Trajectory Planning for Robot Manipulators

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Kinematics: geometrical relationships in terms of position/velocity between the joint- and work-space.

Dynamics: relationships between the torques applied to the joints and the consequent movements of the links.

Control: computation of the control actions (joint torques) necessary to execute a desired motion.

Trajectory planning: planning of the desired movements of the manipulator.

Usually, the user is requested to define some points and general features of the trajectory (e.g. initial/final points, duration, maximum velocity, etc.), and the real computation of the trajectory is demanded to the control system.
Trajectory planning: **IMPORTANT** aspect in robotics, **VERY IMPORTANT** for the dimensioning, control, and use of electric motors in automatic machines (e.g. packaging).

Origin of the interest for the control area was the substitution of mechanical cams with *electric cams* in the design of automatic machines.
Some suggested references:


- G. Canini, C. Fantuzzi, *Controllo del moto per macchine automatiche*, Pitagora Ed., Bo, 2003;


The planning modalities for trajectories may be quite different:

- point-to-point
- with pre-defined path

Or:

- in the joint space;
- in the work space, either defining some points of interest (initial and final points, *via points*) or the whole geometric path $x = x(t)$.

For planning a desired trajectory, it is necessary to specify two aspects:

- *geometric path*
- *motion law*

with constraints on the continuity (smoothness) of the trajectory and on its time-derivatives up to a given degree.
The geometric path can be defined in the work-space or in the joint-space. Usually, it is expressed in a parametric form as

\[ p = p(s) \quad \text{work-space} \]
\[ q = q(\sigma) \quad \text{joint-space} \]

The parameter \( s \) (or \( \sigma \)) is defined as a function of time, and in this manner the motion law \( s = s(t) \) (\( \sigma = \sigma(t) \)) is obtained.

\[ p_x = p_x(s) \]
\[ p_y = p_y(s) \]
\[ p_z = p_z(s) \]
Examples of geometric paths: (in the work space) linear, circular or parabolic segments or, more in general, tracts of analytical functions.

In the joint space, geometric paths are obtained by assigning initial/final (and, in case, also intermediate) values for the joint variables, along with the desired motion law.

Concerning the motion law, it is necessary to specify continuous functions up to a given order of derivation (often at least first and second order, i.e. velocity and acceleration).

Usually, polynomial functions with a proper degree $n$ are employed:

\[ s(t) = a_0 + a_1 t + a_2 t^2 + \ldots + a_n t^n \]

In this manner, a “smooth” interpolation of the points defining the geometric path is achieved.
Input data to an algorithm for trajectory planning are:

- data defining on the path (points),
- geometrical constraints on the path (e.g. obstacles),
- constraints on the mechanical dynamics
- constraints due to the actuation system

Output data is:

- the trajectory in the joint- or work-space, given as a sequence (in time) of the acceleration, velocity and position values:

\[ a(kT), \quad v(kT), \quad p(kT) \quad k = 0, \ldots, N \]

being \( T \) a proper time interval defining the instants in which the trajectory is computed (and in case converted in the joint space) and sent to each actuator.
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 generated by singular configurations, actuation redundancy, velocity/acceleration constraints.
Joint-space trajectories

Trajectories are specified by defining some characteristic points:
- directly assigned by some specifications
- assigned by defining desired configurations $x$ in the work-space, which are then converted in the joint space using the inverse kinematic model.

The algorithm that computes a function $q(t)$ \textit{interpolating} the given points is characterized by the following features:
- trajectories must be computationally efficient
- the position and velocity profiles (at least) must be continuous functions of time
- undesired effects (such as non regular curvatures) must be minimized or completely avoided.

In the following discussion, a single joint is considered.
If more joints are present, a coordinated motion must be planned, e.g. considering for each of them the same initial and final time instant, or evaluating the most stressed joint (with the largest displacement) and then scaling suitably the motion of the remaining ones.
Polynomial trajectories

In the most simple cases, (a segment of) a trajectory is specified by assigning initial and final conditions on: time (duration), position, velocity, acceleration, . . . . Then, the problem is to determine a function

\[ q = q(t) \quad \text{or} \quad q = q(\sigma), \quad \sigma = \sigma(t) \]

so that those conditions are satisfied.

This is a boundary condition problem, that can be easily solved by considering polynomial functions such as:

\[ q(t) = a_0 + a_1 t + a_2 t^2 + \ldots + a_n t^n \]

The degree \( n \) (3, 5, ...) of the polynomial depends on the number of boundary conditions that must be verified and on the desired “smoothness” of the trajectory.
Polynomial trajectories

In general, besides the initial and final values, other constraints could be specified on the values of some time-derivatives (velocity, acceleration, jerk, ...) in generic instants $t_j \in [t_i, \ t_f]$. In other terms, one could be interested in defining a polynomial function $q(t)$ whose $k$-th derivative has a specified value $q^k(t_j)$ at a given instant $t_j$.

Mathematically, these conditions may be expressed as:

$$k! a_k + (k + 1)! a_{k+1} t_j + \ldots + \frac{n!}{(n - k)!} a_n t_j^{n-k} = q^k(t_j)$$

or, in matrix form:

$$M \ a = b$$

where:
- $M$ is a known $(n + 1) \times (n + 1)$ matrix,
- $b$ is the vector with the $n + 1$ constraints on the trajectory (known data),
- $a = [a_0, a_1, \ldots, a_n]^T$ contains the unknown parameters to be computed.

The solution is:

$$a = M^{-1} b$$
Third-order polynomial trajectories

Given an initial and a final instant $t_i, t_f$, a (segment of a) trajectory may be specified by assigning initial and final conditions:

- initial position and velocity $q_i, \dot{q}_i$;
- final position and velocity $q_f, \dot{q}_f$

There are four boundary conditions, and therefore a polynomial of (at least) degree 3 must be considered

$$q(t) = a_0 + a_1 t + a_2 t^2 + a_3 t^3$$  \hspace{1cm} (1)

where the four parameters $a_0, a_1, a_2, a_3$ must be defined so that the boundary conditions are satisfied. From the boundary conditions, it follows that

$$\begin{align*}
q(t_i) &= a_0 + a_1 t_i + a_2 t_i^2 + a_3 t_i^3 &= q_i \\
\dot{q}(t_i) &= a_1 + 2a_2 t_i + 3a_3 t_i^2 &= \dot{q}_i \\
q(t_f) &= a_0 + a_1 t_f + a_2 t_f^2 + a_3 t_f^3 &= q_f \\
\dot{q}(t_f) &= a_1 + 2a_2 t_f + 3a_3 t_f^2 &= \dot{q}_f
\end{align*}$$  \hspace{1cm} (2)
In order to solve these equations, let us assume for the moment that \( t_i = 0 \). Therefore:

\[
\begin{align*}
    a_0 &= q_i \\ 
    a_1 &= \dot{q}_i \\ 
    a_2 &= -3(q_i - q_f) - (2\dot{q}_i + \dot{q}_f)t_f \\ 
    a_3 &= 2(q_i - q_f) + (\dot{q}_i + \dot{q}_f)t_f
\end{align*}
\]
Third-order polynomial trajectories

Position, velocity and acceleration profiles obtained with a cubic polynomial and boundary conditions: \( q_i = 10^\circ, q_f = 30^\circ, \dot{q}_i = \dot{q}_f = 0^\circ/s, \quad t_i = 0, t_f = 1s \):

Obviously:

- position \( \rightarrow \) cubic function
- velocity \( \rightarrow \) parabolic function
- acceleration \( \rightarrow \) linear function
Third-order polynomial trajectories

Non null initial/final velocity: $q_i = 10^\circ$, $q_f = 30^\circ$, $\dot{q}_i = -20^\circ/s$, $\dot{q}_f = -50^\circ/s$, $t_i = 0$, $t_f = 1s$. 

![Diagram of position, velocity, and acceleration graphs for third-order polynomial trajectories.](image-url)
The results obtained with the polynomial (1) and the coefficients (3)-(6) can be generalized to the case in which $t_i \neq 0$. One obtains:

$$q(t) = a_0 + a_1(t - t_i) + a_2(t - t_i)^2 + a_3(t - t_i)^3 \quad t_i \leq t \leq t_f$$

with coefficients

$$a_0 = q_i$$
$$a_1 = \dot{q}_i$$
$$a_2 = \frac{-3(q_i - q_f) - (2\dot{q}_i + \dot{q}_f)(t_f - t_i)}{(t_f - t_i)^2}$$
$$a_3 = \frac{2(q_i - q_f) + (\dot{q}_i + \dot{q}_f)(t_f - t_i)}{(t_f - t_i)^3}$$

In this manner, it is very simple to plan a trajectory passing through a sequence of intermediate points.
Third-order polynomial trajectories

The trajectory is divided in $n$ segments, each of them defined by:
- initial and final point $q_k$ e $q_{k+1}$
- initial and final instant $t_k$, $t_{k+1}$
- initial and final velocity $\dot{q}_k$, $\dot{q}_{k+1}$

$k = 0, \ldots, n - 1$.

The above relationships are then adopted for each of these segments.
Third-order polynomial trajectories

Position, velocity and acceleration profiles with:

\[
\begin{align*}
    t_0 & = 0 & t_1 & = 2 & t_2 & = 4 & t_3 & = 8 & t_4 & = 10 \\
    q_0 & = 10^\circ & q_1 & = 20^\circ & q_2 & = 0^\circ & q_3 & = 30^\circ & q_4 & = 40^\circ \\
    \dot{q}_0 & = 0^\circ/s & \dot{q}_1 & = -10^\circ/s & \dot{q}_2 & = 10^\circ/s & \dot{q}_3 & = 3^\circ/s & \dot{q}_4 & = 0^\circ/s
\end{align*}
\]
Often, a trajectory is assigned by specifying a sequence of desired points (*via-points*) without indication on the velocity in these points. In these cases, the “most suitable” values for the velocities must be automatically computed. This assignment is quite simple with heuristic rules such as:

\[
\begin{align*}
\dot{q}_1 &= 0; \\
\dot{q}_k &= \begin{cases} 
0 & \text{sign}(v_k) \neq \text{sign}(v_{k+1}) \\
\frac{1}{2}(v_k + v_{k+1}) & \text{sign}(v_k) = \text{sign}(v_{k+1})
\end{cases} \\
\dot{q}_n &= 0
\end{align*}
\]

being

\[v_k = \frac{q_k - q_{k-1}}{t_k - t_{k-1}}\]

the ‘slope’ of the tract \([t_{k-1} - t_k]\).
Third-order polynomial trajectories

Automatic computation of the intermediate velocities (data as in the previous example)

\[
\begin{align*}
    t_0 &= 0 & t_1 &= 2 & t_2 &= 4 & t_3 &= 8 & t_4 &= 10 \\
    q_0 &= 10^\circ & q_1 &= 20^\circ & q_2 &= 0^\circ & q_3 &= 30^\circ & q_4 &= 40^\circ
\end{align*}
\]
Fifth-order polynomial trajectories

From the above examples, it may be noticed that both the position and velocity profiles are continuous functions of time.

This is not true for the acceleration, that presents discontinuities among different segments. Moreover, it is not possible to specify for this signal suitable initial/final values in each segment.

In many applications, these aspects do not constitute a problem, being the trajectories “smooth” enough.

On the other hand, if it is requested to specify initial and final values for the acceleration (e.g. for obtaining continuous acceleration profiles), then (at least) fifth-order polynomial functions should be considered

\[ q(t) = a_0 + a_1 t + a_2 t^2 + a_3 t^3 + a_4 t^4 + a_5 t^5 \]

with the six boundary conditions:

\[
\begin{align*}
q(t_i) &= q_i & q(t_f) &= q_f \\
\dot{q}(t_i) &= \dot{q}_i & \dot{q}(t_f) &= \dot{q}_f \\
\ddot{q}(t_i) &= \ddot{q}_i & \ddot{q}(t_f) &= \ddot{q}_f \\
\end{align*}
\]
Fifth-order polynomial trajectories

In this case (if $T = t_f - t_i$) the coefficients of the polynomial are

$$a_0 = q_i$$
$$a_1 = \dot{q}_i$$
$$a_2 = \frac{1}{2} \ddot{q}_i$$
$$a_3 = \frac{1}{2T^3} [20(q_f - q_i) - (8\dot{q}_f + 12\dot{q}_i)T - (3\ddot{q}_f - \ddot{q}_i)T^2]$$
$$a_4 = \frac{1}{2T^4} [30(q_i - q_f) + (14\dot{q}_f + 16\dot{q}_i)T + (3\ddot{q}_f - 2\ddot{q}_i)T^2]$$
$$a_5 = \frac{1}{2T^5} [12(q_f - q_i) - 6(\dot{q}_f + \dot{q}_i)T - (\ddot{q}_f - \ddot{q}_i)T^2]$$

If a sequence of points is given, the same considerations made for third-order polynomials trajectories can be made in computing the intermediate velocity values.
Fifth-order polynomial trajectories

Fifth-order trajectory with the boundary conditions:

\[ q_i = 10^\circ, \quad q_f = 30^\circ, \quad \dot{q}_i = \dot{q}_f = 0 \, \text{°/s}, \quad \ddot{q}_i = \ddot{q}_f = 0 \, \text{°/s}^2, \quad t_i = 0, \quad t_f = 1 \, \text{s}. \]

Obviously:

- **position** \(\rightarrow\) 5-th order function
- **velocity** \(\rightarrow\) 4-th order function
- **acceleration** \(\rightarrow\) 3-rd order function
Fifth-order polynomial trajectories

Comparison of fifth- and third-order trajectories with the boundary conditions:
\[ q_i = 10^\circ, \quad q_f = 30^\circ, \quad \dot{q}_i = \dot{q}_f = 0^\circ/s, \quad \ddot{q}_i = \ddot{q}_f = 0^\circ/s^2, \quad t_i = 0s, \quad t_f = 1s. \]
Fifth-order polynomial trajectories

Position, velocity, acceleration profiles with automatic assignment of the intermediate velocities and null accelerations.

Note that the resulting motion has smoother profiles.
Other functions

Many other families of functions have been used to interpolate a sequence of points with given boundary conditions. Among others:

- Higher order polynomial functions (e.g. 7, 9, 11, ...)
- Harmonic trajectories
- Cycloidal trajectories
- Elliptic trajectories
- Gutman trajectories
- Freudenstein trajectories
- ...

A different approach for planning a trajectory is to use different functions in different segments of the same path. In general, this method increases the flexibility of the overall function, and may adapt the obtained trajectory to different constraints.
Trapezoidal trajectories

Among many other combinations, a possible approach for planning a trajectory is to use linear segments joined with parabolic blends.

In the linear tract, the velocity is constant while, in the parabolic blends, it is a linear function of time: *trapezoidal velocity profiles*, typical of this type of trajectory, are then obtained.

In trapezoidal trajectories, the duration is divided into three parts:

1. in the first part, a constant acceleration is applied, then the velocity is linear and the position results a parabolic function of time
2. in the second, the acceleration is null, the velocity is constant and the position is linear in time
3. in the last part a (negative) acceleration is applied, then the velocity is a negative ramp and the position a parabolic function.
Trapezoidal trajectories

Usually, the acceleration and the deceleration phases have the same duration \((t_a = t_d)\). Therefore, symmetric profiles, with respect to a central instant \((t_f - t_i)/2\), are obtained.

Two trapezoidal profiles with different duration of the acceleration segment.

Notice that both profiles are symmetric with respect to the central point.
The trajectory is computed according to the following equations.

1) **Acceleration phase,** \( t \in [0 \div t_a] \).

The position, velocity and acceleration are described by

\[
\begin{align*}
q(t) &= a_0 + a_1 t + a_2 t^2 \\
\dot{q}(t) &= a_1 + 2a_2 t \\
\ddot{q}(t) &= 2a_2
\end{align*}
\]

The parameters are defined by constraints on the initial position \( q_i \) and velocity \( \dot{q}_i \), and on the desired constant velocity \( \dot{q}_v \) that must be obtained at the end of the acceleration period. Assuming a null initial velocity \( (\dot{q}_i = 0) \) and considering \( t_i = 0 \) one obtains

\[
\begin{align*}
a_0 &= q_i \\
a_1 &= 0 \\
a_2 &= \frac{\dot{q}_v}{2t_a}
\end{align*}
\]

In this phase, the acceleration is constant and equal to \( \dot{q}_v / t_a \).
2) **Constant velocity phase**, \( t \in [t_a \div t_f - t_a] \).

Position, velocity and acceleration are now defined as

\[
\begin{align*}
q(t) &= b_0 + b_1 t \\
\dot{q}(t) &= b_1 \\
\ddot{q}(t) &= 0
\end{align*}
\]

where, because of continuity,

\[b_1 = \dot{q}_v\]

Moreover, the following equation must hold

\[q(t_a) = q_i + \dot{q}_v \frac{t_a}{2} = b_0 + \dot{q}_v t_a\]

and then

\[b_0 = q_i - \dot{q}_v \frac{t_a}{2}\]
3) **Deceleration phase,** $t \in [t_f - t_a \div t_f]$.

The position, velocity and acceleration are given by

\[ q(t) = c_0 + c_1 t + c_2 t^2 \]
\[ \dot{q}(t) = c_1 + 2c_2 t \]
\[ \ddot{q}(t) = 2c_2 \]

The parameters are now defined with constraints on the final position $q_f$ and velocity $\dot{q}_f$, and on the velocity $\dot{q}_v$ at the beginning of the deceleration period.

If the final velocity is null, then:

\[ c_0 = q_f - \frac{\dot{q}_v t_f^2}{2 t_a} \]
\[ c_1 = \dot{q}_v \frac{t_f}{t_a} \]
\[ c_2 = -\frac{\dot{q}_v}{2 t_a} \]
Trapezoidal trajectories

Summarizing, the trajectory is computed as

\[
q(t) = \begin{cases} 
q_i + \frac{\dot{q}_v}{2t_a} t^2 & 0 \leq t < t_a \\
q_i + \dot{q}_v \left( t - \frac{t_a}{2} \right) & t_a \leq t < t_f - t_a \\
q_f - \frac{\dot{q}_v}{t_a} \left( t_f - t \right)^2 & t_f - t_a \leq t \leq t_f
\end{cases}
\]
Trapezoidal trajectories

Typical position, velocity and acceleration profiles of a trapezoidal trajectory.

![Graphs showing position, velocity, and acceleration profiles of a trapezoidal trajectory.](image)
Trapezoidal trajectories

Some additional constraints must be specified in order to solve the previous equations (choice of $t_a$, $\dot{q}_v$, ...).

A typical constraint concerns the duration of the acceleration/deceleration periods $t_a$ that, for symmetry, must satisfy the condition

$$t_a \leq t_f / 2$$

Moreover, the following condition must be verified (for the sake of simplicity, consider $t_i = 0$):

$$\ddot{q}t_a = \frac{q_m - q_a}{t_m - t_a}$$

$$\begin{align*}
q_a &= q(t_a) \\
q_m &= (q_i + q_f)/2 \\
t_m &= t_f/2
\end{align*}$$

$$q_a = q_i + \frac{1}{2} \ddot{q}t_a^2$$

from which

$$\ddot{q}t_a^2 - \ddot{q}t_f t_a + (q_f - q_i) = 0$$

Finally:

$$\dot{q}_v = \frac{q_f - q_i}{t_f - t_a}$$
Trapezoidal trajectories

Any pair of values \((\ddot{q}, t_a)\) verifying (7) can be considered.

Given the acceleration \(\ddot{q}\) (for example \(\ddot{q}_{\text{max}}\)), then

\[
t_a = \frac{t_f}{2} - \frac{\sqrt{\ddot{q}^2 t_f^2 - 4\ddot{q}(q_f - q_i)}}{2\ddot{q}}
\]

from which we have also that the minimum value for the acceleration is

\[
|\ddot{q}| \geq \frac{4|q_f - q_i|}{t_f^2} = \frac{4|L|}{t_f^2}
\]

if the value \(|\ddot{q}| = \frac{4|L|}{t_f^2}\) is assigned, then \(t_a = t_f/2\) and the constant velocity tract does not exist.

If the value \(t_a = t_f/3\) is specified, the following velocity and acceleration values are obtained

\[
\dot{q}_v = \frac{3L}{2t_f} \quad \ddot{q} = \frac{9L}{2t_f^2}
\]
Trapezoidal trajectories

Another way to compute this type of trajectory is to define a maximum value $\ddot{q}_a$ for the desired acceleration and then compute the relative duration $t_a$ of the acceleration and deceleration periods.

If the maximum values ($\ddot{q}_{\text{max}}$ and $\dot{q}_{\text{max}}$, known) for the acceleration and velocity must be reached, it is possible to assign

$$
\begin{align*}
    t_a &= \frac{\dot{q}_{\text{max}}}{\ddot{q}_{\text{max}}} \quad \text{acceleration time} \\
    \dot{q}_{\text{max}}(T - t_a) &= q_f - q_i = L \quad \text{displacement} \\
    T &= \frac{L\ddot{q}_{\text{max}} + \dot{q}_{\text{max}}^2}{\ddot{q}_{\text{max}}q_{\text{max}}} \quad \text{time duration}
\end{align*}
$$

and then ($t_f = t_i + T$)

$$
q(t) = \begin{cases} 
    q_i + \frac{1}{2}\ddot{q}_{\text{max}}(t - t_i)^2 & t_i \leq t \leq t_i + t_a \\
    q_i + \ddot{q}_{\text{max}}t_a(t - t_i - \frac{t_a}{2}) & t_i + t_a < t \leq t_f - t_a \\
    q_f - \frac{1}{2}\ddot{q}_{\text{max}}(t_f - t - t_i)^2 & t_f - t_a < t \leq t_f 
\end{cases}
$$

(8)
In this case, the linear tract exists if and only if
\[ L \geq \frac{q_{\max}^2}{\ddot{q}_{\max}} \]

Otherwise
\[
\begin{aligned}
\frac{a}{c} &= \sqrt{\frac{L}{q_{\max}}} \\
T &= 2t_a
\end{aligned}
\]

acceleration time
total time duration

and (still \( t_f = t_i + T \))

\[
q(t) = \begin{cases} 
q_i + \frac{1}{2} \ddot{q}_{\max} (t - t_i)^2 & \text{for } t_i \leq t \leq t_i + t_a \\
q_f - \frac{1}{2} \ddot{q}_{\max} (t_f - t)^2 & \text{for } t_f - t_a < t \leq t_f 
\end{cases}
\] (9)
Trapezoidal trajectories

With this modality for computing the trajectory, the time duration of the motion from $q_i$ to $q_f$ is not specified. In fact, the period $T$ is computed on the basis of the maximum acceleration and velocity values.

If more joints $q_i, i = \ldots, n$ have to be co-ordinated with the same constraints on the maximum acceleration and velocity ($\ddot{q}_{max}, \dot{q}_{max}$), the joint with the largest displacement $|L_k|$ must be individuated. For this joint, the maximum value $\ddot{q}_{max}$ for the acceleration is assigned, and then the corresponding values $t_a$ and $T$ are computed.

For the remaining joints, the acceleration and velocity values must be computed on the basis of these values of $t_a$ and $T$, and on the basis of the given displacement $L_i$:

$$\ddot{q}_i = \frac{L_i}{t_a(T - t_a)}, \quad \dot{q}_i = \frac{L_i}{T - t_a}, \quad i = 1, \ldots, n, \quad i \neq k$$
Trapezoidal trajectories

Largest displacements:

- 1st segment: second joint $L_{1,2}$
- 2nd segment: second joint $L_{2,2}$
- 3th segment: first joint $L_{3,1}$
Trapezoidal trajectories

The trajectories in the workspace are:

- Motion of a two-dof manipulator
- Cartesian space positions
- Cartesian space velocities
Trapezoidal trajectories

If a trajectory interpolating more consecutive points is computed with the above technique, a motion with null velocities in the via-points is obtained. Since this behavior may be undesirable, it is possible to “anticipate” the actuation of a tract of the trajectory between points $q_k$ and $q_{k+1}$ before the motion from $q_{k-1}$ to $q_k$ is terminated. This is possible by adding (starting at an instant $t_k - t_a'$) the velocity and acceleration contributions of the two segments $[q_{k-1} - q_k]$ and $[q_k - q_{k+1}]$. Obviously, another possibility is to compute the parameters of the functions defining the trapezoidal trajectory in order to have desired boundary conditions (i.e. velocities) for each segment.
A trapezoidal velocity motion profile presents a discontinuous acceleration. For this reason, this trajectory may generate efforts and stresses on the mechanical system that may result detrimental or generate undesired vibrational effects.

Therefore, a smoother motion profile must be defined, for example by adopting a continuous, linear piece-wise, acceleration profile. In this manner, the resulting velocity is composed by linear segments connected by parabolic blends.

The shape of the velocity profile is the reason of the name double S for this trajectory, also known as bell trajectory or seven segments trajectory, because it is composed by seven different tracts with constant jerk.
Other functions can be obtained by properly composing segments defined with polynomial functions of different degree (piecewise polynomial functions). In these cases, it is necessary to define an adequate number of conditions (boundary conditions, point crossing, continuity of velocity, acceleration, ...), as done e.g. for the computation of trapezoidal (linear segments with second or higher degree polynomials blends) and ‘double S’ trajectories.

For example, in pick-and-place operations by an industrial robot it may be of interest to have motions with very smooth initial and final phases. In such a case, one can use a motion profile obtained as the connection of three polynomials \( q_l(t) \), \( q_t(t) \), \( q_s(t) \) (i.e. lift-off, travel, set-down) with (for example):

\[
\begin{align*}
q_l(t) & \quad \implies \quad 4\text{-th degree polyn.} \\
q_t(t) & \quad \implies \quad 3\text{-rd degree polyn.} \\
q_s(t) & \quad \implies \quad 4\text{-th degree polyn.}
\end{align*}
\]
In general, the problem of defining a function interpolating a set of $n$ points can be solved with a polynomial function of degree $n - 1$.

In planning a trajectory, this approach does not give good results since the resulting motions in general present large oscillations.

In general, given:

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

The (unique) polynomial $p(x)$ with degree $n - 1$ interpolating $n$ points $(x_i, y_i)$ can be computed by the *Lagrange* expression:

$$p(x) = \frac{(x - x_2)(x - x_3) \cdots (x - x_n)}{(x_1 - x_2)(x_1 - x_3) \cdots (x_1 - x_n)} y_1 + \frac{(x - x_1)(x - x_3) \cdots (x - x_n)}{(x_2 - x_1)(x_2 - x_3) \cdots (x_2 - x_n)} y_2 + \cdots$$

$$+ \cdots + \frac{(x - x_1)(x - x_2) \cdots (x - x_{n-1})}{(x_n - x_1)(x_n - x_2) \cdots (x_n - x_{n-1})} y_n$$

Other (recursive) expressions have been defined, more efficient from a computational point of view (*Neville formulation*).
Another (less efficient) approach for the computation of the coefficients of the polynomial $p(x)$ interpolating the $n$ points $(x_i, y_i)$ is based on the following procedure:

$$y_i = p(x_i) = a_{n-1}x_i^{n-1} + \cdots + a_1x_i + a_0 \quad i = 1, \ldots, n$$

$$y = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_{n-1} \\ y_n \end{bmatrix} = \begin{bmatrix} x_1^{n-1} & x_1^{n-2} & \cdots & x_1 & 1 \\ x_2^{n-1} & x_2^{n-2} & \cdots & x_2 & 1 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ x_{n-1}^{n-1} & x_{n-1}^{n-2} & \cdots & x_{n-1} & 1 \\ x_n^{n-1} & x_n^{n-2} & \cdots & x_n & 1 \end{bmatrix} \begin{bmatrix} a_{n-1} \\ a_{n-2} \\ \vdots \\ a_1 \\ a_0 \end{bmatrix} = X a$$

and then, by inverting matrix $X$, the parameters are obtained

$$a = X^{-1} y$$

Matrix $X$ contains as elements “1” and “$x_i^{n-1}$” (in general with values having orders of magnitude of difference), and therefore is badly conditioned from a numerical point of view (condition number).

⇒⇒ Numerical problems in computing $X^{-1}$ for high values of $n$!!!
Given $n$ points, in order to avoid the problem of high ‘oscillations’ (and also of the numerical precision):

---

**NO:** one polynomial of degree $n - 1$

---

**YES:** $n - 1$ polynomials with lower degree $p$ ($p < n - 1$): each polynomial interpolates a segment of the trajectory.

---

Usually, the degree $p$ of the $n - 1$ polynomials is chosen so that continuity of the velocity and acceleration profile is achieved. In this case, the choice $p = 3$ is made (cubic polynomials):

\[ q(t) = a_0 + a_1 t + a_2 t^2 + a_3 t^3 \]

There are 4 coefficients for each polynomial, and therefore it is necessary to compute $4(n - 1)$ coefficients.

Obviously, it is possible to choose higher values for $p$ (e.g. $p = 5, 7, \ldots$).
4\(n - 1\) coefficients

On the other hand, there are:

- 2\(n - 1\) conditions on the position (each cubic function interpolates its own initial/final points);
- \(n - 2\) conditions on the continuity of velocity in the intermediate points;
- \(n - 2\) conditions on the continuity of acceleration in the intermediate points.

Therefore, there are

\[
4(n - 1) - 2(n - 1) - 2(n - 2) = 2
\]

degrees of freedom left, that can be used for example for imposing proper conditions on the initial and final velocity.
The function obtained in this manner is a **spline**.

**Property:** Among all the interpolating functions of \( n \) points with the same degree of continuity of derivation, the spline has the smallest *curvature*. 
Mathematically, it is necessary to compute a function

\[
\begin{align*}
q(t) &= \{q_k(t), \quad t \in [t_k, t_{k+1}], \quad k = 1, \ldots, n - 1\} \\
q_k(\tau) &= a_{k0} + a_{k1}\tau + a_{k2}\tau^2 + a_{k3}\tau^3, \quad \tau \in [0, T_k], \\
&\quad (\tau = t - t_k, \quad T_k = t_{k+1} - t_k)
\end{align*}
\]

with the conditions

\[
\begin{align*}
q_k(0) &= q_k, \quad q_k(T_k) = q_{k+1} \quad k = 1, \ldots, n - 1 \\
\dot{q}_k(T_k) &= \dot{q}_{k+1}(0) = v_k \quad k = 1, \ldots, n - 2 \\
\ddot{q}_k(T_k) &= \ddot{q}_{k+1}(0) \quad k = 1, \ldots, n - 2
\end{align*}
\]
Spline - Computation

The parameters $a_{ki}$ are computed according to the following algorithm.

Let assume that the velocities $v_k$, $k = 2, \ldots, n - 1$ in the intermediate points are known.

In this case, we could impose for each cubic polynomial the four boundary conditions on position and velocity:

\[
\begin{align*}
q_k(0) &= a_{k0} & \quad = q_k \\
q_k(0) &= a_{k1} & \quad = v_k \\
q_k(T_k) &= a_{k0} + a_{k1} T_k + a_{k2} T_k^2 + a_{k3} T_k^3 & \quad = q_{k+1} \\
\dot{q}_k(T_k) &= a_{k1} + 2a_{k2} T_k + 3a_{k3} T_k^2 & \quad = v_{k+1}
\end{align*}
\]

and then

\[
\begin{align*}
a_{k0} &= q_k \\
a_{k1} &= v_k \\
a_{k2} &= \frac{1}{T_k} \left[ \frac{3(q_{k+1} - q_k)}{T_k} - 2v_k - v_{k+1} \right] \\
a_{k3} &= \frac{1}{T_k^2} \left[ \frac{2(q_k - q_{k+1})}{T_k} + v_k + v_{k+1} \right]
\end{align*}
\]

... but the velocities $v_k$ are not known...
Spline - Computation

By using the conditions on continuity of the accelerations in the intermediate points, one obtains

\[ \ddot{q}_k(T_k) = 2a_{k2} + 6a_{k3} \quad T_k = 2a_{k+1,2} = \ddot{q}_{k+1}(0) \quad k = 1, \ldots, n - 2 \]

from which, by substituting the expressions of \( a_{k2}, a_{k3}, a_{k+1,2} \) and multiplying by \( (T_k \ T_{k+1})/2 \), one obtains

\[ T_{k+1}v_k + 2(T_k + T_{k+1})v_{k+1} + T_k v_{k+2} = \frac{3}{T_k T_{k+1}} \left[ T_k^2(q_{k+2} - q_{k+1}) + T_{k+1}^2(q_{k+1} - q_k) \right] \]

These equations may be written in matrix form as

\[
\begin{bmatrix}
T_2 & 2(T_1 + T_2) & T_1 & 0 \\
0 & T_3 & 2(T_2 + T_3) & T_2 \\
& & \ddots & \vdots \\
& & & T_{n-1} & 2(T_{n-2} + T_{n-1}) & T_{n-3} & 0 \\
& & & \ddots & \ddots & \ddots & \vdots \\
& & & & T_{n-1} & 2(T_{n-2} + T_{n-1}) & T_{n-3} & 0 \\
& & & & & \ddots & \ddots & \vdots \\
& & & & & & T_{n-1} & 2(T_{n-2} + T_{n-1}) & T_{n-3} & 0 \\
\end{bmatrix}
\begin{bmatrix}
v_1 \\
v_2 \\
\vdots \\
v_{n-1} \\
v_n \\
\end{bmatrix} =
\begin{bmatrix}
c_1 \\
c_2 \\
\vdots \\
c_{n-3} \\
c_{n-2} \\
\end{bmatrix}
\]

where the \( c_k \) are (known) constant terms depending on the intermediate positions and the duration of each segments.
Since the velocities $v_1$ and $v_n$ are known, the corresponding columns can be eliminated from the left-hand side matrix, and then

\[
\begin{bmatrix}
2(T_1 + T_2) & T_1 & T_2 \\
T_3 & 2(T_2 + T_3) & T_3 \\
& \ddots & \ddots & \ddots \\
& 2(T_{n-3} + T_{n-2}) & T_{n-3} & 2(T_{n-2} + T_{n-1}) \\
& T_{n-1} & \ddots & \ddots & \ddots \\
& & \ddots & \ddots & \ddots \\
& & & \ddots & \ddots \\
& & & & \ddots & \ddots \\
& & & & & \ddots & \ddots \\
& & & & & & \ddots & \ddots \\
& & & & & & & \ddots & \ddots \\
& & & & & & & & \ddots & \ddots \\
& & & & & & & & & \ddots & \ddots \\
& & & & & & & & & & \ddots & \ddots \\
\end{bmatrix}
\begin{bmatrix}
v_2 \\
\vdots \\
v_{n-1} \\
\end{bmatrix}
= \begin{bmatrix}
\frac{3}{T_1 T_2} \left[ T_1^2(q_3 - q_2) + T_2^2(q_2 - q_1) \right] - T_2 v_1 \\
\frac{3}{T_2 T_3} \left[ T_2^2(q_4 - q_3) + T_3^2(q_3 - q_2) \right] \\
\vdots \\
\frac{3}{T_{n-2} T_{n-1}} \left[ T_{n-2}^2(q_{n-1} - q_{n-2}) + T_{n-1}^2(q_{n-2} - q_{n-3}) \right] - T_{n-2} v_n \\
\end{bmatrix}
\]

that is

\[
A(T) v = c(T, q, v_1, v_n)
\]

or

\[
A v = c, \quad \text{with } A \in \mathbb{R}^{(n-2) \times (n-2)}
\]
The matrix $A$ is tridiagonal, and is always invertible if $T_k > 0$ ($|a_{kk}| > \sum_{j \neq k} |a_{kj}|$).

Being $A$ tridiagonal, its inverse is computed by efficient numerical algorithms (based on the Gauss-Jordan method).

Once $A^{-1}$ is known, the velocities $v_2, \ldots, v_{n-1}$ are computed as

$$v = A^{-1} c$$

and the problem is solved (the $n - 1$ polynomials are computed with the equations (10)).
The total duration of a spline is

\[ T = \sum_{k=1}^{n-1} T_k = t_n - t_1 \]

It is possible to define an optimality problem aiming at minimizing \( T \). The values of \( T_k \) must be computed so that \( T \) is minimized and the constraints on the velocity and acceleration are satisfied.

Formally the problem is formulated as

\[
\begin{cases}
\min_{T_k} \quad T = \sum_{k=1}^{n-1} T_k \\
\text{such that} \\
|\dot{q}(\tau, T_k)| < v_{\text{max}} \quad \tau \in [0, T] \\
|\ddot{q}(\tau, T_k)| < a_{\text{max}} \quad \tau \in [0, T]
\end{cases}
\]

Non linear optimization problem with linear objective function, solvable with classical techniques from the operational research field.
A spline through the points $q_1 = 0$, $q_2 = 2$, $q_3 = 12$, $q_4 = 5$ must be defined, minimizing the total duration $T$ and with the constraints: $v_{max} = 3$, $a_{max} = 2$.

The non linear optimization problem

$$\min \quad T = T_1 + T_2 + T_3$$

is defined, with the constraints reported in the following slide.

By solving this problem (e.g. with the Matlab Optimization Toolbox) the following values are obtained:

$$T_1 = 1.5549, \quad T_2 = 4.4451, \quad T_3 = 4.5826, \quad \Rightarrow \quad T = 10.5826 \text{ sec}$$
### Constraints on the optimization problem:

\[
\begin{align*}
    a_01 & \
    a_11 & \
    a_21 & \
    a_01 & +2a_02 T_1 + 3a_03 T_1^2 \
    a_11 & +2a_12 T_2 + 3a_13 T_2^2 \
    a_21 & +2a_22 T_3 + 3a_23 T_3^2 \\
    a_01 & +2a_02 \left( -\frac{a_02}{3a_03} \right) + 3a_03 \left( -\frac{a_02}{3a_03} \right)^2 \
    a_11 & +2a_12 \left( -\frac{a_12}{3a_13} \right) + 3a_13 \left( -\frac{a_12}{3a_13} \right)^2 \
    a_21 & +2a_22 \left( -\frac{a_22}{3a_23} \right) + 3a_23 \left( -\frac{a_22}{3a_23} \right)^2 \\
\end{align*}
\]

\[\begin{align*}
    a_{02} & \
    a_{12} & \
    a_{22} & \
    2a_{02} & +6a_{03} T_1 \
    2a_{12} & +6a_{13} T_2 \
    2a_{22} & +6a_{23} T_3 \\
\end{align*}\]

\(\leq v_{\text{max}}\) (init. vel. 1-st tract \(\leq v_{\text{max}}\))
\(\leq v_{\text{max}}\) (init. vel. 2-nd tract \(\leq v_{\text{max}}\))
\(\leq v_{\text{max}}\) (init. vel. 3-rd tract \(\leq v_{\text{max}}\))
\(\leq v_{\text{max}}\) (final vel. 1-st tract \(\leq v_{\text{max}}\))
\(\leq v_{\text{max}}\) (final vel. 2-nd tract \(\leq v_{\text{max}}\))
\(\leq v_{\text{max}}\) (final vel. 3-rd tract \(\leq v_{\text{max}}\))
\(\leq v_{\text{max}}\) (vel. 1-st tract \(\leq v_{\text{max}}\))
\(\leq v_{\text{max}}\) (vel. 2-nd tract \(\leq v_{\text{max}}\))
\(\leq v_{\text{max}}\) (vel. 3-rd tract \(\leq v_{\text{max}}\))
\(\leq a_{\text{max}}\) (init. acc. 1-st tract \(\leq a_{\text{max}}\))
\(\leq a_{\text{max}}\) (init. acc. 2-nd tract \(\leq a_{\text{max}}\))
\(\leq a_{\text{max}}\) (init. acc. 3-rd tract \(\leq a_{\text{max}}\))
\(\leq a_{\text{max}}\) (final acc. 1-st tract \(\leq a_{\text{max}}\))
\(\leq a_{\text{max}}\) (final acc. 2-nd tract \(\leq a_{\text{max}}\))
\(\leq a_{\text{max}}\) (final acc. 3-rd tract \(\leq a_{\text{max}}\)).
Position, velocity and acceleration profiles of the optimal trajectory.
Spline

The above procedure for computing the spline is adopted also for more motion axes (joints). Notice that the matrix $A_i(T) = A(T)$ is the same for all the $i = 1, \ldots, m$ joints (it depends only on the parameters $T_k$), while the vector $c(T, q_i, v_{i1}, v_{in})$ depends on the specific $i$-th joint.

Syncronized splines for two joints:
- same periods $T_k$,
- different points to be interpolated.
From the expressions of matrix $A$ and the vector $c$ ($Av = c$)

$$A = \begin{bmatrix}
2(T_1 + T_2) & T_1 & T_2 \\
T_3 & 2(T_2 + T_3) & T_2 \\
& \ddots & \ddots & \ddots \\
& & T_{n-2} & 2(T_{n-3} + T_{n-2}) & T_{n-3} \\
& & & \ddots & \ddots & \ddots \\
& & & & T_{n-2} & 2(T_{n-3} + T_{n-2}) \\
& & & & & T_{n-1} \\
\end{bmatrix}$$

$$c = \begin{bmatrix}
\frac{3}{T_1 T_2} [T_1^2(q_3 - q_2) + T_2^2(q_2 - q_1)] - T_2 v_1 \\
\frac{3}{T_2 T_3} [T_2^2(q_4 - q_3) + T_3^2(q_3 - q_2)] \\
& \ddots \\
& & \frac{3}{T_{n-3} T_{n-2}} [T_{n-3}^2(q_{n-1} - q_{n-2}) + T_{n-2}^2(q_{n-2} - q_{n-3})] \\
& & & \frac{3}{T_{n-2} T_{n-1}} [T_{n-2}^2(q_n - q_{n-1}) + T_{n-1}^2(q_n - q_{n-2})] - T_{n-2} v_n
\end{bmatrix}$$
If

- the duration $T_k$ of each interval is multiplied by a constant $\lambda$ (linear scaling)
- the initial and final velocities are null

one obtains that the new duration $T'$, the velocities and accelerations of the new trajectory are:

$$T' = \lambda T$$
$$v'_k = \frac{1}{\lambda} v_k$$
$$a'_k = \frac{1}{\lambda^2} a_k$$

The parameter $\lambda$ can then be defined in order to satisfy the constraints on maximum velocities/accelerations and obtain a minimum-time trajectory.
Comparison of a $n - 1$ polynomial, a spline, and a composition of cubic polynomials.

11 points, $v_{in} = v_{fin} = 0$ m/s
Spline - Example

- **Velocità per spline**

- **Velocità per cubica**

- **Accelerazione per spline**

- **Accelerazione per cubica**

---

C. Melchiorri  Trajectory Planning  66 / 140
Trajectory Planning for Robot Manipulators
Scaling Trajectories

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Scaling trajectories

Due to several reasons, like limits on the actuation system (torques, accelerations, velocities, ...) or computational efficiency, it is often requested to scale trajectories and motion laws.

It is possible to adopt

- **Kinematic** scaling procedures
- **Dynamic** scaling procedures
If a trajectory is expressed in parametric form as a function of a parameter \( \sigma = \sigma(t) \), by changing the parameterization it is possible to obtain in a simple manner a trajectory satisfying constraints on velocity or accelerations.

For this purpose, it is convenient to express the trajectory in *normal form*, i.e.:

\[
p(t) = p_0 + (p_1 - p_0)s(\tau) = p_0 + Ls(\tau)
\]

being \( s(\tau) \) a proper parameterization, with

\[
0 \leq s \leq 1, \quad \tau = \frac{t - t_0}{t_1 - t_0} = \frac{t - t_0}{T}, \quad 0 \leq \tau \leq 1
\]

In this manner, it results

\[
\frac{dp}{dt} = \frac{L}{T} s'(\tau)
\]

\[
\frac{d^2p}{dt^2} = \frac{L}{T^2} s''(\tau)
\]

\[
\frac{d^3p}{dt^3} = \frac{L}{T^3} s'''(\tau)
\]

\[
\frac{dnp}{dt^n} = \frac{L}{T^n} s^{(n)}(\tau)
\]

\[
\ldots
\]
Kinematic scaling of trajectories

From

\[
\frac{dp}{dt} = \frac{L}{T} s'(\tau) \quad \frac{d^2 p}{dt^2} = \frac{L}{T^2} s''(\tau) \\
\frac{d^3 p}{dt^3} = \frac{L}{T^3} s'''(\tau) \\
\frac{d^n p}{dt^n} = \frac{L}{T^n} s^{(n)}(\tau)
\]

it follows that the maximum values for the velocity, acceleration, etc. are obtained in correspondence of the maximum values of the functions \( s', s'', \ldots \).

These values and the corresponding time instants \( \tau(t) \) are known from the chosen parameterization \( s(\tau) \).

Notice that if the duration \( T \) of the trajectory is changed, it is possible to satisfy in an exact manner the given constraints or to optimize the trajectory itself (minimum time). Moreover, it is easily possible to co-ordinate more motion axes.
Polynomial trajectories of degree 3

Consider a parameterization expressed by a cubic polynomial

\[ s(\tau) = a_0 + a_1 \tau + a_2 \tau^2 + a_3 \tau^3, \quad 0 \leq s \leq 1, \quad 0 \leq \tau \leq 1, \quad \tau = \frac{t}{T} \]

If the boundary conditions \( s_0 = 0, s_1 = 1, \quad v_0 = 0, v_1 = 0 \) are specified, one obtains

\[ a_0 = 0, \quad a_1 = 0, \quad a_2 = 3, \quad a_3 = -2 \]

Therefore:

\[
\begin{align*}
    s(\tau) &= 3\tau^2 - 2\tau^3 \\
    s'(\tau) &= 6\tau - 6\tau^2 \\
    s''(\tau) &= 6 - 12\tau \\
    s'''(\tau) &= -12
\end{align*}
\]
Then

\[ s'_{\text{max}} = s'(0.5) = \frac{3}{2} \]
\[ s''_{\text{max}} = s''(0) = 6 \]

\[ \Rightarrow \quad \dot{q}_{\text{max}} = \frac{3L}{2T} \]
\[ \ddot{q}_{\text{max}} = \frac{6L}{T^2} \]
Polynomial trajectories of degree 5

The polynomial $s(\tau)$ in normal form is now:

$$s(\tau) = a_0 + a_1 \tau + a_2 \tau^2 + a_3 \tau^3 + a_4 \tau^4 + a_5 \tau^5,$$

$$0 \leq s \leq 1, \quad 0 \leq \tau \leq 1, \quad \tau = \frac{t}{T}$$

With null boundary conditions on accelerations and velocities, the following values for the parameters are obtained (trajectory 3-4-5)

$$a_0 = 0, \quad a_1 = 0, \quad a_2 = 0, \quad a_3 = 10 \quad a_4 = -14, \quad a_5 = 6$$

Then

$$s(\tau) = 10\tau^3 - 15\tau^4 + 6\tau^5$$
$$s'(\tau) = 30\tau^2 - 60\tau^3 + 30\tau^4$$
$$s''(\tau) = 60\tau - 180\tau^2 + 120\tau^3$$
$$s'''(\tau) = 60 - 360\tau + 360\tau^2$$
Therefore

\[ s'_{\text{max}} = s'(0.5) = \frac{15}{8} \]
\[ s''_{\text{max}} = s''(0.2123) = \frac{10\sqrt{3}}{3} \]
\[ s'''_{\text{max}} = s'''(0) = 60 \]

\[ \Rightarrow \quad \dot{q}_{\text{max}} = \frac{15L}{8T} \]
\[ \Rightarrow \quad \ddot{q}_{\text{max}} = \frac{10\sqrt{3}L}{3T^2} \]
\[ \Rightarrow \quad \dddot{q}_{\text{max}} = 60\frac{L}{T^3} \]
Polynomial trajectories of degree 7

If a continuous jerk profile is requested, a polynomial with higher degree must be adopted. The normal form for a polynomial $s(\tau)$ of degree 7 is:

$$s(\tau) = a_0 + a_1 \tau + a_2 \tau^2 + a_3 \tau^3 + a_4 \tau^4 + a_5 \tau^5 + a_6 \tau^6 + a_7 \tau^7$$

If null boundary conditions on velocity, acceleration and jerk are specified, the following parameters are obtained (trajectory 4-5-6-7):

$$a_0 = 0, \ a_1 = 0, \ a_2 = 0, \ a_3 = 0 \ a_4 = 35, \ a_5 = -84, \ a_6 = 70, \ a_7 = -20$$

Therefore

$$s(\tau) = 35\tau^4 - 84\tau^5 + 70\tau^6 - 20\tau^7$$

$$s'(\tau) = 140\tau^3 - 420\tau^4 + 420\tau^5 - 140\tau^6$$

$$s''(\tau) = 420\tau^2 - 1680\tau^3 + 2100\tau^4 - 840\tau^5$$

$$s'''(\tau) = 840\tau - 5040\tau^2 + 8400\tau^3 - 4200\tau^4$$
The maximum velocity and acceleration values are obtained for

\[
s'_{\text{max}} = s'(0.5) = \frac{35}{16}
\]

\[
s''_{\text{max}} = s'' \left( \frac{5 \pm \sqrt{5}}{10} \right) = \frac{84\sqrt{5}}{25}
\]

\[
s'''_{\text{max}} = s''' \left( \frac{1 + \sqrt{3/5}}{2} \right) = 42,\quad s'''_{\text{min}} = s'''(0.5) = -\frac{105}{2}
\]

\[
\Rightarrow \quad \dot{q}_{\text{max}} = \frac{35L}{16T} \quad \Rightarrow \quad \ddot{q}_{\text{max}} = \frac{84\sqrt{5}L}{25T^2} \quad \Rightarrow \quad \max \left| s'''' \right| = \frac{105}{2}
\]
Considerations on limits and durations of trajectories

From the previous examples, it is clear that if the displacement \( L \) and the duration \( T \) of a motion are specified, the profiles of velocity, acceleration and jerk are defined by the parameterization \( s(\tau) \) chosen to generate the motion profile.

In particular, the maximum values for these variables are determined (for the sake of simplicity, consider the case \( L > 0 \)).

<table>
<thead>
<tr>
<th></th>
<th>Polynomial 3</th>
<th>Polynomial 5</th>
<th>Polynomial 7</th>
<th>Cycloidal</th>
<th>Harmonic</th>
</tr>
</thead>
<tbody>
<tr>
<td>Vel. ((*L/T))</td>
<td>(3/2 = 1.5)</td>
<td>(15/8 = 1.875)</td>
<td>(35/16 = 2.1875)</td>
<td>2</td>
<td>(\pi/2 = 1.5708)</td>
</tr>
<tr>
<td>Acc. ((*L/T^2))</td>
<td>6</td>
<td>(10\sqrt{3}/3 = 5.7735)</td>
<td>(84\sqrt{5}/25 = 7.5132)</td>
<td>(2\pi = 6.2832)</td>
<td>(\pi^2/2 = 4.9348)</td>
</tr>
<tr>
<td>Jerk ((*L/T^3))</td>
<td>12</td>
<td>60</td>
<td>(105/2 = 52.25)</td>
<td>(4\pi^2 = 39.4784)</td>
<td>(\pi^3/2 = 15.5031)</td>
</tr>
</tbody>
</table>

Notice that the polynomial of degree 7, originating a very smooth profile, requires higher velocity and acceleration values. Viceversa, the harmonic trajectory has a very good behavior (smoothness with relatively low values for velocity and acceleration).
Example: scaling a trajectory

Trajectory 3-4-5. Polynomial in *normal form*:

\[ s(\tau) = a\tau^5 + b\tau^4 + c\tau^3 + d\tau^2 + e\tau + f \]

with

\[ 0 \leq s \leq 1, \quad 0 \leq \tau \leq 1, \quad \tau = \frac{t}{T} \]

The trajectory is

\[ q(t) = q_0 + (q_1 - q_0) s(\tau) = q_0 + Ls(\tau) \]

and

\[ \dot{q}(t) = Ls'(\tau) \frac{1}{T} \]
\[ \ddot{q}(t) = Ls''(\tau) \frac{1}{T^2} \]
\[ \dot{\cdots} \]
\[ \frac{d^n q}{dt^n} = Ls^{(n)}(\tau) \frac{1}{T^n} \]
Example: scaling a trajectory

Then (trajectory 3-4-5, \( f = e = d = 0 \), and \( a = 6, b = -15, c = 10 \))

\[
\begin{align*}
s'(\tau) &= 30\tau^4 - 60\tau^3 + 30\tau^2 \\
s''(\tau) &= 120\tau^3 - 180\tau^2 + 60\tau \\
s''''(\tau) &= 360\tau^2 - 360\tau + 60
\end{align*}
\]

and

\[
\begin{align*}
s'_{\text{max}} &= s'(0.5) = \frac{15}{8} \quad \implies \quad \dot{q}_{\text{max}} = \frac{15L}{8T} \\
s''_{\text{max}} &= s''(0.2123) = \frac{10\sqrt{3}}{3} \quad \implies \quad \ddot{q}_{\text{max}} = \frac{10\sqrt{3}L}{3T^2}
\end{align*}
\]

Given constraints on maximum acceleration and velocity, it is possible to properly scaling the trajectory.

Co-ordination of more motion axes made on the basis of the “most stressed” actuator.
Example: scaling a trajectory

If:

\[ q_0 = 0; \quad q_1 = 100; \quad t_0 = 0; \quad t_1 = 2; \quad \dot{q}_{max} = 200; \quad \ddot{q}_{max} = 400 \]
Example: scaling a trajectory

\[
T_{\text{min},v} = \frac{15L}{8q_{\text{max}}} = 0.9375 \, \text{s}, \quad T_{\text{min},a} = \sqrt{\frac{10\sqrt{3}L}{3q_{\text{max}}}} = 1.2014 \, \text{s} \quad T_{\text{min}} = \max\{T_{\text{min},v}, \, T_{\text{min},a}\}
\]
Dynamic scaling of trajectories

When a trajectory is specified for a complex mechanical system, because of the dynamics of the actuation system, of the robot manipulator or of the load (dynamic couplings), torques non physically achievable by the actuators could be requested. In these cases, it is possible to scale the trajectory taking into account the dynamics of the system in order to obtain a physically achievable motion.

The dynamic model of a manipulator is

\[
M(q)\ddot{q} + C(q, \dot{q})\dot{q} + g(q) = \tau
\]

Then, for each joint

\[
m_i^T(q)\ddot{q} + \frac{1}{2} \dot{q}^T C_i(q)\dot{q} + g_i(q) = \tau_i \quad i = 1, \ldots, n
\]

If

\[
q = q(\sigma) \quad \sigma = \sigma(t)
\]

is a proper parameterization of the trajectory with a motion law such that

\[
\dot{q} = \frac{d}{d\sigma} q \dot{\sigma}, \quad \ddot{q} = \frac{d^2}{d\sigma^2} q \dot{\sigma}^2 + \frac{d}{d\sigma} q \ddot{\sigma}
\]
Dynamic scaling of trajectories

By substitution in the dynamic model:

\[
\begin{bmatrix}
    m_i^T(q(\sigma)) \frac{d}{d\sigma} q \\
    m_i^T(q(\sigma)) \frac{d^2}{d\sigma^2} q \\
    \frac{1}{2} \frac{d}{d\sigma} C_i(q(\sigma)) \frac{d}{d\sigma} q
\end{bmatrix}
\dot{\sigma} + \begin{bmatrix}
    m_i^T(q(\sigma)) \frac{d}{d\sigma} q \\
    m_i^T(q(\sigma)) \frac{d^2}{d\sigma^2} q \\
    \frac{1}{2} \frac{d}{d\sigma} C_i(q(\sigma)) \frac{d}{d\sigma} q
\end{bmatrix}
\ddot{\sigma} + g_i(q(\sigma)) = \tau_i
\]

from which

\[
\alpha_i(\sigma) \ddot{\sigma} + \beta_i(\sigma) \dot{\sigma}^2 + \gamma_i(\sigma) = \tau_i
\]

Notice that \( \gamma_i(\sigma) \) (gravitational terms) depend on the position only \( \sigma \).
Dynamic scaling of trajectories

Let us suppose to compute the torques $\tau_i$ necessary to achieve the motion defined by $q = q(\sigma), \sigma = \sigma(t)$:

$$\tau_i(t) = \alpha_i(\sigma(t))\ddot{\sigma}(t) + \beta_i(\sigma(t))\dot{\sigma}^2(t) + \gamma_i(\sigma(t)), \quad i = 1, \ldots, n, \quad t \in [0, T]$$

If the time-axis is changed (e.g. in a linear fashion ($x = kt$)), a different parameterization of the trajectory is obtained:

$$t \rightarrow x = kt \quad x \in [0, kT] \quad \sigma(t) \rightarrow \hat{\sigma}(x)$$

Notice that in general even a non linear parameterization $x = x(t)$ could be considered.

With the new parameterization, one obtains:

$$\hat{\sigma}(x) = \sigma(t)$$
$$\dot{\hat{\sigma}}(x) = \frac{\dot{\sigma}(t)}{k}$$
$$\ddot{\hat{\sigma}}(x) = \frac{\ddot{\sigma}(t)}{k^2}$$
Dynamic scaling of trajectories

Therefore

- if $k > 1$ a \textit{slower} motion is obtained \( \left( \dot{\sigma}(x) < \frac{\dot{\sigma}(t)}{k} \right) \)

- if $k < 1$ a \textit{faster} motion is obtained \( \left( \dot{\sigma}(x) > \frac{\dot{\sigma}(t)}{k} \right) \).

With the new parameterization, the torques compute as:

\[
\tau_i(x) = \alpha_i(\hat{\sigma}(x))\ddot{\sigma}(x) + \beta_i(\hat{\sigma}(x))\dot{\sigma}^2(x) + \gamma_i(\hat{\sigma}(x))
\]

\[
= \alpha_i(\sigma(t))\frac{\ddot{\sigma}(t)}{k^2} + \beta_i(\sigma(t))\frac{\dot{\sigma}^2(t)}{k^2} + \gamma_i(\sigma(t))
\]

\[
= \frac{1}{k^2} [\tau_i(t) - \gamma_i(\sigma(t))] + \gamma_i(\sigma(t))
\]

from which

\[
\tau_i(x) - \gamma_i(x) = \frac{1}{k^2} [\tau_i(t) - \gamma_i(t)]
\]
Some considerations:

- It is not necessary to re-compute the whole trajectory.
- Neglecting the gravitational term, the new torques are obtained by scaling by the factor $1/k^2$ the previous torques.
- The motion is slower if $k > 1$, and it is faster if $k < 1$ (total duration equal to $kT$).

\[
\hat{\sigma}(x) = \hat{\sigma}(kt) = \sigma(t)
\]
Dynamic scaling of trajectories

**Example:** Consider a 2 dof manipulator. In order to track a desired motion, the following torques should be generated:

$$
\begin{align*}
U_1 & \quad U_2 \\
- U_1 & \quad - U_2
\end{align*}
$$

By defining

$$
k^2 = \max \left\{ 1, \frac{|\tau_1|}{U_1}, \frac{|\tau_2|}{U_2} \right\} \geq 1
$$

then:

- $$x = kt$$
- total time = $$kT \geq T$$

Then, the new torques are physically achievable, $$\tau(x) = \tau(t)/k^2$$, and at least one of them saturates in a point.

A **variable scaling** can be adopted to avoid slowing down the whole trajectory (saturation usually occurs in a single point).

For the optimal motion law (minimum time), at least one actuator saturates in each segment of the trajectory.
Trajectory Planning for Robot Manipulators
Analysis of Trajectories

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Introduction

**Vibrations** are undesired phenomena often present in automatic machines. They are basically due to the presence of *structural elasticity* in the mechanical system, and may be generated during the normal working cycle of the machine due to several reasons.

In particular, *vibrations may be produced if trajectories with a discontinuous acceleration profile are imposed to the actuation system.*

⇒ Acceleration discontinuities → sudden variation of the inertial forces applied to the system.

⇒ Relevant discontinuities of such forces, applied to an elastic system (i.e. any mechanical device), generate vibrational effects.

⇒ Since every mechanism is characterized by some elasticity, this type of phenomenon must **always** be considered in the design of a trajectory, that therefore should have a *smooth acceleration profile* or, more in general, a *limited bandwidth.*
Example

Let consider a 1-dof mechanical system (output: position $x(t)$):

Output $x(t)$ of the system when the input $y(t)$ is a step or a sinusoidal function: without (top) and with damping (bottom).
Models for Analysis of Vibrations

Analysis of the vibration effects $\Rightarrow$

Models that consider the elastic, inertial and dissipative properties of the elements of the mechanical system.

The complexity level of the model is usually chosen as a compromise between the desired precision and the computational burden.

The simplest criterion is to describe the mechanical devices, that are intrinsically distributed parameter systems, as lumped parameter systems, i.e. as rigid masses (without elasticity) and elastic elements (without mass).

Energy dissipative elements are introduced in order to consider frictional phenomena among moving parts.

The numerical values of the elements that describe inertia, elasticity and dissipative effects have to be determined by energetic considerations, i.e. trying to maintain the equivalence of the kinetic and elastic energy of the model with the energy of the corresponding parts of the mechanism under study.

The description of these phenomena can be either linear or nonlinear.
Some considerations

If \( x(t) \) and \( y(t) \) are the positions of the two masses M and A respectively, and \( z(t) = y(t) - x(t) \), the dynamics of the system is described by:

\[
m \ddot{x} + k_0(x - y) = 0
\]

from which

\[
\ddot{z} + \omega_0^2 z = \dot{y}, \quad \omega_0 = \sqrt{\frac{k_0}{m}}
\]

being \( \omega_0 \) the natural frequency of the mechanical system.

A model with viscous friction (coefficient \( b \)) on the mass M is described as

\[
m \ddot{x} + b \dot{x} + k_0(x - y) = 0
\]
Linear model with one degree of freedom

Output of the two models with a step input

Output with $b = 0$; $b \neq 0$
In the first case, due to oscillations the maximum value of $x$ is twice the value of $y$. Notice that this result does not depend on the stiffness $k_0$ of the mechanism:

- if $k_0$ increases, then the natural frequency $\omega_0$ increases as well, while the amplitude of $x$ remains constant;
- the difference $z(t)$ between the positions of M and A depends on $k_0$ (if $k_0$ increases, then, $z(t)$ decreases).
Linear model with one degree of freedom

Output of the two models with a sinusoidal input:

Sinusoidal input
($b = 0; \quad b \neq 0$)
Linear model with one degree of freedom

Although the acceleration oscillations have not a real influence on the position of the mass $M$, they generate a structural stress to the mechanical device.

This phenomenon may be characterized by an analysis of the frequency content of the acceleration signal given as input to the system: the frequency range of the acceleration signal should be compared with the Bode diagram of the mechanism, and in particular with its natural frequency $\omega_0$. 
Bode diagrams of the mechanical system \((b = 2, m = 1, k_0 = 100)\) and of two acceleration signals:

→ for the unitary impulse \((U(s) = 1)\), the Bode diagram (amplitude) is equal to 1 ∀ \(\omega\),
→ the frequency of the sinusoidal acceleration is \(f = 0.25 \text{ hz}, \ (\omega = 2\pi f)\).
The ‘trapezoidal’ and ‘double S’ trajectories are very common in industrial practice, and therefore it is of interest a comparison of their main features. The criteria for the comparison are:

a) actuator usage;

b) duration of the trajectory;

c) analysis of the frequency content (vibrations induced on the mech. structure).

The trapezoidal, double S and triangular (limit case of trapezoidal) trajectories are considered for the analysis.

The duration of the trajectory is $T$, and the acceleration of the ‘double S’ and of the trapezoidal trajectories are $\sqrt{3}$ times the acceleration of the triangular profile, while the velocity of the trapezoidal trajectory has been set in order to obtain the same duration $T$. 

![Graph showing comparison of trapezoidal and double S trajectories]
a - Use of actuators

From the data sheet, the following characteristics (among others) of an electric drive can be obtained:

- **Continuous torque** ($\tau_c$) (or **rated torque**): torque that the motor can produce continuously without exceeding thermal limits.
- **Peak torque** ($\tau_p$): maximum torque that the motor can generate for short periods.
- **Rated speed** ($\omega_n$): maximum value of the speed at rated torque (and at rated voltage).
- **Maximum power**: maximum amount of output power generated by the motor.
a - Use of actuators

If the motor is in the continuous operation region, it may work for an indefinite period of time, while in the intermittent region it may work only for a limited amount of time.

This limited period depends on the thermal dissipation properties of the motor and of the drive. On the other hand, different trajectories imply different utilizations of the motor, in particular with respect to the intermittent/continuous regions. As a matter of fact, the *double S trajectories allow to use the motor exploiting also the intermittent region*.

Notice that with double S trajectories the maximum torque and the maximum velocities are obtained in different time instants.

Theoretically, it is possible to enter the intermittent region also with the trapezoidal profile, but in this case the thermal power to be dissipated is higher than with double S trajectories.
a - Use of actuators

In comparing trajectories, there are some constitutive constraints for the motors that have to be considered:

- the requested torque cannot in any case exceed the peak torque;
- the RMS torque $\tau_{rms}$ of the trajectory must be not higher than the continuous torque $\tau_{cont}$:

$$\tau_{rms} = \sqrt{\frac{1}{T} \int_0^T \tau^2(t) \, dt} \leq \tau_{cont}$$

The first is a mechanical constraint, since the maximum torque that is possible to generate with an electric motor is limited. The second is related to the thermal energy dissipation capability of the inverter: the trajectory must avoid heating of the motor.

Let define $\tau_{ss}$ and $\tau_{tr}$ respectively as the maximum torque values of the double S and triangular trajectories.
a - Use of actuators

With the triangular profile, the acceleration (and then the torque) is constant. Therefore:

\[ \tau^2(t) = \tau_{tr}^2, \quad t \in [0, T] \]

then

\[ \tau_{rms, tr} = \tau_{tr} \]

Then, from the second constraint on the RMS torque, one gets

\[ \tau_{rms, tr} \leq \tau_{cont} \implies \tau_{tr} = \tau_{cont} \]

The maximum applicable torque is the continuous one.
With the double S trajectory (consider the case without linear velocity segments), the torque is a linear function of time and then $\tau_{ss}^2$ has a parabolic profile.

In this case

$$\tau_{rms,ss} = \frac{\tau_{ss}}{\sqrt{3}}$$

Therefore, the maximum torque achievable with this profile is

$$\tau_{rms,ss} \leq \tau_{cont} \implies \tau_{ss} = \sqrt{3} \tau_{cont}$$

This trajectory allows a better exploitation of the motor with respect to the triangular profile: the maximum torque that it is possible to generate is higher ($\tau_{ss} > \tau_{tr}$).
b - Duration of the trajectory

It is simple to show that the durations $T_{tr}$ and $T_{ss}$ of the triangular and double S (without linear velocity segments) trajectories are:

$$T_{tr} = 2\sqrt{\frac{L}{A_{max}}} \quad \quad \quad T_{ss} = 2\sqrt{2}\sqrt{\frac{L}{A_{max}}}$$

Therefore

$$T_{ss} = \sqrt{2} T_{tr}$$

As expected, being smoother, the duration of the double S trajectory is 1.41 times higher than the duration of the triangular one (with the same maximum acceleration $A_{max}$).
b - Duration of the trajectory

On the other hand, with the double S trajectory it is possible to apply higher acceleration (torque) values (better usage of the actuator):

\[ A_{\text{max, ss}} = \sqrt{3} \ A_{\text{max, tr}} = 1.7321 \ A_{\text{max, tr}} \]

As a consequence, the duration \( T_{\text{ss}} \) of the double trajectory is reduced, and therefore

\[ \frac{T_{\text{ss}}}{T_{\text{tr}}} = \sqrt{2} \sqrt{\frac{A_{\text{max, tr}}}{A_{\text{max, ss}}}} = \sqrt{2} \sqrt{\frac{1}{\sqrt{3}}} = 1.075 \]

Notice that the condition \( T_{\text{ss}} = T_{\text{tr}} \) is obtained with a torque \( \tau_{\text{ss}} = 2\tau_{\text{tr}} \) \((A_{\text{max, ss}} = 2A_{\text{max, tr}})\).

In conclusion, since the double S trajectory allows a better exploitation of the motor, higher values of torques (accelerations) can be generated and therefore the duration is comparable, with the additional positive feature of being the motion smoother!
In many motion control applications, when inertial loads have to be considered, the frequency range interested by the acceleration (torque) profile of the trajectory should be limited in order to avoid resonances or unmodeled dynamics of the mechanical structure.

This aspect, that should always be taken into consideration, is of particular relevance in case of mechanisms with structural elasticities or high inertia. It is then important to evaluate the frequency content of the torque signals in order to understand their influence on the mechanical structure.

With this respect, it is obvious that the smoother the profiles, the better the results are (e.g. the double S trajectory is better than the triangular one because the frequency range is narrower).
c - Frequency analysis

The figure, obtained with accelerations $A_{max,tr} : A_{max,ss} = 1 : \sqrt{3}$, it is possible to notice that the frequency content of the double S profile is lower (already in the second harmonic) than the triangular profile.

The frequency range interested by the double S profile is narrower, and then its effect on resonances and unmodeled dynamics of the mechanical system (if present) is reduced.

![Componenti armonici della acc. per profilo a doppia S](image1)

![Componenti armonici della acc. per profilo triangolare](image2)
c - Frequency analysis

Another example, considering a trapezoidal profile.

In case even double S trajectories are not smooth enough and oscillations are generated on the mechanical structure, smoother profiles should be adopted like, for example, trajectories with a trapezoidal jerk profile.

More in general, motion profiles with derivative up to a given order $n$ should be considered: with a trapezoidal (triangular) velocity the trajectory is a $C^1$ function (continuous first derivative, while the second derivative is discontinuous), a double S trajectory is a $C^2$ function, etc.

An alternative approach is to use Spline functions with a proper order.
Coordination of more motion axes

In many applications, many motion axes are present and need to be coordinated or synchronized. It is therefore necessary to take proper actions for this purpose, ranging from a simple synchronization of the start/stop instants to more complex operations.

Example: automatic machine for packaging medicines (pills)

- **450 motion axes:**
  - 150 electric drives (DC, brushless),
  - 300 step motors
- grouped in **40 blocks** to be synchronized and coordinated
- **specific packages for the single client** (how many pills, what time, ...)

![Image of an automatic machine for packaging medicines](image-url)
Coordination of more motion axes

Example: automatic machine for lifting TGV trains

- Trains up to 200 meters long
- Weight up to 386 tons
- Accuracy between the two extremities: 1 mm
- N. of lifting stations: 13
Coordination of more motion axes

In multi-axis machines based on mechanical cams, the synchronization of the different axes of motion is simply achieved by connecting the slaves to a single master (the coordination is performed at the mechanical level).

In case of electronic cams the problem must be considered in the design of the motion profiles for the different actuators (the synchronization is performed at the software level).

A common solution is to obtain the synchronization of the motors by defining a master motion, that can be either virtual (generated by software) or real (the position of an actuator of the machine), and then by using this master position as “time” (i.e. the variable $\theta(t)$) for the other axes.
An example is reported in the figure, where the variable $\tau$ is computed as a function of the angular position $\theta$ of the master. In the first two cycles ($\theta \in [0^0, 720^0]$) the motion is “slow” ($\tau = \theta$), while in the last one ($\theta \in [720^0, 1080^0]$) the motion is “fast” ($\tau = 2 \theta$).

Two slave axes are present: the first one generates a cycloidal profile from $q_{c0} = 0^0$ to $q_{c1} = 360^0$ (solid), while the second one generates a polynomial profile of degree 5 (dashed) interpolating the points $q_{p0} = 0^0, q_{p1} = 180^0, q_{p2} = 0^0$, in both cases for $\tau = [0^0, 360^0]$.

Note that in the last cycle the velocity values are doubled, while the accelerations are four times those present in the first cycles.
In defining the (constant) velocity $v_c$ of the master axis, i.e. the motion law $\theta(t)$, the most ‘stressed’ axis (in terms of velocity, acceleration, …) should be taken into consideration in order to define profiles that can be generated by each motor:

$$v_c = \min \left\{ \frac{v_{max_1}}{|\dot{q}_1(\theta)|_{max}}, \ldots, \frac{v_{max_n}}{|\dot{q}_n(\theta)|_{max}}, \sqrt{\frac{a_{max_1}}{|\ddot{q}_1(\theta)|_{max}}}, \ldots, \sqrt{\frac{a_{max_n}}{|\ddot{q}_n(\theta)|_{max}}}, \sqrt[3]{\frac{j_{max_1}}{|\vartheta_1(\theta)|_{max}}}, \ldots, \sqrt[3]{\frac{j_{max_n}}{|\vartheta_n(\theta)|_{max}}} \right\}$$

Synchronization of different axis of motion can also be defined analytically, as already briefly discussed for trapezoidal velocity or spline trajectories.
Trajectory Planning for Robot Manipulators
Trajectories in the Workspace

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If trajectories are defined in the workspace, it is necessary to use the inverse kinematic function to translate the motion specification to the joint space (where actuators operate). Since this increases the computational burden for trajectory planning, the operations of computing the trajectory and translating it to the joint space are made at a lower frequency with respect to the control frequency. Therefore, it is necessary to interpolate the data before assigning them to the low-level controllers: usually, a simple linear interpolation is adopted.

Typical values: \( \Delta t_n = 10 \) ms, \( \Delta t_k = 1 \) ms

\[ \implies \] 10 values of \( q(t_k) \) for each value of \( q(t_n) \) (\( x(t_n) \))

\[ \implies \] there is a delay \( \Delta t_n \) between the two sequences \( q(t_k) \) and \( q(t_n) \)
Another problem: the Cartesian positions actually achieved during the motion obtained by interpolating the points $q(t_n)$ are not those originally planned.
Workspace trajectories

For the computation of the workspace trajectories, it is possible to adopt one of the techniques used for the joint space (substituting the joint variable $q(t)$ with $x(t)$, i.e. a position or an orientation in the Cartesian space) or to define analytically the geometric path (e.g. an ellipse) as a function of time (i.e. $p = p(t)$) or, better, in a parametric form $\mathbf{p} = \mathbf{p}(s)$, being $s = s(t)$ a proper parameterization defining the motion law.
Example: planar 2 dof manipulator

Desired trajectory:

- total duration 3s,
- start in $p_i = [-1.0, 1.0]$,
- end in $p_f = [0.7, 1.2]$,
- composed by two linear segments with intermediate point $p_m = [1.1, 0.0]$ at $t_m = 2s$. 
Example: planar 2 dof manipulator

Consider the parametric form

\[
\begin{align*}
  x(t) &= x_0 + (x_1 - x_0) s(\tau) \\
  y(t) &= y_0 + (y_1 - y_0) s(\tau)
\end{align*}
\]

where \( s \in [0, 1] \), \( \tau = \frac{t - t_0}{t_1 - t_0} \). The motion law \( s(\tau) \) is defined so that desired position/velocity profiles are obtained, for example linear segments with parabolic blends (position in the workspace), or polynomial functions with proper degree. The kinematic model is

\[
\begin{align*}
  x &= l_1 C_1 + l_2 C_{12} \\
  y &= l_1 S_1 + l_2 S_{12}
\end{align*}
\]

while the inverse kinematic equations are

\[
\begin{align*}
  C_2 &= \frac{x^2 + y^2 - a_2^2 - a_3^2}{2a_2a_3} \\
  S_2 &= \sqrt{1 - C_2^2} \\
  \theta_2 &= \text{atan2} (S_2, C_2) \\
  \theta_1 &= \text{atan2} (y, x) - \text{atan2} (a_2 S_2, a_1 + a_2 C_2)
\end{align*}
\]
Example: planar 2 dof manipulator

Case of trapezoidal motion:

![Work-space trajectory](image)

![Joint space trajectory](image)
Example: planar 2 dof manipulator

Case of 5-th order polynomial function:
To plan a trajectory in the workspace, usually the geometric path $p$ (line, circle, ellipse, ...) is defined as a function of a parameter $s(t)$: $p = p(s)$.

The parameter $s = s(t)$ is computed by using one of the techniques discussed for joint space trajectories. A classical approach is to plan $s(t)$ as a linear function with parabolic blends, in order to have in the work space acceleration/deceleration tracts (low stress for the mechanical and actuation system).

Notice that for parameterized trajectories the following conditions hold:

$$\dot{p} = \frac{d}{ds} p \dot{s}, \quad \ddot{p} = \frac{d}{ds} \ddot{s} + \frac{d^2}{ds^2} p \dot{s}^2$$
**Curvature of a geometric path**

Consider a path $\Gamma$ in the workspace $\mathbb{R}^3$, expressed in parametric form

$$p = p(r) = \begin{bmatrix} x(r) \\ y(r) \\ z(r) \end{bmatrix}, \quad r \in [r_a, r_b]$$

Assume that the curve is *regular*, i.e.

$$\dot{p} = \frac{d}{dr} p \neq 0, \quad \forall r \in [r_a, r_b]$$

Given a point $p_a$ of $\Gamma$, and a motion direction on the path, the *arc length* of a generic point $p(r)$ is defined as

$$s = \int_{p_a}^{p(r)} \|\dot{p}(\rho)\| d\rho$$
By definition, the arc length represents the length of the arc of $\Gamma$ defined by the two points $p$ and $p_a$ (if $p$ follows $p_a$, or the opposite of such a length if $p$ is before $p_a$). The value $s = 0$ is assigned to point $p_a$. A bijective relationship exists between the values of the arc length $s$ and the points of the path $\Gamma$, and then it is possible to use the arc length for a parametric expression of $\Gamma$.

$$p = p(s)$$

It is possible to assign to each point $p$ of $\Gamma$ a reference frame (Frenet frame) defined by the following unit vectors

$$\begin{align*}
  t &= \frac{\dot{p}}{||\dot{p}||} \quad \text{tangent unit vector} \\
  b &= \frac{\dot{p} \times \ddot{p}}{||\dot{p} \times \ddot{p}||} \quad \text{binormal unit vector} \\
  n &= b \times t \quad \text{normal unit vector}
\end{align*}$$
Position trajectories

- The unit vector \( t \) lies along the direction tangent to \( \Gamma \) in \( p \), and is directed along the positive \( s \) direction.
- The unit vector \( n \) defines, with \( t \), the osculating plane \( O \), defined as the plane containing point \( p \) and a point \( p' \in \Gamma \) when \( p' \to p \).
- The unit vector \( b \) (binormal) is defined so that the frame \((t, n, b)\) is right-handed. Notice that it is not always possible to define uniquely the Frenet frame.
Position trajectories

**Segment of a line**
The linear geometric path between points $p_i$ and $p_f$ has a parametric representation expressed by

$$p(s) = p_i + \frac{s}{||p_f - p_i||}(p_f - p_i), \quad s \in [0, ||p_f - p_i||]$$

Moreover, by deriving $p$ with respect to $s$, one obtains

$$\frac{d\, p}{ds} = \frac{p_f - p_i}{||p_f - p_i||}, \quad \frac{d^2\, p}{ds^2} = 0$$

It is possible to plan a trajectory through a sequence of points with the same modalities seen in the joint space. If it is required to pass exactly through the intermediate points, then it is possible to compute the parameter $s$ using one of the motion laws defined in the joint space (e.g. cubic, trapezoidal, ...). In case it is not required for the manipulator to pass through the intermediate points, the geometric path can be defined for example by linear segments with polynomial blends (position error, but non null velocity in the via points).
A typical profile is shown below. The variable $x$ is defined with a sequence of points interpolated with linear segments, while the real trajectory only approximates (in the vicinity of the via points) the given path.
Arc of a circle
A parametric representation of an arc of a circle is

\[
p = \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} r \cos(\theta) \\ r \sin(\theta) \\ 0 \end{bmatrix}, \quad \theta \in [\theta_{\text{min}}, \theta_{\text{max}}]
\]

where the parameter is the angle \( \theta = \theta(t) \). Notice that if the path must be arbitrarily positioned/oriented in the 3D space, it is sufficient to multiply the (homogeneous) vector \( p \) by a proper transformation matrix \( T \).

A motion law with acceleration/deceleration tracts (in the operational space) is obtained if the parameter (in this case: \( \theta \)) is computed, for example, with a trapezoidal velocity profile.
Position trajectories - Example

Planar 2 dof manipulator with links' length: $a_1 = a_2 = 1$. The desired circular motion is defined by

- Center: $[0.5, 0.5]$, radius $r = 0.7$
- Initial angle: $\theta_i = -60$, final angle: $\theta_f = 90$, time duration: $T = 1$ s
Position trajectories

Trajectory Planning
Scaling trajectories
Analysis of Trajectories
Trajectories in the Workspace

Position trajectories

Traiettoria cartesiana x(t), y(t)

Traiettorie di posizione giunto
Multi-dimensional Trajectories

In many cases, the definition of a trajectory in the task space of a robot or of a multi-axis automatic machine requires also to assign the orientation of the tool in each point of the curve. This can be achieved by specifying the configuration of the frame linked to the end effector (the tool frame) with respect to the base (world) frame. Therefore, in the general case the parametric description of the trajectory is a six dimensional function providing, for each value of the variable $s$, both the position and the orientation of the tool:

$$p = [x, y, z, \alpha, \beta, \gamma]$$

If the multi-dimensional problem can be decomposed in its components, the 3D trajectory planning can be considered as a set of scalar problems, and the techniques reported previously can be adopted for its solution. In this case, each function $p_i(\cdot)$ depends directly on the time $t$, and the synchronization among the different components is performed by imposing interpolation conditions at the same time instants.
In other cases, the trajectories for the position and for the orientation can be defined separately, for example, when it is desirable to track a path in the workspace, e.g. a straight line, with the orientation of the tool specified only at the endpoints of the motion. In fact, with the exception of some applications (e.g. welding, painting, etc.), not always a strict relation between position and orientation exists.

The two problems can then be treated separately, and are conceptually similar, i.e. given a set of via-points $q_k = [x_k, y_k, z_k]^T$ (position) or $q_k = [\alpha_k, \beta_k, \gamma_k]^T$ (orientation) it is necessary to find a parametric curve which passes through or near them.
Rotational trajectories

Example of trajectory with position and orientation displacements. The orientation in each via-point is provided by means of Roll-Pitch-Yaw angles with respect to the world frame.

\[
\begin{bmatrix}
q_x \\
q_y \\
q_z \\
\psi \\
\theta \\
\varphi
\end{bmatrix} =
\begin{bmatrix}
3.31 & -3.01 & -1.07 & 4.48 & 1.52 \\
-2.38 & -3.53 & 5.81 & 2.97 & -1.25 \\
7.14 & 10.89 & 6.72 & 4.54 & 5.81 \\
0 & 0 & -1.11 & 2.11 & 2.46 \\
0 & 0.90 & 0.42 & -0.33 & -0.77 \\
0 & 0 & -0.69 & 0 & -0.87
\end{bmatrix}
\]
Rotational trajectories

Planning trajectories in terms of changes in orientation is somehow more complex than planning in position only. While it is quite simple to plan a motion between points $p_i$ and $p_f$, the same is not true for interpolating the orientation between two rotational matrices $R_i$ and $R_f$: for example if the elements $r_{ij}$ are changed linearly from the initial (in $R_i$) to the final (in $R_f$) value, there is not guarantee that the intermediate matrices are real rotation matrices (orthogonal columns with unit norm).

Usually, the Euler or RPY angles are employed or, alternatively, the angle/axis representation.

With the Euler or RPY angles, two triples $\phi_i, \phi_f$ are defined, and an interpolation based on one of the presented techniques can be adopted (advisable in any case continuity at least of the in rotational velocity).
Rotational trajectories

With the angle/axis representation, if $\mathbf{R}_i$ and $\mathbf{R}_f$ are the initial and final rotation matrices, then a matrix $\mathbf{R}_{i,f}$ exists such that

$$\mathbf{R}_i \mathbf{R}_{i,f} = \mathbf{R}_f$$

or

$$\mathbf{R}_{i,f} = \mathbf{R}_i^T \mathbf{R}_f = \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix}$$

Then, the unit vector $\mathbf{w}$ and the rotational angle $\theta$ are

$$\theta_r = \arccos \frac{r_{11} + r_{22} + r_{33} - 1}{2}$$

(11)

$$\mathbf{w} = \frac{1}{2 \sin \theta_r} \begin{bmatrix} r_{32} - r_{23} \\ r_{13} - r_{31} \\ r_{21} - r_{12} \end{bmatrix}$$

(12)
It is now necessary to define a matrix $R_t(t)$ so that $R_t(0) = I$ and $R_t(t_f) = R_{i,f}$. A choice can be

$$R = \begin{bmatrix}
   w_x^2 (1 - C_{\theta}) + C_{\theta} & w_x w_y (1 - C_{\theta}) - w_z S_{\theta} & w_x w_z (1 - C_{\theta}) + w_y S_{\theta} \\
   w_x w_y (1 - C_{\theta}) + w_z S_{\theta} & w_y^2 (1 - C_{\theta}) + C_{\theta} & w_y w_z (1 - C_{\theta}) - w_x S_{\theta} \\
   w_x w_z (1 - C_{\theta}) - w_y S_{\theta} & w_y w_z (1 - C_{\theta}) + w_x S_{\theta} & w_z^2 (1 - C_{\theta}) + C_{\theta}
\end{bmatrix}$$

where $\theta(t)$ is computed according to one of the previous motion law (cubic, trapezoidal, ...) from $\theta(0) = 0$ to $\theta(t_f) = \theta_r$, while $w$ is defined as in (12).

The following rotation matrix is then obtained

$$R(t) = R_i R_t(\theta(t))$$
When *positioning and orientation problems are coupled* in work space, the orientation of the end effector must be specified on the basis of the orientation of the path at a given point.

As a matter of fact, if the parametric form of the (regular) curve to be tracked is expressed in terms of the curvilinear coordinate $s$ (which measures the arc length)

$$
\Gamma : \quad p = p(s), \quad s \in [0, L]
$$

it is possible to define a coordinate frame directly tied to the curve, *the Frenet frame*, represented by three unit vectors $n, t, b$.

In those applications in which the tool must have a fixed orientation with respect to the motion direction, e.g. in arc welding, the Frenet vectors implicitly define such an orientation. It is therefore sufficient to define the position trajectory function to obtain in each point the orientation of the tool.
A helicoidal trajectory is shown with the associated Frenet frames. The trajectory is described by the parametric form

\[ p = [r \cos(u), \ r \sin(u), \ d \ u]^T \]

with \( u \in [0, 4\pi] \), which leads to the frame

\[
R_F = [e_t, e_n, e_b] = \begin{bmatrix}
-c \sin(u) & -\cos(u) & l \sin(u) \\
c \cos(u) & -\sin(u) & -l \cos(u) \\
l & 0 & c
\end{bmatrix},
\]

\[
c = \frac{r}{\sqrt{r^2 + d^2}} \quad \text{and} \quad l = \frac{d}{\sqrt{r^2 + d^2}}.
\]
Final considerations

Some techniques for planning trajectories in the joint and in the work space have been illustrated.

If the trajectory is planned in the work space, the end-effector moves along well defined paths, a very important aspect in many industrial applications.

On the other hand, the computational burden is higher in case of work-space trajectories. For this reason, the frequency at which the trajectory in computed in lower than the control frequency, and an interpolation is then necessary.

Moreover, since the velocity/acceleration/torque limits required in the work-space may result non physically achievable in the joint space (i.e. in the actuation space) a re-computation of the trajectory might be necessary.
Final considerations

Finally, singular configurations may generate problems if the trajectory is planned in the work space.

As a matter of fact, if a motion defined in the work space reaches points close to singular configuration, it should be avoided. Therefore, the trajectory should be checked in advance and, in case, not actuated or modified.

Clearly all these problems are not present if the trajectory is planned in the joint space.