

### △ Inverse Kinematics

\* Problem: forward kinematics  
 $T_{SE}(t) = [C_{SE}(t) \ J_{SE}(t)] \ SE(3)$   
 pseudocode

$$X_0 = \begin{pmatrix} X_0(t) \\ p_0(t) \end{pmatrix} = f(t)$$

$$X_k = \begin{pmatrix} X_k(t) \\ p_k(t) \end{pmatrix}$$

$$\therefore Q: \dot{\theta} = f^{-1}(x_k) ?$$

give you  $x_k$

one me  $\dot{\theta}$

△ Soln

\* Analytic:

- for 3 intersecting neighboring rays

\* Geometric:

- use length, then... geometric info!

\* Algebraic:

- use TPs to get joints

\* Numerical!

△ Numerical method:

inverse differential kinematics

- recall

$$w = J_{SE} \dot{\theta}$$

singularities

- occur if  $\dot{\theta} = 0$   $J_{SE}(\dot{\theta})$  is column

$$\therefore \dot{\theta} = J_{SE} w$$

if  $w$  - small

- constraints

[bounding]: when algorithm is at its minimum; hard to move

- normal: hard to move

- dampened version of Moore-Penrose pseudo inverse

$$\dot{\theta} = J^T w \rightarrow \arg\min \|J\dot{\theta} - w\|_2$$

$$\therefore w = J^T(J^T J)^{-1} J \dot{\theta}$$

$$\dot{\theta} = J^T(J^T J + \lambda I)^{-1} J \dot{\theta}$$

\* Redundancy

$$w = J \dot{\theta}$$

$$N \in R^n$$

$$J \in R^{n \times n}$$

$n > n$   
 ↳ redundancy

$$\therefore J \dot{\theta} = w$$

$$\Rightarrow J_{SE}(J_{SE}^T W_{SE} + N_{SE}) = w$$

$$\therefore \dot{\theta} = J_{SE}^T W_{SE} + N \dot{\theta}_0$$

$$w = N(J_{SE}) \text{ null space}$$

$$J \dot{\theta} = N \dot{\theta}_0$$

- get  $N$ ?

$$N = J - J_{SE} J_{SE}$$

↳ end-up of different basis they span Null space

$$\therefore Q: \dot{\theta} = J^T w$$

\* multi-task control

- reach desired position & orientation

- hand driven: tasks:  $t = f(\dot{\theta}, w)$

$$\therefore \dot{\theta} = J_{SE}^T W_{SE} + N \dot{\theta}_0$$

$$J = \begin{bmatrix} J_1 & J_2 & \dots & J_n \end{bmatrix}$$

↳ min  $\|J \dot{\theta} - t\|_2$

$$\Rightarrow \min \|J \dot{\theta}\|_2$$

$$\therefore \dot{\theta} = J \dot{\theta}$$

- weighting

$$J \in R^{m \times n}$$

$$\therefore J^T W = (J^T W J)^{-1} J^T W$$

\* Parameterization

$$\text{recall: } \dot{\theta} = J^T W \dot{\theta} + N \dot{\theta}_0$$

$$\Rightarrow w = J \dot{\theta}$$

$$= J_{SE}(J_{SE}^T W_{SE} + N_{SE})$$

$$\Rightarrow \dot{\theta} = J_{SE}^T W_{SE} + N(J_{SE}^T W_{SE})$$

$$\therefore \dot{\theta} = \frac{1}{m} N \dot{\theta}_0$$

$$\therefore w = (J_{SE}^T W_{SE})^T / (W_{SE}^T N_{SE})$$

but due to log-concavity of norm of moving least squares

### △ Back to inverse Kinematics

$$\begin{aligned} \dot{\theta} &= J_{SE}(\dot{\theta}) \dot{\theta} \\ w &= J \dot{\theta} \quad \text{angular} \\ &\quad \text{geometric} \end{aligned}$$

$$\therefore \text{now } \dot{\theta} = J_{SE}(\dot{\theta}) \dot{\theta}$$

- tracking a point  $X_0^*$ ,  $\theta = \theta^*$

pseudo-rate

$$\therefore \dot{\theta} = \frac{1}{m} (X_0^* - X_0(\dot{\theta}))$$

→ make  $\|X_0^* - X_0(\dot{\theta})\| \approx \text{not do}$

$$\therefore J_{SE} \leftarrow \frac{\partial X_0}{\partial \theta}$$

(usually joint-space wagon after task-space)

→ redundant system

△ Inverse method

- problem 1

$$\dot{\theta} = \dot{\theta}^* + J_{SE}^T A X$$

$A$  fix  $\dot{\theta}$

$X$  is a column

∴  $\dot{\theta} = \dot{\theta}^* + J_{SE}^T A X$

△ Orientation

- depends on parameterization

$$\therefore GSO(\theta)$$

-  $A$   $X$  matrix in  $A \in \mathbb{R}^{n \times n}$

$$GSO(\theta) = C_{SE}(\theta^*) C_{SE}(\theta^*)^T$$

$$\Rightarrow \dot{\theta} = \dot{\theta}^* + J_{SE}^T (A \in \mathbb{R}^{n \times n}) \dot{X}$$

or  $\dot{\theta} = \dot{\theta}^* + J_{SE}^T A X$

△ Derivatives

- depends on parameterization

$$\therefore GSO(\theta)$$

-  $A$   $X$  matrix in  $A \in \mathbb{R}^{n \times n}$

$$GSO(\theta) = C_{SE}(\theta^*) C_{SE}(\theta^*)^T$$

$$\Rightarrow \dot{\theta} = \dot{\theta}^* + J_{SE}^T (A \in \mathbb{R}^{n \times n}) \dot{X}$$

△ Draj. calc.

- given

$$J^T \dot{\theta}^*$$

$$J^T \dot{\theta}^*$$

- feedback

$$\therefore \partial \dot{\theta} = J^T (\dot{\theta}^* - J^T \dot{\theta}^* + k_p A \dot{p} + k_d A \dot{\theta})$$

$$\therefore \dot{\theta} = J^T \dot{\theta}^* + k_p A \dot{p} + k_d A \dot{\theta}$$

△ L-AI: invert

- determinants

$$\therefore \det(A) = 0$$

spare unif. same space, no linear elements

$$\therefore \det(A) = c$$

△ Rank

$$\text{rank}(A) = n$$

where  $n$  is the dimension of the spanned space of the output column space

△ Nullspace (kernel)

$$A \dot{\theta} = 0 \quad A \in \mathbb{R}^{m \times n}$$

↳ a set of vectors, such that no space for "spans" no the origin

more than all  $A$  has linear null space  $\mathbb{R}^{n-m}$

null space  $\mathbb{R}^{n-m}$  of  $A$  in  $\mathbb{R}^n$

internal rank

△ Ill/well-condition of matrix

$$Ax = c$$

↳ ill-conditioned? small change in well-conditioned?

small change in  $c$ ?

well-well → small change in  $c$

small change in  $A$ ?

well-conditioned?

well-well → large  $c$

well-well → well-conditioned?

