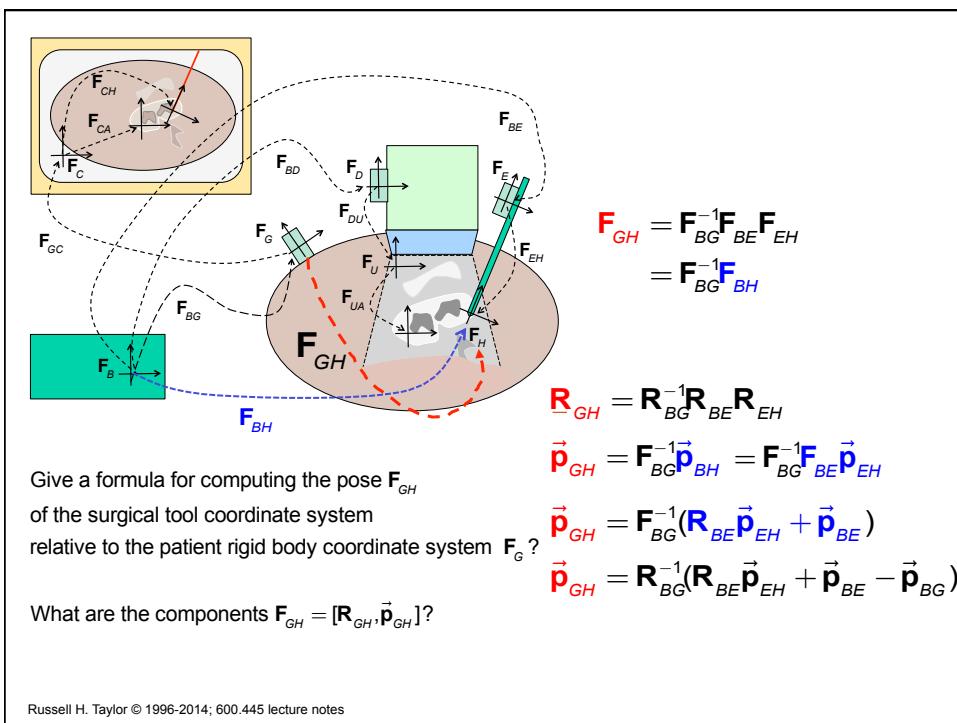
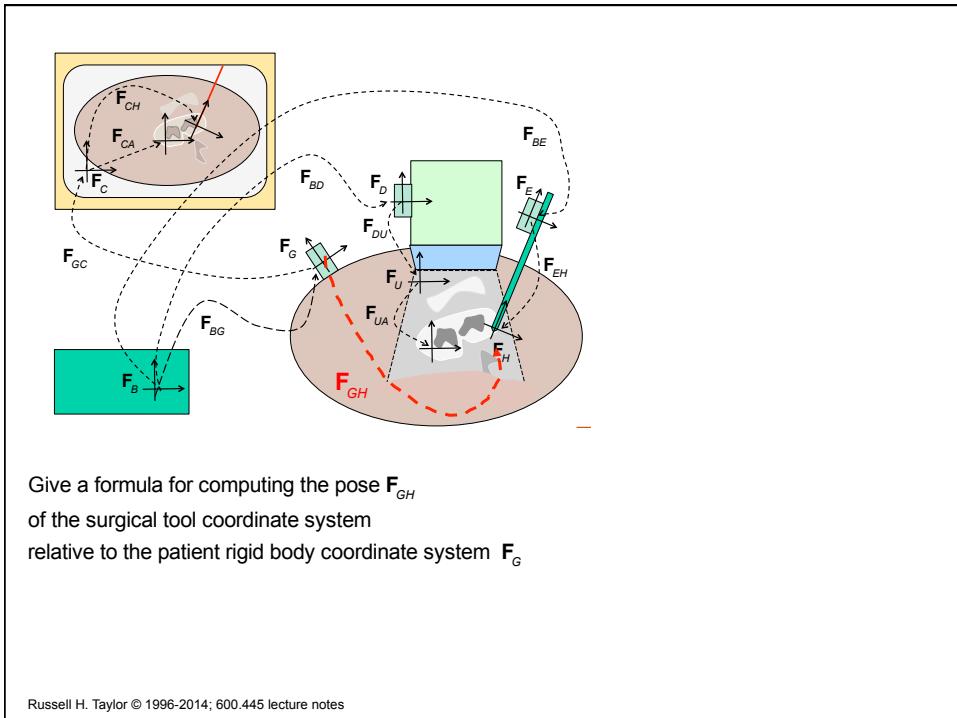
$$\mathbf{R} \bullet \mathbf{v} \rightarrow \begin{bmatrix} \mathbf{R} & \mathbf{0} \\ \mathbf{0} & 1 \end{bmatrix} \begin{bmatrix} \mathbf{v} \\ 1 \end{bmatrix}$$

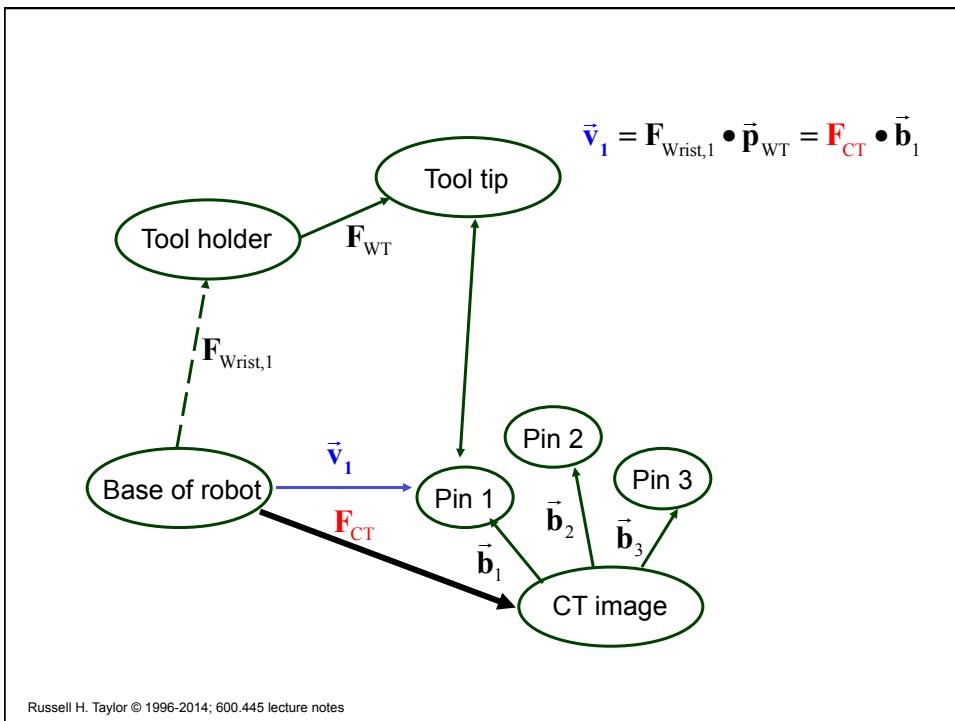
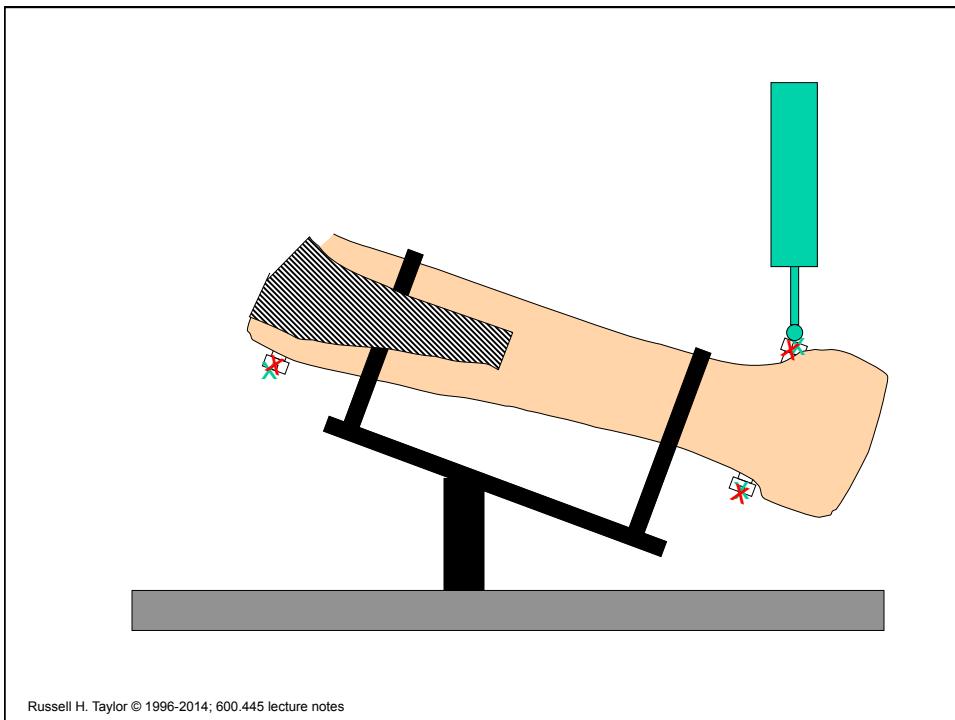
$$\mathbf{P} \bullet \mathbf{R} \rightarrow \begin{bmatrix} \mathbf{I} & \mathbf{p} \\ \mathbf{0} & 1 \end{bmatrix} \bullet \begin{bmatrix} \mathbf{R} & \mathbf{0} \\ \mathbf{0} & 1 \end{bmatrix} = \begin{bmatrix} \mathbf{R} & \mathbf{p} \\ \mathbf{0} & 1 \end{bmatrix} = [\mathbf{R}, \mathbf{p}] = \mathbf{F}$$

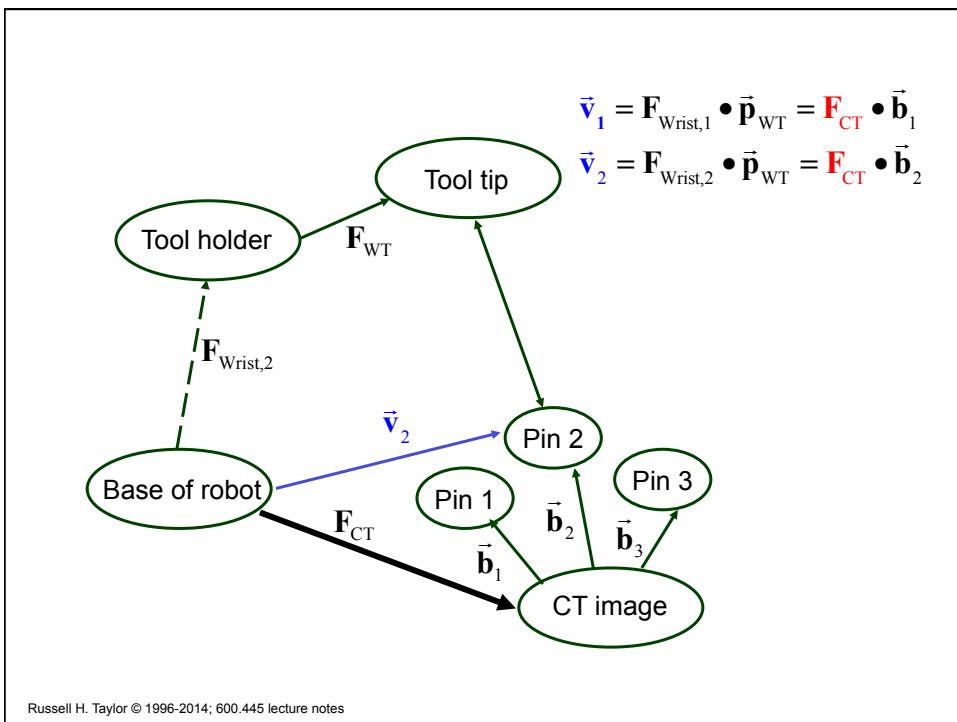
$$\mathbf{F} \bullet \mathbf{v} \rightarrow \begin{bmatrix} \mathbf{R} & \mathbf{p} \\ \mathbf{0} & 1 \end{bmatrix} \begin{bmatrix} \mathbf{v} \\ 1 \end{bmatrix} = \begin{bmatrix} (\mathbf{R} \bullet \mathbf{v}) + \mathbf{p} \\ 1 \end{bmatrix}$$

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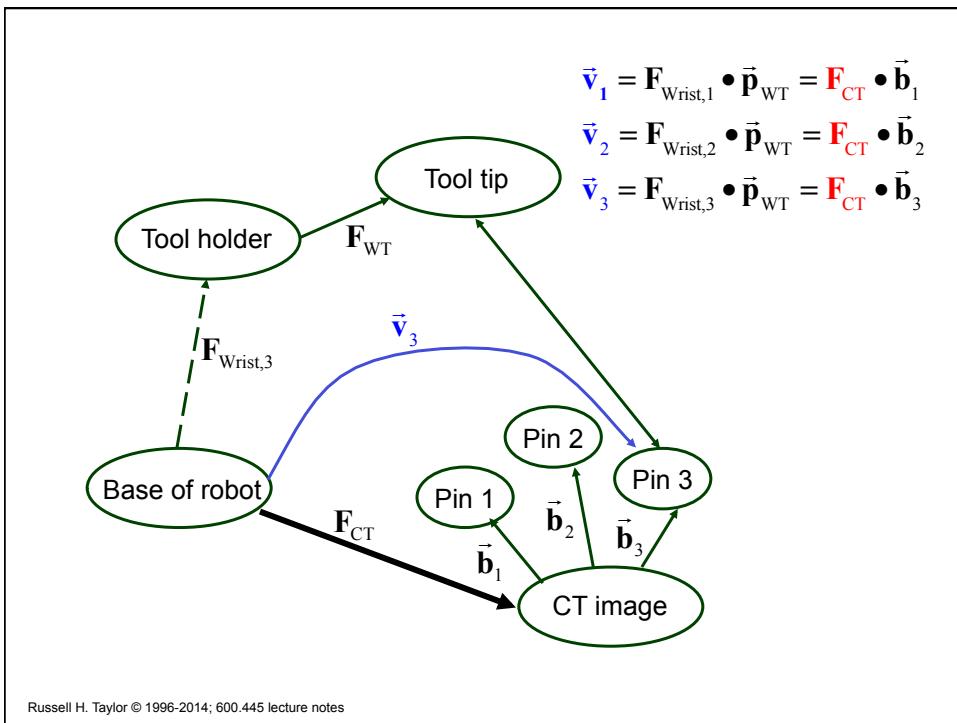






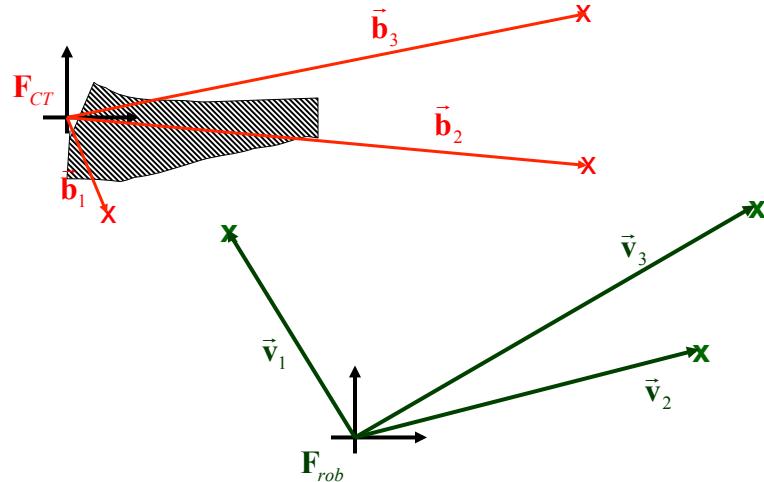


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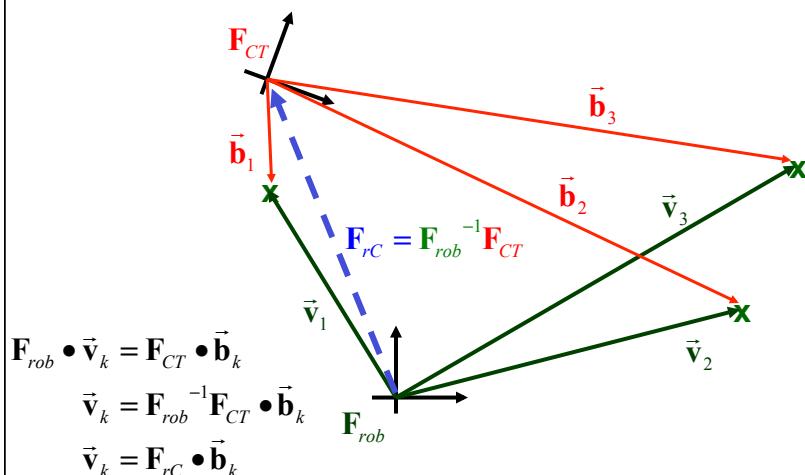
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Frame transformation from 3 point pairs



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Frame transformation from 3 point pairs



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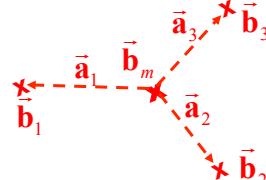
Frame transformation from 3 point pairs

$$\vec{v}_k = \mathbf{F}_{rC} \vec{b}_k = \mathbf{R}_{rC} \vec{b}_k + \vec{p}_{rC}$$

Define

$$\vec{v}_m = \frac{1}{3} \sum_1^3 \vec{v}_k \quad \vec{b}_m = \frac{1}{3} \sum_1^3 \vec{b}_k$$

$$\vec{u}_k = \vec{v}_k - \vec{v}_m \quad \vec{a}_k = \vec{b}_k - \vec{b}_m$$

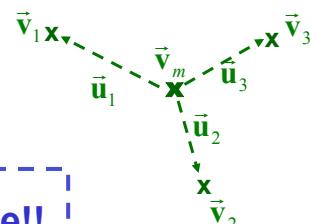


$$\mathbf{F}_{rC} \vec{a}_k = \mathbf{R}_{rC} \vec{a}_k + \vec{p}_{rC}$$

$$\mathbf{R}_{rC} \vec{a}_k + \vec{p}_{rC} = \mathbf{R}_{rC} (\vec{b}_k - \vec{b}_m) + \vec{p}_{rC}$$

$$\mathbf{R}_{rC} \vec{a}_k = \mathbf{R}_{rC} \vec{b}_k + \vec{p}_{rC} - \mathbf{R}_{rC} \vec{b}_m - \vec{p}_{rC}$$

$$\boxed{\begin{aligned} \mathbf{R}_{rC} \vec{a}_k &= \vec{v}_k - \vec{v}_m = \vec{u}_k \\ \vec{p}_{rC} &= \vec{v}_m - \mathbf{R}_{rC} \vec{b}_m \end{aligned}}$$



Solve These!!

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Rotation from multiple vector pairs

Given a system $\mathbf{R} \vec{a}_k = \vec{u}_k$ for $k = 1, \dots, n$ the problem is to estimate \mathbf{R} . This will require at least three such point pairs. Later in the course we will cover some good ways to solve this system. Here is a not-so-good way that will produce roughly correct answers:

Step 1: Form matrices $\mathbf{U} = [\vec{u}_1 \ \dots \ \vec{u}_n]$ and $\mathbf{A} = [\vec{a}_1 \ \dots \ \vec{a}_n]$

Step 2: Solve the system $\mathbf{RA} = \mathbf{U}$ for \mathbf{R} . E.g., by $\mathbf{R} = \mathbf{UA}^{-1}$

Step 3: Renormalize \mathbf{R} to guarantee $\mathbf{R}^T \mathbf{R} = \mathbf{I}$.

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Renormalizing Rotation Matrix

Given "rotation" matrix $\mathbf{R} = [\vec{\mathbf{r}}_x \mid \vec{\mathbf{r}}_y \mid \vec{\mathbf{r}}_z]$, modify it so $\mathbf{R}^T \mathbf{R} = \mathbf{I}$.

Step 1: $\vec{\mathbf{a}} = \vec{\mathbf{r}}_y \times \vec{\mathbf{r}}_z$

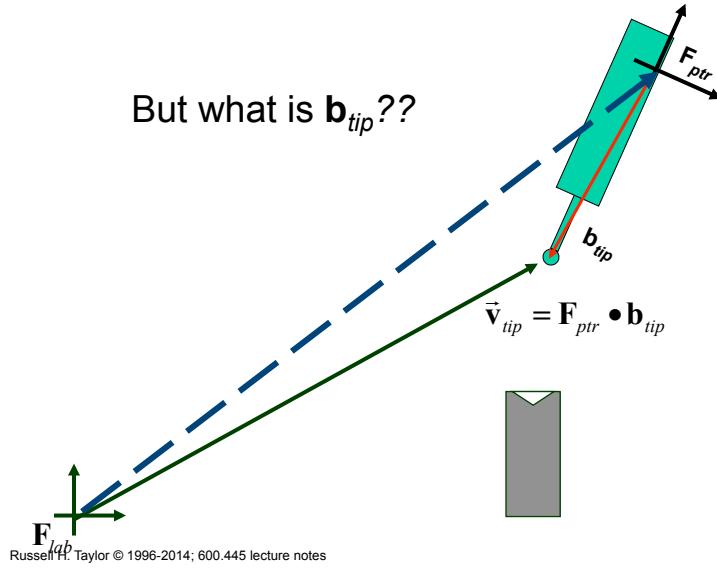
Step 2: $\vec{\mathbf{b}} = \vec{\mathbf{r}}_z \times \vec{\mathbf{a}}$

Step 3: $\mathbf{R}_{\text{normalized}} = \left[\frac{\vec{\mathbf{a}}}{\|\vec{\mathbf{a}}\|} \mid \frac{\vec{\mathbf{b}}}{\|\vec{\mathbf{b}}\|} \mid \frac{\vec{\mathbf{r}}_z}{\|\vec{\mathbf{r}}_z\|} \right]$

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Calibrating a pointer

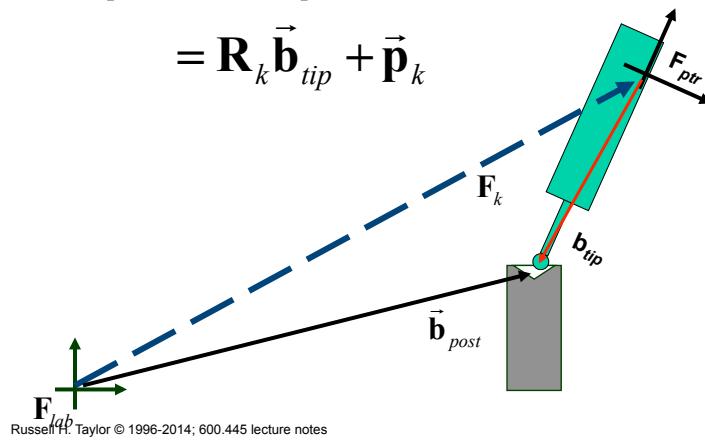
But what is \mathbf{b}_{tip} ??



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Calibrating a pointer

$$\begin{aligned}\vec{b}_{post} &= \mathbf{F}_k \vec{b}_{tip} \\ &= \mathbf{R}_k \vec{b}_{tip} + \vec{p}_k\end{aligned}$$



Calibrating a pointer

For each measurement k , we have

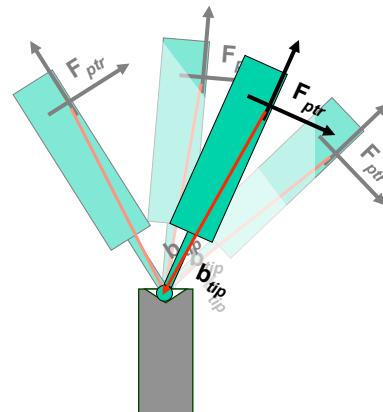
$$\vec{b}_{post} = \mathbf{R}_k \vec{b}_{tip} + \vec{p}_k$$

i.e.,

$$\mathbf{R}_k \vec{b}_{tip} - \vec{b}_{post} = -\vec{p}_k$$

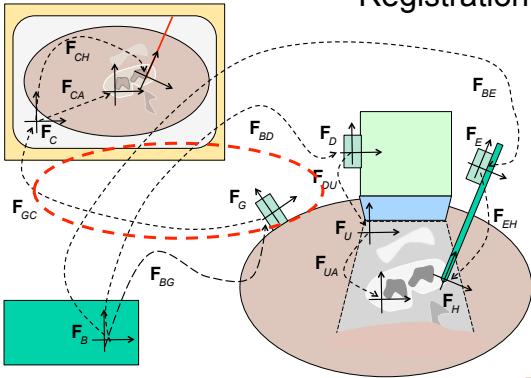
Set up a least squares problem

$$\left[\begin{array}{c|c} \vdots & \vdots \\ \hline \mathbf{R}_k & -\mathbf{I} \\ \vdots & \vdots \end{array} \right] \left[\begin{array}{c} \vec{b}_{tip} \\ \hline \vec{b}_{post} \end{array} \right] \cong \left[\begin{array}{c} \vdots \\ -\vec{p}_k \\ \vdots \end{array} \right]$$



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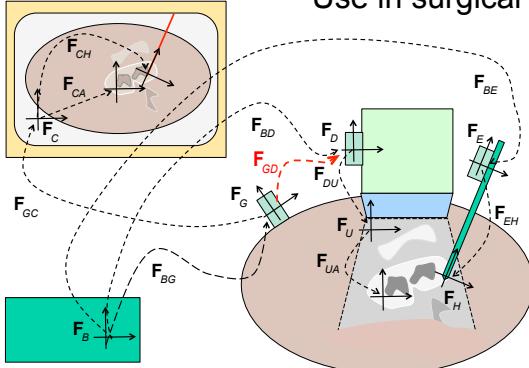
“Registration Transformations”



Given a coordinate system F_C and another coordinate system F_G (e.g., a CT scan and a tracked "rigid body" attached to the patient, and points \vec{c}_i in the coordinate system F_C and points \vec{g}_i in the coordinate system F_G , then the "registration transformation" F_{GC} between F_G and F_C one in which for $F_{GC} \vec{c}_i = \vec{g}_i$ if and only if \vec{c}_i and \vec{g}_i refer to the same or corresponding points.

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Use in surgical navigation



$$F_{BA} = F_{BD} F_{DU} F_{UA}$$

$$F_{BA} = F_{BG} F_{GC} F_{CA}$$

$$F_{BG} F_{GC} F_{CA} = F_{BD} F_{DU} F_{UA}$$

$$F_{CA} = (F_{BG} F_{GC})^{-1} F_{BD} F_{DU} F_{UA}$$

$$F_{GC} = F_{BG}^{-1} F_{BD} F_{DU} F_{UA} F_{CA}^{-1}$$

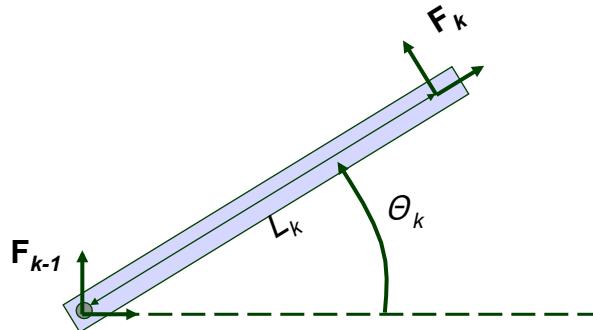
$$F_{GC} = F_{GD} F_{DU} F_{UA} F_{CA}^{-1}$$

where $F_{GD} = F_{BG}^{-1} F_{BD}$

If an anatomic structure is identified at pose F_{UA} in ultrasound image coordinates give the formula for computing the corresponding pose F_{CA} in CT coordinates

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Kinematic Links

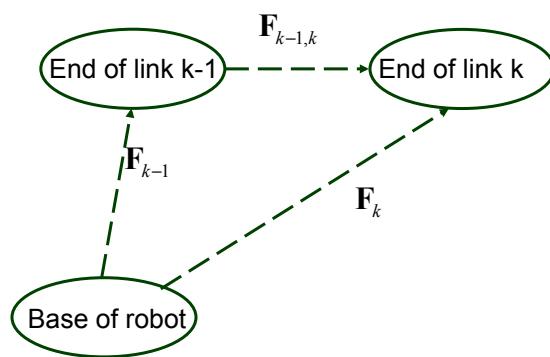


$$\mathbf{F}_k = \mathbf{F}_{k-1} \bullet \mathbf{F}_{k-1,k}$$

$$\begin{aligned} [\mathbf{R}_k, \vec{\mathbf{p}}_k] &= [\mathbf{R}_{k-1}, \vec{\mathbf{p}}_{k-1}] \bullet [\mathbf{R}_{k-1,k}, \vec{\mathbf{p}}_{k-1,k}] \\ &= [\mathbf{R}_{k-1}, \vec{\mathbf{p}}_{k-1}] \bullet [Rot(\vec{\mathbf{r}}_k, \theta_k), L_k Rot(\vec{\mathbf{r}}_k, \theta_k) \bullet \vec{\mathbf{x}}] \end{aligned}$$

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Kinematic Links



$$\mathbf{F}_k = \mathbf{F}_{k-1} \bullet \mathbf{F}_{k-1,k}$$

$$\begin{aligned} [\mathbf{R}_k, \vec{\mathbf{p}}_k] &= [\mathbf{R}_{k-1}, \vec{\mathbf{p}}_{k-1}] \bullet [\mathbf{R}_{k-1,k}, \vec{\mathbf{p}}_{k-1,k}] \\ &= [\mathbf{R}_{k-1}, \vec{\mathbf{p}}_{k-1}] \bullet [Rot(\vec{\mathbf{r}}_k, \theta_k), L_k Rot(\vec{\mathbf{r}}_k, \theta_k) \bullet \vec{\mathbf{x}}] \end{aligned}$$

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Kinematic Chains

$$\mathbf{F}_0 = [\mathbf{I}, \mathbf{0}]$$

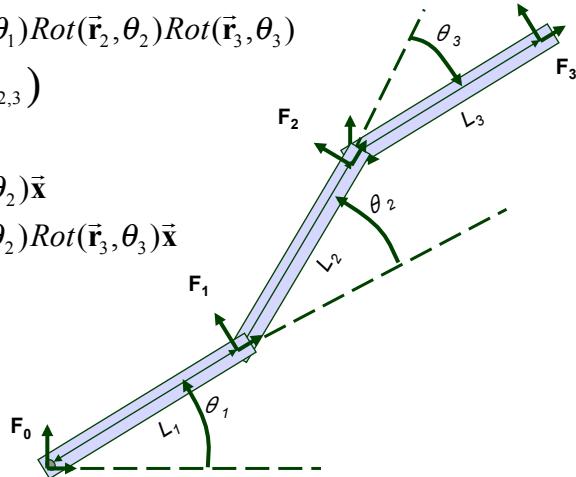
$$\mathbf{R}_3 = \mathbf{R}_{0,1} \mathbf{R}_{1,2} \mathbf{R}_{2,3} = \text{Rot}(\vec{\mathbf{r}}_1, \theta_1) \text{Rot}(\vec{\mathbf{r}}_2, \theta_2) \text{Rot}(\vec{\mathbf{r}}_3, \theta_3)$$

$$\vec{\mathbf{p}}_3 = \vec{\mathbf{p}}_{0,1} + \mathbf{R}_{0,1} (\vec{\mathbf{p}}_{1,2} + \mathbf{R}_{1,2} \vec{\mathbf{p}}_{2,3})$$

$$= L_1 \text{Rot}(\vec{\mathbf{r}}_1, \theta_1) \vec{\mathbf{x}}$$

$$+ L_2 \text{Rot}(\vec{\mathbf{r}}_1, \theta_1) \text{Rot}(\vec{\mathbf{r}}_2, \theta_2) \vec{\mathbf{x}}$$

$$+ L_3 \text{Rot}(\vec{\mathbf{r}}_1, \theta_1) \text{Rot}(\vec{\mathbf{r}}_2, \theta_2) \text{Rot}(\vec{\mathbf{r}}_3, \theta_3) \vec{\mathbf{x}}$$



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Kinematic Chains

$$\text{If } \vec{\mathbf{r}}_1 = \vec{\mathbf{r}}_2 = \vec{\mathbf{r}}_3 = \vec{\mathbf{z}},$$

$$\mathbf{R}_3 = \text{Rot}(\vec{\mathbf{z}}, \theta_1) \text{Rot}(\vec{\mathbf{z}}, \theta_2) \text{Rot}(\vec{\mathbf{z}}, \theta_3)$$

$$= \text{Rot}(\vec{\mathbf{z}}, \theta_1 + \theta_2 + \theta_3)$$

$$\vec{\mathbf{p}}_3 = \vec{\mathbf{p}}_{0,1} + \mathbf{R}_{0,1} (\vec{\mathbf{p}}_{1,2} + \mathbf{R}_{1,2} \vec{\mathbf{p}}_{2,3})$$

$$= L_1 \text{Rot}(\vec{\mathbf{z}}, \theta_1) \vec{\mathbf{x}}$$

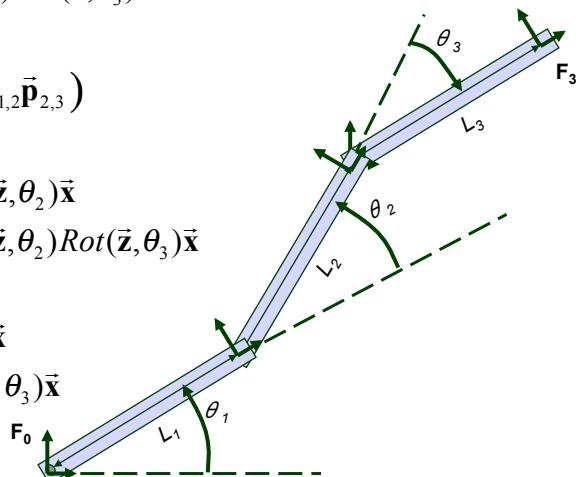
$$+ L_2 \text{Rot}(\vec{\mathbf{z}}, \theta_1) \text{Rot}(\vec{\mathbf{z}}, \theta_2) \vec{\mathbf{x}}$$

$$+ L_3 \text{Rot}(\vec{\mathbf{z}}, \theta_1) \text{Rot}(\vec{\mathbf{z}}, \theta_2) \text{Rot}(\vec{\mathbf{z}}, \theta_3) \vec{\mathbf{x}}$$

$$= L_1 \text{Rot}(\vec{\mathbf{z}}, \theta_1) \vec{\mathbf{x}}$$

$$+ L_2 \text{Rot}(\vec{\mathbf{z}}, \theta_1 + \theta_2) \vec{\mathbf{x}}$$

$$+ L_3 \text{Rot}(\vec{\mathbf{z}}, \theta_1 + \theta_2 + \theta_3) \vec{\mathbf{x}}$$



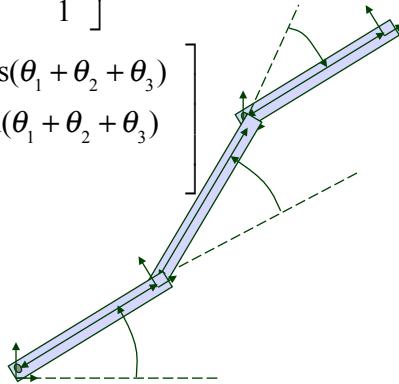
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Kinematic Chains

If $\vec{r}_1 = \vec{r}_2 = \vec{r}_3 = \vec{z}$,

$$\mathbf{R}_3 = \begin{bmatrix} \cos(\theta_1 + \theta_2 + \theta_3) & -\sin(\theta_1 + \theta_2 + \theta_3) & 0 \\ \sin(\theta_1 + \theta_2 + \theta_3) & \cos(\theta_1 + \theta_2 + \theta_3) & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$\vec{\mathbf{p}}_3 = \begin{bmatrix} L_1 \cos(\theta_1) + L_2 \cos(\theta_1 + \theta_2) + L_3 \cos(\theta_1 + \theta_2 + \theta_3) \\ L_1 \sin(\theta_1) + L_2 \sin(\theta_1 + \theta_2) + L_3 \sin(\theta_1 + \theta_2 + \theta_3) \\ 0 \end{bmatrix}$$



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“Small” Transformations

- A great deal of CIS is concerned with computing and using geometric information based on imprecise knowledge
- Similarly, one is often concerned with the effects of relatively small rotations and displacements
- Essentially, we will be using fairly straightforward linearizations to model these situations, but a specialized notation is often useful

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“Small” Frame Transformations

Represent a "small" pose shift consisting of a small rotation $\Delta\mathbf{R}$ followed by a small displacement $\Delta\vec{\mathbf{p}}$ as

$$\Delta\mathbf{F} = [\Delta\mathbf{R}, \Delta\vec{\mathbf{p}}]$$

Then

$$\Delta\mathbf{F} \bullet \vec{\mathbf{v}} = \Delta\mathbf{R} \bullet \vec{\mathbf{v}} + \Delta\vec{\mathbf{p}}$$

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Small Rotations

$\Delta\mathbf{R}$ = a small rotation

$\mathbf{R}_{\vec{\mathbf{a}}}(\Delta\alpha)$ = a rotation by a small angle $\Delta\alpha$ about axis $\vec{\mathbf{a}}$

$\text{Rot}(\vec{\mathbf{a}}, \|\vec{\mathbf{a}}\|) \bullet \vec{\mathbf{b}} \approx \vec{\mathbf{a}} \times \vec{\mathbf{b}} + \vec{\mathbf{b}}$ for $\|\vec{\mathbf{a}}\|$ sufficiently small

$\Delta\mathbf{R}(\vec{\mathbf{a}})$ = a rotation that is small enough so that any error introduced by this approximation is negligible

$$\Delta\mathbf{R}(\lambda\vec{\mathbf{a}}) \bullet \Delta\mathbf{R}(\mu\vec{\mathbf{b}}) \cong \Delta\mathbf{R}(\lambda\vec{\mathbf{a}} + \mu\vec{\mathbf{b}}) \quad (\text{Linearity for small rotations})$$

Exercise: Work out the linearity proposition by substitution

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Approximations to “Small” Frames

$$\begin{aligned}\Delta \mathbf{F}(\bar{\mathbf{a}}, \Delta \bar{\mathbf{p}}) &\approx [\Delta \mathbf{R}(\bar{\mathbf{a}}), \Delta \bar{\mathbf{p}}] \\ \Delta \mathbf{F}(\bar{\mathbf{a}}, \Delta \bar{\mathbf{p}}) \bullet \bar{\mathbf{v}} &= \Delta \mathbf{R}(\bar{\mathbf{a}}) \bullet \bar{\mathbf{v}} + \Delta \bar{\mathbf{p}} \\ &\approx \bar{\mathbf{v}} + \bar{\mathbf{a}} \times \bar{\mathbf{v}} + \Delta \bar{\mathbf{p}}\end{aligned}$$

$$\begin{aligned}\bar{\mathbf{a}} \times \bar{\mathbf{v}} &= \text{skew}(\bar{\mathbf{a}}) \bullet \bar{\mathbf{v}} \\ &= \begin{bmatrix} 0 & -a_z & a_y \\ a_z & 0 & -a_x \\ -a_y & a_x & 0 \end{bmatrix} \bullet \begin{bmatrix} v_x \\ v_y \\ v_z \end{bmatrix} \\ \text{skew}(\bar{\mathbf{a}}) \bullet \bar{\mathbf{a}} &= \bar{0}\end{aligned}$$

$$\begin{aligned}\Delta \mathbf{R}(\bar{\mathbf{a}}) &\approx \mathbf{I} + \text{skew}(\bar{\mathbf{a}}) \\ \Delta \mathbf{R}(\bar{\mathbf{a}})^{-1} &\approx \mathbf{I} - \text{skew}(\bar{\mathbf{a}}) = \mathbf{I} + \text{skew}(-\bar{\mathbf{a}})\end{aligned}$$

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Approximations to “Small” Frames

Notational NOTE:

We often use $\vec{\alpha}$ to represent a vector of small angles
and $\vec{\epsilon}$ to represent a vector of small displacements

In using these approximations, we typically ignore second order terms. I.e.,

$$\vec{\alpha}_A \vec{\alpha}_B \approx \bar{0}, \vec{\alpha}_A \vec{\epsilon}_B \approx \bar{0}, \vec{\epsilon}_A \vec{\epsilon}_B \approx \bar{0}, \text{etc.}$$

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Errors & sensitivity

Often, we do not have an accurate value for a transformation, so we need to model the error. We model this as a composition of a "nominal" frame and a small displacement

$$\mathbf{F}_{\text{actual}} = \mathbf{F}_{\text{nominal}} \bullet \Delta\mathbf{F}$$

Often, we will use the notation \mathbf{F}^* for $\mathbf{F}_{\text{actual}}$ and will just use \mathbf{F} for $\mathbf{F}_{\text{nominal}}$. Thus we may write something like

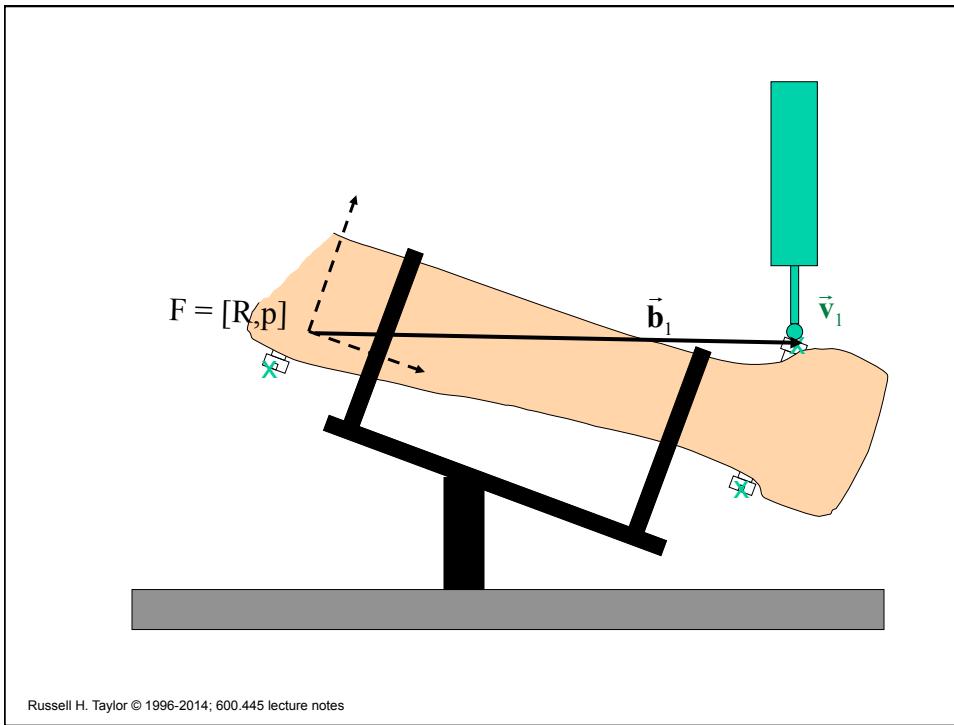
$$\mathbf{F}^* = \mathbf{F} \bullet \Delta\mathbf{F}$$

or (less often) $\mathbf{F}^* = \Delta\mathbf{F} \bullet \mathbf{F}$. We also use $\vec{\mathbf{v}}^* = \vec{\mathbf{v}} + \Delta\vec{\mathbf{v}}$, etc.

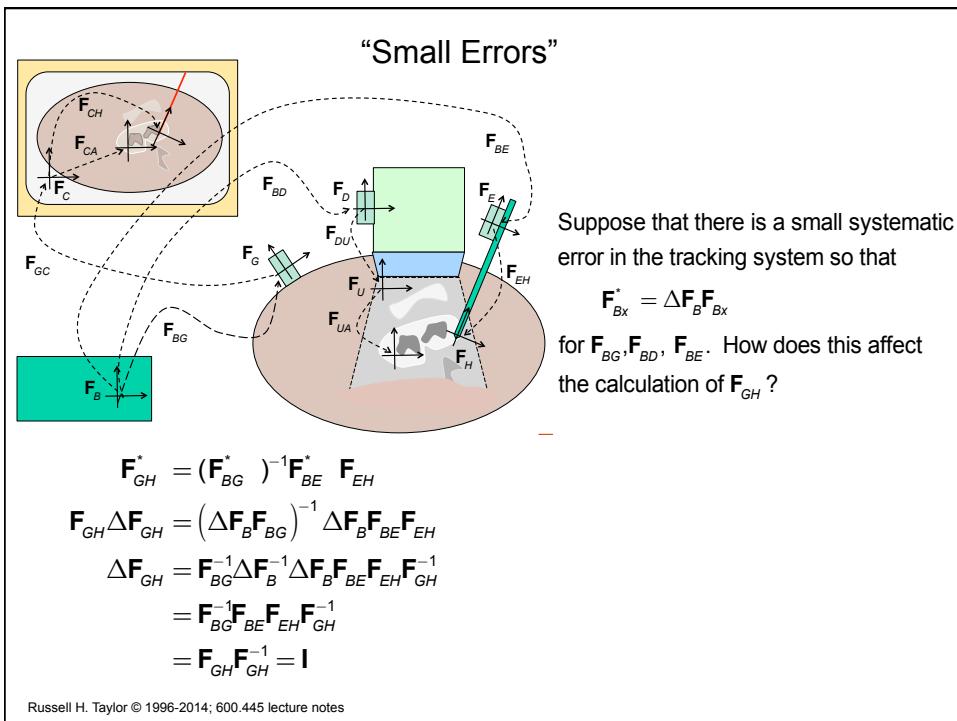
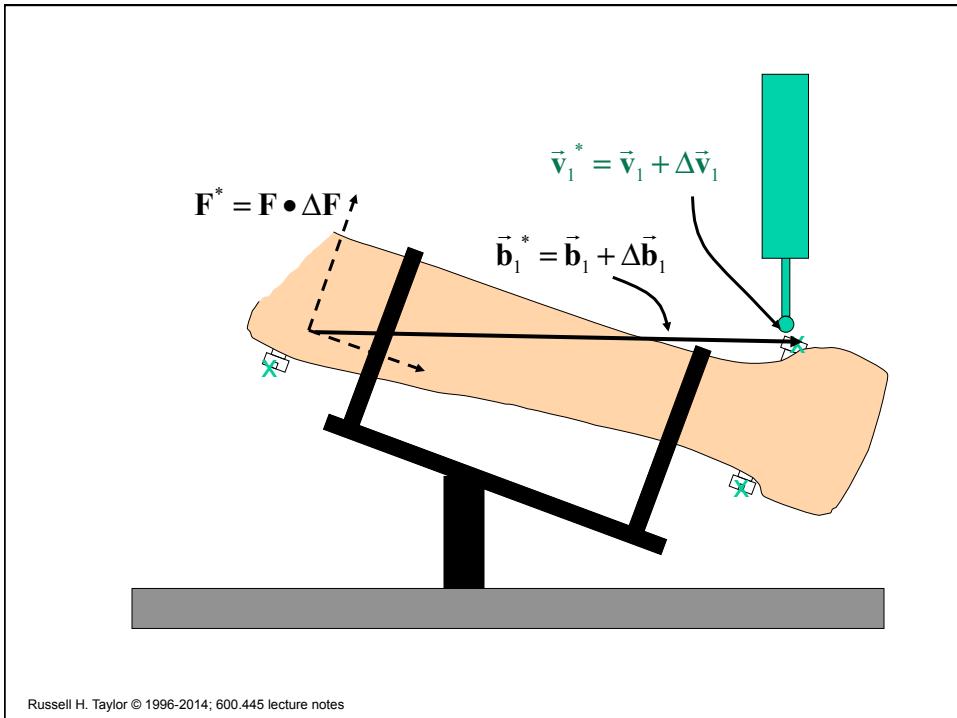
Thus, if we use the former form (error on the right), and have nominal relationship $\vec{\mathbf{v}} = \mathbf{F} \bullet \vec{\mathbf{b}}$, we get

$$\begin{aligned}\vec{\mathbf{v}}^* &= \mathbf{F}^* \bullet \vec{\mathbf{b}}^* \\ &= \mathbf{F} \bullet \Delta\mathbf{F} \bullet (\vec{\mathbf{b}} + \Delta\vec{\mathbf{b}})\end{aligned}$$

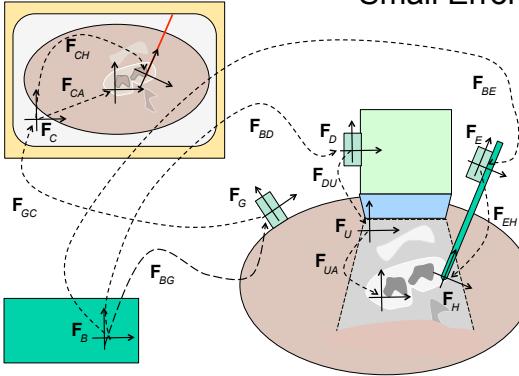
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“Small Errors”



Suppose that there are additional errors in the tracking of each tracker body so that

$$\mathbf{F}_{Bx}^* = \Delta \mathbf{F}_B \mathbf{F}_{Bx} \Delta \mathbf{F}_{Bx}$$

for $\mathbf{F}_{BG}, \mathbf{F}_{BD}, \mathbf{F}_{BE}$. How does this affect the calculation of \mathbf{F}_{GH} ?

$$\begin{aligned}\mathbf{F}_{GH}^* &= \mathbf{F}_{GH} \Delta \mathbf{F}_{GH} = (\mathbf{F}_{BG}^*)^{-1} \mathbf{F}_{BE}^* \mathbf{F}_{EH} \\ \Delta \mathbf{F}_{GH} &= \mathbf{F}_{GH}^{-1} (\Delta \mathbf{F}_B \mathbf{F}_{BG} \Delta \mathbf{F}_{BG})^{-1} (\Delta \mathbf{F}_B \mathbf{F}_{BE} \Delta \mathbf{F}_{BE}) \mathbf{F}_{EH} \\ \Delta \mathbf{F}_{GH} &= (\mathbf{F}_{BG}^{-1} \mathbf{F}_{BE} \mathbf{F}_{EH})^{-1} \Delta \mathbf{F}_{BG}^{-1} \mathbf{F}_{BG}^{-1} \Delta \mathbf{F}_B^{-1} \Delta \mathbf{F}_B \mathbf{F}_{BE} \Delta \mathbf{F}_{BE} \mathbf{F}_{BG} \mathbf{F}_{EH} \\ &= \mathbf{F}_{EH}^{-1} \mathbf{F}_{BE}^{-1} \mathbf{F}_{BG} \Delta \mathbf{F}_{BG}^{-1} \mathbf{F}_{BG}^{-1} \mathbf{F}_{BE} \Delta \mathbf{F}_{BE} \mathbf{F}_{EH}\end{aligned}$$

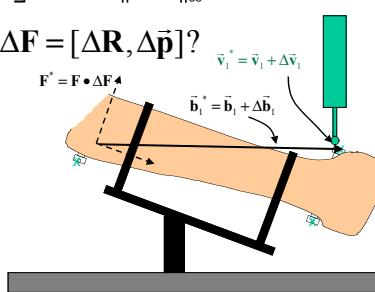
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Errors & Sensitivity

Suppose that we know nominal values for \mathbf{F} , $\bar{\mathbf{b}}$, and $\vec{\mathbf{v}}$ and that

$$[-\varepsilon, -\varepsilon, -\varepsilon]^T \leq \Delta \vec{\mathbf{v}}_1 \leq [\varepsilon, \varepsilon, \varepsilon]^T \quad (\text{i.e., } \|\Delta \vec{\mathbf{v}}_1\|_\infty \leq \varepsilon)$$

What does this tell us about $\Delta \mathbf{F} = [\Delta \mathbf{R}, \Delta \bar{\mathbf{p}}]$?



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Errors & Sensitivity

$$\begin{aligned}
 \vec{v}^* &= F^* \bullet \vec{b}^* \\
 &= F \bullet \Delta F \bullet (\vec{b} + \Delta \vec{b}) \\
 &= R \bullet (\Delta R(\vec{\alpha}) \bullet (\vec{b} + \Delta \vec{b}) + \Delta \vec{p}) + \vec{p} \\
 &\approx R \bullet (\vec{b} + \Delta \vec{b} + \vec{\alpha} \times \vec{b} + \vec{\alpha} \times \Delta \vec{b} + \Delta \vec{p}) + \vec{p} \\
 &= R \bullet \vec{b} + \vec{p} + R \bullet (\Delta \vec{b} + \vec{\alpha} \times \vec{b} + \vec{\alpha} \times \Delta \vec{b} + \Delta \vec{p}) \\
 &\approx \vec{v} + R \bullet (\Delta \vec{b} + \vec{\alpha} \times \vec{b} + \Delta \vec{p})
 \end{aligned}$$

if $\|\vec{\alpha} \times \Delta \vec{b}\| \leq \|\vec{\alpha}\| \|\Delta \vec{b}\|$ is negligible (it usually is)

so

$$\Delta \vec{v} = \vec{v}^* - \vec{v} \approx R \bullet (\Delta \vec{b} + \vec{\alpha} \times \vec{b} + \Delta \vec{p}) = R \bullet \Delta \vec{b} + \color{red}{R \bullet \vec{\alpha} \times \vec{b}} + R \bullet \Delta \vec{p}$$

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Digression: “rotation triple product”

Expressions like $R \bullet \vec{a} \times \vec{b}$ are linear in \vec{a} , but are not always convenient to work with. Often we would prefer something like $M(R, \vec{b}) \bullet \vec{a}$.

$$\begin{aligned}
 R \bullet \vec{a} \times \vec{b} &= -R \bullet \vec{b} \times \vec{a} \\
 &= R \bullet skew(-\vec{b}) \bullet \vec{a} \\
 &= [R \bullet skew(\vec{b})^T] \bullet \vec{a}
 \end{aligned}$$

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Digression: “rotation triple product”

Here are a few more useful facts:

$$\begin{aligned}\mathbf{R} \bullet (\vec{\mathbf{a}} \times \vec{\mathbf{b}}) &= (\mathbf{R} \bullet \vec{\mathbf{a}}) \times (\mathbf{R} \bullet \vec{\mathbf{b}}) \\ \vec{\mathbf{a}} \times (\mathbf{R} \bullet \vec{\mathbf{b}}) &= \mathbf{R} \bullet ((\mathbf{R}^{-1} \bullet \vec{\mathbf{a}}) \times \vec{\mathbf{b}})\end{aligned}$$

Consequently

$$\begin{aligned}skew(\vec{\mathbf{a}}) \bullet \mathbf{R} &= \mathbf{R} \bullet skew(\mathbf{R}^{-1} \bullet \vec{\mathbf{a}}) \\ \mathbf{R}^{-1} skew(\vec{\mathbf{a}}) \bullet \mathbf{R} &= skew(\mathbf{R}^{-1} \bullet \vec{\mathbf{a}})\end{aligned}$$

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Errors & Sensitivity

Previous expression was

$$\Delta \vec{\mathbf{v}}_1 \approx \mathbf{R} \bullet (\Delta \vec{\mathbf{b}}_1 + \vec{\mathbf{a}} \times \vec{\mathbf{b}} + \Delta \vec{\mathbf{p}}_1)$$

Substituting triple product and rearranging gives

$$\Delta \vec{\mathbf{v}}_1 \approx \left[\begin{array}{c|c|c} \mathbf{R} & \mathbf{R} & \mathbf{R} \bullet skew(-\vec{\mathbf{b}}) \end{array} \right] \bullet \left[\begin{array}{c} \Delta \vec{\mathbf{b}}_1 \\ \Delta \vec{\mathbf{p}} \\ \vec{\mathbf{a}} \end{array} \right]$$

So

$$\left[\begin{array}{c} -\varepsilon \\ -\varepsilon \\ -\varepsilon \end{array} \right] \leq \left[\begin{array}{c|c|c} \mathbf{R} & \mathbf{R} & \mathbf{R} \bullet skew(-\vec{\mathbf{b}}) \end{array} \right] \left[\begin{array}{c} \Delta \vec{\mathbf{b}}_1 \\ \Delta \vec{\mathbf{p}} \\ \vec{\mathbf{a}} \end{array} \right] \leq \left[\begin{array}{c} \varepsilon \\ \varepsilon \\ \varepsilon \end{array} \right]$$

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Errors & Sensitivity

Now, suppose we know that $|\Delta \vec{b}_1| \leq \beta$, this will give us a system of linear constraints

$$\begin{bmatrix} -\varepsilon \\ -\varepsilon \\ -\varepsilon \\ -\frac{\varepsilon}{\beta} \\ -\beta \\ -\beta \\ -\beta \end{bmatrix} \leq \begin{bmatrix} \mathbf{R} & | & \mathbf{R} & | & \mathbf{R} \bullet skew(-\vec{b}) \\ \hline \mathbf{I} & | & \mathbf{0} & | & \mathbf{0} \end{bmatrix} \begin{bmatrix} \Delta \vec{b}_1 \\ \Delta \vec{p}_1 \\ \vec{a} \end{bmatrix} \leq \begin{bmatrix} \varepsilon \\ \varepsilon \\ \varepsilon \\ \frac{\varepsilon}{\beta} \\ \beta \\ \beta \\ \beta \end{bmatrix}$$

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Error from frame composition

Consider $\mathbf{R}_1^* \mathbf{R}_2^* = \mathbf{R}_3^*$ where $\mathbf{R}_1^* = \mathbf{R}_1 \Delta \mathbf{R}_1$, $\mathbf{R}_2^* = \mathbf{R}_2 \Delta \mathbf{R}_2$, $\mathbf{R}_3^* = \mathbf{R}_3 \Delta \mathbf{R}_3$ and $\Delta \mathbf{R}_1 \approx \mathbf{I} + sk(\vec{\alpha}_1)$, $\Delta \mathbf{R}_2 \approx \mathbf{I} + sk(\vec{\alpha}_2)$, estimate $\Delta \mathbf{R}_3 \approx \mathbf{I} + sk(\vec{\alpha}_3)$

$$\begin{aligned} \mathbf{R}_1 \Delta \mathbf{R}_1 \mathbf{R}_2 \Delta \mathbf{R}_2 &= \mathbf{R}_1 \mathbf{R}_2 \Delta \mathbf{R}_3 \\ \mathbf{R}_1 (\mathbf{I} + sk(\vec{\alpha}_1)) \mathbf{R}_2 (\mathbf{I} + sk(\vec{\alpha}_2)) &\approx \mathbf{R}_1 \mathbf{R}_2 (\mathbf{I} + sk(\vec{\alpha}_3)) \\ (\mathbf{R}_1 \mathbf{R}_2)^{-1} \mathbf{R}_1 (\mathbf{I} + sk(\vec{\alpha}_1)) \mathbf{R}_2 (\mathbf{I} + sk(\vec{\alpha}_2)) &\approx \mathbf{I} + sk(\vec{\alpha}_3) \\ \cancel{\mathbf{R}_2^{-1} \mathbf{R}_1^{-1} \mathbf{R}_1} (\mathbf{I} + sk(\vec{\alpha}_1)) \mathbf{R}_2 (\mathbf{I} + sk(\vec{\alpha}_2)) &\approx \mathbf{I} + sk(\vec{\alpha}_3) \\ \mathbf{I} + \mathbf{R}_2^{-1} sk(\vec{\alpha}_1) \mathbf{R}_2 + sk(\vec{\alpha}_2) + \cancel{\mathbf{R}_2^{-1} sk(\vec{\alpha}_1) \mathbf{R}_2 sk(\vec{\alpha}_2)} &\approx \mathbf{I} + sk(\vec{\alpha}_3) \\ \mathbf{R}_2^{-1} sk(\vec{\alpha}_1) \mathbf{R}_2 + sk(\vec{\alpha}_2) &\approx sk(\vec{\alpha}_3) \end{aligned}$$

Since $\mathbf{R}^{-1} \bullet (\vec{a} \times \vec{R} \vec{b}) = (\mathbf{R}^{-1} \vec{a}) \times \vec{b}$ for all $\mathbf{R}, \vec{a}, \vec{b}$ we get $\mathbf{R}_2^{-1} sk(\vec{\alpha}_1) \mathbf{R}_2 = sk(\mathbf{R}_2^{-1} \vec{\alpha}_1)$

$$\begin{aligned} sk(\vec{\alpha}_3) &\approx sk(\mathbf{R}_2^{-1} \vec{\alpha}_1) + sk(\vec{\alpha}_2) = sk(\mathbf{R}_2^{-1} \vec{\alpha}_1 + \vec{\alpha}_2) \\ \vec{\alpha}_3 &\approx \mathbf{R}_2^{-1} \vec{\alpha}_1 + \vec{\alpha}_2 \end{aligned}$$

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Error from frame composition

Consider $\mathbf{F}_1^*\mathbf{F}_2^* = \mathbf{F}_3^*$ where $\mathbf{F}_1^* = \mathbf{F}_1\Delta\mathbf{F}_1$, $\mathbf{F}_2^* = \mathbf{F}_2\Delta\mathbf{F}_2$, $\mathbf{F}_3^* = \mathbf{F}_3\Delta\mathbf{F}_3$
and $\Delta\mathbf{F}_1 \approx [\mathbf{I} + sk(\vec{\alpha}_1), \vec{\varepsilon}_1]$, $\Delta\mathbf{F}_2 \approx [\mathbf{I} + sk(\vec{\alpha}_2), \vec{\varepsilon}_2]$,
estimate $\Delta\mathbf{F}_3 \approx [\mathbf{I} + sk(\vec{\alpha}_3), \vec{\varepsilon}_3]$

From before, we have $\vec{\alpha}_3 \approx \mathbf{R}_2^{-1}\vec{\alpha}_1 + \vec{\alpha}_2$. So now we just need $\vec{\varepsilon}_3$.

$$\begin{aligned}\vec{\mathbf{p}}_3^* &= \mathbf{R}_1^*\vec{\mathbf{p}}_2^* + \vec{\mathbf{p}}_1^* \\ \vec{\mathbf{p}}_3 + \vec{\varepsilon}_3 &\approx \mathbf{R}_1(\mathbf{I} + sk(\vec{\alpha}_1))(\vec{\mathbf{p}}_2 + \vec{\varepsilon}_2) + (\vec{\mathbf{p}}_1 + \vec{\varepsilon}_1) \\ &= \mathbf{R}_1\vec{\mathbf{p}}_2 + \mathbf{R}_1\vec{\varepsilon}_2 + \mathbf{R}_1 \cdot (\vec{\alpha}_1 \times \vec{\mathbf{p}}_2 + \vec{\alpha}_1 \times \vec{\varepsilon}_2) + \vec{\mathbf{p}}_1 + \vec{\varepsilon}_1 \\ &= \vec{\mathbf{p}}_3 + \mathbf{R}_1\vec{\varepsilon}_2 + \mathbf{R}_1 \cdot (\vec{\alpha}_1 \times \vec{\mathbf{p}}_2 + \vec{\alpha}_1 \times \vec{\varepsilon}_2) + \vec{\varepsilon}_1 \\ \vec{\varepsilon}_3 &\approx \mathbf{R}_1\vec{\varepsilon}_2 + \mathbf{R}_1 \cdot \vec{\alpha}_1 \times \vec{\mathbf{p}}_2 + \vec{\varepsilon}_1 \\ &= \mathbf{R}_1\vec{\varepsilon}_2 - \mathbf{R}_1 \cdot \vec{\mathbf{p}}_2 \times \vec{\alpha}_1 + \vec{\varepsilon}_1 \\ &= \vec{\varepsilon}_1 - \mathbf{R}_1 sk(\vec{\mathbf{p}}_2)\vec{\alpha}_1 + \mathbf{R}_1\vec{\varepsilon}_2\end{aligned}$$

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Inverse of frame transformation with errors

$$\begin{aligned}\mathbf{F}_i &= \mathbf{F}^{-1} = [\mathbf{R}^{-1}, -\mathbf{R}^{-1}\vec{\mathbf{p}}] \\ \mathbf{F}_i^* &= (\mathbf{F}\Delta\mathbf{F})^{-1} \\ \mathbf{F}_i\Delta\mathbf{F}_i &= [\mathbf{R}\Delta\mathbf{R}, \mathbf{R}\Delta\vec{\mathbf{p}} + \vec{\mathbf{p}}]^{-1} \\ &= [(\mathbf{R}\Delta\mathbf{R})^{-1}, -(\mathbf{R}\Delta\mathbf{R})^{-1}(\mathbf{R}\Delta\vec{\mathbf{p}} + \vec{\mathbf{p}})] \\ &= [\Delta\mathbf{R}^{-1}\mathbf{R}^{-1}, -\Delta\mathbf{R}^{-1}\mathbf{R}^{-1}(\mathbf{R}\Delta\vec{\mathbf{p}} + \vec{\mathbf{p}})] \\ &= [\Delta\mathbf{R}^{-1}\mathbf{R}^{-1}, -\Delta\mathbf{R}^{-1}\Delta\vec{\mathbf{p}} - \Delta\mathbf{R}^{-1}\vec{\mathbf{p}}] \\ \Delta\mathbf{F}_i &= (\mathbf{F}^{-1})^{-1}[\Delta\mathbf{R}^{-1}\mathbf{R}^{-1}, -\Delta\mathbf{R}^{-1}\Delta\vec{\mathbf{p}} - \Delta\mathbf{R}^{-1}\vec{\mathbf{p}}] \\ &= [\mathbf{R}, \vec{\mathbf{p}}] \cdot [\Delta\mathbf{R}^{-1}\mathbf{R}^{-1}, -\Delta\mathbf{R}^{-1}\Delta\vec{\mathbf{p}} - \Delta\mathbf{R}^{-1}\vec{\mathbf{p}}] \\ &= [\mathbf{R}\Delta\mathbf{R}^{-1}\mathbf{R}^{-1}, -\mathbf{R}\Delta\mathbf{R}^{-1}\Delta\vec{\mathbf{p}} - \mathbf{R}\Delta\mathbf{R}^{-1}\vec{\mathbf{p}} + \vec{\mathbf{p}}]\end{aligned}$$

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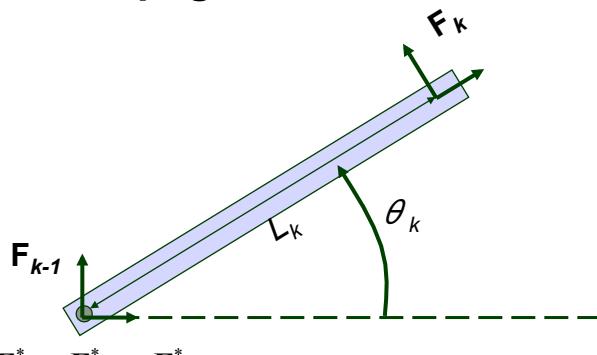
Inverse of frame transformation with errors

Suppose we know that $\Delta\mathbf{R}$ is "small", i.e., $\Delta\mathbf{R} \approx \mathbf{I} + sk(\vec{\alpha})$, and for notational convenience we write $\Delta\vec{p} = \vec{\varepsilon}$, we get

$$\begin{aligned}
 \Delta\mathbf{R}_i &= \mathbf{R}\Delta\mathbf{R}^{-1}\mathbf{R}^{-1} \approx \mathbf{R}(\mathbf{I} + sk(\vec{\alpha}))^{-1}\mathbf{R}^{-1} \\
 &\approx \mathbf{R}(\mathbf{I} - sk(\vec{\alpha}))\mathbf{R}^{-1} \\
 &= \mathbf{R}\mathbf{R}^{-1} - \mathbf{R}sk(\vec{\alpha})\mathbf{R}^{-1} \\
 &= \mathbf{I} - \mathbf{R}sk(\vec{\alpha})\mathbf{R}^{-1} \\
 &= \mathbf{I} - sk\left((\mathbf{R}^{-1})^{-1}\vec{\alpha}\right) = \mathbf{I} - sk(\mathbf{R}\vec{\alpha}) \\
 \Delta\vec{p}_i &= -\mathbf{R}\Delta\mathbf{R}^{-1}\Delta\vec{p} - \mathbf{R}\Delta\mathbf{R}^{-1}\mathbf{R}^{-1}\vec{p} + \vec{p} \\
 &\approx -\mathbf{R}(\mathbf{I} - sk(\vec{\alpha}))\vec{\varepsilon} - (\mathbf{I} - sk(\mathbf{R}\vec{\alpha}))\vec{p} + \vec{p} \\
 &= -\mathbf{R}\vec{\varepsilon} + \mathbf{R}(\vec{\alpha} \times \vec{\varepsilon}) - \vec{p} + (\mathbf{R}\vec{\alpha}) \times \vec{p} + \vec{p} \\
 &\approx -\mathbf{R}\vec{\varepsilon} + (\mathbf{R}\vec{\alpha}) \times \vec{p} = -\mathbf{R}\vec{\varepsilon} - \vec{p} \times (\mathbf{R}\vec{\alpha}) = -\mathbf{R}\vec{\varepsilon} - \mathbf{R}(\vec{p} \times (\mathbf{R}^{-1}\mathbf{R}\vec{\alpha})) \\
 &= -\mathbf{R}(\vec{\varepsilon} + \vec{p} \times \vec{\alpha}) = -\mathbf{R}\vec{\varepsilon} + \mathbf{R}sk(\vec{p})\vec{\alpha}
 \end{aligned}$$

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Error Propagation in Chains



$$\begin{aligned}
 \mathbf{F}_k^* &= \mathbf{F}_{k-1}^* \bullet \mathbf{F}_{k-1,k}^* \\
 \mathbf{F}_k \Delta \mathbf{F}_k &= \mathbf{F}_{k-1} \Delta \mathbf{F}_{k-1} \mathbf{F}_{k-1,k} \Delta \mathbf{F}_{k-1,k} \\
 \Delta \mathbf{F}_k &= (\mathbf{F}_k^{-1} \mathbf{F}_{k-1}) \Delta \mathbf{F}_{k-1} \mathbf{F}_{k-1,k} \Delta \mathbf{F}_{k-1,k} \\
 &= (\mathbf{F}_{k-1,k}^{-1} \Delta \mathbf{F}_{k-1} \mathbf{F}_{k-1,k}) \Delta \mathbf{F}_{k-1,k}
 \end{aligned}$$

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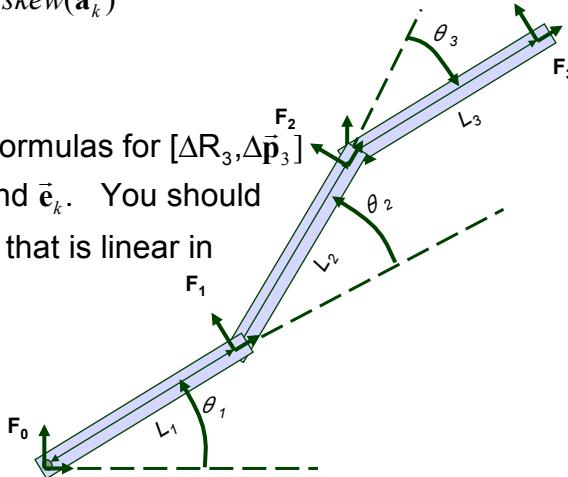
Exercise

Suppose that you have

$$\Delta \mathbf{R}_{k-1,k} = \Delta \mathbf{R}(\vec{\mathbf{a}}_k) \equiv \mathbf{I} + skew(\vec{\mathbf{a}}_k)$$

$$\Delta \vec{\mathbf{p}}_{k-1,k} = \vec{\mathbf{e}}_k$$

Work out approximate formulas for $[\Delta \mathbf{R}_3, \Delta \vec{\mathbf{p}}_3]$ in terms of $L_k, \vec{\mathbf{r}}_k, \theta_k, \vec{\mathbf{a}}_k$ and $\vec{\mathbf{e}}_k$. You should come up with a formula that is linear in $L_k, \vec{\mathbf{a}}_k$, and $\vec{\mathbf{e}}_k$.



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Exercise

Suppose we want to know error in $\mathbf{F}_{0,3} = \mathbf{F}_0^{-1}\mathbf{F}_3$

$$\mathbf{F}_{0,3} = \mathbf{F}_0^{-1}\mathbf{F}_0\mathbf{F}_{0,1}\mathbf{F}_{1,2}\mathbf{F}_{2,3}$$

$$\mathbf{F}_{0,3}^* = \mathbf{F}_{0,1}^{-1}\mathbf{F}_{1,2}^{-1}\mathbf{F}_{2,3}^*$$

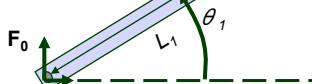
$$\mathbf{F}_{0,3}\Delta\mathbf{F}_{0,3} = \mathbf{F}_{0,1}^{-1}\mathbf{F}_{1,2}^{-1}\mathbf{F}_{2,3}^*$$

$$\Delta\mathbf{F}_3 = \mathbf{F}_{0,3}^{-1}\mathbf{F}_{0,1}\Delta\mathbf{F}_{0,1}\mathbf{F}_{1,2}\Delta\mathbf{F}_{1,2}\mathbf{F}_{2,3}\Delta\mathbf{F}_{2,3}$$

$$= \mathbf{F}_{2,3}^{-1}\mathbf{F}_{1,2}^{-1}\mathbf{F}_{0,1}^{-1}\Delta\mathbf{F}_{0,1}\mathbf{F}_{1,2}\Delta\mathbf{F}_{1,2}\mathbf{F}_{2,3}\Delta\mathbf{F}_{2,3}$$

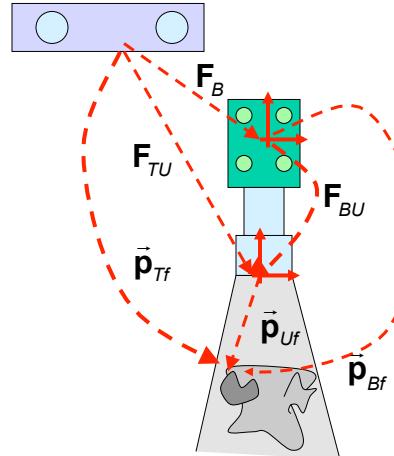
$$= \mathbf{F}_{2,3}^{-1}\mathbf{F}_{1,2}^{-1}\Delta\mathbf{F}_{0,1}\mathbf{F}_{1,2}\Delta\mathbf{F}_{1,2}\mathbf{F}_{2,3}\Delta\mathbf{F}_{2,3}$$

Now substitute and simplify



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Another Example



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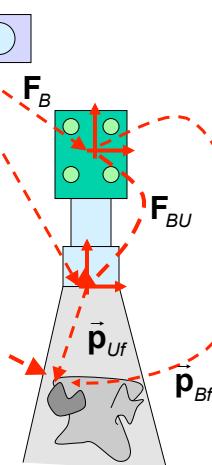
Another Example

$$\begin{aligned}\vec{p}_{Tf} &= \mathbf{F}_{TU} \bullet \vec{p}_{Uf} \\ \mathbf{F}_{TU} &= \mathbf{F}_B \bullet \mathbf{F}_{BU} \\ &= [\mathbf{R}_B \bullet \mathbf{R}_{BU}, \mathbf{R}_B \bullet \vec{p}_{BU} + \vec{p}_B] \\ \vec{p}_{Tf} &= \mathbf{R}_B \bullet \mathbf{R}_{BU} \bullet \vec{p}_{Uf} + \mathbf{R}_B \bullet \vec{p}_{BU} + \vec{p}_B\end{aligned}$$

Also

$$\begin{aligned}\vec{p}_{Tf} &= \mathbf{F}_B \bullet \vec{p}_{Bf} \\ \vec{p}_{Bf} &= \mathbf{F}_{BU} \bullet \vec{p}_{Uf} \\ &= \mathbf{R}_{BU} \bullet \vec{p}_{Uf} + \vec{p}_{BU} \\ \vec{p}_{Tf} &= \mathbf{R}_B \bullet \mathbf{R}_{BU} \bullet \vec{p}_{Uf} + \mathbf{R}_B \bullet \vec{p}_{BU} + \vec{p}_B\end{aligned}$$

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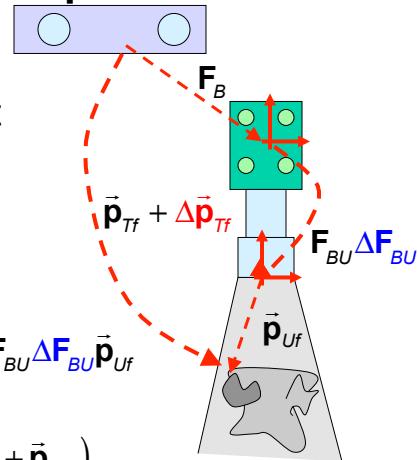


Another Example

Suppose that the track body to US calibration is not perfect

$$\begin{aligned}\mathbf{F}_{BU}^* &= \mathbf{F}_{BU} \Delta \mathbf{F}_{BU} \\ &= [\mathbf{R}_{BU} \Delta \mathbf{R}_{BU}, \mathbf{R}_{BU} \Delta \vec{\mathbf{p}}_{BU} + \vec{\mathbf{p}}_{BU}]\end{aligned}$$

$$\begin{aligned}\vec{\mathbf{p}}_{Bf}^* &= \mathbf{F}_{BU}^* \bullet \vec{\mathbf{p}}_{Uf} \quad \text{i.e.,} \quad \vec{\mathbf{p}}_{Bf} + \Delta \vec{\mathbf{p}}_{Bf} = \mathbf{F}_{BU} \Delta \mathbf{F}_{BU} \vec{\mathbf{p}}_{Uf} \\ \Delta \vec{\mathbf{p}}_{Bf} &= \mathbf{F}_{BU} \Delta \mathbf{F}_{BU} \vec{\mathbf{p}}_{Uf} - \vec{\mathbf{p}}_{Bf} \\ &= \mathbf{F}_{BU} (\Delta \mathbf{R}_{BU} \vec{\mathbf{p}}_{Uf} + \Delta \vec{\mathbf{p}}_{BU}) - (\mathbf{R}_{BU} \vec{\mathbf{p}}_{Uf} + \vec{\mathbf{p}}_{BU}) \\ &= \mathbf{R}_{BU} \Delta \mathbf{R}_{BU} \vec{\mathbf{p}}_{Uf} + \mathbf{R}_{BU} \Delta \vec{\mathbf{p}}_{BU} + \vec{\mathbf{p}}_{BU} - \mathbf{R}_{BU} \vec{\mathbf{p}}_{Uf} - \vec{\mathbf{p}}_{BU} \\ &= \mathbf{R}_{BU} \Delta \mathbf{R}_{BU} \vec{\mathbf{p}}_{Uf} + \mathbf{R}_{BU} \Delta \vec{\mathbf{p}}_{BU} - \mathbf{R}_{BU} \vec{\mathbf{p}}_{Uf}\end{aligned}$$



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Another Example

Continuing ...

$$\begin{aligned}\Delta \vec{\mathbf{p}}_{Bf} &= \mathbf{R}_{BU} \Delta \mathbf{R}_{BU} \vec{\mathbf{p}}_{Uf} + \mathbf{R}_{BU} \Delta \vec{\mathbf{p}}_{BU} - \mathbf{R}_{BU} \vec{\mathbf{p}}_{Uf} \\ &\approx \mathbf{R}_{BU} (\mathbf{I} + \text{skew}(\vec{\alpha}_{BU})) \vec{\mathbf{p}}_{Uf} + \mathbf{R}_{BU} \Delta \vec{\mathbf{p}}_{BU} - \mathbf{R}_{BU} \vec{\mathbf{p}}_{Uf} \\ &= \mathbf{R}_{BU} \vec{\mathbf{p}}_{Uf} + \mathbf{R}_{BU} \bullet \vec{\alpha}_{BU} \times \vec{\mathbf{p}}_{Uf} + \mathbf{R}_{BU} \Delta \vec{\mathbf{p}}_{BU} - \mathbf{R}_{BU} \vec{\mathbf{p}}_{Uf} \\ &= \mathbf{R}_{BU} \bullet \vec{\alpha}_{BU} \times \vec{\mathbf{p}}_{Uf} + \mathbf{R}_{BU} \Delta \vec{\mathbf{p}}_{BU} \\ &= -\mathbf{R}_{BU} \bullet \vec{\mathbf{p}}_{Uf} \times \vec{\alpha}_{BU} + \mathbf{R}_{BU} \Delta \vec{\mathbf{p}}_{BU} \\ &= \mathbf{R}_{BU} \text{skew}(-\vec{\mathbf{p}}_{Uf}) \vec{\alpha}_{BU} + \mathbf{R}_{BU} \Delta \vec{\mathbf{p}}_{BU}\end{aligned}$$

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Another Example

$$\begin{aligned}
 \vec{\mathbf{p}}_{Tf} + \Delta\vec{\mathbf{p}}_{Tf} &= \mathbf{F}_B \Delta\mathbf{F}_B (\vec{\mathbf{p}}_{Bf} + \Delta\vec{\mathbf{p}}_{Bf}) \\
 \Delta\vec{\mathbf{p}}_{Tf} &= \mathbf{F}_B \Delta\mathbf{F}_B (\vec{\mathbf{p}}_{Bf} + \Delta\vec{\mathbf{p}}_{Bf}) - \mathbf{F}_B \vec{\mathbf{p}}_{Bf} \\
 \Delta\mathbf{F}_B (\vec{\mathbf{p}}_{Bf} + \Delta\vec{\mathbf{p}}_{Bf}) &= \Delta\mathbf{R}_B (\vec{\mathbf{p}}_{Bf} + \Delta\vec{\mathbf{p}}_{Bf}) + \Delta\vec{\mathbf{p}}_B \\
 &\approx (\mathbf{I} + \text{skew}(\vec{\alpha}_B)) (\vec{\mathbf{p}}_{Bf} + \Delta\vec{\mathbf{p}}_{Bf}) + \Delta\vec{\mathbf{p}}_B \\
 &= (\vec{\mathbf{p}}_{Bf} + \Delta\vec{\mathbf{p}}_{Bf}) + \vec{\alpha}_B \times \vec{\mathbf{p}}_{Bf} + \vec{\alpha}_B \times \Delta\vec{\mathbf{p}}_{Bf} + \Delta\vec{\mathbf{p}}_B \\
 &\approx \vec{\mathbf{p}}_{Bf} + \Delta\vec{\mathbf{p}}_{Bf} + \vec{\alpha}_B \times \vec{\mathbf{p}}_{Bf} + \Delta\vec{\mathbf{p}}_B \\
 \Delta\vec{\mathbf{p}}_{Tf} &\approx \mathbf{F}_B (\vec{\mathbf{p}}_{Bf} + \Delta\vec{\mathbf{p}}_{Bf} + \vec{\alpha}_B \times \vec{\mathbf{p}}_{Bf} + \Delta\vec{\mathbf{p}}_B) - \mathbf{F}_B \vec{\mathbf{p}}_{Bf} \\
 &= \mathbf{R}_B (\vec{\mathbf{p}}_{Bf} + \Delta\vec{\mathbf{p}}_{Bf} + \vec{\alpha}_B \times \vec{\mathbf{p}}_{Bf} + \Delta\vec{\mathbf{p}}_B) + \vec{\mathbf{p}}_B - (\mathbf{R}_B \vec{\mathbf{p}}_{Bf} + \vec{\mathbf{p}}_B) \\
 &= \mathbf{R}_B (\Delta\vec{\mathbf{p}}_{Bf} + \vec{\alpha}_B \times \vec{\mathbf{p}}_{Bf} + \Delta\vec{\mathbf{p}}_B)
 \end{aligned}$$

$\Delta\vec{\mathbf{p}}_{Bf} \approx \mathbf{R}_{BU} \text{skew}(-\vec{\mathbf{p}}_{BU}) \vec{\alpha}_{BU} + \mathbf{R}_{BU} \Delta\vec{\mathbf{p}}_{BU}$

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Another Example

$$\begin{aligned}
 \Delta\vec{\mathbf{p}}_{Tf} &\approx \mathbf{R}_B (\Delta\vec{\mathbf{p}}_{Bf} + \vec{\alpha}_B \times \vec{\mathbf{p}}_{Bf} + \Delta\vec{\mathbf{p}}_B) \\
 \Delta\vec{\mathbf{p}}_{Bf} &\approx \mathbf{R}_{BU} \text{skew}(-\vec{\mathbf{p}}_{BU}) \vec{\alpha}_{BU} + \mathbf{R}_{BU} \Delta\vec{\mathbf{p}}_{BU} \\
 \Delta\vec{\mathbf{p}}_{Tf} &\approx \mathbf{R}_B (\mathbf{R}_{BU} \text{skew}(-\vec{\mathbf{p}}_{BU}) \vec{\alpha}_{BU} + \mathbf{R}_{BU} \Delta\vec{\mathbf{p}}_{BU} + \vec{\alpha}_B \times \vec{\mathbf{p}}_{Bf} + \Delta\vec{\mathbf{p}}_B) \\
 &= \begin{pmatrix} \mathbf{R}_{BU} \text{skew}(-\vec{\mathbf{p}}_{BU}) \vec{\alpha}_{BU} + \mathbf{R}_{BU} \Delta\vec{\mathbf{p}}_{BU} \\ + \mathbf{R}_B \text{skew}(-\vec{\mathbf{p}}_{Bf}) \vec{\alpha}_B + \Delta\vec{\mathbf{p}}_B \end{pmatrix} \\
 &= \begin{bmatrix} \mathbf{R}_{BU} \text{skew}(-\vec{\mathbf{p}}_{BU}) & | & \mathbf{R}_{BU} & | & \mathbf{R}_B \text{skew}(-\vec{\mathbf{p}}_{Bf}) & | & \mathbf{I} \end{bmatrix} \begin{bmatrix} \vec{\alpha}_{BU} \\ \Delta\vec{\mathbf{p}}_{BU} \\ \vec{\alpha}_B \\ \Delta\vec{\mathbf{p}}_B \end{bmatrix}
 \end{aligned}$$

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Parametric Sensitivity

Suppose you have an explicit formula like

$$\vec{p}_3 = \begin{bmatrix} L_1 \cos(\theta_1) + L_2 \cos(\theta_1 + \theta_2) + L_3 \cos(\theta_1 + \theta_2 + \theta_3) \\ L_1 \sin(\theta_1) + L_2 \sin(\theta_1 + \theta_2) + L_3 \sin(\theta_1 + \theta_2 + \theta_3) \\ 0 \end{bmatrix}$$

and know that the only variation is in parameters like L_k and θ_k . Then you can estimate the variation in \vec{p}_3 as a function of variation in L_k and θ_k by remembering your calculus.

$$\Delta \vec{p}_3 \cong \begin{bmatrix} \frac{\partial \vec{p}_3}{\partial \vec{L}} & \frac{\partial \vec{p}_3}{\partial \vec{\theta}} \end{bmatrix} \begin{bmatrix} \Delta \vec{L} \\ \Delta \vec{\theta} \end{bmatrix}$$

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Parametric Sensitivity

Grinding this out gives:

$$\Delta \vec{p}_3 \cong \begin{bmatrix} \frac{\partial \vec{p}_3}{\partial \vec{L}} & \frac{\partial \vec{p}_3}{\partial \vec{\theta}} \end{bmatrix} \begin{bmatrix} \Delta \vec{L} \\ \Delta \vec{\theta} \end{bmatrix}$$

where

$$\vec{L} = [L_1, L_2, L_3]^T$$

$$\vec{\theta} = [\theta_1, \theta_2, \theta_3]^T$$

$$\frac{\partial \vec{p}_3}{\partial \vec{L}} = \begin{bmatrix} \cos(\theta_1) & \cos(\theta_1 + \theta_2) & \cos(\theta_1 + \theta_2 + \theta_3) \\ \sin(\theta_1) & \sin(\theta_1 + \theta_2) & \sin(\theta_1 + \theta_2 + \theta_3) \\ 0 & 0 & 0 \end{bmatrix}$$

$$\frac{\partial \vec{p}_3}{\partial \vec{\theta}} = \begin{bmatrix} -L_1 \sin(\theta_1) - L_2 \sin(\theta_1 + \theta_2) - L_3 \sin(\theta_1 + \theta_2 + \theta_3) & -L_2 \sin(\theta_1 + \theta_2) - L_3 \sin(\theta_1 + \theta_2 + \theta_3) & -L_3 \sin(\theta_1 + \theta_2 + \theta_3) \\ L_1 \cos(\theta_1) + L_2 \cos(\theta_1 + \theta_2) + L_3 \cos(\theta_1 + \theta_2 + \theta_3) & L_2 \cos(\theta_1 + \theta_2) + L_3 \cos(\theta_1 + \theta_2 + \theta_3) & L_3 \cos(\theta_1 + \theta_2 + \theta_3) \\ 0 & 0 & 0 \end{bmatrix}$$

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More generally ...

Suppose that we have a vector function

$$\bar{\mathbf{v}} = \bar{\mathbf{g}}(\bar{\mathbf{q}}) = [g_1(\bar{\mathbf{q}}), \dots, g_m(\bar{\mathbf{q}})]^T$$

of parameters $\bar{\mathbf{q}} = [q_1, \dots, q_n]$. Then we can estimate the value of

$$\bar{\mathbf{v}} + \Delta\bar{\mathbf{v}} = \bar{\mathbf{g}}(\bar{\mathbf{q}} + \Delta\bar{\mathbf{q}})$$

by

$$\bar{\mathbf{v}} + \Delta\bar{\mathbf{v}} \approx \bar{\mathbf{g}}(\bar{\mathbf{q}}) + \mathbf{J}_G(\bar{\mathbf{q}}) \bullet \Delta\bar{\mathbf{q}}$$

where

$$\mathbf{J}_G(\bar{\mathbf{q}}) = \begin{bmatrix} \frac{\partial g_1}{\partial q_1} & \frac{\partial g_1}{\partial q_j} & \frac{\partial g_1}{\partial q_n} \\ \vdots & \vdots & \vdots \\ \frac{\partial g_i}{\partial q_1} & \dots & \frac{\partial g_i}{\partial q_j} & \dots & \frac{\partial g_i}{\partial q_n} \\ \vdots & & \vdots & & \vdots \\ \frac{\partial g_m}{\partial q_1} & \frac{\partial g_m}{\partial q_j} & \frac{\partial g_m}{\partial q_n} \end{bmatrix}$$

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