

Robot trajectory generation

Václav Hlaváč

Czech Technical University in Prague

Czech Institute of Informatics, Robotics and Cybernetics

160 00 Prague 6, Jugoslávských partyzánů 1580/3, Czech Republic

<http://people.ciirc.cvut.cz/hlavac>, vaclav.hlavac@cvut.cz

also Center for Machine Perception, <http://cmp.felk.cvut.cz>

Courtesy: Alessandro De Luca, Claudio Melchiorri; other presentations from the web.

Outline of the talk:

- ◆ Trajectory \times path.
- ◆ Trajectory generation, problem formulation.
- ◆ Trajectory in operational and joint spaces.
- ◆ Curve approximation.
- ◆ Trajectory classification; Dubins curves.
- ◆ Trajectory approximation by polynomials.

Practical requirements related to trajectories

Note: We consider a robotic manipulator (open kinematic chain) for simplicity. Generalization of methods for mobile robots is not difficult.

- ◆ Endow the robot with the capability to move the manipulator arm and its end effector or the mobile robot from the initial posture to the final posture.
- ◆ Motion laws have to be considered in order not to:
 - violate saturation limits of joint drives;
 - excite the resonant modes of the actuator mechanical structure, being driven by electric/hydraulic/pneumatic or other more exotic drives.
- ◆ Explore path planning and trajectory generation methods providing smooth trajectories solving the practical robotics task.

Question: Why are smooth trajectories preferred over jerky ones?

Terminology: path vs. trajectory

- ◆ *Note: Terms **path**, **trajectory** are often exchanged because they are perceived as synonyms informally.*
- ◆ **Path** consists of ordered locii of points in the space (either the joint space or the operational space), which the robot should follow.
 - The path provides a pure geometric description of motion.
 - The path is usually planned globally taking into account obstacle avoidance, traversing a complicated maze, etc.
- ◆ **Trajectory** is a path plus velocities and accelerations in its each point.
 - The design of a trajectory does not need global information, which simplifies the task significantly.
 - The trajectory is specified and designed locally (divide and conquer methodology). Parts of the path are covered by individual trajectories.
 - It is often required that pieces of trajectories join smoothly, which induces that a single trajectory design takes into account only neighboring trajectories from the path.

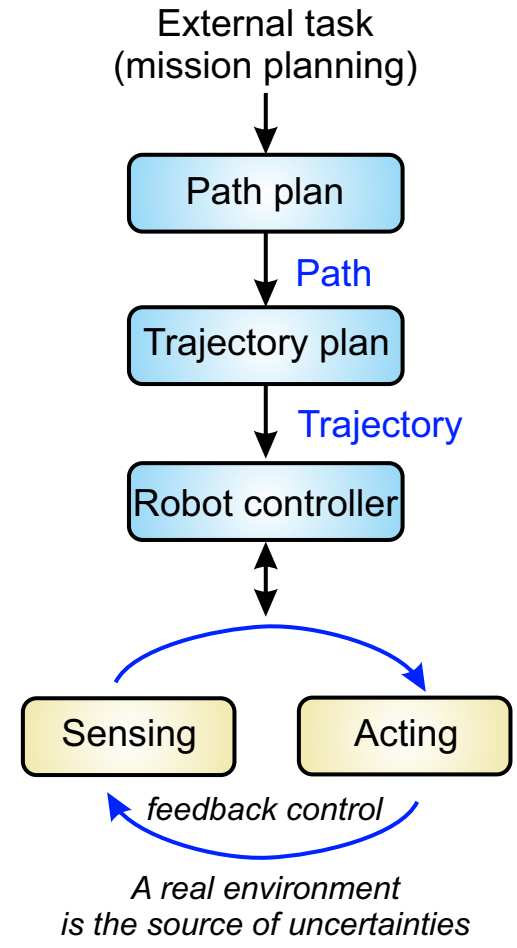
Robot motion planning, an overview

Path planning (global)

- ◆ The (geometric) path is a sequence of waypoints defining the trajectory coarsely.
- ◆ Issues solved at this level: obstacle avoidance, shortest path.

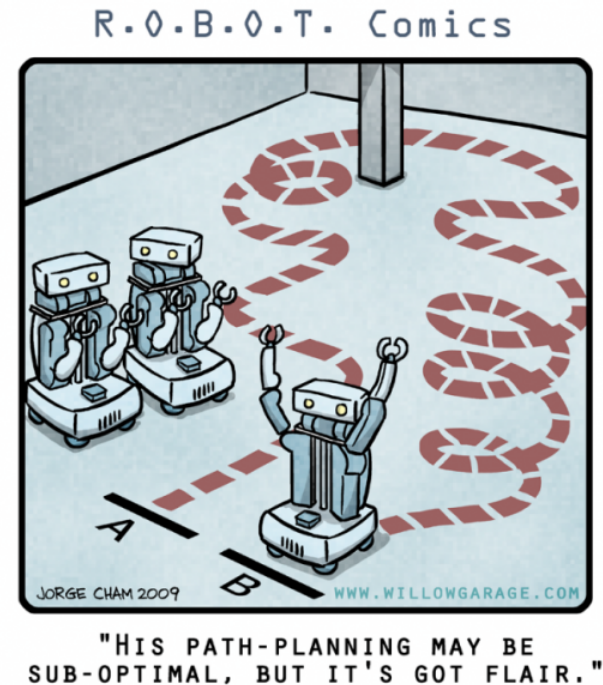
Trajectory generating (local)

- ◆ The path provided by path planning constitutes the input to the trajectory generator.
- ◆ Trajectory generator “approximates” the desired path waypoints by a class of polynomial functions and
- ◆ generates a time-based control sequence moving the manipulator/mobile platform from its initial configuration to the destination.



Path planning, the problem formulation

- ◆ The path planning task:
Find a collision free path for the robot from one configuration to another configuration.
- ◆ Path planning is an algorithmically difficult search problem.
 - The involved task has an exponential complexity with respect to the degrees of freedom (controllable joints).
 - With industrial robots, the path planning is often replaced by a path/trajectory taught in by a human operator.



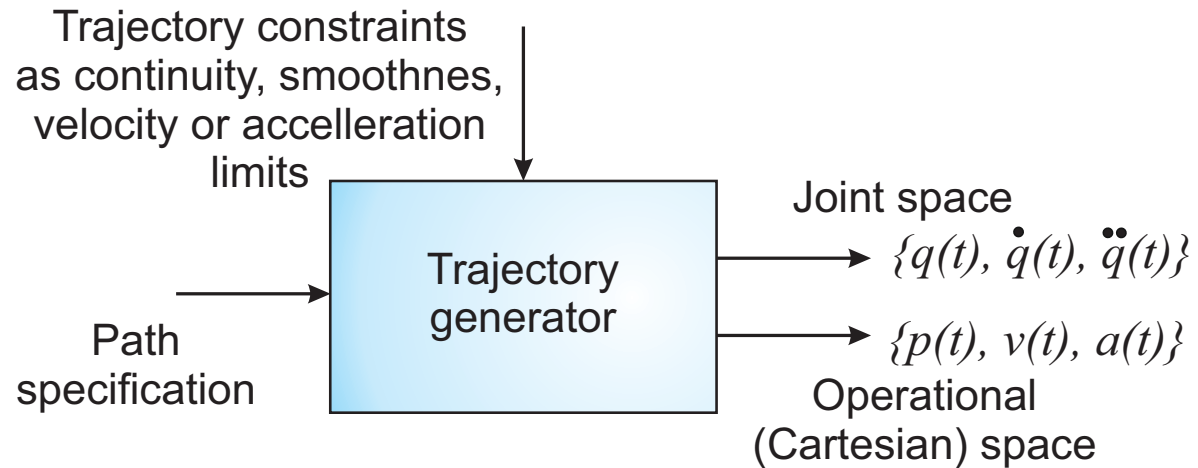
Note: A separate (next) lecture will be devoted to the path planning.

Trajectory generation, the problem formulation

- ◆ Trajectory generation aims at creating inputs to the motion control system, which ensures that the planned trajectory is smoothly executed.
- ◆ The planned path is typically represented by way-points, which is the sequence of points (or end-effector poses) along the path.
- ◆ Trajectory generating = creating a trajectory connecting two or more way-points.
 - In the industrial settings, a trajectory is taught-in by a human expert and later played back (by teach-and-playback).

A more recent approach utilizes several tens of trajectories performed by human experts as the input. They vary statistically. Machine learning techniques are used to create the final trajectory.
 - In general, e.g., with mobile robots and more and more with industrial robots, the path points generated by the path planner are smoothly approximated using function approximation methods from mathematics.

Trajectory generating, illustration



- ◆ Trajectory generating = finding the desired joint space trajectory $\mathbf{q}(t)$ given the desired operational (Cartesian) path inverse kinematics.
- ◆ \mathbf{q} is a vector of joint parameters. Its dimension matches to the number of DOFs.
- ◆ $\mathbf{p}(t) = (x(t), y(t), z(t))$ is the position, $\mathbf{v}(t) = (x'(t), y'(t), z'(t))$ is the velocity, $\mathbf{a}(t) = (x''(t), y''(t), z''(t))$ is the acceleration.

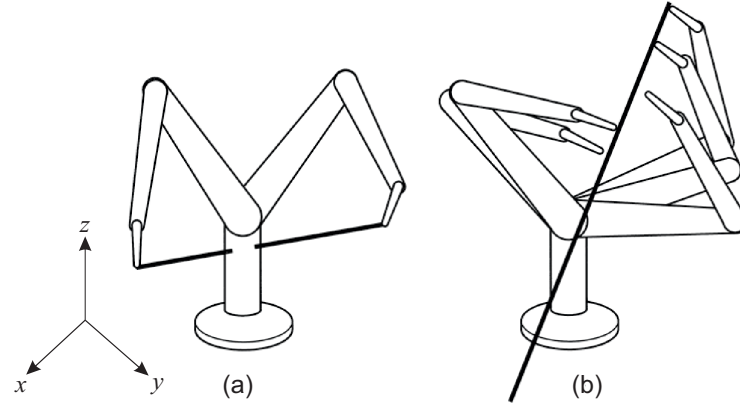
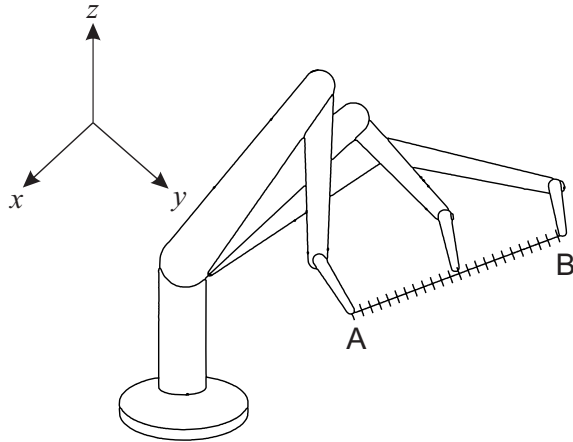
Joint space vs. operational space

- ◆ Joint-space description:
 - The description of the motion to be made by the robot by its joint values.
 - The motion between the two points is unpredictable.
- ◆ Operational space description:
 - Operational space = (Cartesian space or working space) in many cases.
 - The motion between the two points is known and controllable at all times.
 - It is easy to visualize the trajectory, but it is difficult to ensure that singularity does not occur.

The path planning/trajectory generation can be done in both the joint space or operational spaces.

Example: Troubles with the operational (Cartesian) space

Task: Create a trajectory of the manipulator end effector to follow a straight line.



- (a) The trajectory specified in Cartesian coordinates may force the robot to run into itself, see Figure (a) above;
- (b) The trajectory may requires a sudden change in the joint angles due to singularities, see Figure (b) above.

Trajectory in the operational space

- ◆ Calculate the path from the initial point to the final point.
- ◆ Assign a total time T_{path} to traverse the path.
- ◆ Discretize points in time and space.
- ◆ Blend a continuous time function between these points.
- ◆ Solve inverse kinematics at each step.

Advantages

- ◆ Collision free path can be obtained.

Disadvantages

- ◆ Computationally expensive due to involved inverse kinematics.
- ◆ It is unknown how to set the total time T_{path} .

Trajectory in the joint space

- ◆ Calculate the inverse kinematics solution from the initial point to the final point.
- ◆ Assign the total time T_{path} using maximal velocities in joints.
- ◆ Discretize the individual joint trajectories in time.
- ◆ Blend a continuous function between these point.

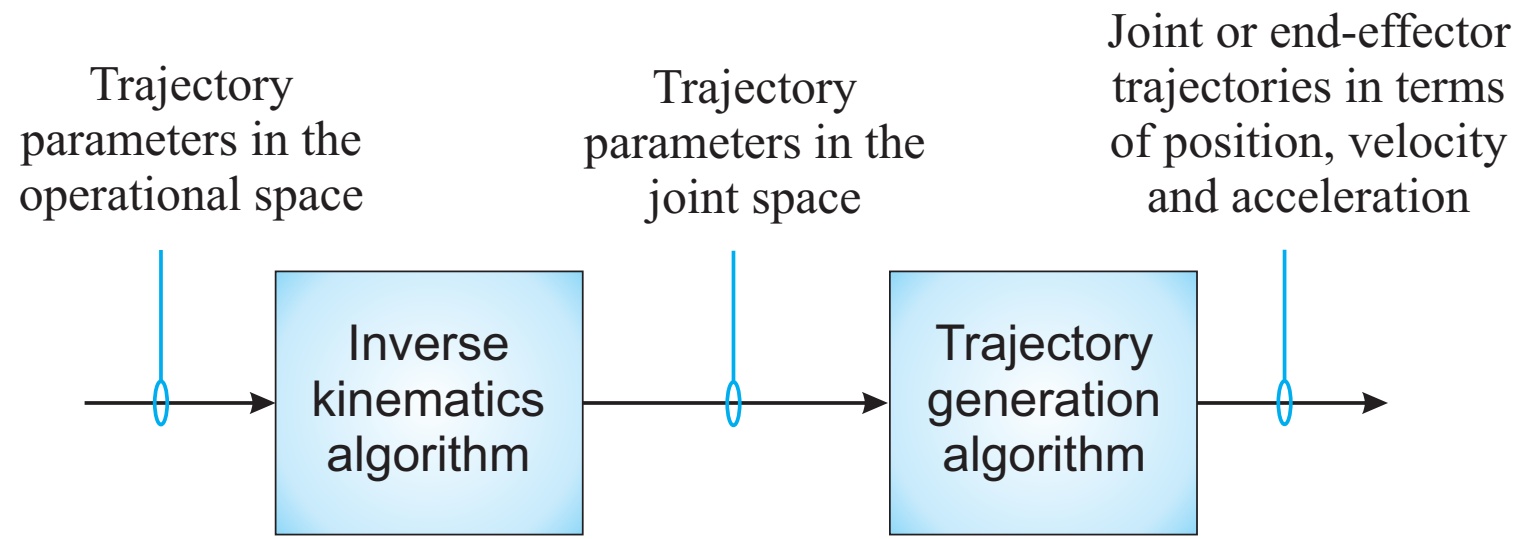
Advantages

- ◆ The inverse kinematics is computed only once.
- ◆ The approach can take into account joint angles limits and velocity constraints easily.

Disadvantages

- ◆ It is more difficult to deal with obstacles represented in the operational space. Nevertheless, it is doable.

Trajectory from the operational space to the joint space



E.g., initial and final end-effector location, travelling time.

Types of the trajectory control

- ◆ **Displacement control** = control the end effector/mobile robot displacement, i.e., angles or positions in space, maybe including dynamics of motion.

Examples:

- Moving payloads.
- Painting objects.

- ◆ **Force control** = control of both displacement (as above) and applied force.

Examples:

- Machining.
- Grinding.
- Sanding.

Curve approximation, the motivation

- ◆ A draftsman uses “ducks” and strips of wood (splines) to draw curves.
- ◆ “Wooden splines” provide the second-order continuity
- ◆ and pass through control points.



Examples of functions used for trajectory interpolation:

- ◆ Polynomials of different orders.
- ◆ Linear functions with parabolic blends.
- ◆ Splines.

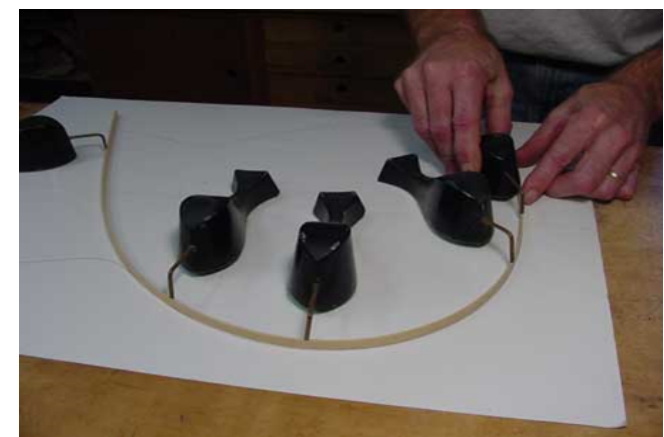


Illustration of trajectory generation issues

Different approaches will be demonstrated on a simple example.

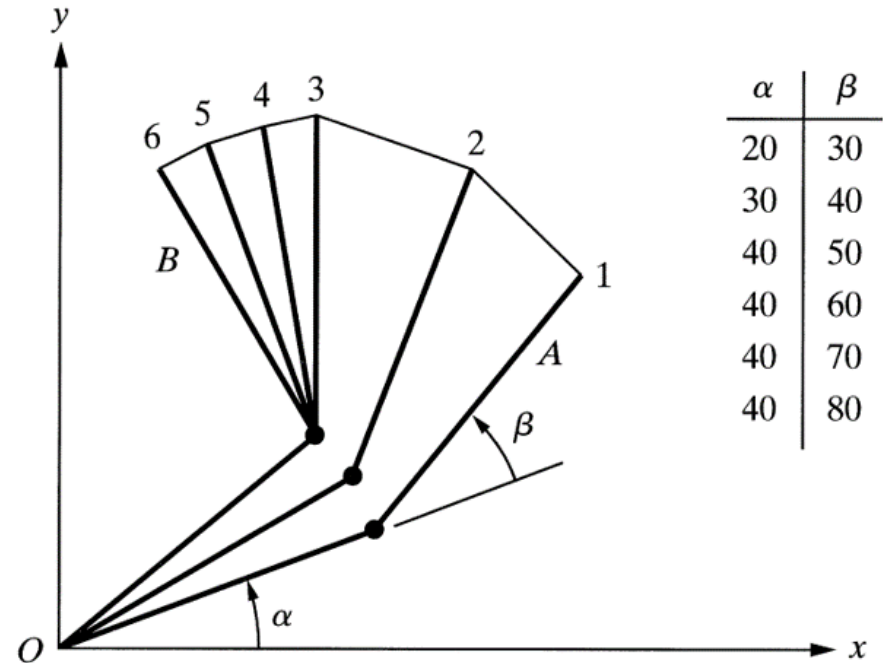
- ◆ Let us consider a simple two degrees of freedom robot.
- ◆ We desire to move the robot from Point A to Point B.
- ◆ Let us assume that both joints of the robot can move at the maximum rate of 10 degree/sec.

Note:

- ◆ Jerk (informally) = thrashing of the mechanism.
- ◆ Jerk (mechanics, formally) = the rate of velocity (time derivative of a trajectory) change.

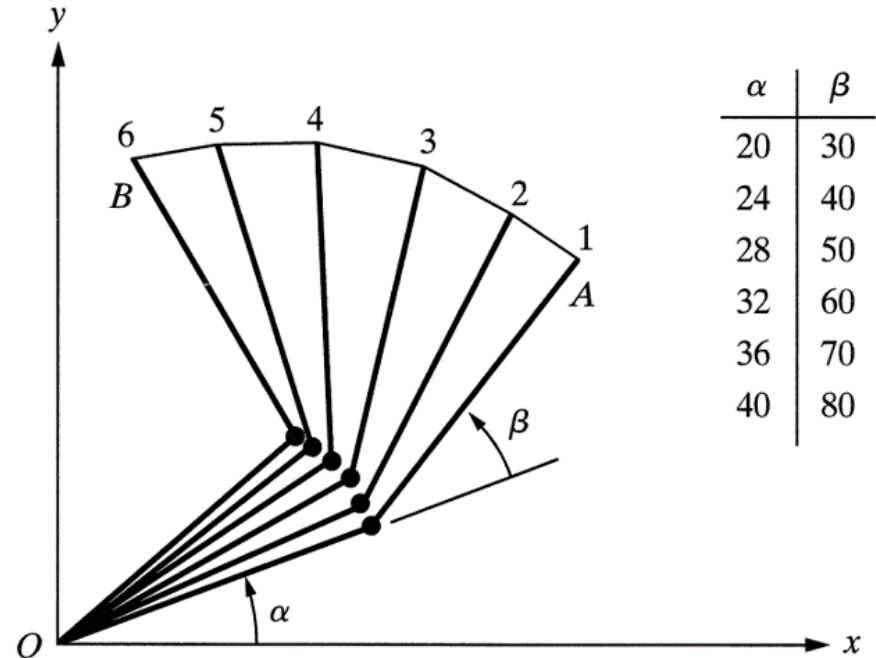
Non-normalized trajectory

- ◆ Move the robot from A to B, to run both joints at their maximum angular velocities.
- ◆ After 2 [sec], the lower link will have finished its motion, while the upper link continues for another 3 [sec]
- ◆ The path is irregular and the distances traveled by the robot end point are not uniform.



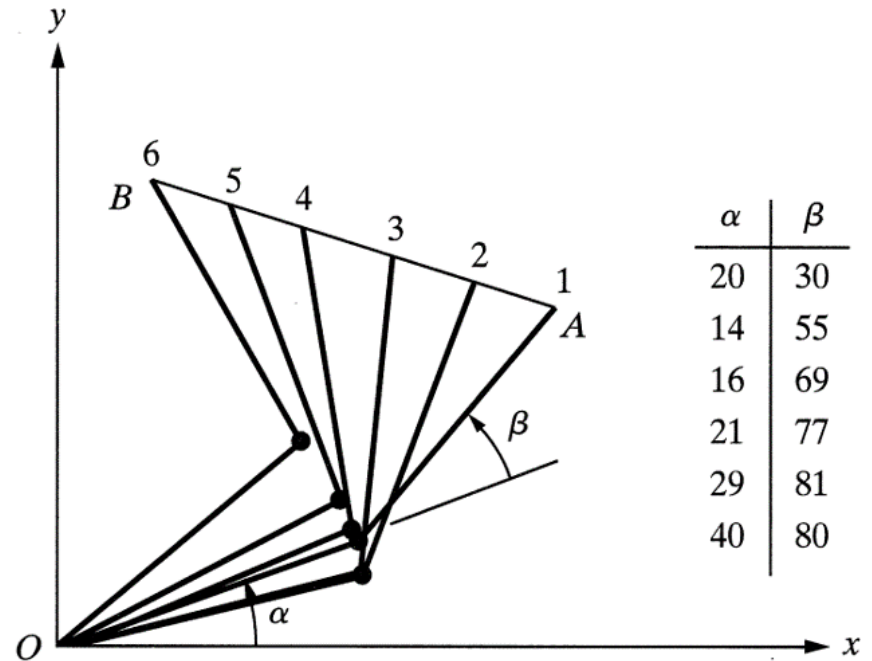
Normalized trajectory

- ◆ Let us assume that the motions of both joints are normalized by a common factor such that the joint with smaller motion will move proportionally slower and the both joints will start and stop their motion simultaneously.
- ◆ Both joints move at different speeds, but move continuously together.
- ◆ The resulting trajectory will be different.



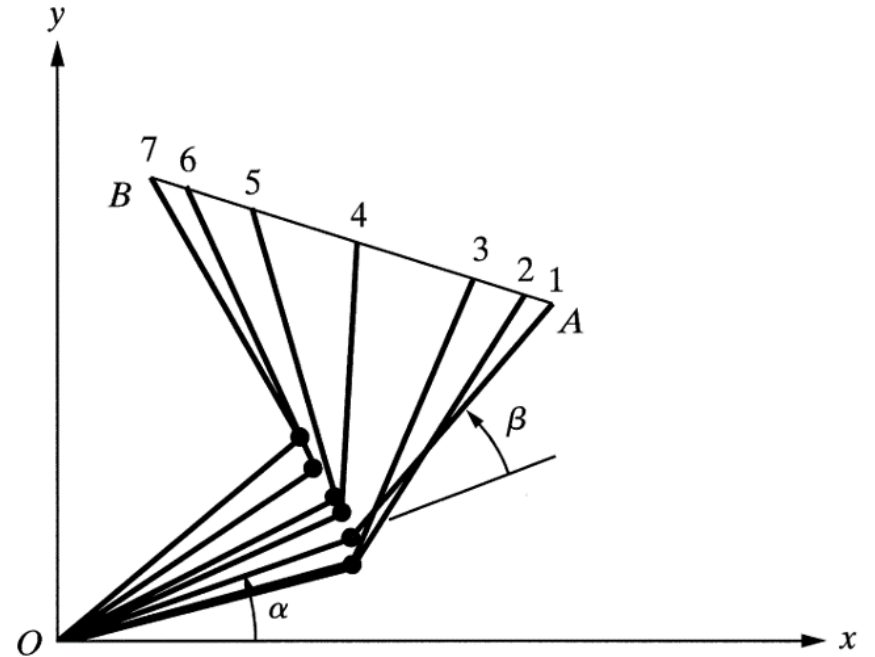
Straight line trajectory, version A

- ◆ Let us assume that the robot gripper follows a straight line between points A and B.
- ◆ The simplest way is to draw a line (interpolate) between A, B.
- ◆ Divide the line into five segments and solve for necessary angles α and β at each point.
- ◆ The joint angles do not change uniformly.



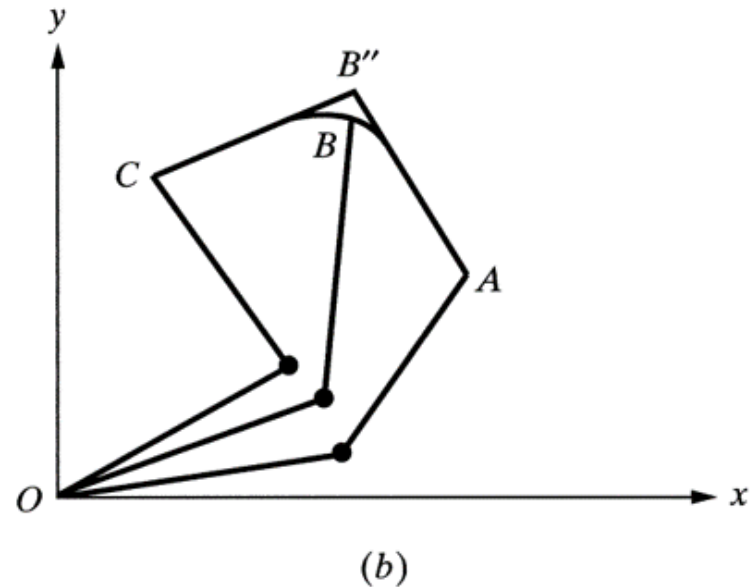
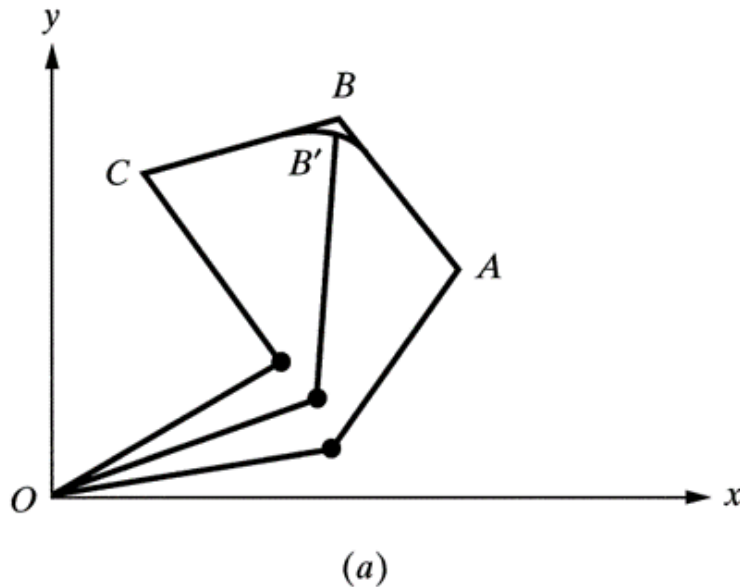
Straight line trajectory, version B

- ◆ Again interpolation between A, B by a straight line.
- ◆ The aim is to accelerate at the beginning and decelerate at the end.
- ◆ Divide the segments differently.
 - The arm moves at smaller segments as we speed up at the beginning;
 - Go at a constant cruising rate;
 - Decelerate with smaller segments as approaching Point B.



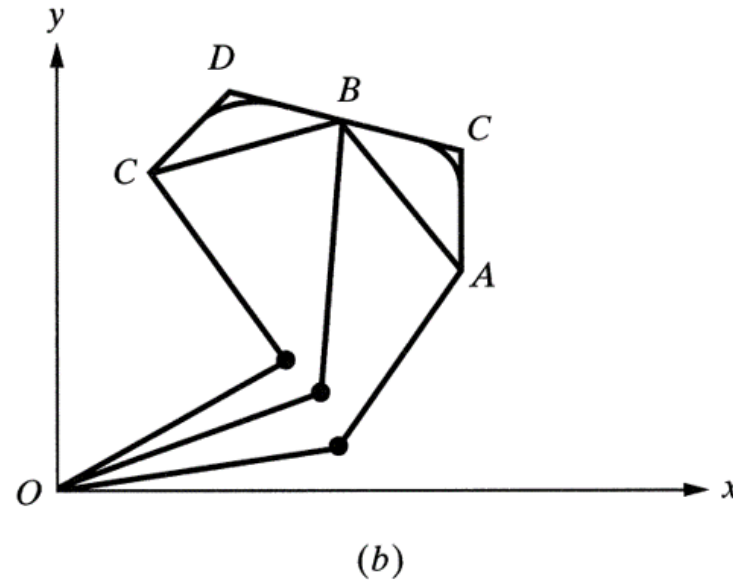
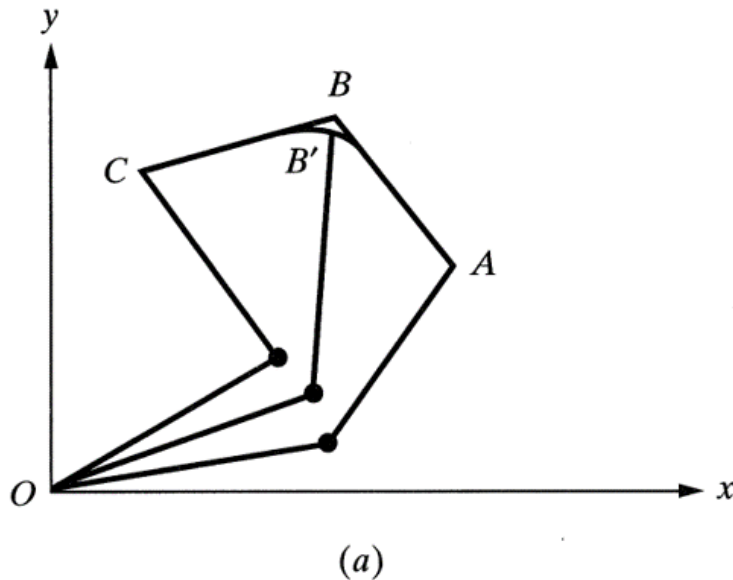
Continuous transition, version A

- ◆ Stop-and-go motion through the way-point list creates jerky motions with unnecessary stops.
- ◆ Stop-and-go motion through the way-point list creates jerky motions with unnecessary stops.
- ◆ How? Blend the two portions of the motion at Point B.

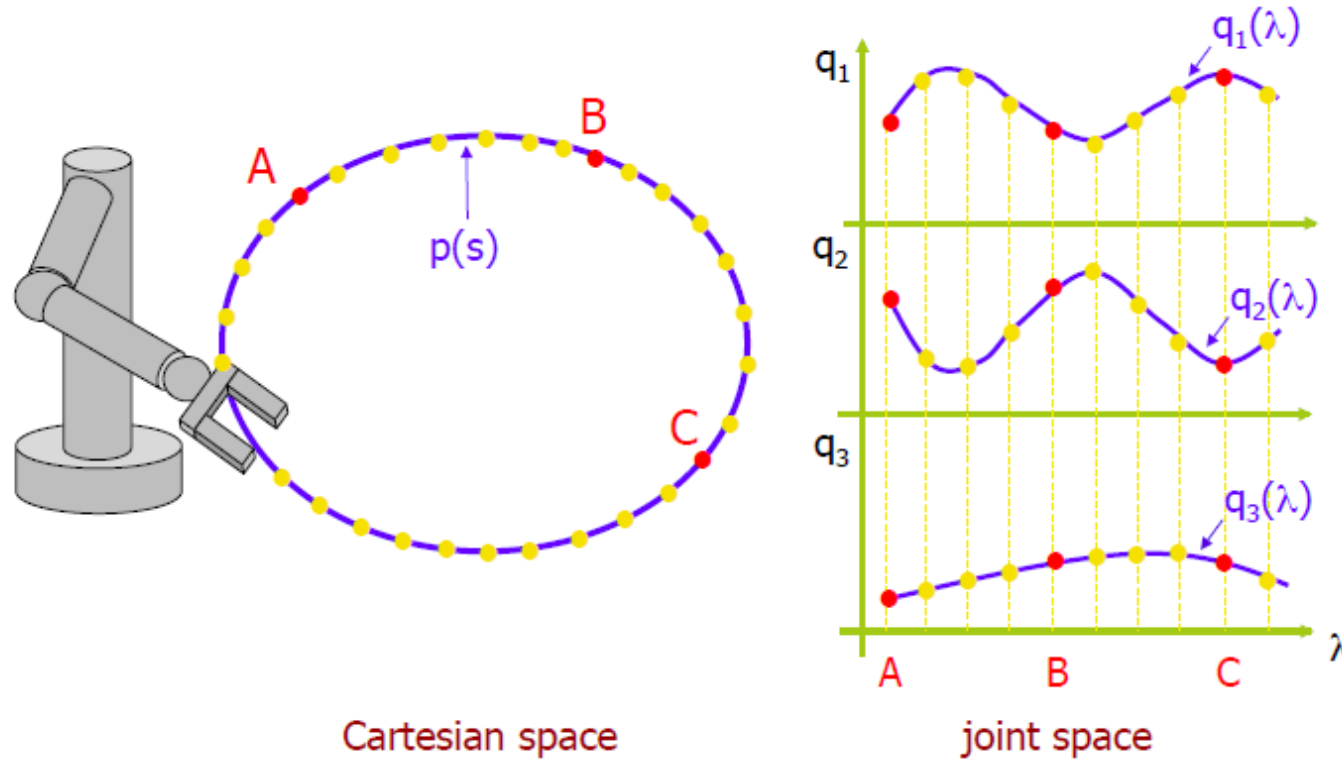


Continuous transition, version B

- ◆ Alternative scheme ensuring that the trajectory passes through control points.
- ◆ Two way-points D and E are picked such that Point B will fall on the straight-line segment of the segment ensuring that the robot will pass through Point B .



Motivating example of trajectories in operational and joint spaces



Courtesy for the image: Alessandro De Luca, Universita di Roma, La Sapienza.

Trajectory = path + timing law (1)

Having the path (approximated way-points), the trajectory is completed by a choice of a timing law.

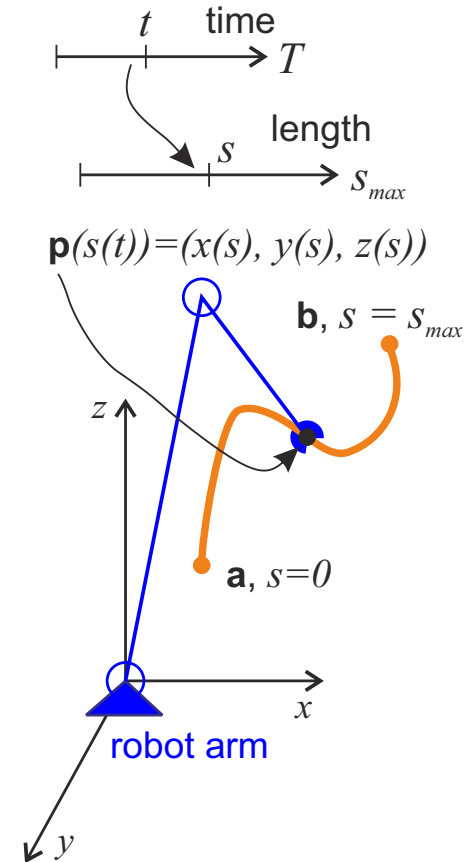
Consider the geometric path in a parametric form:

- ◆ Operational (Cartesian) space:

$\mathbf{p}(s) = (x(s), y(s), z(s))$, where the motion law is $s = s(t)$, $t \in [0, T]$.

- ◆ Joint space: $\mathbf{q}(\lambda) = (q_1(\lambda), q_2(\lambda), \dots, q_n(\lambda))$, where $n = \text{DOFs} \Rightarrow$ motion law $\lambda = \lambda(t)$.

If $s(t) = t$, the trajectory parametrization is the natural one given by the time.



Trajectory = path + timing law (2)

The timing law:

- ◆ is chosen based on task specifications (stop in a point, move at a constant velocity, etc.);
- ◆ It may consider optimality criteria (min transfer time, min energy, etc.);
- ◆ Constraints are imposed by actuator capabilities (e.g. max torque, max velocity) and/or by the task (e.g., the maximal allowed acceleration on a payload).

Space-time decomposition on parametrized path

E.g., in the operational (Cartesian) space:

$$\mathbf{p}(s) = (x(s), y(s), z(s)) \text{ with the motion law } s = s(t)$$

$$\text{velocity } \dot{\mathbf{p}}(t) = \frac{d\mathbf{p}}{ds} \dot{s}(t), \quad \text{acceleration } \ddot{\mathbf{p}}(t) = \frac{d\mathbf{p}}{ds} \ddot{s}(t) + \frac{d^2\mathbf{p}}{ds^2} \dot{s}(t)$$

Polynomial functions of a degree n are employed usually for way-points approximation.

$$s(t) = a_0 + a_1 t + a_2 t^2 + \dots + a_n t^n$$

Trajectories classification

- ◆ **Space of the definition**: the operational (Cartesian) space or the joint space.
- ◆ **Task type**: point-to-point (PTP), multiple points (knots), continuous, concatenated.
- ◆ **Path geometry**: rectilinear, Dubins, polynomial, exponential, cycloid, ...
- ◆ **Timing law**: bang-bang in the acceleration, trapezoidal in the velocity, polynomial, ...
- ◆ **Coordinated or independent**:
 - Motion of all joints (or of all Cartesian components) start and ends at the same instants (say, $t = 0$ and $t = T$) \Rightarrow the single timing law.
 - Motions are timed independently, e.g. according to requested displacement and robot capabilities. Such trajectory is performed in the joint space mostly.

Relevant characteristics of the trajectory

- ◆ Computational **efficiency** and memory space, e.g. store only coefficients of a polynomial function.
- ◆ **Predictability** vs. “wandering” out of the knots.
- ◆ **Accuracy** vs. the “overshot” on the final position.
- ◆ **Flexibility** allowing concatenation, over-fly, ...
- ◆ **Continuity** in space and in time.
Continuity C^1 at least. Sometimes also up to C^2 , i.e. up to jerk = $\frac{da}{dt}$, where a is the acceleration.

Note: Continuity of the class C^k means that the time derivative up to the order k is smooth.

Trajectory planning

Usually, the user has to specify only a minimum amount of information about the trajectory, such as initial and final points, duration of the motion, maximum velocity, and so on.

- ◆ **Work-space trajectories**

allow to consider directly possible constraints on the path (obstacles, path geometry, ...) that are more difficult to take into consideration in the joint space (because of the non linear kinematics)

- ◆ **Joint space trajectories**

are computationally simpler and allow to consider problems due to singular configurations, actuation redundancy, velocity/acceleration constraints.

Trajectory planning in the joint space

- ◆ The curve parametrization in time $q = q(t)$ or in joint space $q = q(\lambda)$, $\lambda = \lambda(t)$
- ◆ It is sufficient to work component-wise, (q_i in vector \mathbf{q}).
- ◆ An implicit definition of the trajectory is obtained by solving a problem with specified boundary conditions in a **given class of functions**.
- ◆ Typical classes of functions: **polynomials** (cubic, quintic, ...), (co)sinusoids, clothoids, etc.
- ◆ **Imposed conditions**
 - Passage through points \Rightarrow interpolation.
 - Initial, final, intermediate velocity or geometric tangent to the path.
 - Initial, final, intermediate velocity or geometric curvature.
 - Continuity up to C^k .

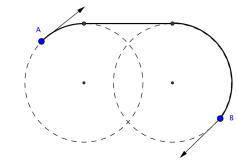
Many of the following methods and remarks can be applied directly also to Cartesian trajectory planning (and vice versa).

Dubins curves

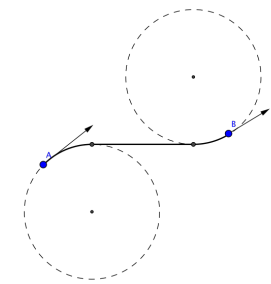
Lester Eli Dubins (1920–2010); result from 1957

The task is explained in 2D for simplicity; works in 3D as well.

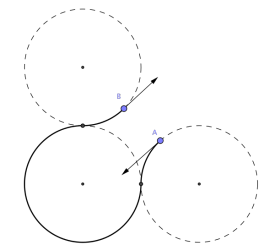
- ◆ Given: two points (start, target, both with prescribe tangent of the trajectory); Assumption: forward motion only.
- ◆ The optimization task is to find the shortest path between the start and the target when the maximum trajectory curvature is prescribed.
- ◆ L.E. Dubins proved that such path consists in a general case of at most (a) two maximum curvature two circular arcs and a straight line segment, or (b) three circular arcs.
- ◆ Let R - right turn; L - left turn; S - straight. The optimal trajectory will be one of six combinations RSR, RSL, LSR, LSL, RLR, LRL, see the figure.
- ◆ The methods are simple using algebraic equations.



RSR



RSL



LRL

A cubic polynomial

- ◆ Four boundary constraints:

$$q(0) = q_{\text{ini}}; q(T) = q_{\text{fin}}; \dot{q}(0) = v_{\text{ini}}; \dot{q}(T) = v_{\text{fin}}.$$

- ◆ $\Delta q = q_{\text{fin}} - q_{\text{ini}}$; the curve parametrization $\tau = t/T$, $\tau \in [0, 1]$.

- ◆ A cubic polynomial $q(\tau) = q_{\text{ini}} + \Delta q (a\tau^3 + b\tau^2 + c\tau + d)$.

- ◆ A “doubly normalized” polynomial $q_N(\tau)$ such that

- $q_N(0) = 0 \Leftrightarrow d = 0$.

- $q_N = 1 \Leftrightarrow a + b + c = 1$.

- $\dot{q}_N(0) = \left. \frac{dq_N}{d\tau} \right|_{\tau=0} = c = \frac{v_{\text{ini}}T}{\Delta q}$.

- $\dot{q}_N(1) = \left. \frac{dq_N}{d\tau} \right|_{\tau=1} = 3a + 2b + c = \frac{v_{\text{fin}}T}{\Delta q}$.

Cubic polynomial, a special case, rest-to-rest

- ◆ The boundary constraints and the parametrization remains as above.
- ◆ A cubic polynomial as well: $q(\tau) = q_{\text{ini}} + \Delta q (a\tau^3 + b\tau^2 + c\tau + d)$.
- ◆ A polynomial $q_N(\tau)$ such that
 - $q_N(0) = 0 \Leftrightarrow d = 0$.
 - $\dot{q}_N(0) = 0 \Leftrightarrow c = 0$.
 - $q_N = 1 \Leftrightarrow a + b = 1$.
 - $\dot{q}_N(1) = 0 \Leftrightarrow 3a + 2b = 0$.

From previous two equations, $a = -2$, $b = 3$.

Quintic polynomial

- ◆ $q(\tau) = a\tau^5 + b\tau^4 + c\tau^3 + d\tau^2 + e\tau + f$, i.e., 6 coefficients,
- ◆ satisfying constraints, e.g., in the normalized time τ :

$$q(0) = q_0; q(1) = q_1; \dot{q}(0) = v_0T; \dot{q}(1); \ddot{q}(0) = a_0T^2; \ddot{q}(1) = a_1T^2;$$

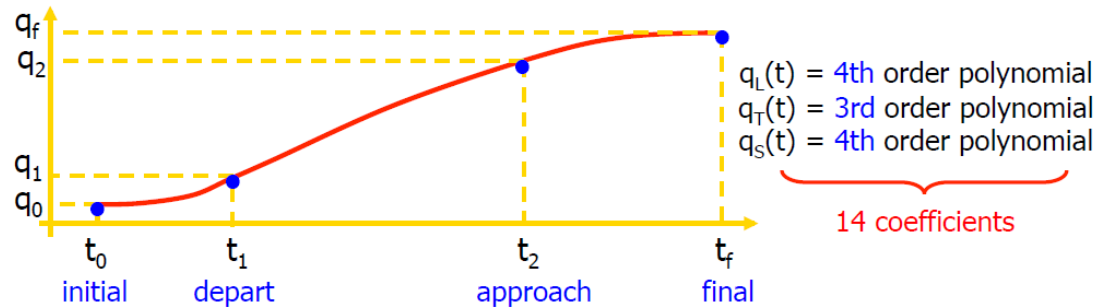
$$\begin{aligned} q(\tau) &= (1 - \tau)^3 \left(q_0 + (3q_0 + v_0T) \tau + (a_0T^2 + 6v_0T + 12q_0) \frac{\tau^2}{2} \right) \\ &= +\tau^3 \left(q_1 + (3q_1 - v_1T)(1 - \tau) + \frac{(a_1T^2 - 6v_1T + 12q_1)(1 - \tau)^2}{2} \right) \end{aligned}$$

A special case, rest-to-rest:

- ◆ $v_0 = v_1 = a_0 = a_1 = 0$.
- ◆ $q(\tau) = q_0 + \Delta q(6\tau^5 - 15\tau^4 + 10\tau^3)$; $\Delta q = q_1 - q_0$.

4-3-4 polynomials

Three phases in pick-and-place operations: Lift off, Travel, Set down.



Boundary constraints:

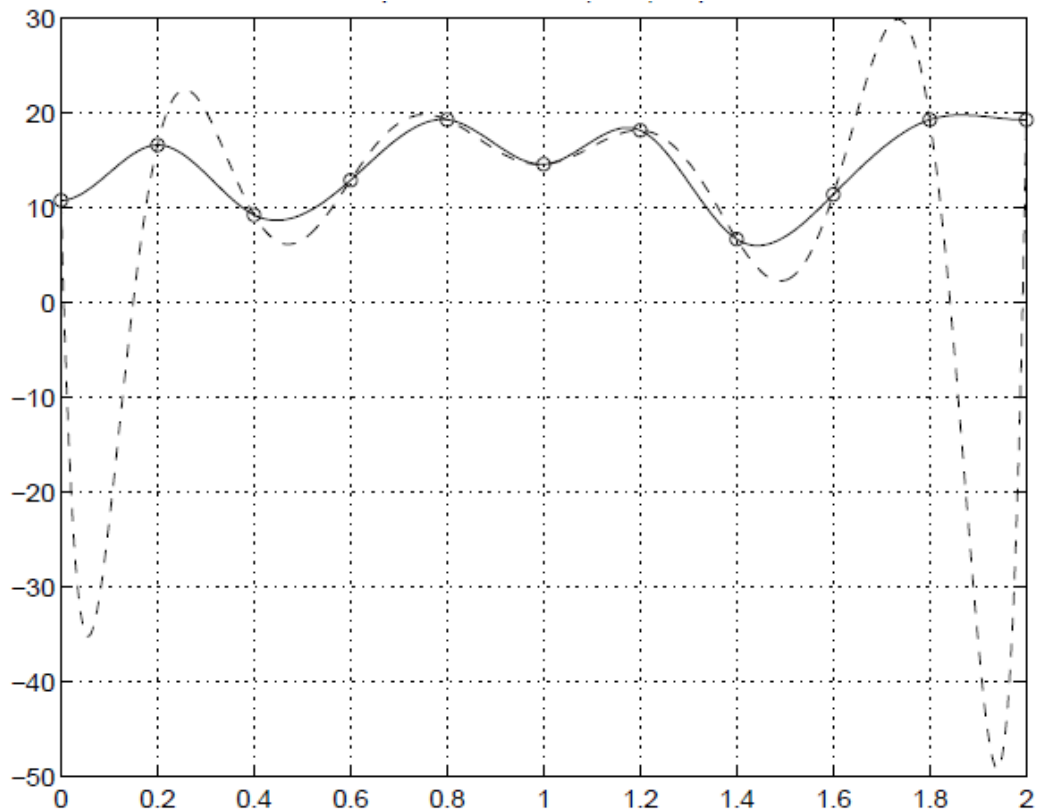
$$\begin{aligned}
 q(t_0) &= q_0, & q(t_1^-) &= q(t_1^+) = q_1, & q(t_2^-) &= q(t_2^+) = q_2, & q(t_f) &= q_f \\
 \dot{q}(t_0) &= \dot{q}(t_f) = 0, & \ddot{q}(t_0) &= \ddot{q}(t_f) = 0 \\
 \dot{q}(t_0^-) &= \dot{q}(t_0^+), & \ddot{q}(t_i^-) &= \ddot{q}(t_i^+), & i &= 1, 2
 \end{aligned}$$

The first equation corresponds to six passages; the second equation to four initial/final velocities/accelerations; the third equation to four continuity constraints.

Higher-order polynomials

- ◆ Higher-order polynomials provide a suitable solution class for satisfying symmetric boundary conditions in a point-to-point motion that imposes zero values on higher-order derivatives.
 - The interpolating polynomial is always of the odd degree.
 - The coefficients of such (doubly normalized) polynomial are always integers, alternate in sign, sum up to unity, and are zero for all term up to the power $= \frac{(\text{degree}-1)}{2}$.
- ◆ In all other cases (e.g., for interpolating a large number N of points), the use of higher-order polynomials is not recommended.
 - N -th order polynomials have $N - 1$ maximum and minimum points.
 - Oscillations arise out of the interpolation points (wandering).

Higher-order polynomials, oscillations

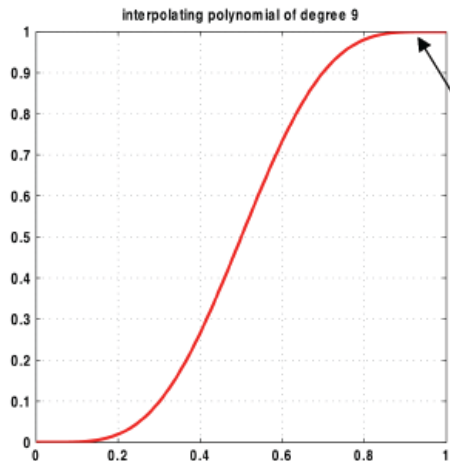


In general, given:

- 2 points \implies unique line
- 3 points \implies unique quadric
- ...
- n points \implies unique polynomial with degree $n - 1$

Numerical examples

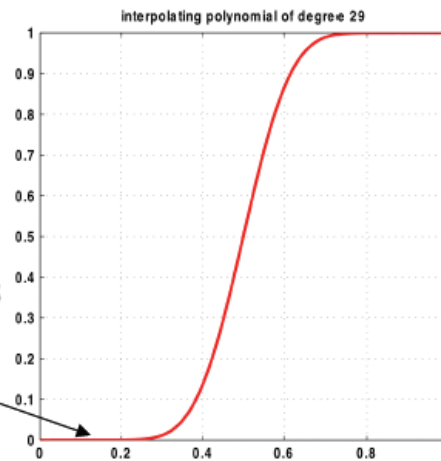
9th degree



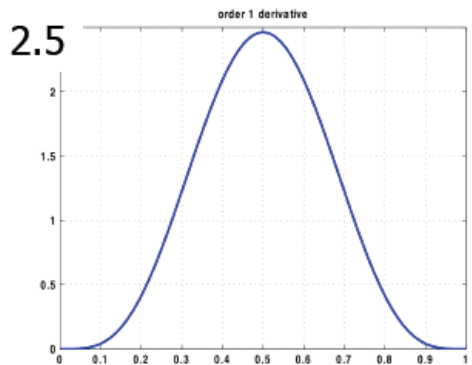
4 derivatives are zero

14 derivatives are zero!

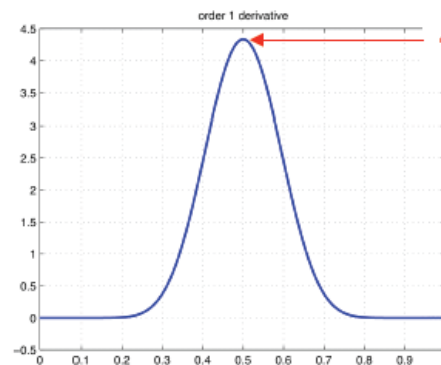
29th degree



no overshoot nor wandering



normalized velocity



4.5!!

velocity peaking at midpoint

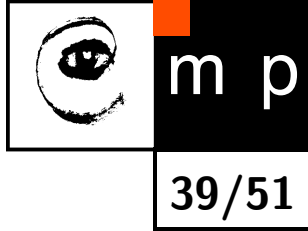
Several polynomials instead of one high-degree polynomial

- ◆ Given N points, in order to avoid the problem of high ‘oscillations’ and troubles with the numerical precision avoid a single high-degree $N - 1$ polynomial.
- ◆ Instead, use $N - 1$ polynomials with lower degree p , $p < N - 1$. Each polynomial interpolates a segment of the trajectory.
- ◆ Often $p = 3$ is chosen so that continuity of the velocity and acceleration is achieved.

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

- ◆ There are 4 coefficients for each polynomial, and thus it is necessary to compute $3(N - 1)$ coefficients.

Spline

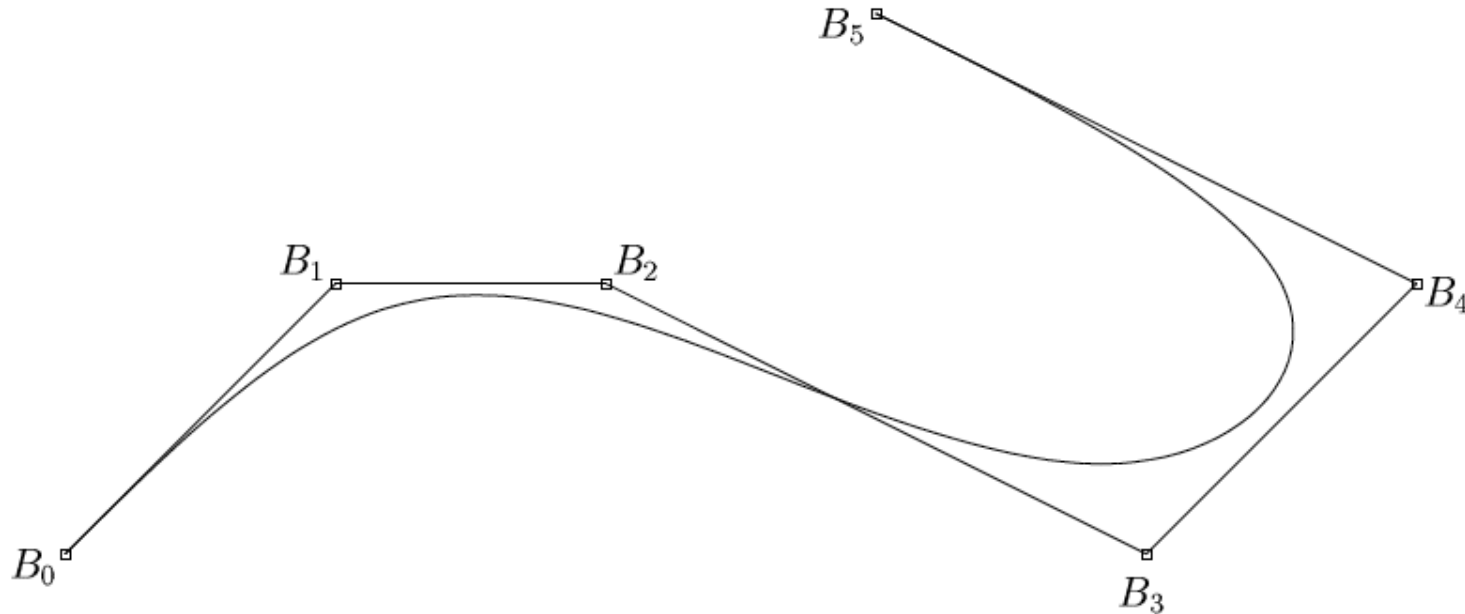


- ◆ The word “spline” refers to thin strip of wood or metal. At one time, curves were designed for ships or planes by mounting actual strips so that they went through a desired points but were free to move otherwise.
- ◆ Definition:
A cubic spline curve is a piecewise cubic curve with continuous second derivation.
- ◆ Definition (a special case):
A cubic spline curve is relaxed if its second derivative is zero at each endpoint.
- ◆ An easy way of making a controlled-design curve with many control points is to use B-spline curves.

B-spline, construction by hand (1)

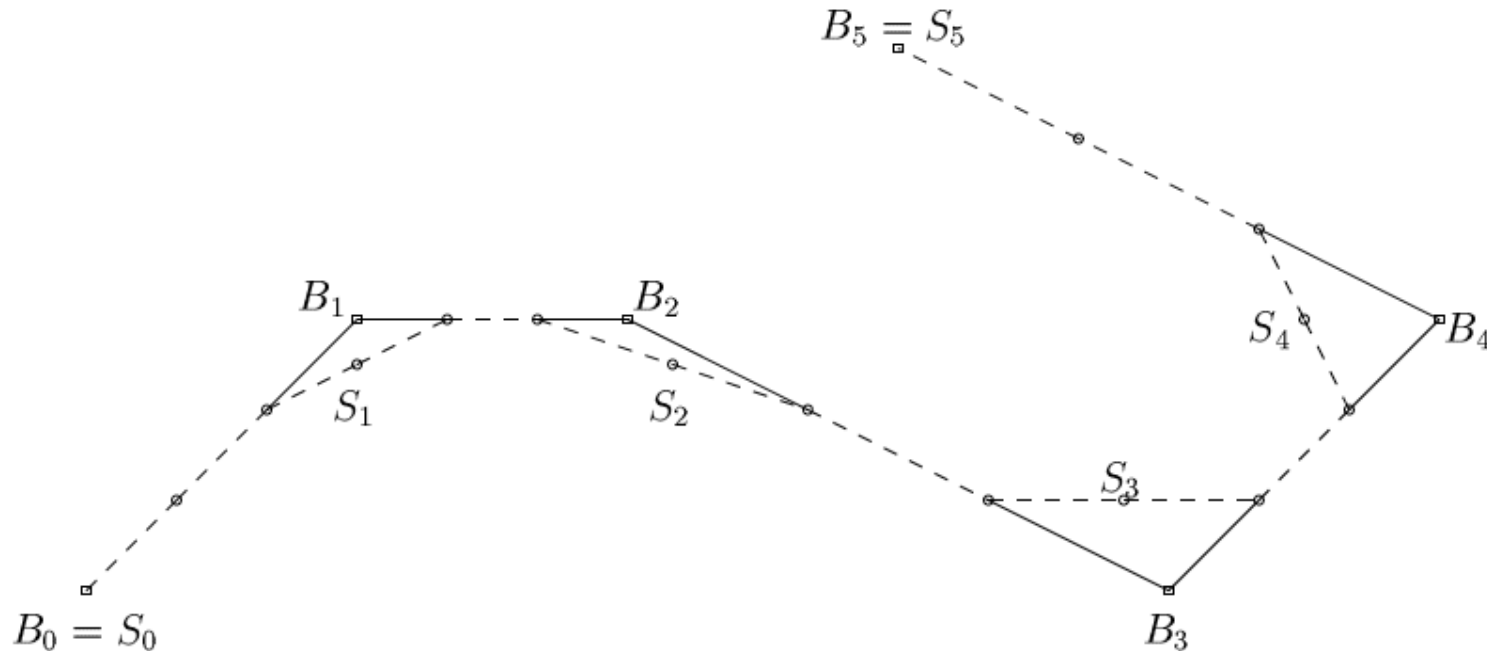
We start the description with relaxed uniform B-spline curves for simplicity.

Assume starting by specifying of a control polygon of points B_0, B_1, \dots, B_n .



B-spline, construction by hand (2)

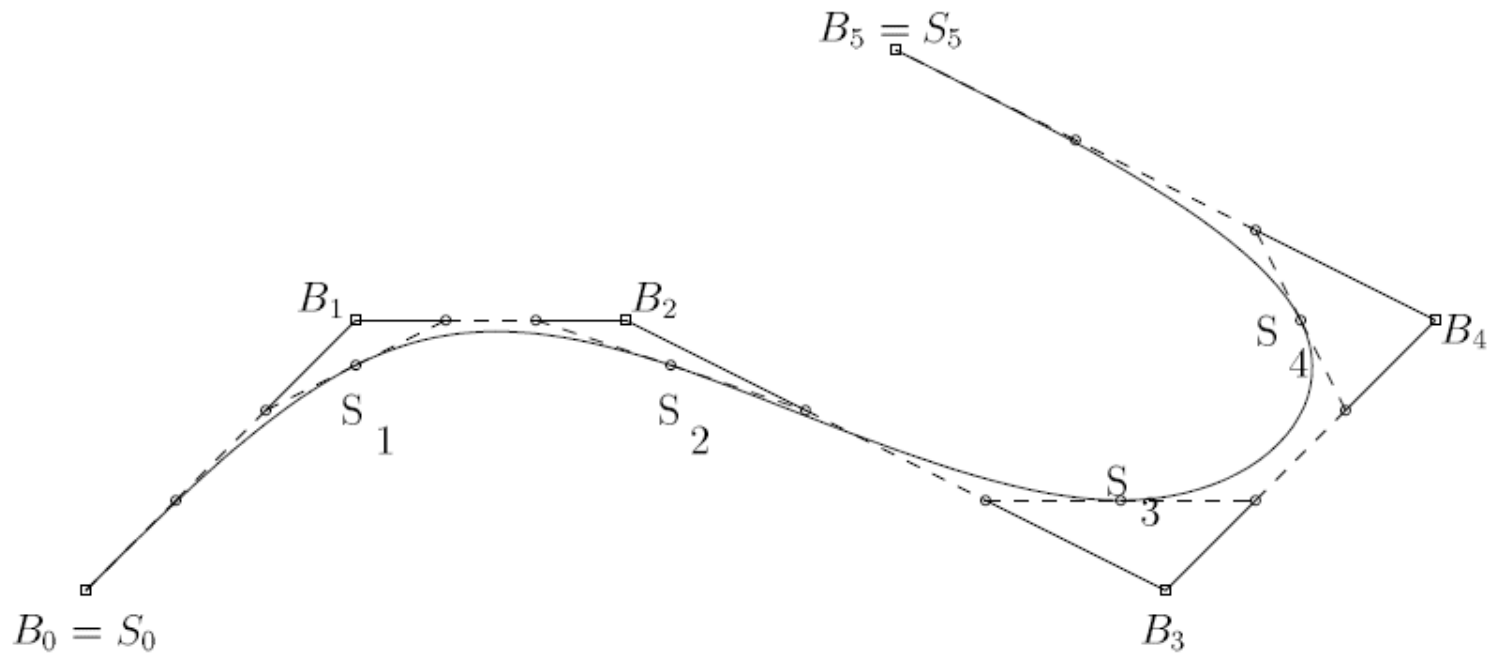
The B-spline construction method if done by hand: Divide each leg of a control polygon in thirds by marking two “division” points. At each B_i except the first and last, draw a line segments between the two nearest “division” points. Call the midpoint S_i . Then you make an A-frame with B_i at the apex. Let $S_0 = B_0$ and $S_n = B_n$ for completeness.





B-spline, construction by hand (3)

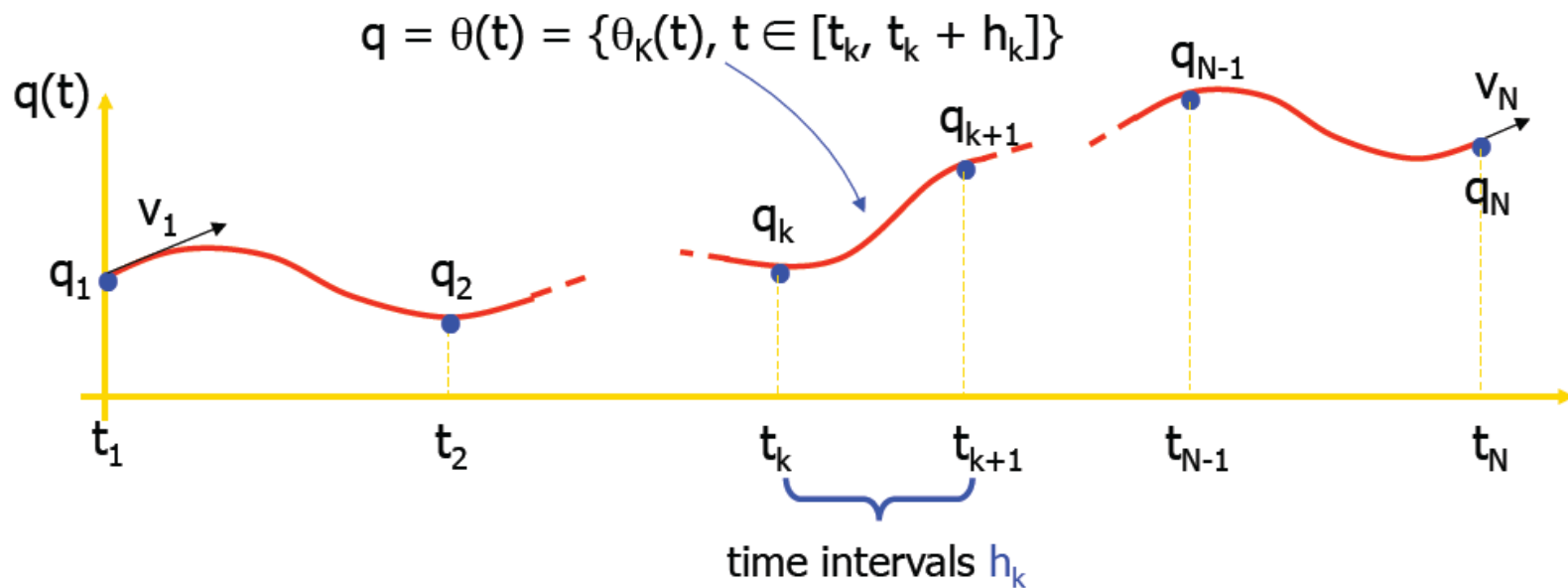
Sketch a cubic (Bézier) curve from each point S_i to the next using as Bézier control points the four points S_i , two “division” points, and S_{i+1} .



Interpolation using splines

- ◆ **Task formulation:** Interpolate N knots along the desired curve, keep C^2 continuity, i.e. continuity up to the second derivative.
- ◆ **Solution:** Approximate by splines.
Splines are $N - 1$ cubic polynomials concatenated to pass through knots (also control points) and being continuous in velocity and acceleration in the $N - 2$ internal knots.
- ◆ $4(N - 1)$ coefficients.
- ◆ $4(N - 1) - 2$ conditions, more specifically
 - $2(N - 1)$ passages, pro each cubic in the two knots at its ends.
 - $N - 2$ continuities for velocities at internal knots.
 - $N - 2$ continuities for accelerations at internal knots.
- ◆ Two free parameters are still left over. These parameters can be used to, e.g., assign the initial and final velocities v_1, v_N .
- ◆ We present curves in terms of time t . It is similar for space λ .

Building a cubic spline



- ◆ The polynomial $\Theta_k(\tau) = a_{k0} + a_{k1}\tau + a_{k2}\tau^2 + a_{k3}\tau^3$; $\tau \in [0, h_k]$, $\tau = t - t_k$, $k = 1, \dots, N - 1$.
- ◆ Continuity conditions for velocity and acceleration:
 $\dot{\Theta}_k(h_k) = \dot{\Theta}_{k+1}(0)$; $\ddot{\Theta}_k(h_k) = \ddot{\Theta}_{k+1}(0)$; $k = 1, \dots, N - 2$.

An efficient algorithm

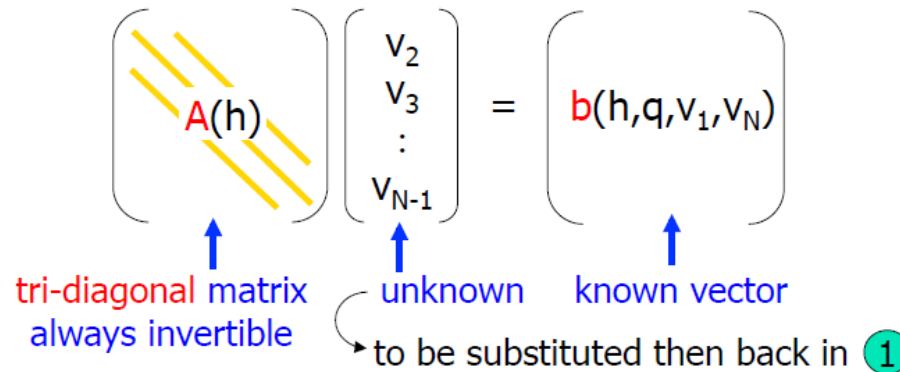
1. If all velocities v_k at internal knots were known then each cubic in the spline would be uniquely determined by

$$\begin{aligned} \Theta_k(0) &= q_k = a_{k0} \\ \dot{\Theta}(0) &= v_k = a_{k1} \end{aligned} \begin{bmatrix} h_{k2} & h_{k3} \\ 2h_k & 3h_{k2} \end{bmatrix} \begin{bmatrix} a_{k2} \\ a_{k3} \end{bmatrix} = \begin{bmatrix} q_{k+1} - q_k - v_k h_k \\ v_{k+1} - v_k \end{bmatrix}$$

2. Impose $N - 2$ acceleration continuity constraints

$$\ddot{\Theta}_k(h_k) = 2a_{k2} + 6a_{k3}h_k = \ddot{\Theta}_{k+1}(0) = 2a_{k+1,2}$$

3. Expressing the coefficients a_{k2} , a_{k3} , $a_{k+1,2}$ in terms of **still unknown** knot velocities, see step 1, yields a linear system of equations, which are always solvable



Structure of $b(h, q, v_1, v_N)$

$$\left(\begin{array}{c}
 \frac{3}{h_1 h_2} [h_1^2(q_3 - q_2) + h_2^2(q_2 - q_1)] - h_2 v_1 \\
 \frac{3}{h_2 h_3} [h_2^2(q_4 - q_3) + h_3^2(q_3 - q_2)] \\
 \vdots \\
 \frac{3}{h_{N-3} h_{N-2}} [h_{N-3}^2(q_{N-1} - q_{N-2}) + h_{N-2}^2(q_{N-2} - q_{N-3})] \\
 \frac{3}{h_{N-2} h_{N-1}} [h_{N-2}^2(q_N - q_{N-1}) + h_{N-1}^2(q_{N-1} - q_{N-2})] - h_{N-2} v_N
 \end{array} \right)$$

Properties of splines

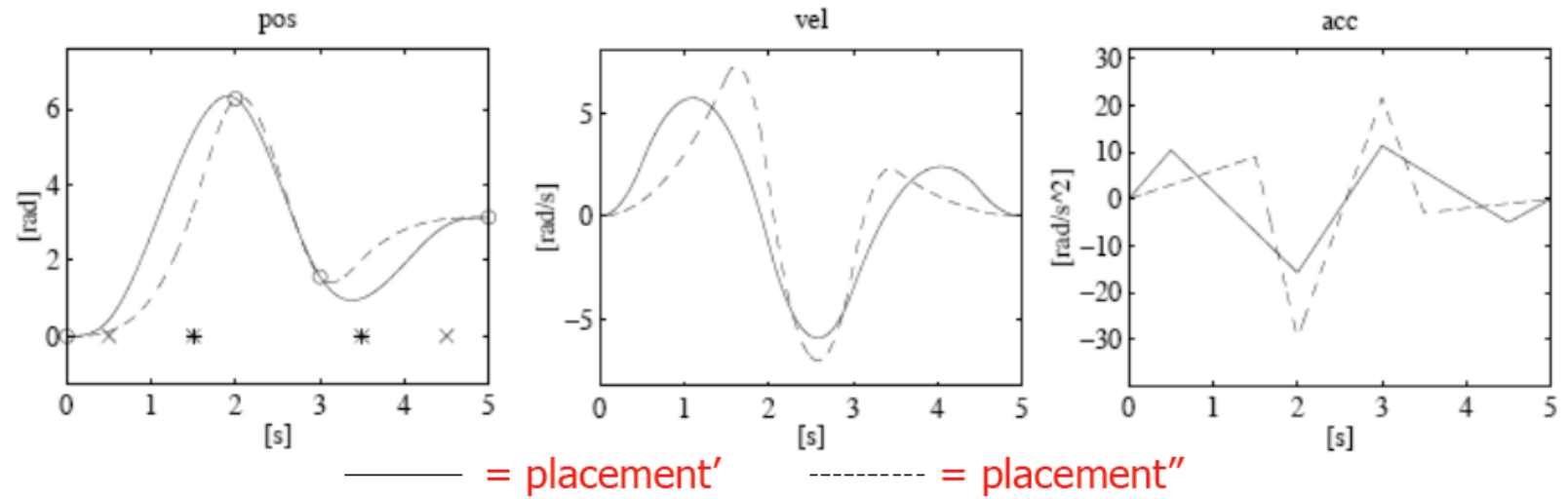
- ◆ The spline is the solution with the minimum curvature among all interpolating functions having continuous derivatives up to the second one.
- ◆ A spline is uniquely determined from the data $q_1, \dots, q_n, h_1, \dots, h_{N-1}, v_1, \dots, v_n$.
- ◆ The total transfer time is $T = \sum_{k=1}^K h_k = t_N - t_1$.
- ◆ The time intervals h_k can be chosen to minimize T (linear objective function) under (nonlinear) bounds on velocity and acceleration in $[0, T]$.
- ◆ For cyclic tasks ($q_1 = q_N$), it is preferable to impose simply the continuity of velocity and acceleration at $t_1 = t_N$ as the “squaring” conditions
 - in fact, even choosing $v_1 = v_N$ does not guarantee the acceleration continuities;
 - in this way, the first = last knot will be handled as all other internal knots
- ◆ When initial and final accelerations are also assigned, the spline construction can be suitably modified.

A modification handling assigned initial and final accelerations

- ◆ Two more parameters are needed in order to impose also the initial acceleration α_1 and final acceleration α_N .
- ◆ Two “fictious knots” are inserted in the first and last original intervals, increasing the number of cubic polynomials from $N - 1$ to $N + 1$.
- ◆ In these two knots only continuity conditions on position, velocity and acceleration are imposed \Rightarrow two free parameters are left over (one in the first cubic and one in the last cubic), which are used to satisfy the boundary constraints on acceleration.
- ◆ Depending on the (time) placement of the two additional knots, the resulting spline changes.

A numerical example

- $N = 4$ knots (3 cubic polynomials)
 - joint values $q_1 = 0, q_2 = 2\pi, q_3 = \pi/2, q_4 = \pi$
 - at $t_1 = 0, t_2 = 2, t_3 = 3, t_4 = 5$ (thus, $h_1 = 2, h_2 = 1, h_3 = 2$)
 - boundary velocities $v_1 = v_4 = 0$
- 2 added knots to **impose accelerations** at both ends (5 cubic polynomials)
 - boundary accelerations $\alpha_1 = \alpha_4 = 0$
 - **two** placements: at $t_1' = 0.5$ and $t_4' = 4.5$ (\times), or $t_1'' = 1.5$ and $t_4'' = 3.5$ ($*$)



References

- ◆ B. Siciliano et al. Robotics (Modeling, planning and control), Springer, Berlin, 2009, chapter 4: Trajectory planning, pages 161-189.