

MIN Faculty Department of Informatics



Introduction to Robotics Lecture 6

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Technical Aspects of Multimodal Systems

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Introduction Coordinate systems Kinematic Equations Robot Description Inverse Kinematics for Manipulators Differential motion with homogeneous transformations Jacobian Trajectory planning Trajectory generation Generation of trajectories Trajectories in multidimensional space Cubic polynomials between two configurations

Optimizing motion

Trajectory generation



Dynamics Principles of Walking Robot Control Task-Level Programming and Trajectory Generation Task-level Programming and Path Planning Task-level Programming and Path Planning Architectures of Sensor-based Intelligent Systems Summary Conclusion and Outlook



Definition

A trajectory is a time history of position, velocity and acceleration for each DOF

Describes motion of TCP frame relative to base frame

abstract from joint configuration

Series of discrete poses (TCP or joint configuration)

- usually fixed temporal intervals
- possibly fixed distances, key frames



Problem

- I am at point A and want move to point B.
 - How do I get to point B?
 - How long does it take me to get to point B?
 - Which constraints exist for moving from A to B?

Solution

- generate a possible trajectory
- trajectory planning
- describe intermediate poses (waypoints)



The methods for path generation should be applicable for

- calculation of cartesian trajectories for the TCP
- calculation for trajectories in joint space



Trajectory planning - Trajectory generation

Naive approach

Set the pose for the next time step (e.g. 10 ms later) to B.

- possible only in simulation
- ► the moving distance for a manipulator at the next time step may be too large (velocity approaches ∞)



Trajectory planning - Trajectory generation

Next best approach

- divide distance between A and B to shorter (sub-)distances
- use linear interpolation for these (sub-)distances
- respect the maximum velocity constraint



Linear interpolation – visualization

Trajectory planning - Trajectory generation

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Linear interpolation – constraints

Trajectory planning - Trajectory generation

Problem

The physical constraints are violated

- joint velocity is limited by maximum motor rotation speed
- joint acceleration is limited by maximum motor torque

Implicitly these contraints are valid for motion in cartesian space.

 robot dynamics (joint moments resulting from the robot motion) affect the boundary condition

Solution

- dynamical trajectory planning
- \blacktriangleright advanced optimization methods \rightarrow current topic of research

Linear interpolation – improvement

Trajectory planning - Trajectory generation

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Next best approach

- Limitation of joint velocity and acceleration
- Two different methods
 - trapezoidal interpolation
 - polynomial interpolation



Trapezoidal interpolation – visualization

Trajectory planning - Trajectory generation

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Trajectory planning - Trajectory generation

- consider joint velocity and acceleration contraints
- optimal time usage (move with maximum acceleration and velocity)
- acceleration is not differentiable (the jerk is not continuous)
- start and end velocity equals 0
 - not sensible for concatenating trajectories
 - improved by polynomial interpolation

Trapezoidal interpolation – constraints

Trajectory planning - Trajectory generation

Problem

Multidimensional trapezoidal interpolations

- different run time for joints (or cartesian dimensions)
- multiple velocity and acceleration contraints
- results in various time switch points
 - from acceleration to continuous velocity
 - from continuous velocity to deceleration
 - moving along a line in joint/cartesian space is impossible.

Solution

- Normalization to the slowest joint
- ▶ Use jerk and arrival time of the slowest joint instead of velocity.

Trapezoidal interpolation – normalization

Trajectory planning - Trajectory generation

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Normalize to the slowest joint



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Trapezoidal interpolation – normalization (cont.)

Trajectory planning - Trajectory generation

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Normalize to the slowest joint





Trajectory planning - Trajectory generation

Consider velocity and acceleration boundary conditions

- calculation of extremum and duration of trajectory
- Acceleration differentiable
 - continous jerk
 - smooth trajectory
 - interesting only in the theory for momentum control
- Start and end velocity may be $\neq 0$
 - sensible for concatenating trajectories



Trajectory planning - Trajectory generation

- Usually a polynom with degree of 3 (cubic spline) or 5
- Calculation of coefficient with respect to boundary constraints
 - ▶ 3rd-degree polynomial: consider 4 boundary constraints
 - position and velocity; start and goal
 - ▶ 5th-degree polynomial: consider 6 boundary constraints
 - position, velocity and acceleration; start and goal

Polynomial interpolation (cont.)

Trajectory planning - Trajectory generation

Example 5th-degree

$$f(x) = a_0 + a_1x + a_2x^2 + a_3x^3 + a_4x^4 + a_5x^5$$

Boundary conditions for start $(x = t_0)$ and goal $(x = t_d)$:

t: formal time from the interval [0;1]

Proper position interpolation from start (A) to goal (B)

$$P(t) = Af(t) + Bf(1-t)$$

Polynomial interpolation (cont.)

Trajectory planning - Trajectory generation

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Trajectory planning - Trajectory generation

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Boundary constraints (cont.) Pick-and-Place example

Trajectory planning - Trajectory generation

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Pick $pos_{Start} = object$, $vel_{Start} = 0$, $acc_{Start} = 0$ Lift-off limited velocity and acceleration Motion continuous via waypoints, full velocity and acceleration Set-down similar to Lift-off Place similar to Pick



Trajectory planning - Generation of trajectories

Task

find trajectory for moving the robot from start to goal pose

- calculate
- interpolate
- approximate
- use continous functions of time

Solution:

- Cartesian space
- Joint Space



Trajectory planning - Generation of trajectories

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Cartesian space:

- near to the task specification
- advantageous for collision avoidance

Trajectory planning - Generation of trajectories

Joint space:

- no inverse kinematics in joint space required
- the planned trajectory can be immediately applied
- physical joint constraints can be considered





Trajectory planning - Trajectories in multidimensional space

- Changes in position, velocity and acceleration of all joints are analyzed over a period of time
- Trajectory with n DOF is a parameterized function q(t) with values in its motion region.
- ► Trajectory q(t) of a robot with n DOF is then a vector of n parameterized functions q_i(t), i ∈ {1...n} with one common parameter t:

$$q(t) = [q_1(t), q_2(t), \dots, q_n(t)]^T$$



Trajectory planning - Trajectories in multidimensional space

- A trajectory is C^k-continuous, if all derivatives up to the k-th (including) exist and are continuous.
- A trajectory is called *smooth*, if it is at least C^2 -continuous
- q(x) is the trajectory,
- $\dot{q}(x)$ is the velocity,
- $\ddot{q}(x)$ is the acceleration,
- $\ddot{q}(x)$ is the jerk

Remarks on generation of trajectories

Trajectory planning - Trajectories in multidimensional space

- The smoothest curves are generated by infinitly often differentiable functions.
 - ► e^x
 - sin(x), cos(x)
 - ▶ log(x) (for x > 0)
 - ...
- Polynomials are suitable for interpolation
 - Problem: oscillations caused by a degree which is too high
- Piecewise polynomials with specified degree are applicable
 - cubic polynomial
 - splines
 - B-Splines

▶ ...

Cubic polynomials between two configurations

Trajectory planning - Cubic polynomials between two configurations

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• third-degree polynomial \Rightarrow four constraints:

$$\theta(t) = a_0 + a_1t + a_2t^2 + a_3t^3$$

if the start and end velocity is 0 then

$$\theta(0) = \theta_0 \tag{70}$$
$$\theta(t_f) = \theta_f \tag{71}$$
$$\dot{\theta}(0) = 0 \tag{72}$$
$$\dot{\theta}(t_f) = 0 \tag{73}$$

Cubic polynomials between two configurations (cont.)

Trajectory planning - Cubic polynomials between two configurations

eq.

eq.

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The solution

(70)
$$a_0 = \theta_0$$

(72) $a_1 = 0$
 $a_2 = \frac{3}{t_f^2}(\theta_f - \theta_0)$
 $a_3 = -\frac{2}{t_f^3}(\theta_f - \theta_0)$

Cubic polynomials with waypoints and velocities

Trajectory planning - Cubic polynomials between two configurations

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- Similar to the previous example:
 - positions of waypoints are given (same)
 - velocities of waypoints are different from 0 (different)

$\theta(0) = \theta_0$	(74)

$$\theta(t_f) = \theta_f \tag{75}$$

$$\dot{\theta}(0) = \dot{\theta}_0 \tag{76}$$

$$\dot{\theta}(t_f) = \dot{\theta}_f \tag{77}$$

Cubic polynomials with waypoints and velocities (cont.)

Trajectory planning - Cubic polynomials between two configurations

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► The solution

eq. (74)
$$a_0 = \theta_0$$

eq. (76) $a_1 = \dot{\theta}_0$
 $a_2 = \frac{3}{t_f^2}(\theta_f - \theta_0) - \frac{2}{t_f}\dot{\theta}_0 - \frac{1}{t_f}\dot{\theta}_f$
 $a_3 = -\frac{2}{t_f^3}(\theta_f - \theta_0) + \frac{1}{t_f^2}(\dot{\theta}_f + \dot{\theta}_0)$

Velocity calculation at the waypoints

Trajectory planning - Optimizing motion

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- Manually specify waypoints
 - based on cartesian linear and angle velocity of the tool frame
- Automatic calculation of waypoints in cartesian or joint space
 - based on heuristics
- Automatic determination of the parameters
 - based on continous acceleration at the waypoints

Factors for time optimal motion – Arc Length

Trajectory planning - Optimizing motion

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If the curve in the *n*-dimensional K space is given by

$$\mathbf{q}(t) = [q^1(t), q^2(t), \dots, q^n(t)]^T$$

then the arc length can be defined as follows:

$$s = \int_0^t \left\| \dot{\mathbf{q}}(t) \right\|_2 dt$$

where $\|\dot{\mathbf{q}}(t)\|_2$ is the euclidean norm of vector $d\mathbf{q}(t)/dt$ and is labeled as a flow velocity along the curve.

$$\|\mathbf{x}\|_2 := \sqrt{x_1^2 + \dots + x_n^2}$$

Factors for time optimal motion – Arc Length (cont.)

Trajectory planning - Optimizing motion

Introduction to Robotics

With the following two points given $\mathbf{p}_0 = \mathbf{q}(t_s)$ und $\mathbf{p}_1 = \mathbf{q}(t_f)$, the arc length *L* between \mathbf{p}_s and \mathbf{p}_s is the int

the arc length L between \mathbf{p}_0 and \mathbf{p}_1 is the integral:

$$L = \int_{\mathbf{p}_1}^{\mathbf{p}_0} ds = \int_{t_s}^{t_f} \|\dot{\mathbf{q}}(t)\|_2 dt$$

"The trajectory parameters should be calculated in the way that the arc length L under the given constraints has the shortest possible value."

Factors for time optimal motion – Arc Length (cont.)

Trajectory planning - Optimizing motion

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trajectory of circle

$$q(t) = c(t) = [r\cos(t), r\sin(t)]^T$$

arc length L of circle (circumference)

$$L = \int_{0}^{2\pi} \|\dot{\mathbf{c}}(t)\|_{2} dt$$
(78)
= $\int_{0}^{2\pi} \left\| [-r\sin(t), r\cos(t)]^{T} \right\|_{2} dt$ (79)
= $\int_{0}^{2\pi} \sqrt{r^{2}(\sin^{2}(t) + \cos^{2}(t))} dt$ (80)
= $\int_{0}^{2\pi} r dt$ (81)
 $L = 2\pi r$ (82)

Factors for time optimal motion – Curvature

Trajectory planning - Optimizing motion

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Curvature

Defines the sharpness of a curve. A straight line has zero curvature. Curvature of large circles is smaller than of small circles.

At first the *unit vector* of a curve $\mathbf{q}(t)$ can be defined as

$$\mathbf{U} = \frac{d\mathbf{q}(t)}{ds} = \frac{d\mathbf{q}(t)/dt}{ds/dt} = \frac{\dot{\mathbf{q}}(t)}{|\dot{\mathbf{q}}(t)|}$$

If s is the parameter of the arc length and **U** as the unit vector is given, the **curvature** of curve $\mathbf{q}(t)$ can be defined as

$$\kappa(s) = \left| \frac{d\mathbf{U}}{ds} \right|$$

Factors for time optimal motion – Curvature (cont.)

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with
$$\kappa(s) = \left|rac{d {\sf U}}{ds}
ight| o {\sf curvature}$$

If the parameter t, the first derivative $\dot{\mathbf{q}} = d\mathbf{q}(t)/dt$ and the second derivative $\ddot{\mathbf{q}} = d\dot{\mathbf{q}}(t)/dt$ for the curve $\mathbf{q}(t)$ are given, then the *curvature* can be calculated from the following representation

$$\kappa(t) = rac{|\dot{\mathbf{q}} imes \ddot{\mathbf{q}}|}{|\dot{\mathbf{q}}^3|} = rac{\left(\dot{\mathbf{q}}^2 \cdot \ddot{\mathbf{q}}^2 - (\dot{\mathbf{q}} \cdot \ddot{\mathbf{q}})^2
ight)^{1/2}}{|\dot{\mathbf{q}}|^3}$$

where $\dot{\mathbf{q}} \times \ddot{\mathbf{q}}$ is the cross product and $\dot{\mathbf{q}} \cdot \ddot{\mathbf{q}}$ is the dot product

Factors for time optimal motion – Curvature (cont.)

with
$$q(t) = c(t) = [r\cos(t), r\sin(t)]^T \rightarrow \text{trajectory of a circle}$$

 $\dot{c}(t) = [-r\sin(t), r\cos(t)]^T$
 $\ddot{c}(t) = [-r\cos(t), -r\sin(t)]^T$
 $\dot{c}^2(t) = r^2\sin^2(t) + r^2\cos^2(t) = r^2$
 $\ddot{c}^2(t) = r^2\cos^2(t) + r^2\sin^2(t) = r^2$
 $\dot{c}(t) \cdot \ddot{c}(t) = r^2\sin(t)\cos(t) - r^2\cos(t)\sin(t) = 0$

Curvature of a circle

$$\kappa(t) = rac{\left(\dot{\mathbf{c}}^2 \cdot \ddot{\mathbf{c}}^2 - (\dot{\mathbf{c}} \cdot \ddot{\mathbf{c}})^2\right)^{1/2}}{|\dot{\mathbf{c}}|^3} = rac{\sqrt{r^4}}{r^3} = rac{1}{r}$$

Factors for time optimal motion – Bending Energy

Trajectory planning - Optimizing motion

The **bending energy** of a smooth curve $\mathbf{q}(t)$ over the interval $t \in [0, T]$ is defined as

$$E = \int_0^L \kappa(s)^2 ds = \int_0^T \kappa(t)^2 |\dot{\mathbf{q}}(t)| dt$$

where $\kappa(t)$ is the curvature of $\mathbf{q}(t)$.

"The bending energy E of a trajectory should be as small as possible under consideration of the arc length."

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Factors for time optimal motion – Motion Time

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If a motion consists of *n* successive segments

$$q_j, j \in \{1 \dots n\}$$

then

$$u_j = t_{j+1} - t_j$$

is the required time for the motion in the segment \mathbf{q}_j . The total motion time is

$$T = \sum_{j=1}^{n-1} u_j$$

The borders for the minimum motion time T_{min} for the trajectory $\mathbf{q}_i^i(t)$ are defined over dynamical parameters of all joints.

For joint $i \in \{1 \dots n\}$ of trajectory part $j \in \{1 \dots m\}$ this kind of constraint can be described as follows

$$\begin{aligned} |\dot{q}_{j}^{i}(t)| &\leq \dot{q}_{max}^{i} \end{aligned} \tag{83} \\ |\ddot{q}_{i}^{i}(t)| &\leq \ddot{q}_{max}^{i} \end{aligned} \tag{84}$$

$$|m_{i}^{i}(t)| \leq m_{max}^{i}$$
(85)

- *mⁱ* is the torque (moment of force) for the joint *i* and can be calculated from the dynamical equation (motion equation).
- ▶ qⁱ_{max}, qⁱ_{max} and mⁱ_{max} represent the important parameters of the dynamical capacity of the robot.



Trajectory planning - Optimizing motion

Introduction to Robotics

- Waypoints cannot be realized
 - workspace boundaries, object collision, self-collision
- Velocities in the vicinity of singular configurations are too high
- Start and end configurations can be achieved, but there are different solutions
 - ambiguous solutions



- The following algorithm should create the smallest set of waypoints in the joint space under a predefined deviation e > 0.
- ► Therefore the deviation between the trajectory q(t) and the given line < w₀, w₁ > must be smaller than ε.

Algorithm(Bounded_Deviation)

- 1. Calculation of possible configurations $\mathbf{q}_0, \mathbf{q}_1$ from $\mathbf{w}_0, \mathbf{w}_1$ with the help of the inverse kinematics.
- 2. Calculation of the center in joint space:

$$\mathbf{q}_m = \frac{\mathbf{q}_0 + \mathbf{q}_1}{2}$$



Trajectory planning - Optimizing motion

 Calculation of the corresponding point of q_m in the workspace with usage of direct kinematics:

$$\mathbf{w}_m = W(\mathbf{q}_m)$$

4. Calculation of the center in the workspace:

$$\mathbf{w}_M = \frac{\mathbf{w}_0 + \mathbf{w}_1}{2}$$

- 5. If the deviation $||\mathbf{w}_{\mathbf{m}} \mathbf{w}_{\mathbf{M}}|| \ge \epsilon$, then cancel; else add the \mathbf{w}_{M} as node point between \mathbf{w}_{0} and \mathbf{w}_{1} .
- Recursive application of the algorithm for two new segments (w₀, w_M) und (w_M, w₁).



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