



## ***Robotics 1***

# **Trajectory planning**

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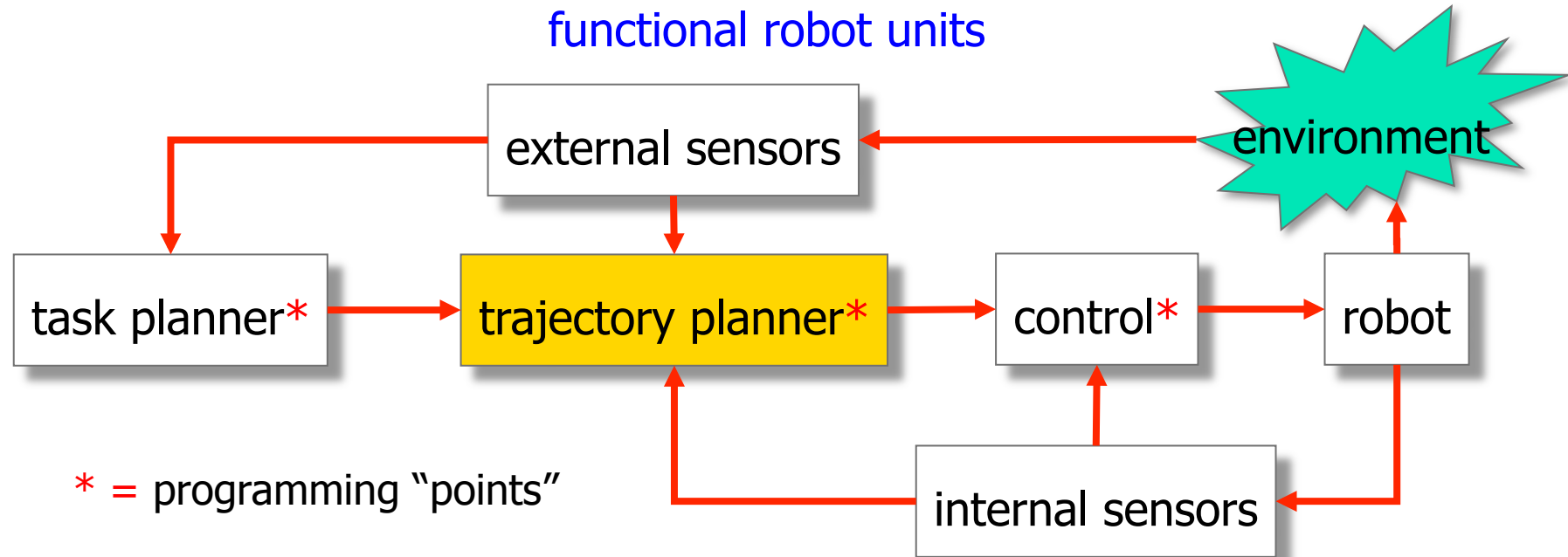
DIPARTIMENTO DI INGEGNERIA INFORMATICA  
AUTOMATICA E GESTIONALE ANTONIO RUBERTI



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# Trajectory planner interfaces



robot **action** described as a sequence of **poses** or **configurations** (with possible exchange of **contact** forces)



TRAJECTORY PLANNER



**reference profile/values** (continuous or discrete) for the **robot controller**



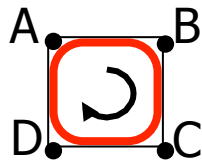
# Trajectory definition

a standard procedure for industrial robots

1. define Cartesian pose points (position+orientation) using the teach-box
2. program an (average) velocity between these points, as a 0-100% of a maximum system value (different for Cartesian- and joint-space motion)
3. linear interpolation in the joint space between points sampled from the built trajectory

## examples of additional features

a) over-fly



b) sensor-driven STOP

c) circular path  
through 3 points

## main drawbacks

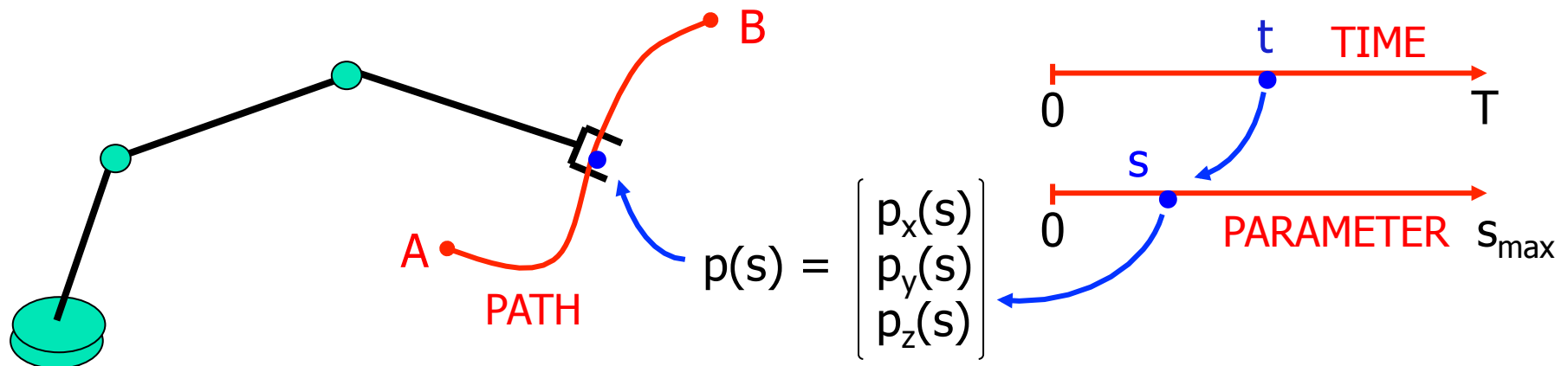
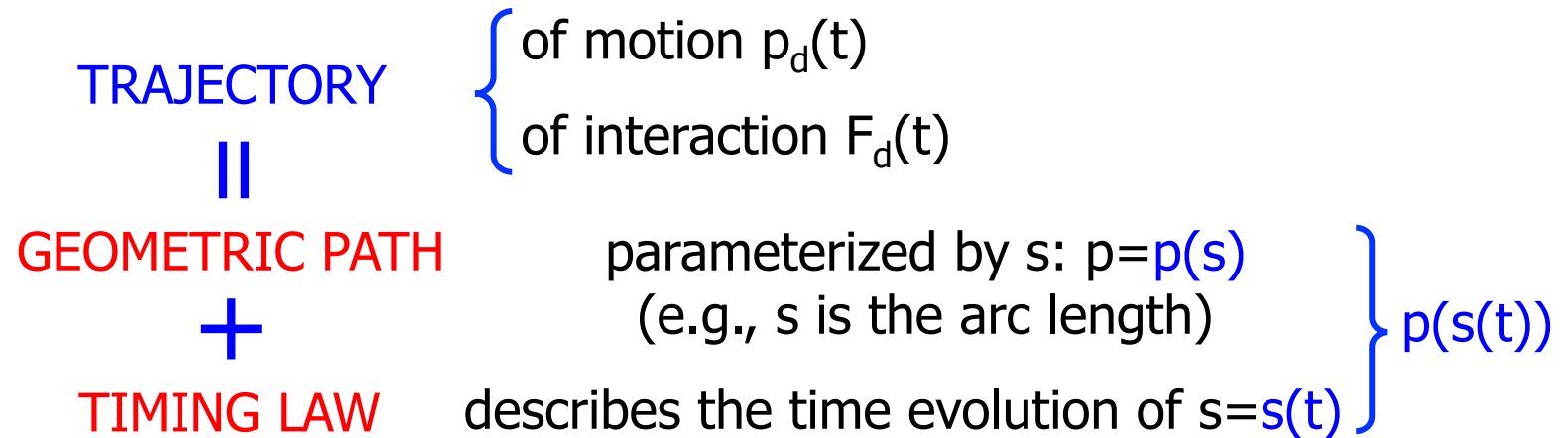
- semi-manual programming (as in "first generation" robot languages)
- limited visualization of motion



a mathematical formalization of trajectories is useful/needed



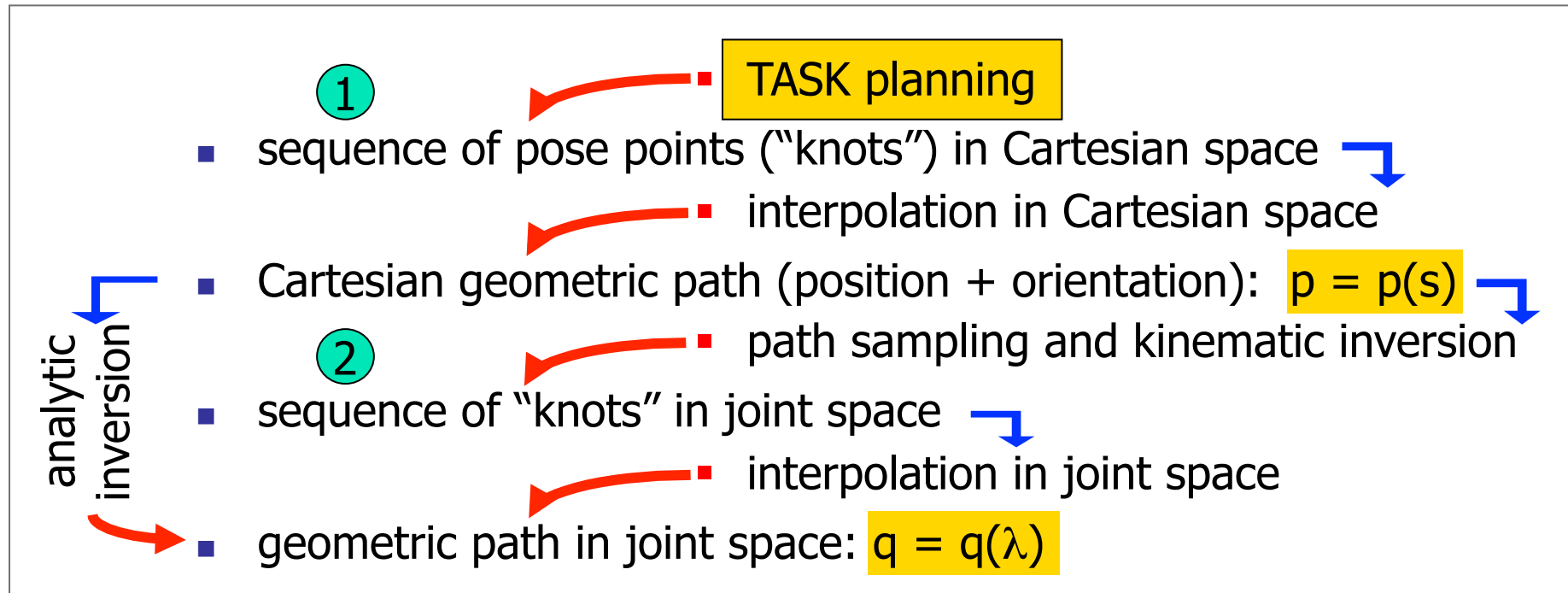
# From task to trajectory



example: TASK planner provides A, B  
TRAJECTORY planner generates  $p(t)$



# Trajectory planning operative sequence

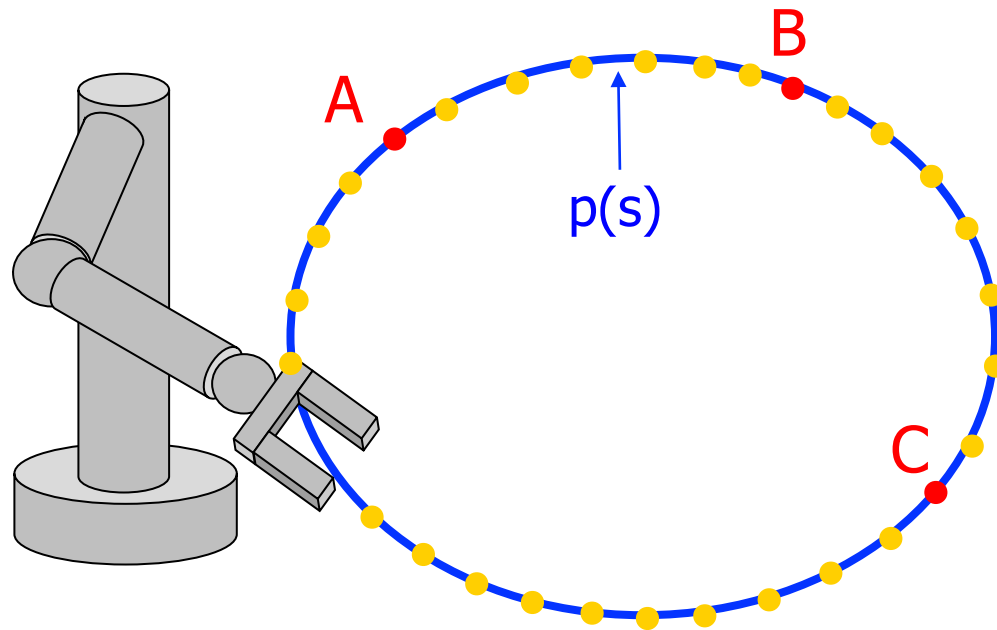


## additional issues to be considered in the planning process

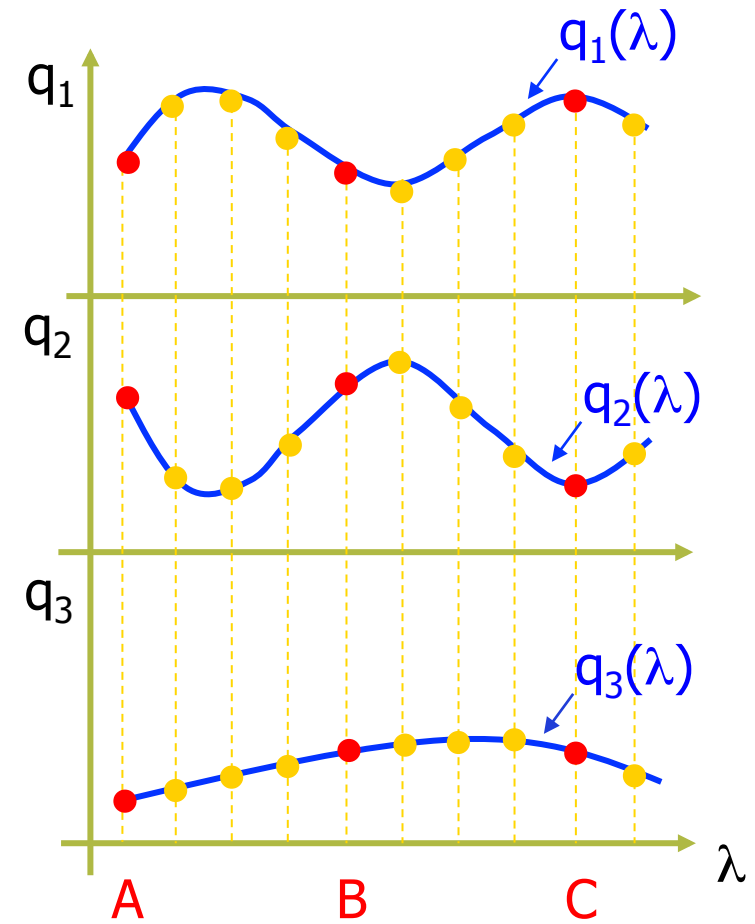
- obstacle avoidance
- on-line/off-line computational load
- sequence **2** is more "dense" than **1**



# Example



Cartesian space



joint space



# Cartesian vs. joint trajectory planning

- planning in **Cartesian space**
  - allows a more direct visualization of the generated path
  - obstacle avoidance, lack of “wandering”
- planning in **joint space**
  - does not need on-line kinematic inversion
- issues in kinematic inversion
  - $\dot{q}$  e  $\ddot{q}$  (or higher-order derivatives) may also be needed
    - Cartesian task specifications involve the geometric path, but also bounds on the associated timing law
  - for redundant robots, choice among  $\infty^{n-m}$  inverse solutions, based on optimality criteria or additional auxiliary tasks
  - off-line planning in advance is not always feasible
    - e.g., when interaction with the environment occurs or sensor-based motion is needed



# Path and timing law

- after choosing a **path**, the trajectory definition is completed by the choice of a timing law

$$p = p(s) \quad \Rightarrow \quad s = s(t) \quad (\text{Cartesian space})$$

$$q = q(\lambda) \quad \Rightarrow \quad \lambda = \lambda(t) \quad (\text{joint space})$$

- if  $s(t) = t$ , path parameterization is the **natural** one given by time
- the **timing law**
  - is chosen based on **task specifications** (stop in a point, move at constant velocity, and so on)
  - may consider **optimality criteria** (min transfer time, min energy,...)
  - **constraints** are imposed by actuator capabilities (max torque, max velocity,...) and/or by the task (e.g., max acceleration on payload)

**note:** on parameterized paths, a **space-time decomposition** takes place

e.g., in Cartesian space

$$\dot{p}(t) = \frac{dp}{ds} \dot{s} \quad \ddot{p}(t) = \frac{dp}{ds} \ddot{s} + \frac{d^2p}{ds^2} \dot{s}^2$$





# Trajectory classification

- space of definition
  - Cartesian, joint
- task type
  - point-to-point (PTP), multiple points (knots), continuous, concatenated
- path geometry
  - rectilinear, polynomial, exponential, cycloid, ...
- timing law
  - bang-bang in acceleration, trapezoidal in 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  $t=T$ ) = **single timing law**  
or
  - motions are timed **independently** (according to the requested displacement and robot capabilities) – mostly only in **joint space**



# Relevant characteristics

- computational **efficiency** and memory space
  - e.g., store only the coefficients of a polynomial function
- **predictability** (vs. “wandering” out of the knots) and **accuracy** (vs. “overshoot” on final position)
- **flexibility** (allowing concatenation, over-fly, ...)
- **continuity** (in space and in time)  
(at least **C<sup>1</sup>**, but also up to jerk =  $\frac{da}{dt}$ )



# Trajectory planning in joint space

- $q = q(t)$  or  $q = q(\lambda)$ ,  $\lambda = \lambda(t)$
- it is sufficient to work **component-wise** ( $q_i$  in vector  $q$ )
- an **implicit** definition of the trajectory, by solving a problem with specified **boundary conditions** in a given **class of functions**
- typical classes: **polynomials** (cubic, quintic,...), (co)sinusoids, clothoids, ...
- **imposed conditions**
  - passage through points = interpolation
  - initial, final, intermediate velocity (or **geometric tangent for paths**)
  - initial, final acceleration (or **geometric curvature**)
  - continuity up to the  $k$ -th order time (or **space**) derivative: class  $\mathbf{C}^k$

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



# Cubic polynomial

$$\boxed{q(0) = q_{in}} \quad \boxed{q(T) = q_{fin}} \quad \boxed{\dot{q}(0) = v_{in}} \quad \boxed{\dot{q}(T) = v_{fin}} \quad \leftarrow 4 \text{ conditions}$$

$$q(\tau) = q_{in} + \Delta q [a\tau^3 + b\tau^2 + c\tau + d]$$

$$\Delta q = q_{fin} - q_{in}$$
$$\tau = t/T, \tau \in [0, 1]$$

4 coefficients  $\rightarrow$  "doubly normalized" polynomial  $q_N(\tau)$

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

$$q_N(1) = 1 \Leftrightarrow a + b + c = 1$$

$$q_N'(0) = dq_N/d\tau|_{\tau=0} = c = v_{in}T/\Delta q$$

$$q_N'(1) = dq_N/d\tau|_{\tau=1} = 3a + 2b + c = v_{fin}T/\Delta q$$

special case:  $v_{in} = v_{fin} = 0$  (rest-to-rest)

$$q_N'(0) = 0 \Leftrightarrow c = 0$$

$$q_N(1) = 1 \Leftrightarrow a + b = 1$$

$$q_N'(1) = 0 \Leftrightarrow 3a + 2b = 0$$

$$\left. \begin{array}{l} a + b = 1 \\ 3a + 2b = 0 \end{array} \right\} \Leftrightarrow \begin{array}{l} a = -2 \\ b = 3 \end{array}$$



# Quintic polynomial

$$q(\tau) = a\tau^5 + b\tau^4 + c\tau^3 + d\tau^2 + e\tau + f$$

6 coefficients

$$\tau = t/T, \tau \in [0, 1]$$

allows to satisfy 6 conditions, for example (in normalized time  $\tau$ )

$$q(0) = q_0$$

$$q(1) = q_1$$

$$q'(0) = v_0T$$

$$q'(1) = v_1T$$

$$q''(0) = a_0T^2$$

$$q''(1) = a_1T^2$$

$$q(\tau) = (1 - \tau)^3[q_0 + (3q_0 + v_0T)\tau + (a_0T^2 + 6v_0T + 12q_0)\tau^2/2] \\ + \tau^3[q_1 + (3q_1 - v_1T)(1 - \tau) + (a_1T^2 - 6v_1T + 12q_1)(1 - \tau)^2/2]$$

special case:  $v_0 = v_1 = a_0 = a_1 = 0$

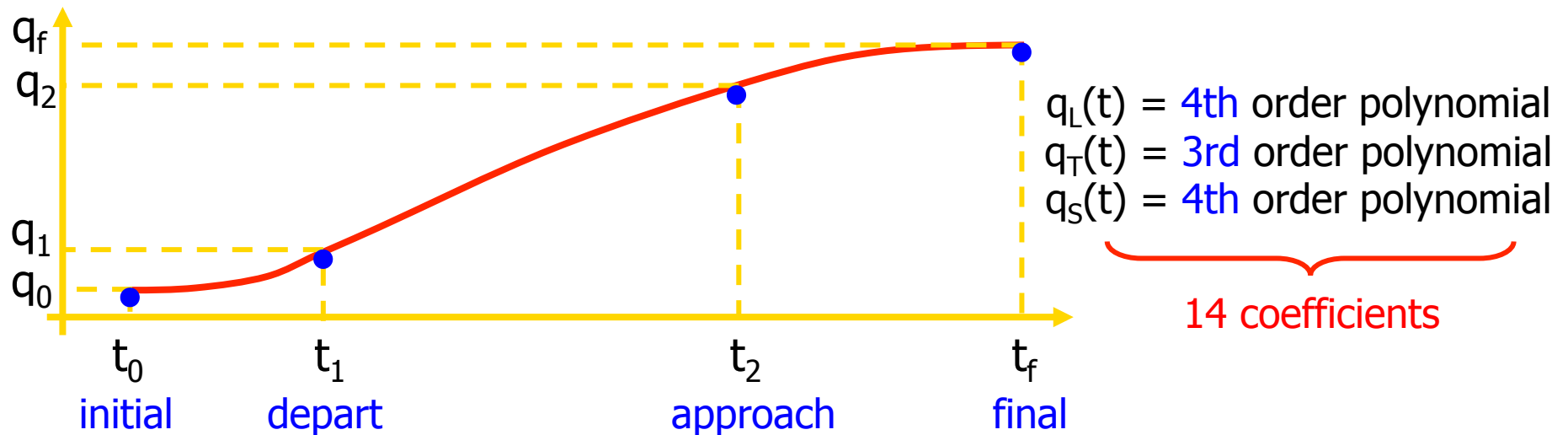
$$q(\tau) = q_0 + \Delta q [6\tau^5 - 15\tau^4 + 10\tau^3]$$

$$\Delta q = q_1 - q_0$$



# 4-3-4 polynomials

three phases (Lift off, Travel, Set down) in pick-and-place operations



boundary conditions

$$\begin{array}{l}
 q(t_0) = q_0 \quad q(t_1^-) = q(t_1^+) = q_1 \quad q(t_2^-) = q(t_2^+) = q_2 \quad q(t_f) = q_f \quad \left. \vphantom{q(t_0)} \right\} 6 \text{ passages} \\
 \dot{q}(t_0) = \dot{q}(t_f) = 0 \quad \ddot{q}(t_0) = \ddot{q}(t_f) = 0 \quad \left. \vphantom{\dot{q}(t_0)} \right\} 4 \text{ initial/final} \\
 \dot{q}(t_i^-) = \dot{q}(t_i^+) \quad \ddot{q}(t_i^-) = \ddot{q}(t_i^+) \quad i = 1,2 \quad \left. \vphantom{\dot{q}(t_i^-)} \right\} \text{velocity/acceleration} \\
 \qquad \qquad \qquad \qquad \qquad \qquad \qquad \qquad \qquad \qquad \qquad \qquad \qquad \qquad \qquad \qquad \qquad \qquad \left. \vphantom{\dot{q}(t_i^-)} \right\} 4 \text{ continuity}
 \end{array}$$



# Higher-order polynomials

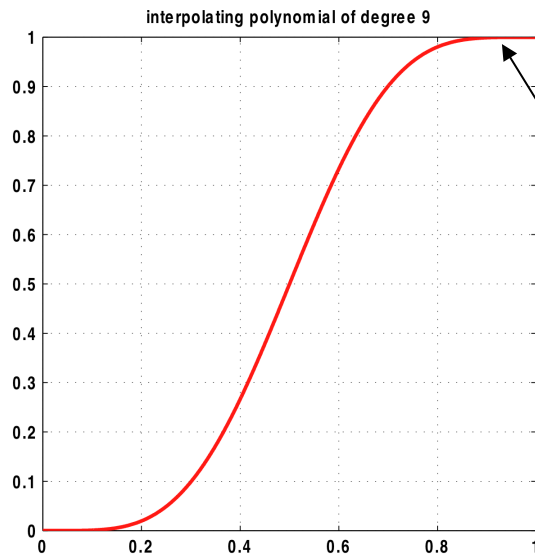
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- a suitable solution class for satisfying **symmetric** boundary conditions (in a PTP motion) that **impose zero** values on higher-order derivatives
  - the interpolating polynomial is always of **odd** degree
  - the coefficients of such (**doubly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



# Numerical examples

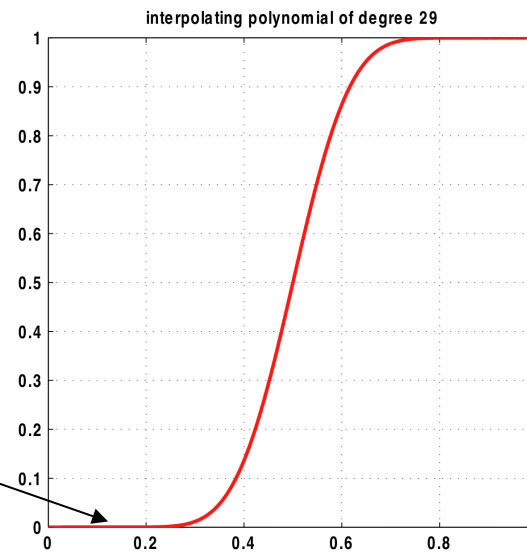
9th degree



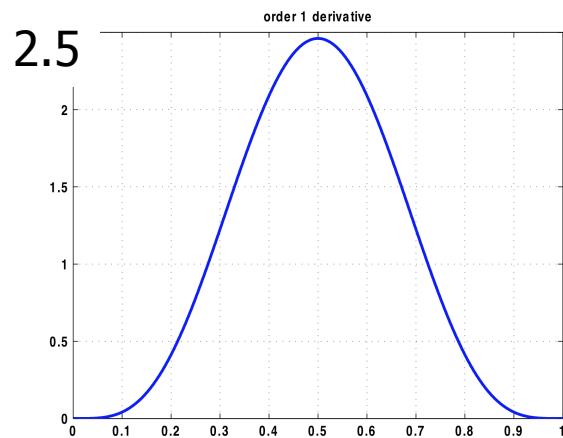
4 derivatives are zero

14 derivatives are zero!

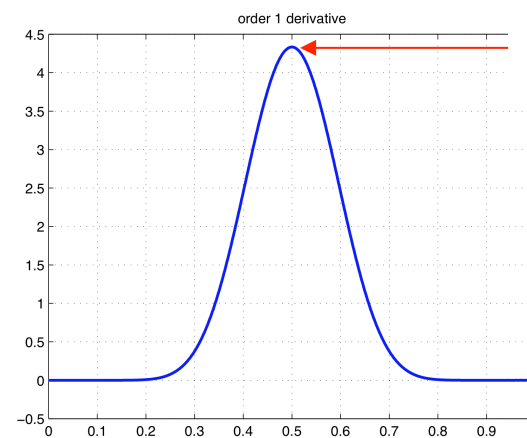
29th degree



no overshoot nor wandering



normalized velocity



velocity peaking at midpoint



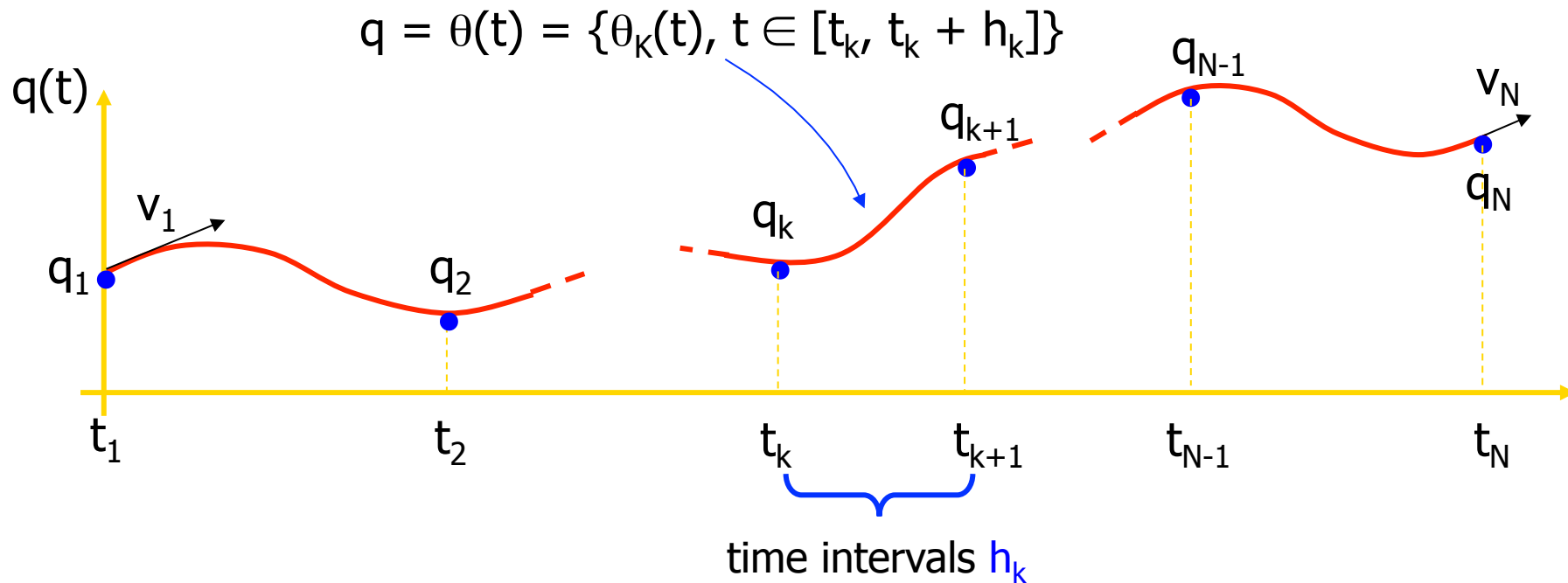


# Interpolation using splines

- **problem**
  - interpolate  $N$  knots, with continuity up to the second derivative
- **solution**
  - **spline**:  $N-1$  cubic polynomials, concatenated so as to pass through  $N$  knots and being continuous in velocity and acceleration in the  $N-2$  internal knots
- **$4(N-1)$  coefficients**
- **$4(N-1)-2$  conditions**, or
  - $2(N-1)$  of passage (for each cubic, in the two knots at its ends)
  - $N-2$  of continuity for velocity (at the internal knots)
  - $N-2$  of continuity for acceleration (at the internal knots)
- **2 free parameters** are still left over
  - can be used, e.g., to assign initial and final velocities,  $v_1$  and  $v_N$
- presented next in terms of **time  $t$** , but similar in terms of **space  $\lambda$**



# Building a cubic spline



$$\theta_k(\tau) = a_{k0} + a_{k1} \tau + a_{k2} \tau^2 + a_{k3} \tau^3 \quad \tau \in [0, h_k], \tau = t - t_k \quad (k = 1, \dots, N-1)$$

continuity conditions  
for velocity and acceleration



$$\begin{aligned} \dot{\theta}_k(h_k) &= \dot{\theta}_{k+1}(0) \\ \ddot{\theta}_k(h_k) &= \ddot{\theta}_{k+1}(0) \end{aligned} \quad k = 1, \dots, N-2$$



# 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

$$\begin{aligned} \theta_k(0) &= q_k = a_{k0} \\ \dot{\theta}_k(0) &= v_k = a_{k1} \end{aligned} \quad \begin{pmatrix} h_k^2 & h_k^3 \\ 2h_k & 3h_k^2 \end{pmatrix} \begin{pmatrix} a_{k2} \\ a_{k3} \end{pmatrix} = \begin{pmatrix} q_{k+1} - q_k - v_k h_k \\ v_{k+1} - v_k \end{pmatrix} \quad \textcircled{1}$$

2. impose the **continuity for accelerations** (N-2 conditions)

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

3. expressing the coefficients  $a_{k2}$ ,  $a_{k3}$ ,  $a_{k+1,2}$  in terms of the **still unknown** knot velocities (see step 1.) yields a linear system of equations that is always (easily) solvable

$$\begin{pmatrix} \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \end{pmatrix} \begin{pmatrix} v_2 \\ v_3 \\ \vdots \\ v_{N-1} \end{pmatrix} = \begin{pmatrix} \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \end{pmatrix} \begin{matrix} \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \end{matrix}$$

$A(h)$                        $b(h, q, v_1, v_N)$

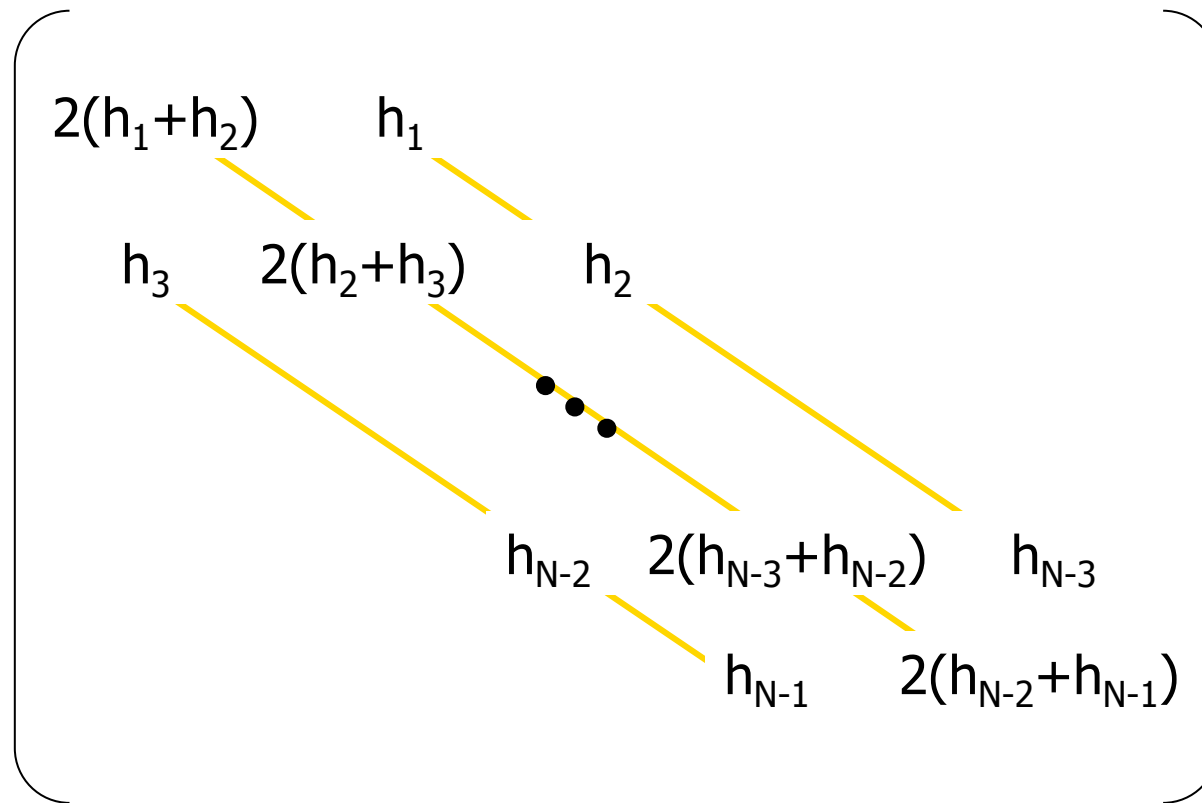
↑                                      ↑                                      ↑

tri-diagonal matrix                      unknown                      known vector

always invertible                      to be substituted then back in                       $\textcircled{1}$



# Structure of $A(h)$



diagonally dominant matrix (for  $h_k > 0$ )  
[the **same** matrix for all joints]



## 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 **minimum curvature** among all interpolating functions having continuous second derivative
- a spline is **uniquely** determined from the set of data  $q_1, \dots, q_N$ ,  $h_1, \dots, h_{N-1}$ ,  $v_1$ ,  $v_N$
- the total transfer time is  $T = \sum h_k = t_N - t_1$
- the time intervals  $h_k$  can be chosen so as 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 simply impose continuity of velocity and acceleration at  $t_1 = t_N$  as the “squaring” conditions
  - in fact, even choosing  $v_1 = v_N$  doesn't guarantee acceleration continuity
  - 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

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- two more parameters are needed in order to impose also the initial acceleration  $\alpha_1$  and final acceleration  $\alpha_N$
- two “fictitious 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
  - ⇒ **two** free parameters are left over (one in the first cubic and the other in the last cubic), which are used to satisfy the boundary conditions 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  $t_4'' = 3.5$  ( $*$ )

