

Configuration Space (Pre Lecture)

Dr. Neil T. Dantam

CSCI-534, Colorado School of Mines

Spring 2020



Configuration Space

- ▶ Defines positions of all points in the system
- ▶ The space which we search in motion planning

Outcomes

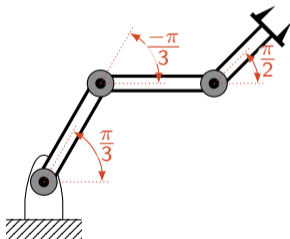
- ▶ Know definitions of configuration space
- ▶ Identify the degrees-of-freedom (DoF) of a robot
- ▶ Compute transforms for various joint types
- ▶ Relate:
 - ▶ Physical mechanism
 - ▶ Joints
 - ▶ Frames and transforms
 - ▶ Configurations

Configuration Space

Definition: Configuration

A specification for the position of all points in the system (robot).

Typically a real vector: $q \in \mathbb{R}^n$.



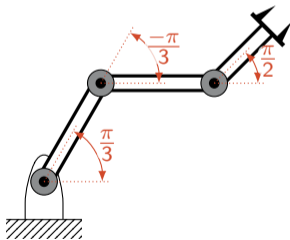
$$q = \begin{bmatrix} \frac{\pi}{3} \\ -\frac{\pi}{3} \\ \frac{\pi}{2} \end{bmatrix}$$

Degrees of Freedom

Definition: Degrees of Freedom

The smallest number of real-valued coordinates necessary to represent a configuration.

Typically a natural number: $n \in \mathbb{N}$



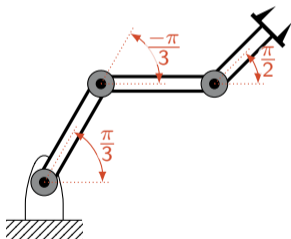
$$n = 3$$

Configuration Space

Definition: Configuration Space

The n -dimensional space containing all possible configurations of the robot.

Typically a real vector (sub)space: $Q \subseteq \mathbb{R}^n$



$$Q \subseteq \mathbb{R}^3$$

Motion Planning: Search in the configuration space

Outline

Manipulator DoF

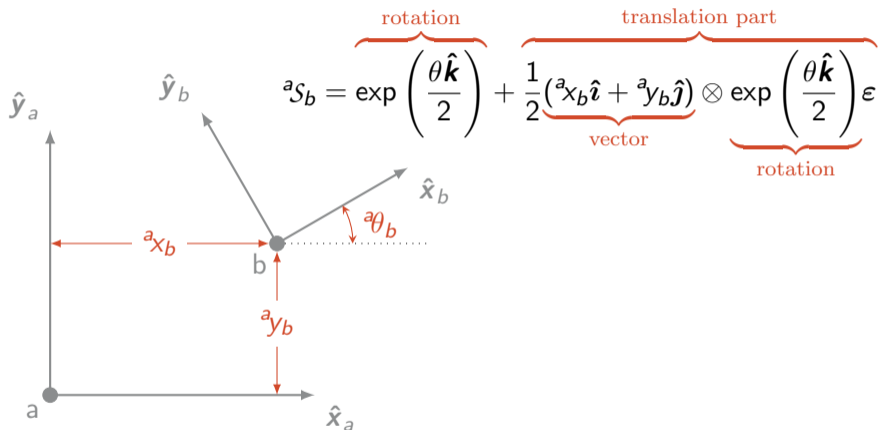
Manipulator Kinematics

Joint Transforms

Examples

The Motion Planning Problem

Planar Rigid Bodies



Three parameters: x , y , θ

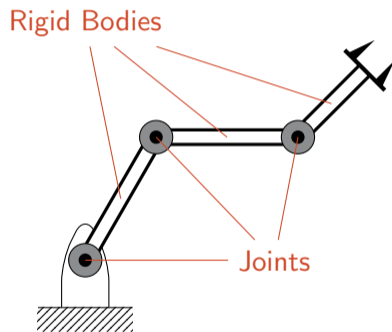
3D Rigid Bodies

$${}^a S_b = \overbrace{\exp\left(\frac{\vec{u}}{2}\right)}^{\text{rotation}} + \frac{1}{2} \underbrace{\left({}^a x_b \hat{i} + {}^a y_b \hat{j} + {}^a z_b \hat{k}\right)}_{\text{vector } {}^a v_b} \otimes \underbrace{\exp\left(\frac{\vec{u}}{2}\right)}_{\text{rotation}}$$

$\theta \hat{u} = \vec{u} = u_x \hat{i} + u_y \hat{j} + u_z \hat{k}$

Six parameters: (x, y, z) and (u_x, u_y, u_z)

Robot DoF



Joint Types

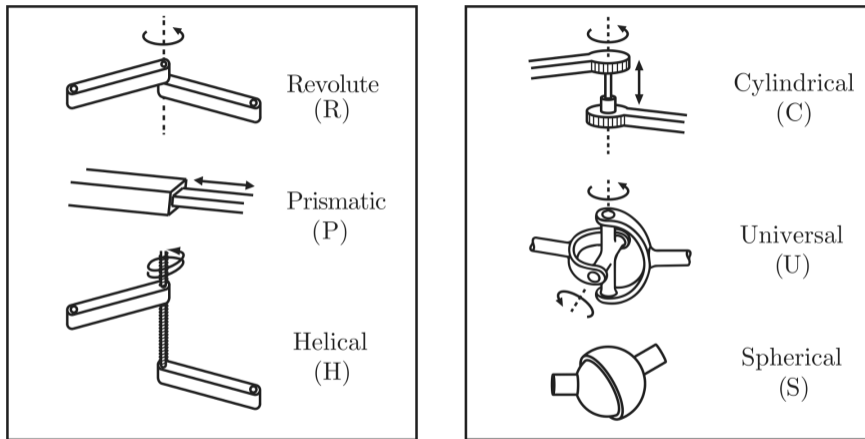


Figure 2.3: Typical robot joints.
(Lynch and Park. Modern Robotics.)

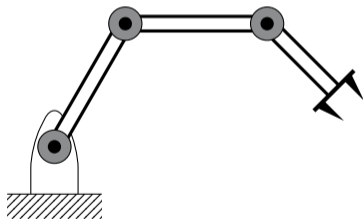
Joint Constraints DoF

Joint Type	Constraints (2D)	Constraints (3D)	Net DoF
Revolute	2	5	1
Prismatic	2	5	1
Helical	N/A	5	1
Cylindrical	N/A	4	2
Universal	N/A	4	2
Spherical	N/A	3	3

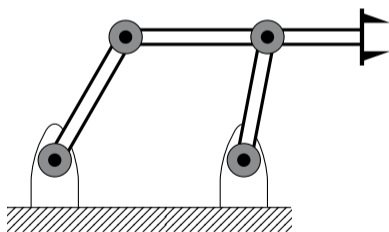
Open vs. Closed Chains

Serial vs. Parallel Manipulators

Serial



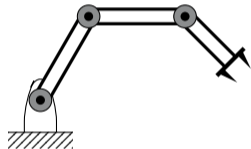
Parallel



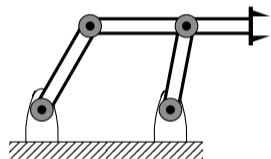
Grübler's Formula

Mechanism DoF

$$\text{dof} = \underbrace{m(N - 1)}_{\text{links}} - \underbrace{\sum_{i=1}^J c_i}_{\text{joint constraints}}$$



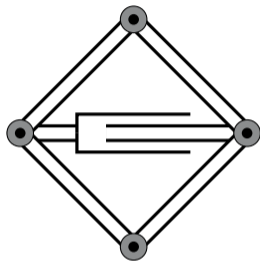
- ▶ Planar: $m = 3$
- ▶ 4 links: $N = 4$
- ▶ $\text{dof} = 3(4 - 1) - (2 + 2 + 2)$
 $= 3$



- ▶ Planar: $m = 3$
- ▶ 4 links: $N = 4$
- ▶ $\text{dof} = 3(4 - 1) - (2 + 2 + 2 + 2)$
 $= 1$

Exercise: Scissor Jack Mechanism

Simplified Planar Model



- ▶ Links:
- ▶ Joints:

Hint: look for constraints between pairs of links

Exercise: Human Arm

Kinematic Model

Outline

Manipulator DoF

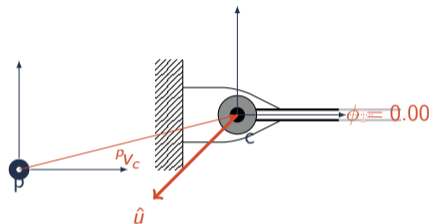
Manipulator Kinematics

Joint Transforms

Examples

The Motion Planning Problem

Revolute Joint Animation

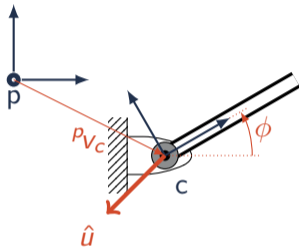
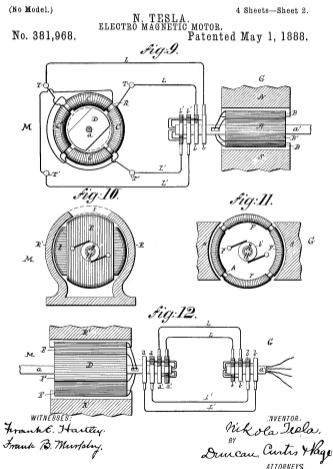


$${}^p S_c = \exp\left(\frac{0.00}{2} \hat{u}\right) + \frac{1}{2} p_{V_c} \otimes \exp\left(\frac{0.00}{2} \hat{u}\right) \epsilon$$



Revolute Joints

Rotating Motion



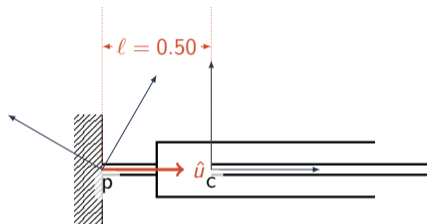
► P_{Vc} : Translation from parent p to child c (fixed)

► \hat{u} : Axis of rotation (fixed)

► ϕ : Rotation angle (varying)

► $P_{Sc}(\phi) = \underbrace{\exp\left(\frac{\phi}{2}\hat{u}\right)}_{\text{rotation}} + \underbrace{\frac{1}{2}P_{Vc} \otimes \exp\left(\frac{\phi}{2}\hat{u}\right)}_{\text{translation}} \epsilon$

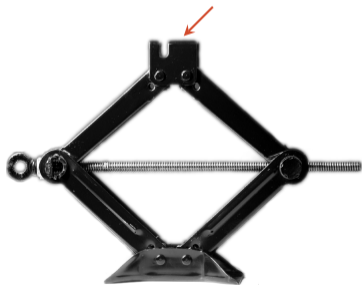
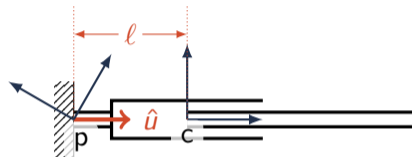
Prismatic Joint Animation



$${}^P S_C = {}^P h_c + \frac{1}{2} \begin{pmatrix} 0 & 0 & 0.50 \end{pmatrix} \hat{u} \otimes {}^P h_c \epsilon$$

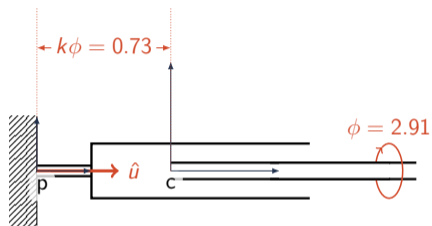
Prismatic Joints

Linear / Sliding Motion



- ▶ P_{h_c} : Rotation from parent p to child c (fixed)
- ▶ \hat{u} : Axis of translation (fixed)
- ▶ l : Translation length (varying)
- ▶ $P_{S_c}(l) = \underbrace{P_{h_c}}_{\text{rotation}} + \underbrace{\frac{1}{2}l\hat{u} \otimes P_{h_c}\epsilon}_{\text{translation}}$

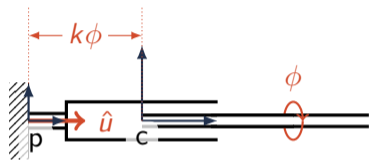
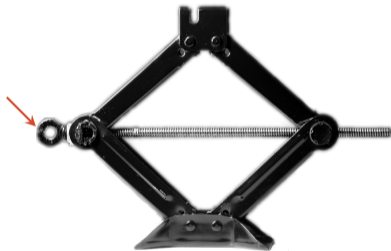
Helical Joint Animation



$${}^pS_c = \exp\left(\frac{2.91}{2}\hat{u}\right) + \frac{1}{2}(0.73)\hat{u} \otimes \exp\left(\frac{2.91}{2}\hat{u}\right)$$

Helical Joints

Coupled Rotation and Linear Motion



- ▶ k : thread pitch
- ▶ \hat{u} : Axis (fixed)
- ▶ ϕ : Rotation angle (varying)

$$\text{▶ } {}^p S_c(\phi) = \underbrace{\exp\left(\frac{\phi}{2}\hat{u}\right)}_{\text{rotation}} + \underbrace{\frac{1}{2}(k\phi\hat{u}) \otimes \exp\left(\frac{\phi}{2}\hat{u}\right)}_{\text{translation}} \epsilon$$

Multi-DoF Joints

Cylindrical: Revolute \otimes Prismatic

$${}^P S_c(l, \phi) = {}^P S_{c1}(l) \otimes {}^{c1} S_c(\phi)$$

Universal: Revolute \otimes Revolute

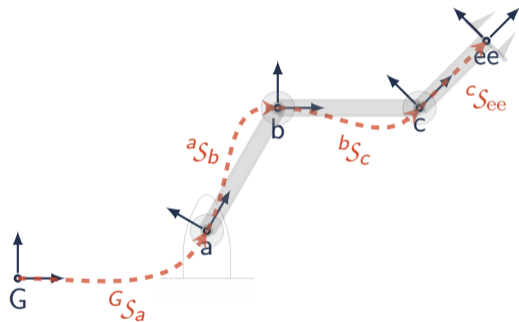
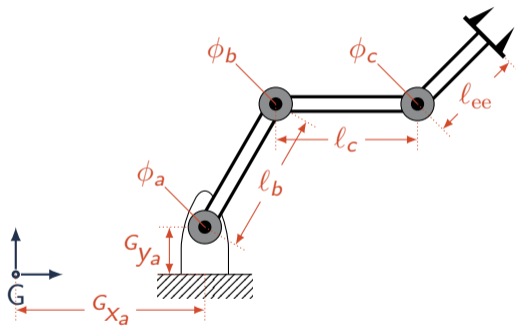
$${}^P S_c(\phi_0, \phi_1) = {}^P S_{c0}(\phi_0) \otimes {}^{c0} S_c(\phi_1)$$

Spherical: Revolute \otimes Revolute \otimes Revolute

$${}^P S_c(\phi_0, \phi_1, \phi_2) = {}^P S_{c0}(\phi_0) \otimes {}^{c0} S_{c1}(\phi_1) \otimes {}^{c1} S_c(\phi_2)$$

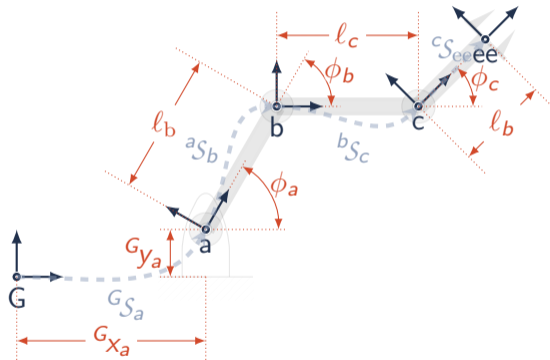
Products of revolute and prismatic joints

Serial Manipulator



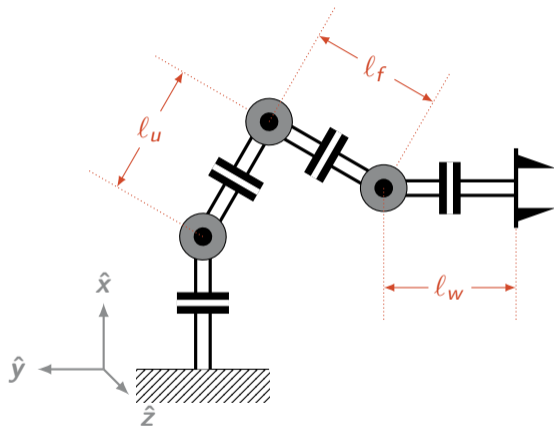
Serial Manipulator

Transforms



- ▶ **Relative:** $S = h + \frac{1}{2}\vec{v} \otimes h\epsilon$
 - ▶ ${}^G S_a = \exp\left(\frac{\phi_a}{2}\hat{k}\right) + \frac{1}{2}G_{v_a} \otimes \exp\left(\frac{\phi_a}{2}\hat{k}\right)\epsilon$
 - ▶ ${}^a S_b = \exp\left(\frac{\phi_b}{2}\hat{k}\right) + \frac{1}{2}l_b\hat{i} \otimes \exp\left(\frac{\phi_b}{2}\hat{k}\right)\epsilon$
 - ▶ ${}^b S_c = \exp\left(\frac{\phi_c}{2}\hat{k}\right) + \frac{1}{2}l_c\hat{i} \otimes \exp\left(\frac{\phi_c}{2}\hat{k}\right)\epsilon$
 - ▶ ${}^c S_{ee} = 1 + \frac{1}{2}l_{ee}\hat{i}\epsilon$
- ▶ **Absolute:** ${}^G S_n = {}^G S_m \otimes {}^m S_n$
 - ▶ ${}^G S_b = {}^G S_a \otimes {}^a S_b$
 - ▶ ${}^G S_c = {}^G S_b \otimes {}^b S_c$
 - ▶ ${}^G S_{ee} = {}^G S_c \otimes {}^c S_{ee}$

Anthropomorphic arm



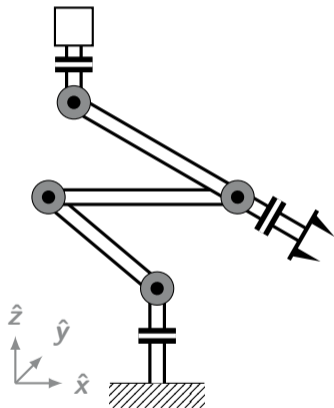
How do LWA4 joints map to human joints?

EOD 510 Packbot

Endeavor Robotics



<http://endeavorrobotics.com/products>



What's wrong (kinematically) with this design?

Outline

Manipulator DoF

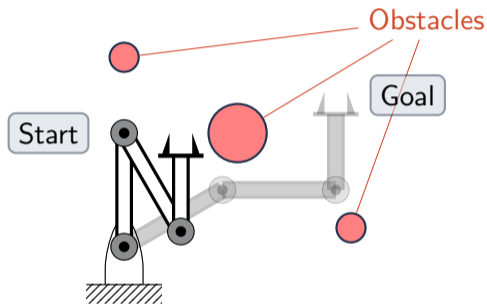
Manipulator Kinematics

Joint Transforms

Examples

The Motion Planning Problem

Motion Planning Illustration



Find collision free path from start to goal.

Paths

Definition: Path (Mathematical)

A path through configuration space \mathcal{Q} is a continuous function defining the configuration along a time-independent parameter:

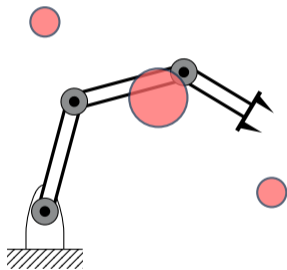
$$\tau : [0, 1] \mapsto \mathcal{Q}$$

Definition: Path (Programming Implementation)

A path is a sequence of points in configuration space \mathcal{Q} such that valid transitions exist between subsequent points.

$$\tau \in \mathcal{Q}^*$$

Collisions and Free Configuration Space



- ▶ Want paths in *Collision Free Space*
- ▶ Usually no explicit representation / parameterization
- ▶ Blackbox collision checking: $\text{is-valid} : \mathcal{Q} \mapsto \mathbb{B}$
- ▶ Obstacle Region: $\mathcal{Q}_{\text{obs}} = \{q \in \mathcal{Q} \mid \neg \text{is-valid}(q)\}$
- ▶ Free Space: $\mathcal{Q}_{\text{free}} = \{q \in \mathcal{Q} \mid \text{is-valid}(q)\}$
- ▶ Total Space:
 - ▶ $\mathcal{Q}_{\text{obs}} \cap \mathcal{Q}_{\text{free}} = \emptyset$
 - ▶ $\mathcal{Q}_{\text{obs}} \cup \mathcal{Q}_{\text{free}} = \mathcal{Q}$

Dynamics vs. Combinatorics

- ▶ What is the challenge?
 - ▶ Stability?
 - ▶ Dimensionality?
- ▶ Trade-offs:
 - ▶ Algorithmic completeness
 - ▶ Computational efficiency
 - ▶ Plan optimality
- ▶ Different perspectives:
 - ▶ Solving differential equations
 - ▶ State-space search

Understand the problem to solve.

Piano Mover's Problem

Given: Environment, Robot, Start and Goal configurations

- ▶ World \mathcal{W} in \mathbb{R}^2 or \mathbb{R}^3
- ▶ A robot in \mathcal{W} : either a single or collection of rigid bodies
- ▶ Configuration space \mathcal{Q} for the robot, from which \mathcal{Q}_{obs} and $\mathcal{Q}_{\text{free}}$ are derived.
- ▶ Initial configuration $q_0 \in \mathcal{Q}_{\text{free}}$
- ▶ Goal configuration $q_G \in \mathcal{Q}_{\text{free}}$

Find: A valid path from start q_0 to goal q_G .

Overview of Motion Planning Approaches

High-dimensional / Manipulation Problems

Local Search

1. Start from initial configuration or (possibly invalid) path
2. Progressively “improve” the configuration/path, via:
 - ▶ Gradient descent
 - ▶ Hill-Climbing
 - ▶ Sequential Optimization
 - ▶ Etc.
3. Terminate when:
 - ▶ We reach the goal/ collision free path
 - ▶ Reach a local minimum

Sampling-Based Planning

1. Start from initial configuration
2. Sample new states in the configuration space
3. Add valid states to tree or graph
4. Terminate when:
 - ▶ We find a path to the goal in the tree/graph
 - ▶ We exhaust a timeout