



THE UNIVERSITY *of* EDINBURGH
informatics

Advanced Robotics

Forward Kinematics and Inverse Kinematics

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University of Edinburgh

Reading for this week

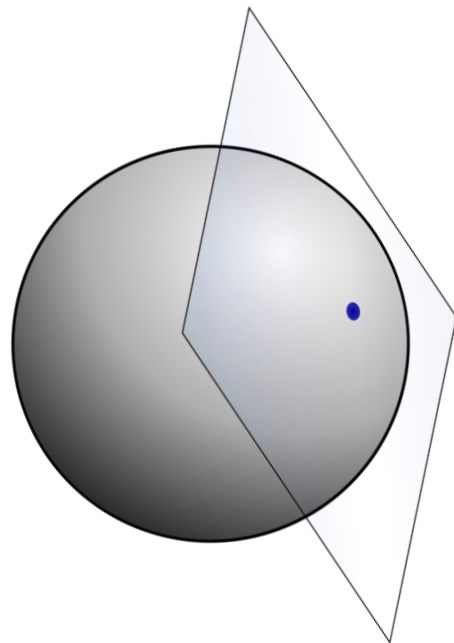
- ❑ Siciliano, B., et al., Robotics: Modelling, Planning and Control.

Chapter 3.2 , 3.3, 3.5

- ❑ (If you are really motivated) Roy Featherstone, Rigid body dynamics algorithms

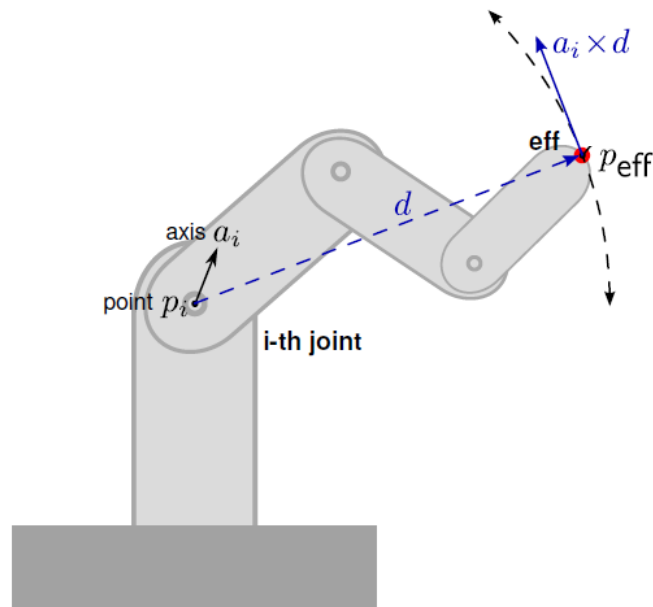
Chapter 2.1, 2.2

An informal introduction



An informal introduction

How does an infinitesimal change in configuration affects end-effector linear velocity ?



$$\delta p_{\text{eff}} = \delta q_i [a_i \times (p_{\text{eff}} - p_i)]$$

$$J_{\text{pos}}(q) = \begin{pmatrix} [a_1 \times (p_{\text{eff}} - p_1)] \\ [a_2 \times (p_{\text{eff}} - p_2)] \\ \vdots \\ [a_n \times (p_{\text{eff}} - p_n)] \end{pmatrix} \in \mathbb{R}^{3 \times n}$$

Our objectives for today

- Forward **kinematics** consists, given configuration and velocity in configuration space, in computing the velocity of a rigid body in the cartesian space:

$$FK : \mathbf{q}, \mathbf{v}_q \longrightarrow \nu$$

- Inverse **kinematics** consists, given a desired velocity ν^* in the cartesian space (and the current configuration), in computing a velocity in the configuration space result in a velocity as close as possible to ν^* :

$$IK : \nu^*, \mathbf{q} \longrightarrow \mathbf{v}_q$$

But before, minimal info on spatial velocity

An element of the group....

Can be represented as ...

$$r \in SO(3)$$

rotation

$$\simeq$$

$$\mathbf{R} \in \mathbb{R}^{3 \times 3}$$

Rotation matrix

$$\mathbf{q} \in \mathbb{H} \simeq \mathbb{R}^4, \|\mathbf{q}\| = 1 \quad \text{quaternion}$$

$$\omega \in so(3) \simeq \mathbb{R}^3 \quad \text{A velocity?}$$

$$m \in SE(3)$$

displacement

$$\simeq \mathbb{R}^3 \times SO(3)$$

translation rotation

$$\simeq \mathbb{R}^3 \times \mathbb{R}^{3 \times 3} \simeq \mathbb{R}^{4 \times 4} \quad \text{Homogeneous matrix}$$

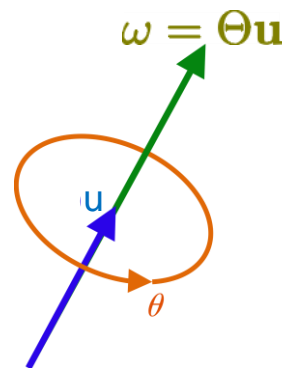
$$\mathbb{R}^3 \times \mathbb{H} \simeq \mathbb{R}^7$$

$$\mathcal{V} \in se(3) = \begin{bmatrix} \mathbf{v} \\ \omega \end{bmatrix} \simeq \mathbb{R}^6 \quad \text{A spatial velocity?}$$

Spatial velocity vector

- \mathbf{v} linear velocity
- $\boldsymbol{\omega}$ angular velocity

$$\mathcal{V} \in se(3) = \begin{bmatrix} \mathbf{v} \\ \boldsymbol{\omega} \end{bmatrix} \simeq \mathbb{R}^6$$



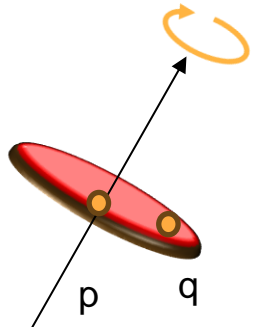
Spatial velocity vector

□ \mathbf{v} linear velocity

□ Not completely intuitive

$$\mathcal{V} \in se(3) = \begin{bmatrix} \mathbf{v} \\ \boldsymbol{\omega} \end{bmatrix} \simeq \mathbb{R}^6$$

$\mathbf{v}_q = \mathbf{v}_p + \boldsymbol{\omega} \times \overrightarrow{pq}$: velocity of the point of the rigid body currently coinciding with \mathbf{q} seen from p



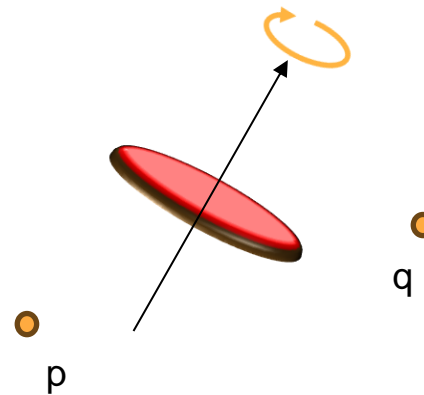
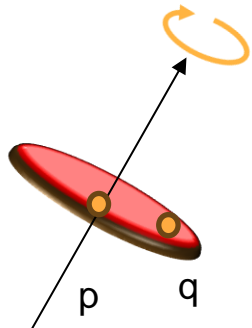
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Also true. Defined beyond a rigid body: vector field

Spatial velocity vector

- \mathbf{v} linear velocity
- $\boldsymbol{\omega}$ angular velocity

$${}^A \mathcal{V} \in se(3) = \begin{bmatrix} {}^A \mathbf{v}_A \\ {}^A \boldsymbol{\omega} \end{bmatrix} \simeq \mathbb{R}^6$$



Spatial velocity vector

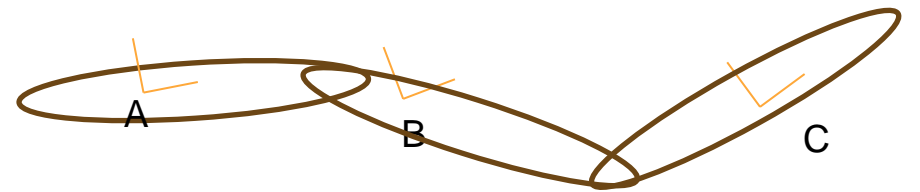
□ \mathbf{v} linear velocity

□ $\boldsymbol{\omega}$ angular velocity

□ ${}^A\mathcal{V}_{BC}$: spatial velocity of frame C wrt to frame B, expressed in frame A

$${}^A\mathcal{V}_{AC} = {}^A\mathcal{V}_{AB} + {}^A\mathcal{V}_{BC}$$

$${}^A\mathcal{V} \in se(3) = \begin{bmatrix} {}^A\mathbf{v}_A \\ {}^A\boldsymbol{\omega} \end{bmatrix} \simeq \mathbb{R}^6$$



Spatial velocity vector

□ \mathbf{v} linear velocity

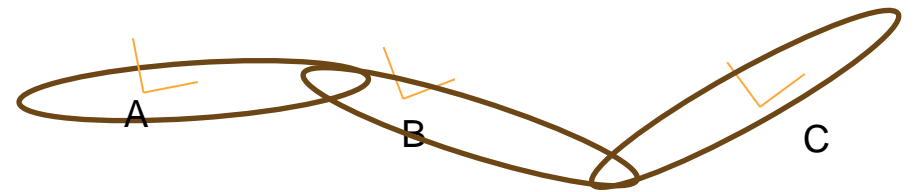
□ $\boldsymbol{\omega}$ angular velocity

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$${}^A \mathcal{V}_{AC} = {}^A \mathcal{V}_{AB} + {}^A \mathbf{X}_B {}^B \mathcal{V}_{BC}$$



Spatial velocity vector

□ \mathbf{v} linear velocity

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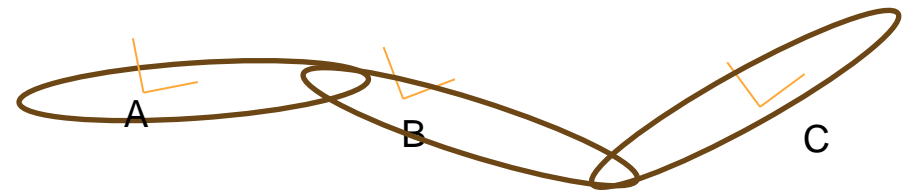
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with ${}^A \mathbf{X}_B$ the « adjoint » or « action » matrix

$${}^A \mathbf{X}_B = \begin{bmatrix} {}^A \mathbf{R}_B & {}^A \overrightarrow{\mathbf{AB}} \times {}^A \mathbf{R}_B \\ 0 & {}^A \mathbf{R}_B \end{bmatrix}$$

Let's just accept this for now

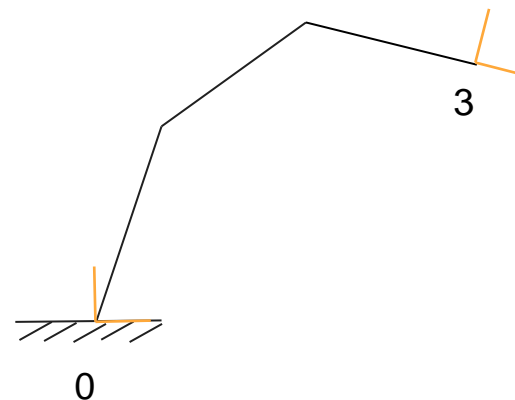


Forward kinematics

- The spatial velocity vector between the root frame {0} of a kinematic tree and the end effector frame {3} can depend both on the position and velocity in configuration space.

$${}^0\nu_{03}(\mathbf{q}, v_q) \quad \text{and} \quad {}^3\nu_{03}(\mathbf{q}, v_q)$$

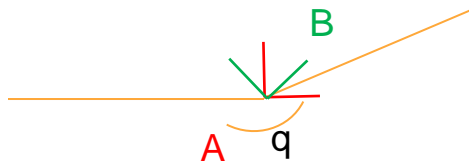
- However for most classical joints it only depends on \mathbf{v}_q



Example: revolute joint around x axis

$${}^A M_B(q) = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & \cos(q) & \sin(q) & 0 \\ 0 & -\sin(q) & \cos(q) & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

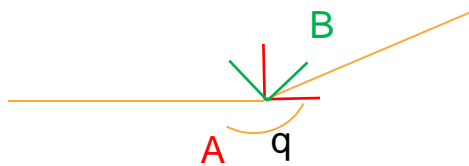
$$v_q = \dot{q}$$



Example: revolute joint around x axis

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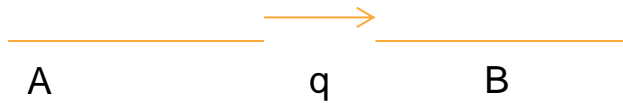
$$v_q = \dot{q}$$



Example: prismatic z joint

$${}^A M_B(q) = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & q \\ 0 & 0 & 0 & 1 \end{bmatrix} \quad {}^A \nu_{AB}(\dot{q}) = \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \\ 0 \end{bmatrix} \dot{q}$$

$$v_q = \dot{q}$$



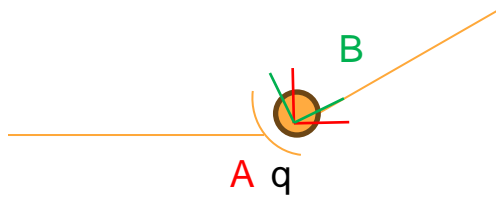
Example: spherical joint

$$\underline{q} \in \mathbb{R}^4, \|\underline{q}\| = 1$$

$${}^A M_B(q) = \begin{bmatrix} & \mathcal{R}(\underline{q}) & & 0 \\ & & & 0 \\ & & & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$v_q = \omega$$

It's a choice



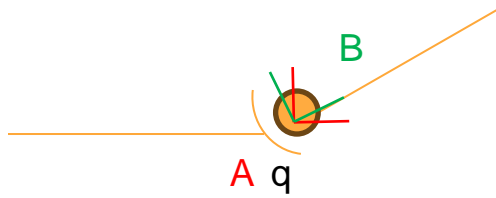
Example: spherical joint

$$\underline{q} \in \mathbb{R}^4, \|\underline{q}\| = 1$$

$${}^A M_B(q) = \begin{bmatrix} \mathcal{R}(\underline{q}) & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \quad {}^A \mathcal{V}_{AB}(v_q) = \begin{bmatrix} 0_{3 \times 3} \\ I_{3 \times 3} \end{bmatrix} v_q$$

$$v_q = \omega$$

It's a choice



In general, forward kinematics for one joint

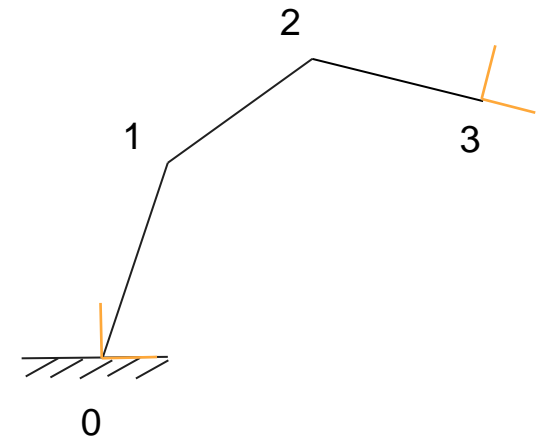
$${}^B \nu_{AB}(q, v_q) = {}^B J(q) v_q$$

$\frac{\partial f}{\partial t}$ $\frac{\partial f}{\partial q}$ $\frac{\partial q}{\partial t}$

J is a joint Jacobian (or at least behaves as such):

Forward kinematics for a chain of joints:

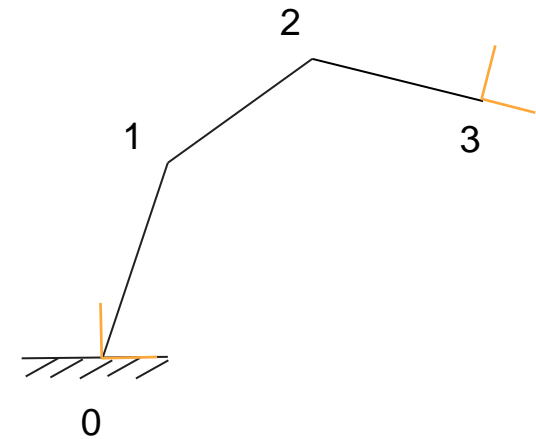
$${}^0\nu_{03}(q, v_q) = {}^0\nu_{01} + {}^0\nu_{12} + {}^0\nu_{23}$$



Forward kinematics for a chain of joints:

$${}^0\nu_{03}(q, v_q) = {}^0\nu_{01} + {}^0\nu_{12} + {}^0\nu_{23}$$

$${}^0\nu_{03}(q, v_q) = {}^0\mathbf{X}_1 {}^1\nu_{01} + {}^0\mathbf{X}_2 {}^2\nu_{12} + {}^0\mathbf{X}_3 {}^3\nu_{23}$$

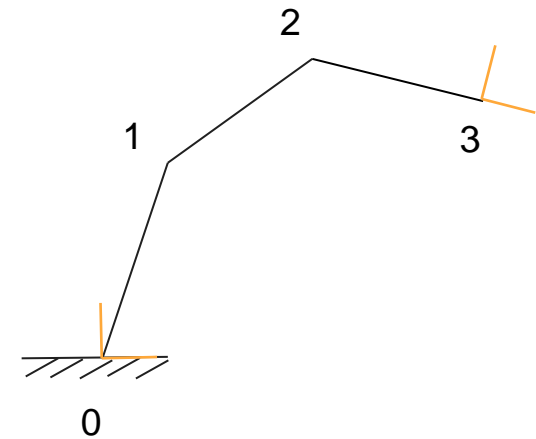


Forward kinematics for a chain of joints:

$${}^0\nu_{03}(q, v_q) = {}^0\nu_{01} + {}^0\nu_{12} + {}^0\nu_{23}$$

$${}^0\nu_{03}(q, v_q) = {}^0\mathbf{X}_1^1 \nu_{01} + {}^0\mathbf{X}_2^2 \nu_{12} + {}^0\mathbf{X}_3^3 \nu_{23}$$

$${}^0\nu_{03}(q, v_q) = {}^0\mathbf{X}_1^1 J_1(q_1)v_{q_1} + {}^0\mathbf{X}_2^2 J_2(q_2)v_{q_2} + {}^0\mathbf{X}_3^3 J_3(q_3)v_{q_3}$$



Forward kinematics for a chain of joints:

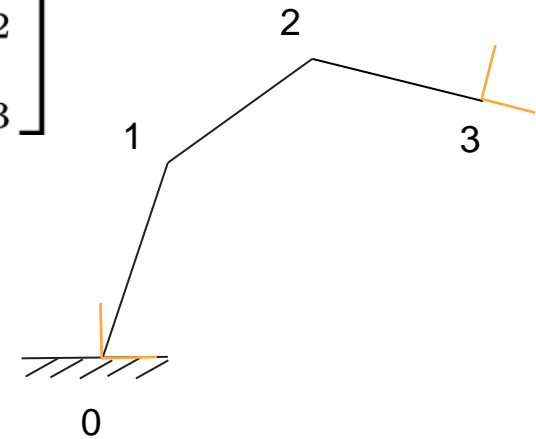
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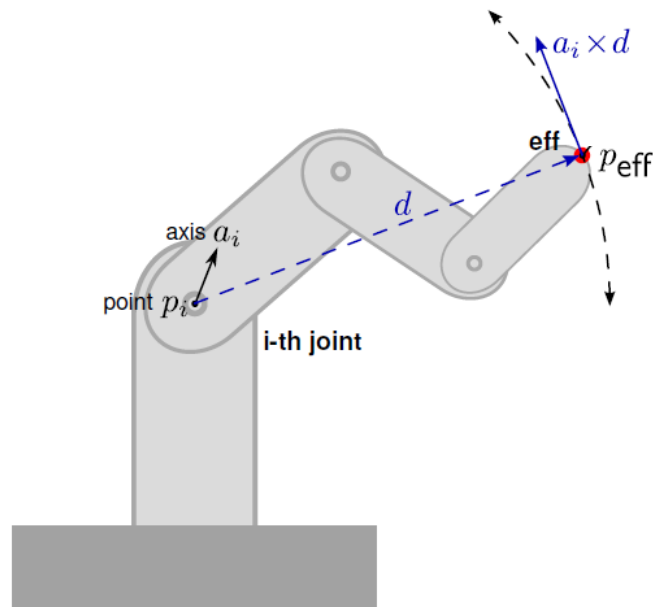
$${}^0\nu_{03}(q, v_q) = \begin{bmatrix} {}^0\mathbf{X}_1^1 J_1(q_1) & {}^0\mathbf{X}_2^2 J_2(q_2) & {}^0\mathbf{X}_3^3 J_3(q_3) \end{bmatrix} \begin{bmatrix} v_{q_1} \\ v_{q_2} \\ v_{q_3} \end{bmatrix}$$

$$\nu(\mathbf{q}, v_{\mathbf{q}}) = \mathbf{J}(\mathbf{q}) v_{\mathbf{q}}$$



An informal introduction

How does an infinitesimal change in configuration affects end-effector linear velocity ?

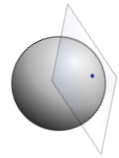


$$\delta p_{\text{eff}} = \delta q_i [a_i \times (p_{\text{eff}} - p_i)]$$

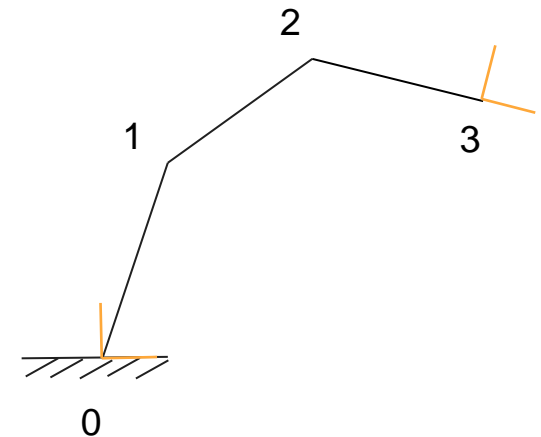
$$J_{\text{pos}}(q) = \begin{pmatrix} [a_1 \times (p_{\text{eff}} - p_1)] \\ [a_2 \times (p_{\text{eff}} - p_2)] \\ \vdots \\ [a_n \times (p_{\text{eff}} - p_n)] \end{pmatrix} \in \mathbb{R}^{3 \times n}$$

Forward kinematics for a chain of joints:

- ❑ For any given \mathbf{q} , $\mathbf{J}(\mathbf{q})$ is straightforward to compute
- ❑ If \mathbf{J} is known, there is a linear mapping from a velocity in the configuration to a velocity in the cartesian space
- ❑ \mathbf{J} is only valid **locally**

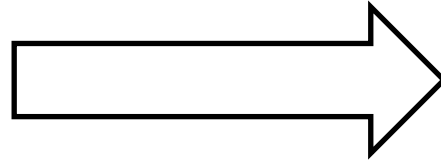


$$\nu(\mathbf{q}, v_{\mathbf{q}}) = \mathbf{J}(\mathbf{q})v_{\mathbf{q}}$$



Inverse kinematics is an inversion problem

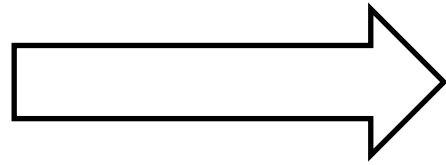
$$\nu(\mathbf{q}, v_{\mathbf{q}}) = \mathbf{J}(\mathbf{q})v_{\mathbf{q}}$$



$$v_{\mathbf{q}} = \mathbf{J}(\mathbf{q})^{-1}\nu(\mathbf{q}, v_{\mathbf{q}})$$

Inverse kinematics is an inversion problem

$$\nu(\mathbf{q}, v_{\mathbf{q}}) = \mathbf{J}(\mathbf{q})v_{\mathbf{q}}$$



$$v_{\mathbf{q}} = \mathbf{J}(\mathbf{q})^{-1}\nu(\mathbf{q}, v_{\mathbf{q}})$$

J is most likely not invertible, so what?

We can formulate IK as optimisation problem

- ❑ Also known in literature as differential inverse kinematics of closed loop inverse kinematics (click)
- ❑ Given q and a target end-effector velocity ν^* , find a joint velocity v_q that results in a end-effector velocity ν as close as possible to ν^*

$$\min_{v_q} \|\nu(q, v_q) - \nu^*\|^2$$

How is this similar to inverse geometry ?

□ IG:

$$\min_q \text{dist}({}^0M_e(q), {}^0M_*)$$

$$\min_{v_q} \text{dist}({}^0M_e(q_0 \oplus v_q), {}^0M_*)$$

□ IK:

$$\min_{v_q} \|\nu(q, v_q) - \nu^*\|^2$$

$$\min_{v_q} \|J(q)v_q - \nu^*\|^2$$

How is this similar to inverse geometry ?

□ IG:

$$\min_q \text{dist}({}^0M_e(q), {}^0M_*)$$

$$\min_{v_q} \text{dist}({}^0M_e(q_0 \oplus v_q), {}^0M_*)$$

← Non-linear

□ IK:

$$\min_{v_q} \|\nu(q, v_q) - \nu^*\|^2$$

$$\min_{v_q} \|J(q)v_q - \nu^*\|^2$$

← Linear

Solution to the unconstrained IK problem:

$$v_q^* = J^\dagger v^*$$

With J^\dagger Moore Penrose pseudo-inverse

However, we could consider additional constraints to our problem: joint limits, velocity limits, etc:

IK with constraints

- Velocity bounds
(element-wise)

$$\begin{aligned} \min_{v_q} & \|J(q)v_q - \nu^*\|^2 \\ \text{s.t.} & v_q^- \leq v_q \leq v_q^+ \end{aligned}$$

- Joint bounds
(using euler integration over a time step)

$$\mathbf{q}^- \leq \mathbf{q} + \Delta t v_q \leq \mathbf{q}^+$$

- Can also add other cost functions...

- $v_q^* = J^\dagger \nu^*$ no longer optimal solution

However, easy to solve using a Quadratic Program solver (e.g. quadprog)

IK vs IG

- ❑ Inverse kinematics (also called differential IK) is a linear, convex problem, very easy to solve
- ❑ Inverse geometry (also called IK) is a non-linear problem, very hard to solve
- ❑ When trying to solve IG iteratively, we can use the pseudo-inverse of the jacobian to locally update a configuration towards one that is closer to the goal. This is similar to performing one step of gradient descent (See example after)

A really simple articulated robot

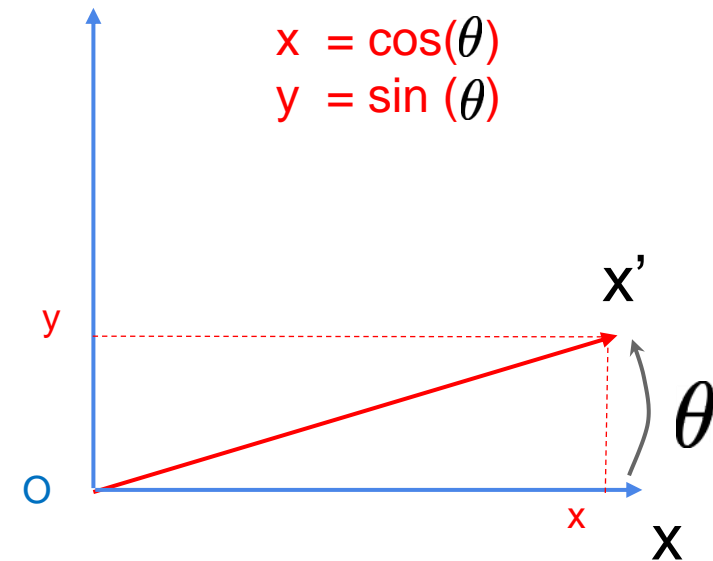
- 1 Dof, unit norm link (clock)

- FG:

$$\phi(q) = \phi(\theta) = [\cos(\theta), \sin(\theta)]^T$$

- IG:

$$\phi^{-1}([x, y]^T) = \text{atan2}(y, x)$$



- This problem has an analytical solution. **When solution exists, analytical IG defined if num (dof) \leq dim(task)** (necessary condition)
- What if we did not have an analytical solution?

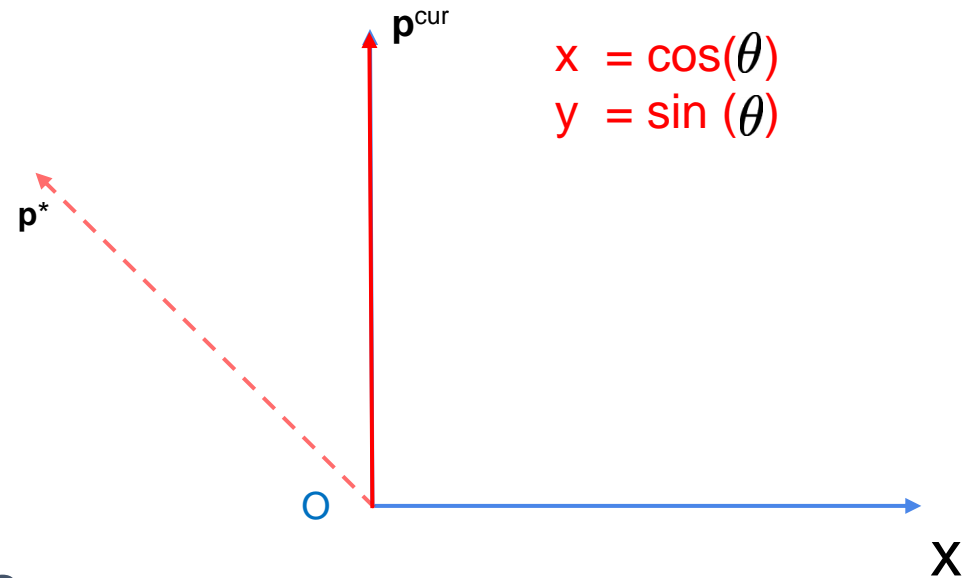
A toy problem

□ Current $\theta = \pi/2 \Leftrightarrow \mathbf{p}^{\text{cur}} = [0, 1]$

□ Target: $\mathbf{p}^* = [-\sqrt{2}, \sqrt{2}]$

□ Idea: 'nudge' θ to see what happens

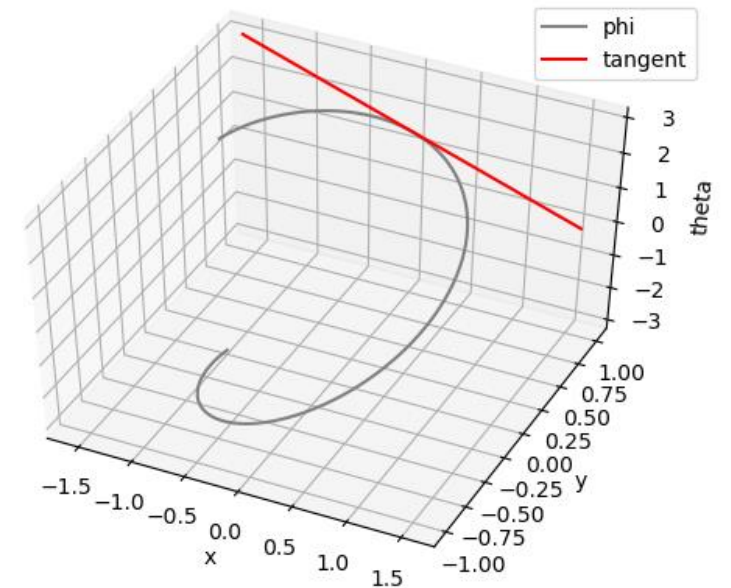
Finite differences?



A toy problem

- ❑ Current $\theta = \pi/2 \Leftrightarrow \mathbf{p}^{\text{cur}} = [0, 1]$
- ❑ Target: $\mathbf{p}^* = [-\sqrt{2}, \sqrt{2}]$
- ❑ The partial derivatives might give us information

$$\delta p = \frac{\partial}{\partial q} \phi(q) \delta q$$



Local derivative indicate how an infinitesimal change in configuration affects \mathbf{p}

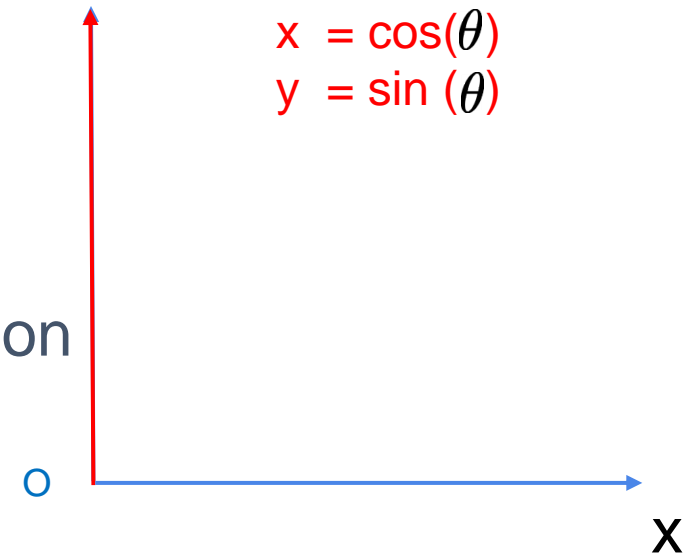
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- ❑ Target: $\mathbf{p}^* = [-\sqrt{2}, \sqrt{2}]$
- ❑ The partial derivatives might give us information

$$\frac{\partial}{\partial \theta} \phi(\theta) = \begin{bmatrix} \frac{\partial x}{\partial \theta} \\ \frac{\partial y}{\partial \theta} \end{bmatrix}$$

$$\frac{\partial x}{\partial \theta} = -\sin(\theta)$$

$$\frac{\partial y}{\partial \theta} = \cos(\theta)$$



Toy problem

- ❑ Current $\theta = \pi/2 \Leftrightarrow \mathbf{p}^{\text{cur}} = [0, 1]$
- ❑ Target: $\mathbf{p}^* = [-\sqrt{2}, \sqrt{2}]$
- ❑ The partial derivatives might give us information

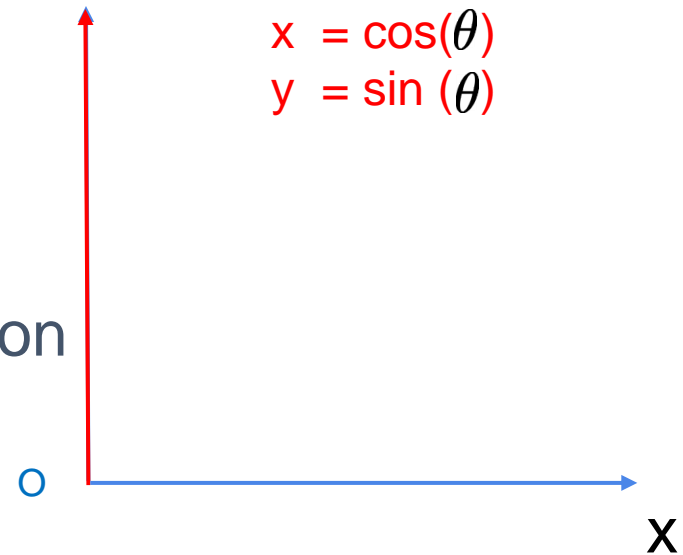
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$$\frac{\partial x}{\partial \theta} = -\sin(\theta)$$

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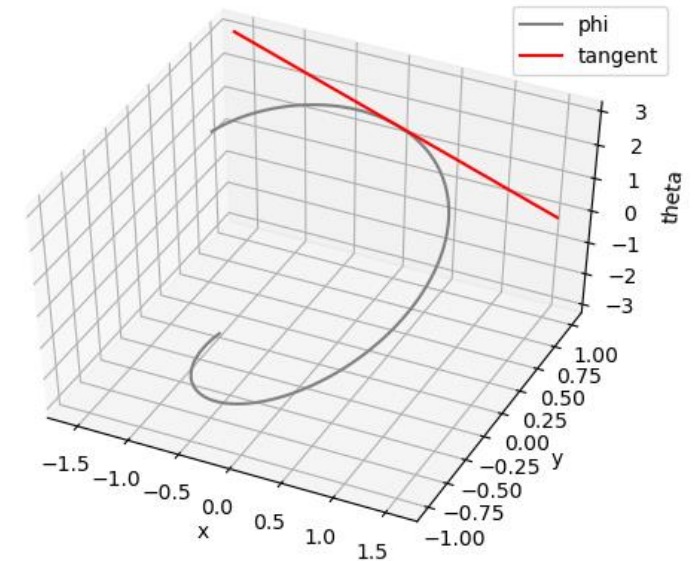
$$\frac{\partial}{\partial \theta} \phi\left(\frac{\pi}{2}\right) = [-1, 0]^T$$



A gradient descent approach

$$\frac{\partial}{\partial \theta} \phi\left(\frac{\pi}{2}\right) = [-1, 0]^T$$

- ❑ Slope of the tangent at $\pi/2$
- ❑ Thus, local linear approximation of Φ
 - ❑ Locally, if $\theta \nearrow$, $x \searrow$ (not true if we increase θ too much)
 - ❑ Nothing to say about y ?
- ❑ To reach $[-\sqrt{2}, \sqrt{2}]$, we have an idea of a baby step to make:
 - ❑ Increase θ a little. If target is reached, we won, otherwise...
 - ❑ ...start again: compute the new partial derivatives, slope etc



Gradient descent => find the minimum of a function

- ❑ Here distance between current position / target can be used as such function
- ❑ Trying to find the **global minima** will not always work...

