

Trajectory construction

Robotics and Computer vision (RoVi)

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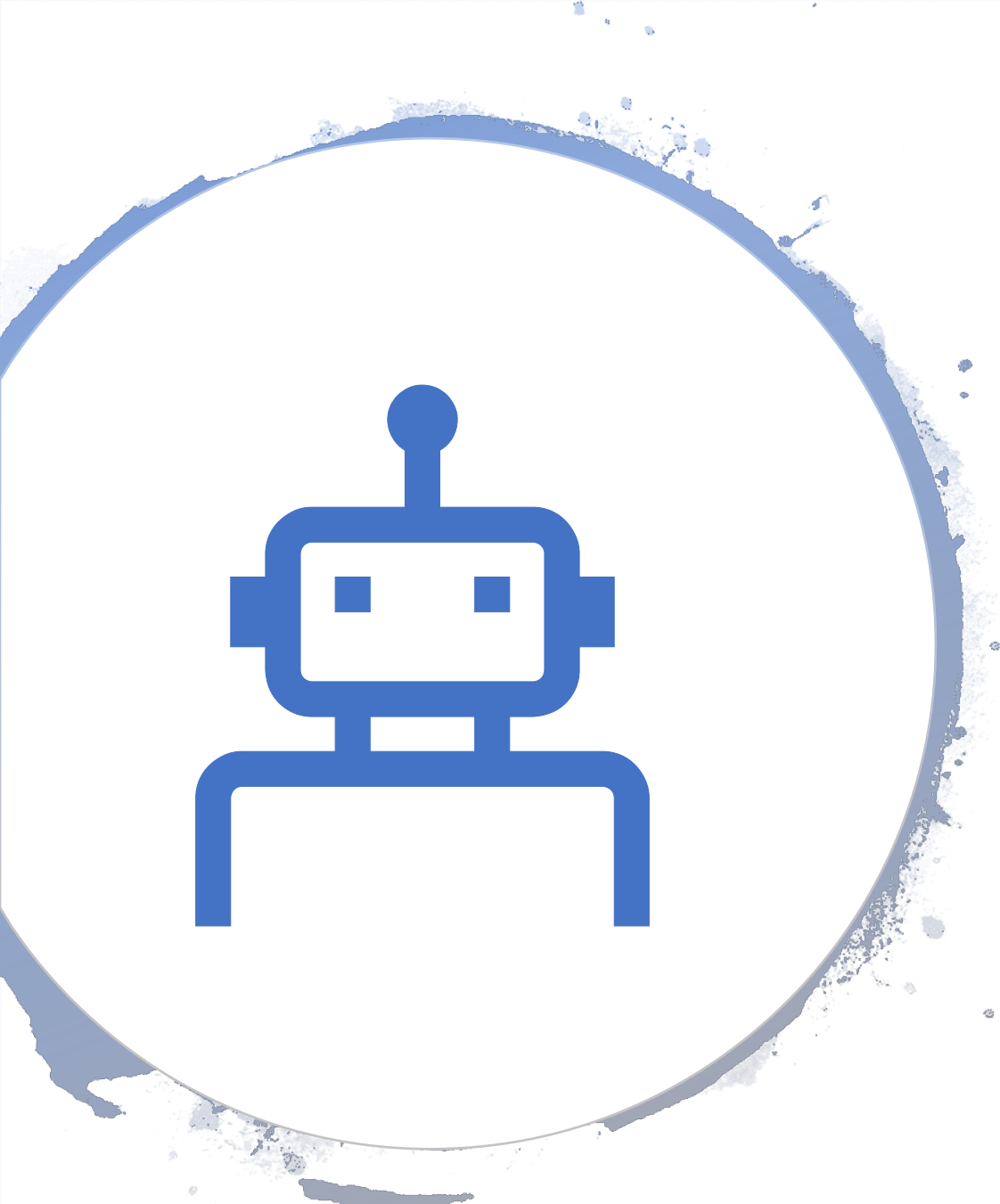
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Based on M. Mihelj, T. Bajd. et al. Robotics – second edition, Springer, 2018.

Siciliano, Bruno, Lorenzo Sciavicco, Luigi Villani, and Giuseppe Oriolo. Robotics: modelling, planning and control. Springer Science & Business Media, 2010.

A Geometrical Introduction to Robotics and Manipulation, Richard Murray and Zexiang Li and Shankar S. Sastry 2011, CRC Press, Hongkong



Agenda


- Introduction to trajectory's
- general point to point movement,
 - linear interpolation between two frames,
 - quaternion interpolation,
 - parabolic blends,
 - example.

Trajectory's

General we distinguish...



Joint
trajectories



Tool
trajectories

Differences

- Joint trajectories are presented as:
 - $Trj_{joint} = \{q_i^1, q_i^2, \dots, q_i^n, t_i\}_{i=1}^{T,n}$, t_i , $i = 1, \dots, T_i$ are the number of samples and n number of joints.
 - can directly be send to the robot.
 - precise deterministic execution – (better tracking),
 - the movement is not necessary linear although it was constructed such.
- Tool trajectories are presented as:
 - $Trj_{joint} = \{p_i, o_i, t_i\}_{i=1}^T$ where $p = (p_x, p_y, p_z)$ is the positional part and $o = \mathbf{R}$ orientational part of the trajectory.
 - Invers kinematics is needed in order to execute the trajectory.

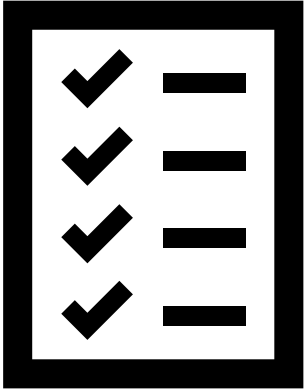
Trajectory construction

- Trajectories can be constructed in different ways:
 - trajectories can be acquired with demonstration,
 - simple construction with point to point interpolation,
 - splines and polynomials,
 - using dynamic systems,
 - etc.

Point-to-point movement

Exercise

Exercise



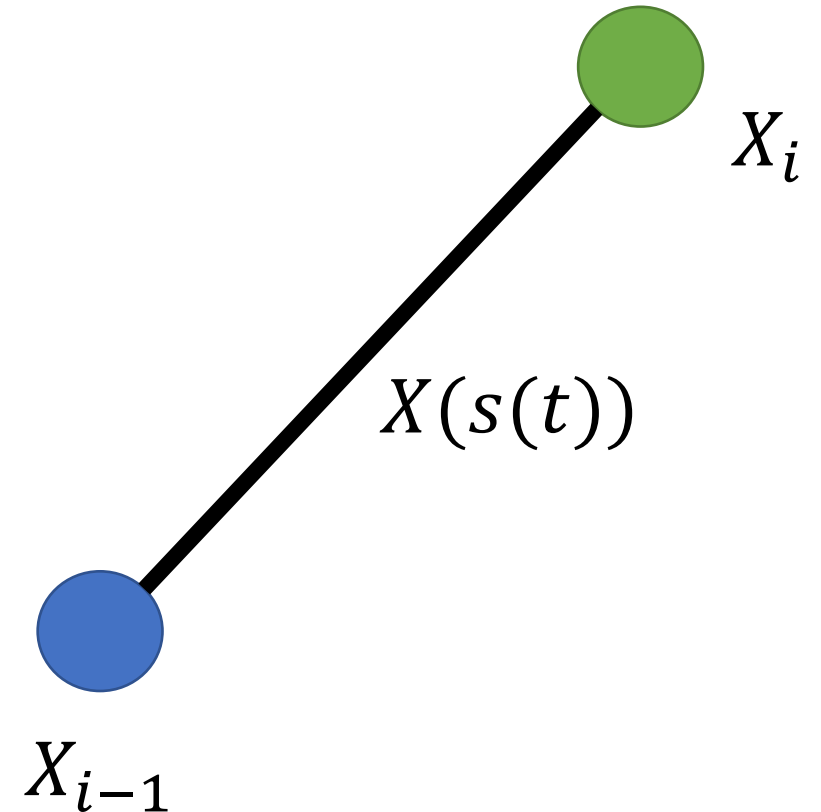
In groups discuss which are the necessary poses in 3D space, to satisfy a **pick and place** operation.

Linear interpolation

- Let's consider two vectors X_{i-1} and X_i . A linear interpolation between them is given as:

$$X(s(t)) = X_{i-1} + (X_i - X_{i-1})s(t) \quad 0 \leq s \leq 1$$

- where $s(t)$ denotes a non decreasing continuous time profile satisfying:
 - $s(t_{i-1}) = 0$ at start,
 - $s(t_i) = 1$ at the end of the movement.



Linear interpolation

- For velocity we will consider the following:

$$s(t) = \frac{t - t_{i-1}}{t_i - t_{i-1}} \text{ (constant velocity)}$$

$$s(t) = \left(\frac{t - t_{i-1}}{t_i - t_{i-1}} \right)^2 \text{ (linear ramp-up of velocity)}$$

$$s(t) = 1 - \left(\frac{t_i - t}{t_i - t_{i-1}} \right)^2 \text{ (linear ramp-down of velocity)}$$

Linear interpolation

- In tool frame we can write the definition of the interpolated path with constant velocity as such:

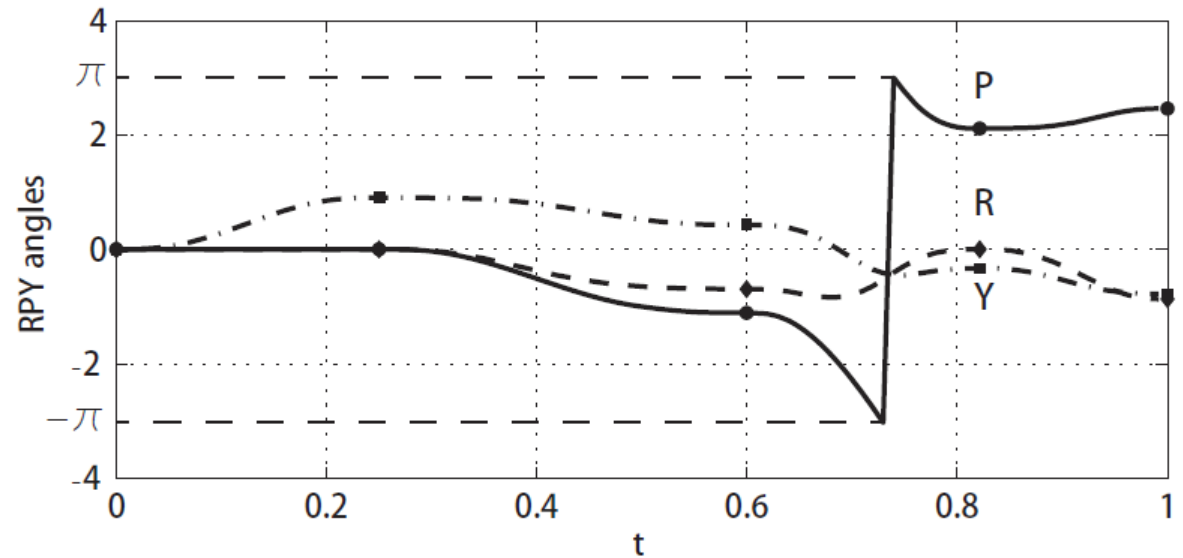
$$\mathbf{p}(t) = \mathbf{p}^{i-1} + \frac{t - t_{i-1}}{t_i - t_{i-1}} (\mathbf{p}^i - \mathbf{p}^{i-1})$$

$$\mathbf{R}(t) = \mathbf{R}_{\text{eaa}} \left(\frac{t - t_{i-1}}{t_i - t_{i-1}} \mathbf{W}_{\text{rot}} (\mathbf{R}^i [\mathbf{R}^{i-1}]^T) \right) \mathbf{R}^{i-1}$$

- where $p(t)$ represents the position part of the tool trajectory and $R(t)$ the orientation part of the trajectory.

Linear interpolation of rotations

- Interpolation using Euler angles, gives problems:
 - Parametrization singularity, in other words RPY angles defined on the interval $[-\pi, \pi]^3$ encounter a parametrization singularity.
 - **Derivatives** of the Euler angles have no physical meaning!



Linear interpolation of rotations

- A more meaningful approach:
 - Choose physically meaningful coordinates,
 - Add via- points to avoid parametrization singularity,
 - Generate trajectory and use inverse kinematics to obtain joint trajectory and then execute on the robot.
- For orientations specifically use a different representation:
 - equivalent axis angles,
 - rotation matrixes,
 - or quaternions.

Linear interpolation of rotations EEA

- With EEA we can interpolate relative changes in orientation.
- Axis-angle interpolation how-to:
 - Remember Euler's theorem:
 - All orientations \mathbf{A} , \mathbf{B} are related by a rotation \mathbf{R} around an axis \mathbf{v} by an angle γ : $\mathbf{B} = \mathbf{AR}$.
 - Compute \mathbf{R} as $\mathbf{A}^{-1}\mathbf{B}$
 - Extract EEA representation $r = \mathbf{W}_{rot}(\mathbf{R})$ from \mathbf{R} (see lecture 3)
 - Interpolate linearly between zero and \mathbf{r} to yield $\mathbf{r}(t)$, i.e., $\mathbf{r}(t) = t\mathbf{r}$
 - Convert $\mathbf{r}(t)$ back to matrix $\mathbf{R}(t)$
 - Get final orientation by computing $\mathbf{AR}(t)$

Linear interpolation of rotations

- We cannot identify orientations/rotations with 3D points and have all of the nice features such as:
 - Nice interpolation
 - Smoothness (no gimbal lock)
 - Simple composition of rotations
 - No complicated multiple-valuedness
 - However we can, interpolate **relative** changes in orientation using the axis-angle representation.
- **A better approximation can be achieved with Quaternions**, a 4D construction that does exactly what we want, both absolutely and relatively.

Quaternion interpolation

Linear interpolation

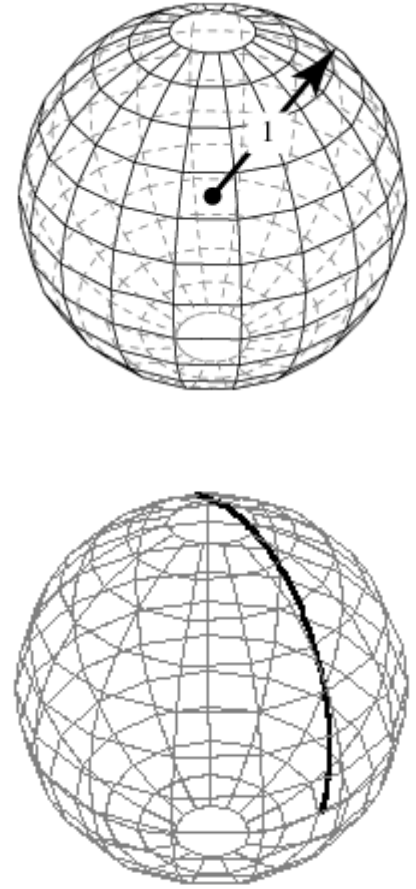
- Interpolation is used to determine intermediate points between two given points Q_1 and Q_2
- Linear interpolation can be used to do interpolation in normal space
- The standard linear interpolation formula is

$$Q^i = Q_1 + t(Q_2 - Q_1)$$

- The general steps to apply this equation are:
 - Compute the difference between Q_1 and Q_2
 - Take the fractional part of that difference
 - Adjust the original value by the fractional difference between the two points

Unit sphere

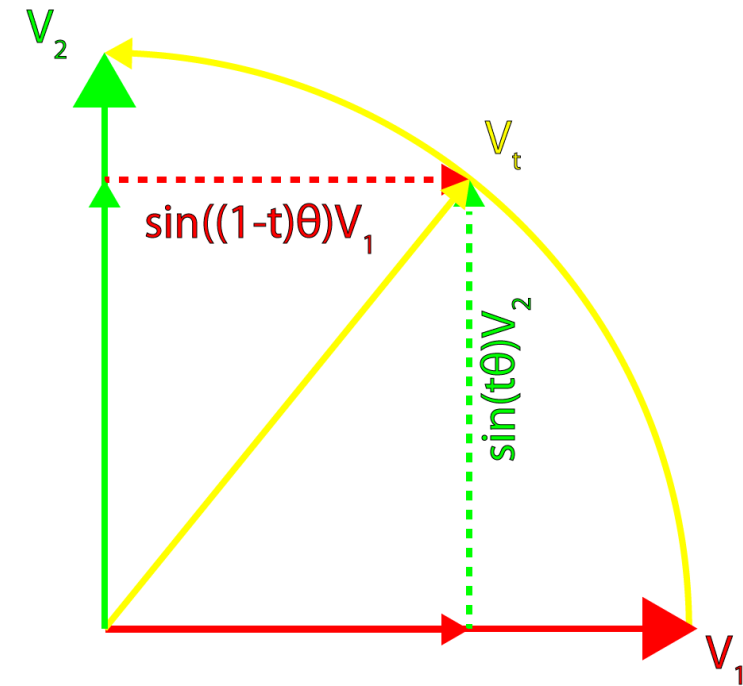
- A quaternion is a point on the 4-D unit sphere
- If we want to interpolate between two points on a sphere, we do not just want to Lerp between them
- Instead, we will travel across the surface of the sphere by following a great arc
- Move with constant angular velocity along the great circle between the two points
- There are two SLERP choices on a sphere – we pick the shortest



Spherical interpolation of vectors

- We will apply a spherical interpolation of vectors to quaternions
- The general form of a spherical interpolation for vectors is defined as:

$$\mathbf{v}_t = \frac{\sin(1-t)\gamma}{\sin \gamma} \mathbf{v}_1 + \frac{\sin t\gamma}{\sin \gamma} \mathbf{v}_2$$



Spherical interpolation of quaternions

- The formula for spherical interpolation of vectors can be applied unmodified to quaternions:

$$Q_t = \frac{\sin(1-t)\gamma}{\sin \gamma} Q_2 + \frac{\sin t\gamma}{\sin \gamma} Q$$

- Angle γ need to be calculated

Spherical interpolation - angle θ

- We can obtain the angle γ by computing the dot-product between q_1 and q_2

$$\cos \gamma = \frac{Q_1 \cdot Q_2}{|Q_1||Q_2|} = \frac{w_1 w_2 + x_1 x_2 + y_1 y_2 + z_1 z_2}{|Q_1||Q_2|}$$

$$\gamma = \cos^{-1} \left(\frac{w_1 w_2 + x_1 x_2 + y_1 y_2 + z_1 z_2}{|Q_1||Q_2|} \right)$$

Spherical interpolation - considerations

- The quaternion dot-product can result in a negative value
-> the resulting interpolation will travel the long-way around the 4D sphere
- If the result of the dot product is negative, then we can negate one of the orientations
- Negating the scalar and the vector part of the quaternion does not change the orientation that it represents

Spherical interpolation - considerations

- If the angular difference γ between Q_1 and Q_2 is very small then $\sin \gamma$ becomes 0 -> undefined result when we divide by $\sin \gamma$.
- In this case, we can fall-back to using linear interpolation between Q_1 and Q_2 .
- As changes are small a straight line is close to the arc on the sphere

Quaternion difference

- quaternion logarithm which maps $S^3 \rightarrow \mathbb{R}^3$ is defined as:

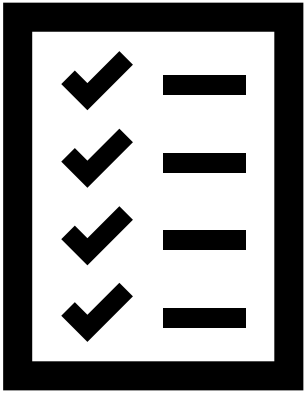
$$\mathbf{r} = \log(Q) = \log(v + \mathbf{u}) = \begin{cases} \arccos(v) \frac{\mathbf{u}}{\|\mathbf{u}\|}, \mathbf{u} \neq 0 \\ [0,0,0]^T, \text{ otherwise} \end{cases}$$

- which can be interpreted as a difference vector $\log(\mathbf{Q}_2 * \overline{\mathbf{Q}_1})$ between 2 unit quaternions.
- its inverse, the exponential map, which maps from $\mathbb{R}^3 \rightarrow S^3$ is defined:

$$\exp(\mathbf{r}) = \begin{cases} \cos(\|\mathbf{r}\|) + \sin(\|\mathbf{r}\|) \frac{\mathbf{r}}{\|\mathbf{r}\|}, \mathbf{r} \neq 0 \\ 0, \text{ otherwise} \end{cases}$$

Exercise

Python example 1



Calculate a slerp (in python use the quaternion interp. function) curve in quaternion space between:

$Q_1 = [0.7071, 0.7071, 0.0000, 0.0000]$

$Q_2 = [0.5000, -0.0000, 0.8660, -0.0000]$

Interpolation coefficient = 100.

plot the results as interpolated curves on a sphere with the provided function.

Linear interpolation

Linear interpolation

- We can also define a linear interpolator in joint space:

$$q^i(t) = q_o^i + \frac{t - t_0}{t_f - t_0} (q_f^i - q_o^i), i = 1, \dots, \text{numDOF}$$

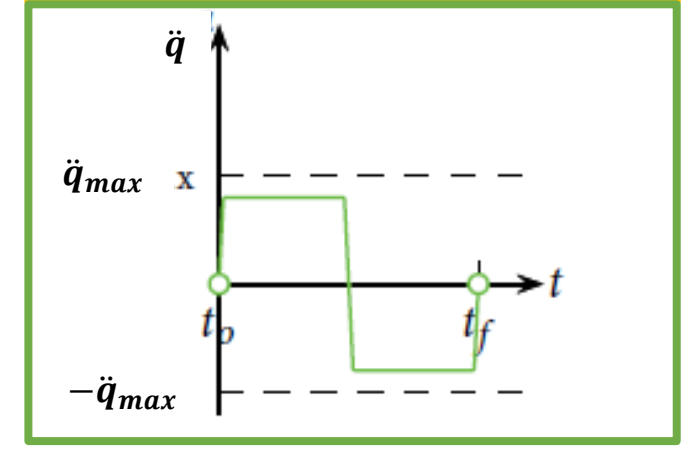
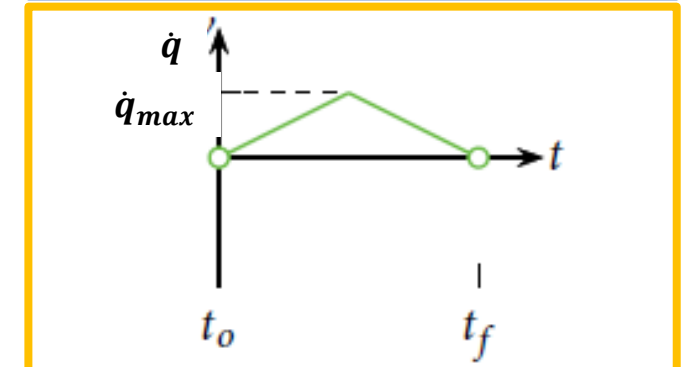
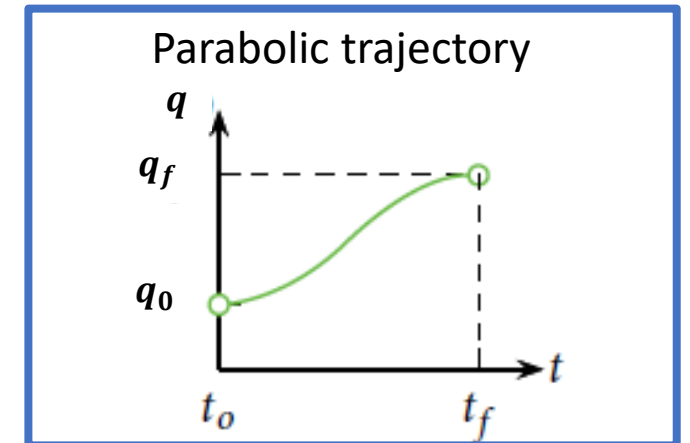
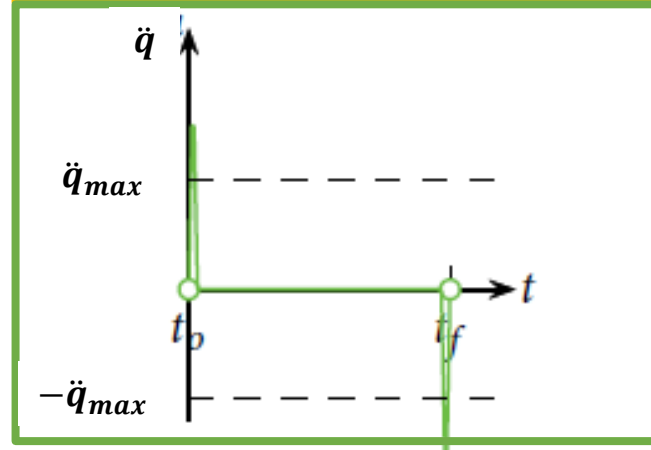
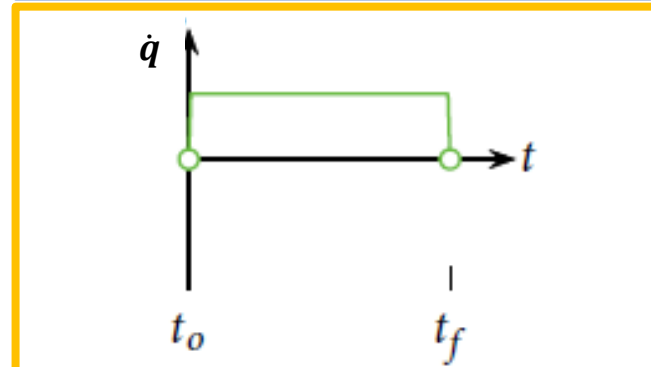
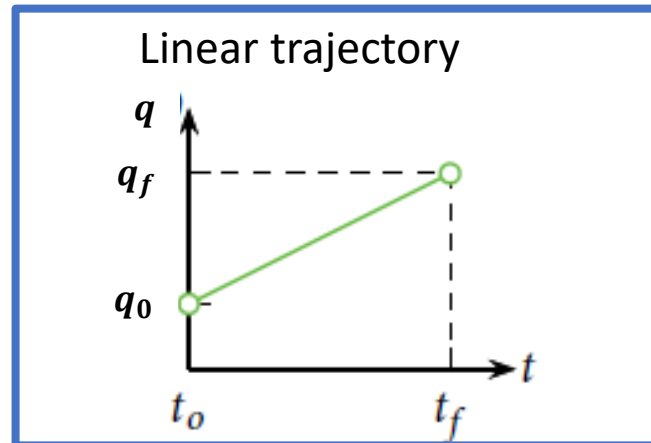
- It is given by a first order polynomial, for better results higher order polynomials work better:
 - the problem of velocity – discontinuity and unbounded acceleration in both ends.
- the outcome, compared with tool space interpolation is not equal!!!!
- therefore, we have to carefully consider which type of interpolation is appropriate for the desired task/application.

Trajectories

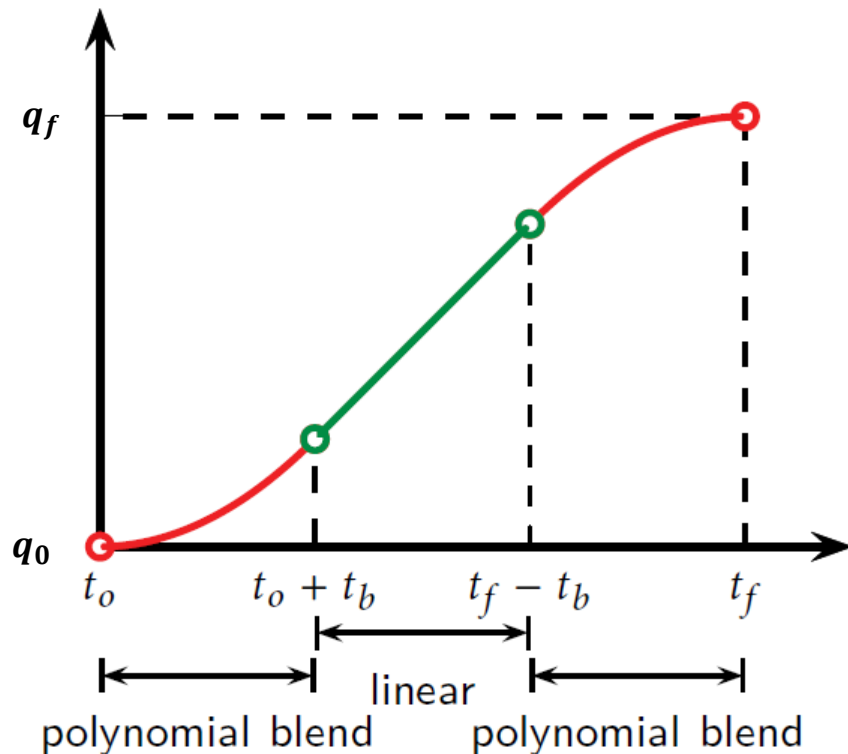
POSITION

VELOCITY

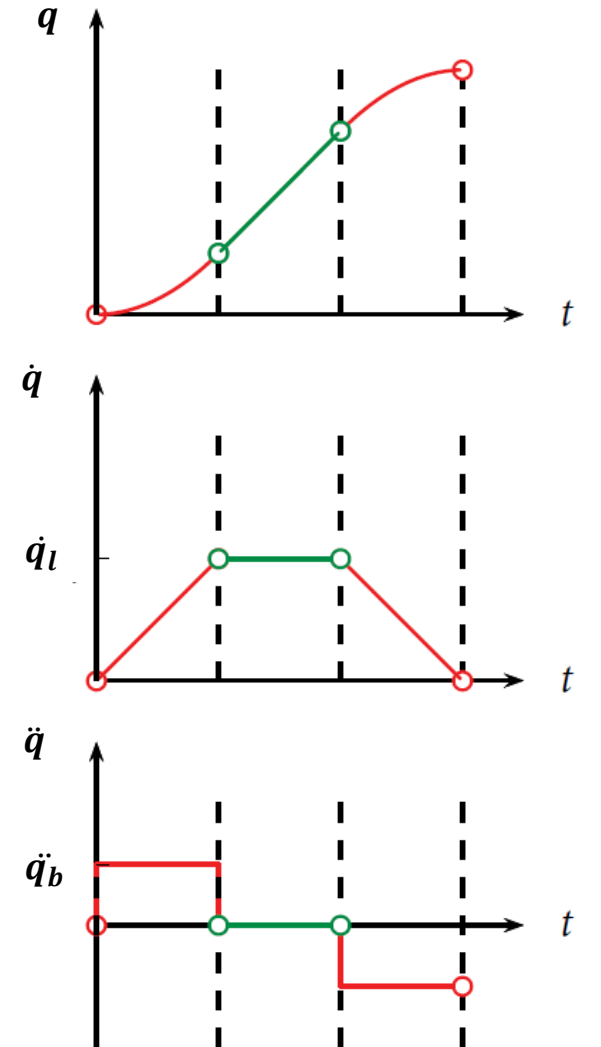
ACCELERATION



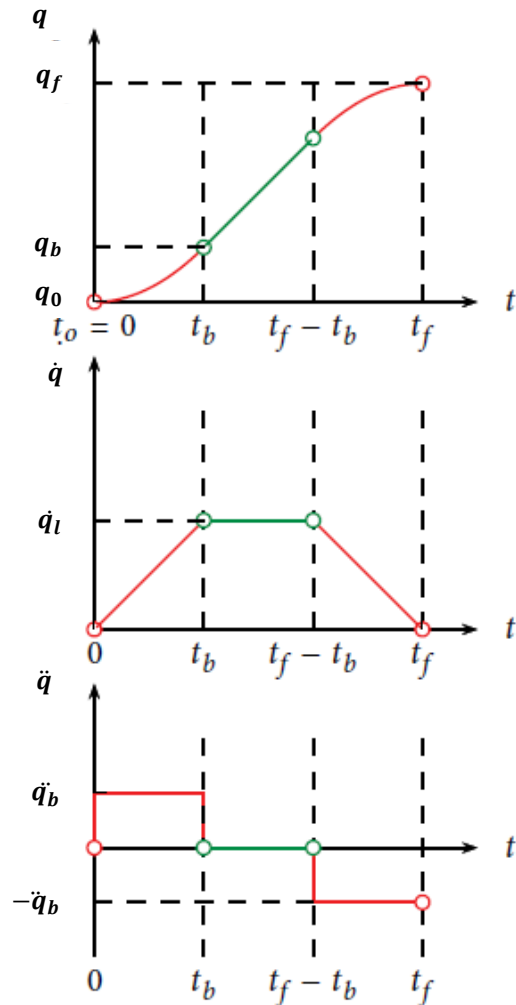
Let's first look at joint trajectories



- Parabolic blend
 - composite trajectory of a linear and a parabolic function.



Linear function with parabolic blend



- Input parameters:

$$q(t_o) = q(0) = q_o$$

$$q(t_f) = q_f$$

$$t_d = t_f - t_o: \text{Duration of travel}$$

- Control parameters:

$$\dot{q}_l (\leq \dot{q}_{\max}) \quad : \text{linear velocity}$$

$$q_b (\leq \ddot{q}_{\max}) \quad : \text{blend acceleration}$$

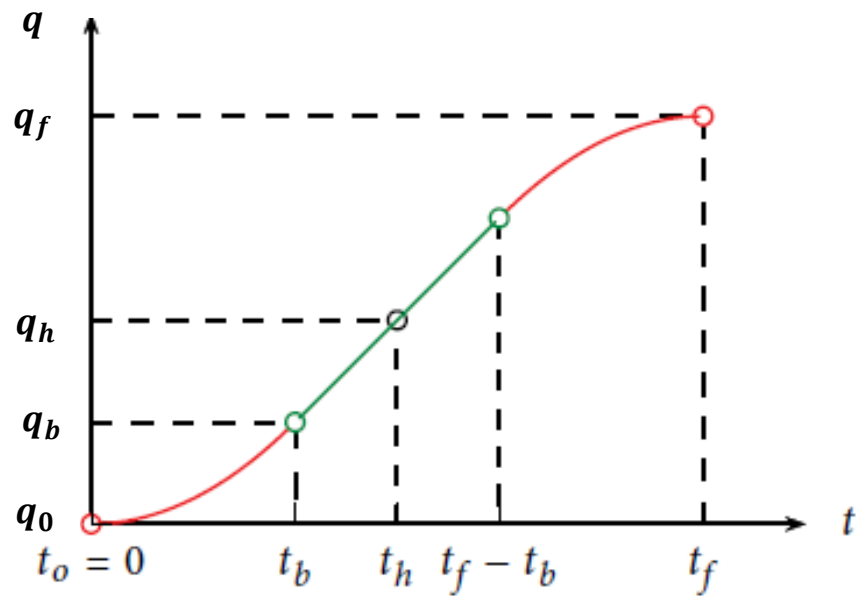
$$t_b \left(0 < t_b \leq \frac{t_d}{2} \right) \quad : \text{blend time}$$

$$t_d - 2t_b \quad : \text{linear time}$$

- Trajectory

$$q(t) = \begin{cases} q_o + \frac{1}{2} \ddot{q}_b (t - t_o)^2 & t_o \leq t < t_o + t_b \\ q_o + \dot{q}_b t_b \left(t - t_o - \frac{t_b}{2} \right) & t_o + t_b \leq t < t_f - t_b \\ q_f - \frac{1}{2} \ddot{q}_b (t_f - t)^2 & t_f - t_b \leq t \leq t_f \end{cases}$$

Derivation:



- Velocity match condition

- Blend region

- Constraints on \ddot{q}_b :

$$\ddot{q}_b t_b = \frac{q_h - q_b}{t_h - t_b}, \left(t_h \triangleq \frac{t_d}{2}, q_h \triangleq q(t_h) \right)$$

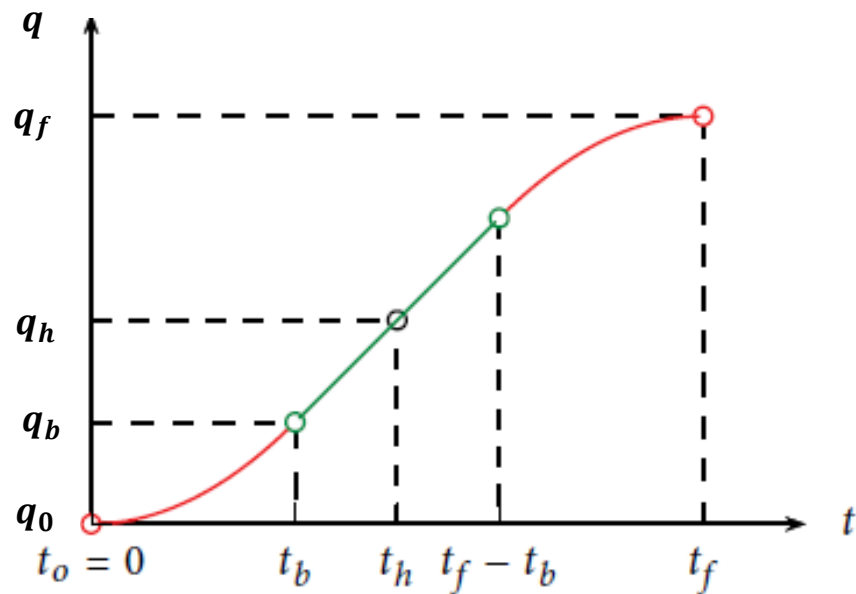
$$q_b \triangleq q(t_b) = q_o + \frac{1}{2} \ddot{q}_b t_b^2$$

$$\Rightarrow \ddot{q}_b t_b^2 - \ddot{q}_b t_d t_b + (q_f - q_o) = 0$$

$$\Rightarrow t_b = \frac{t_d}{2} - \frac{\sqrt{\ddot{q}_b^2 t_d^2 - 4 \ddot{q}_b (q_f - q_o)}}{2 \ddot{q}_b}$$

$$\ddot{q}_b \geq \frac{4(q_f - q_o)}{t_d^2}$$

Derivation:

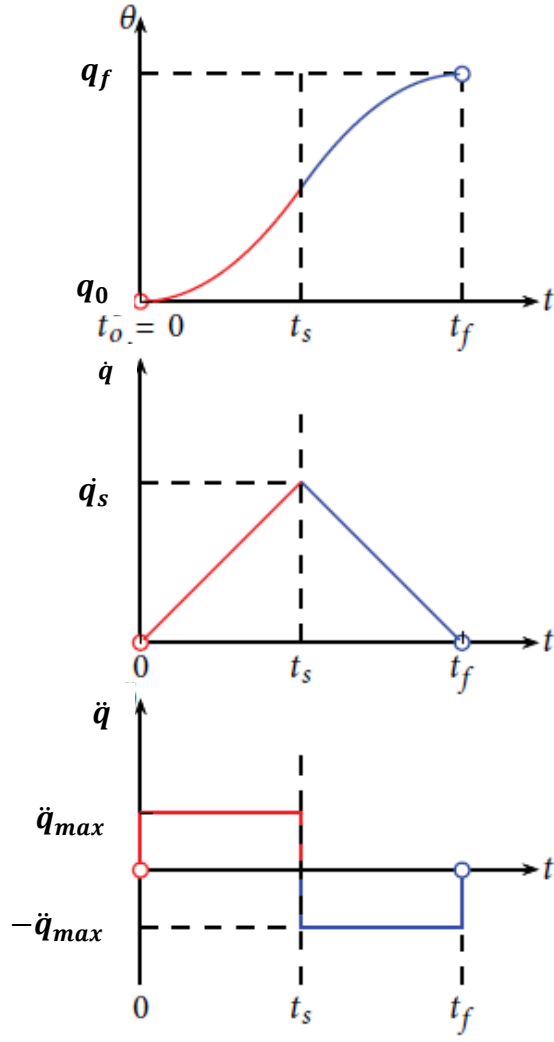


Observation:

$$\ddot{q}_b \geq \frac{4(q_f - q_o)}{t_d^2}$$

- As \ddot{q}_b increases, t_b decreases and linear time $t_b - 2t_b$ increases, $\ddot{q}_b \rightarrow \infty$
- if the above equation expresses equality the, the linear part of the parabolic blend goes to zero.

Minimum time trajectory



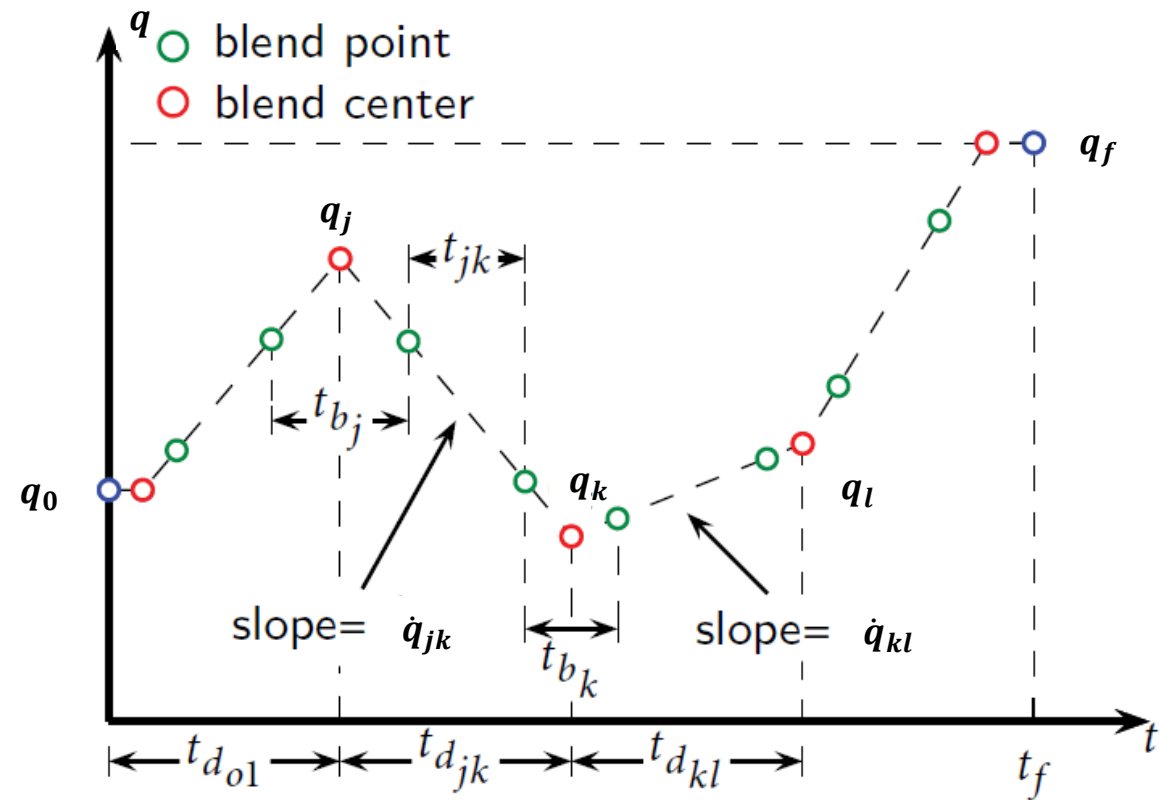
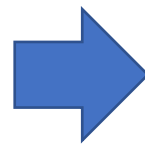
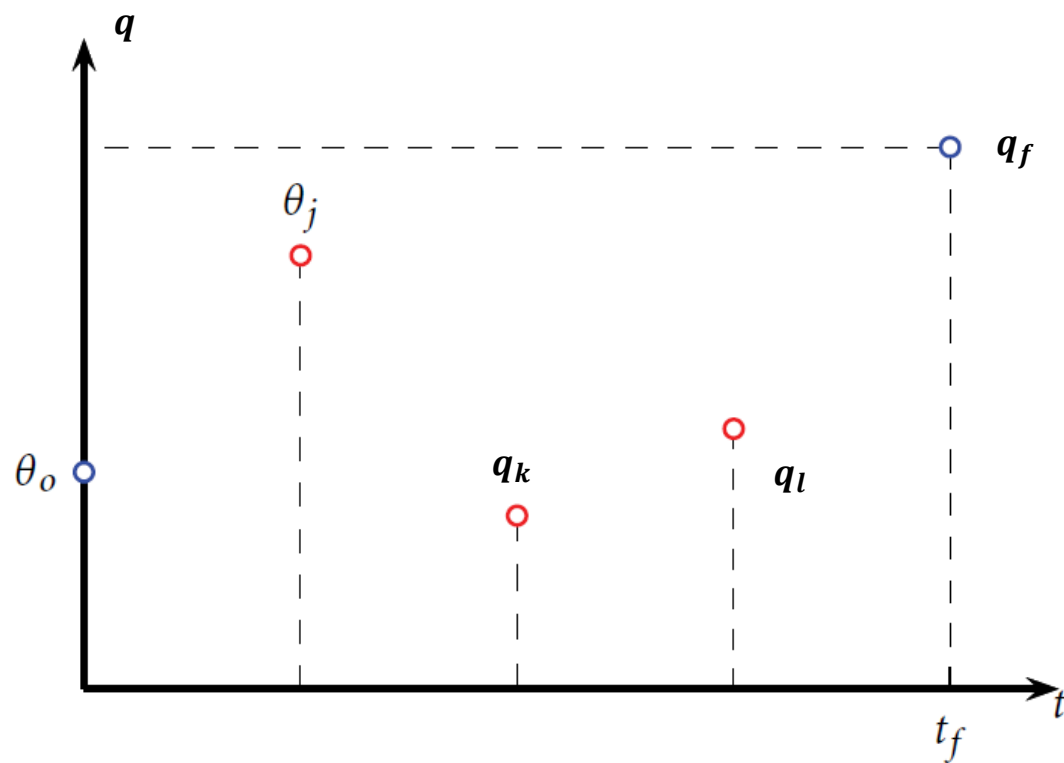
- Given: q_0, q_f and \ddot{q}_{max}
- objective to minimize: t_d
- The outcome is a solution without the linear part
e.g. **Bang-bang trajectory**

$$\ddot{q}(t) = \begin{cases} \ddot{q}_{max} & 0 \leq t \leq t_s \\ -\ddot{q}_{max} & t_s \leq t \leq t_d \end{cases}$$

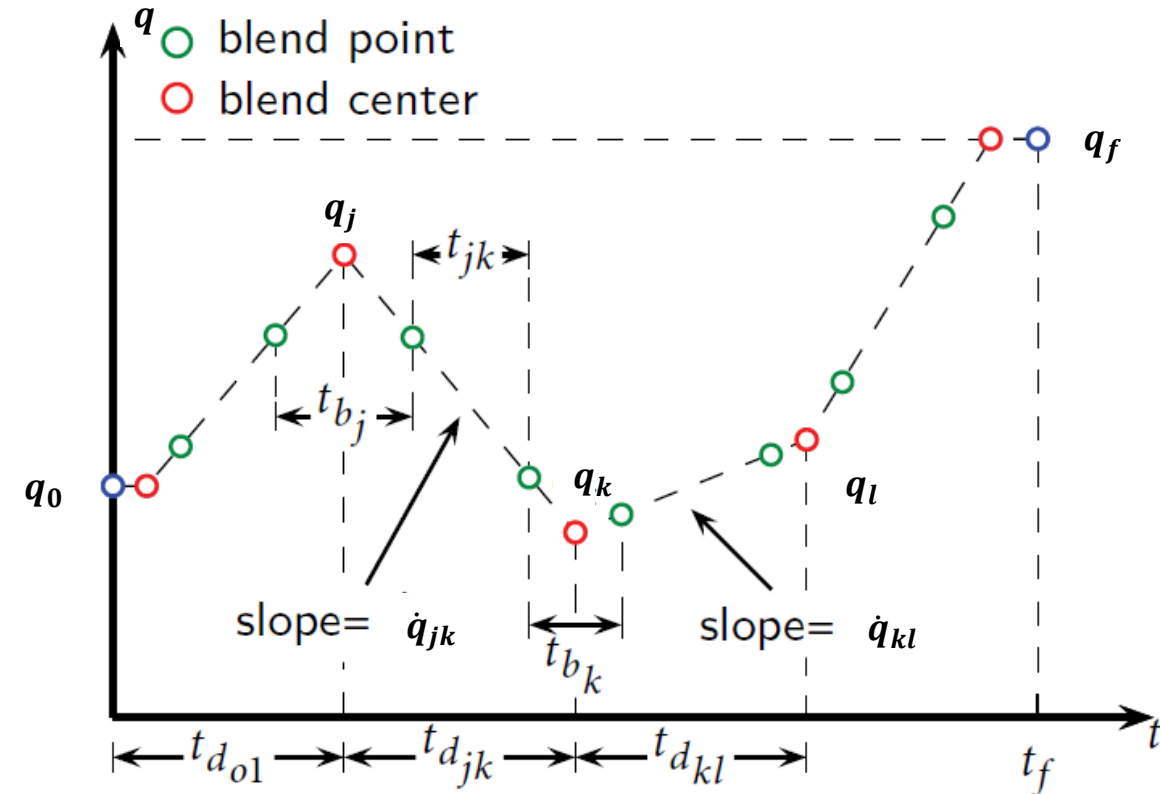
- where the switching time t_s can be calculated by

$$t_s = \frac{t_d}{2} = \sqrt{\frac{q_f - q_0}{\ddot{q}_{max}}}$$

Parabolic blend for a path with via points



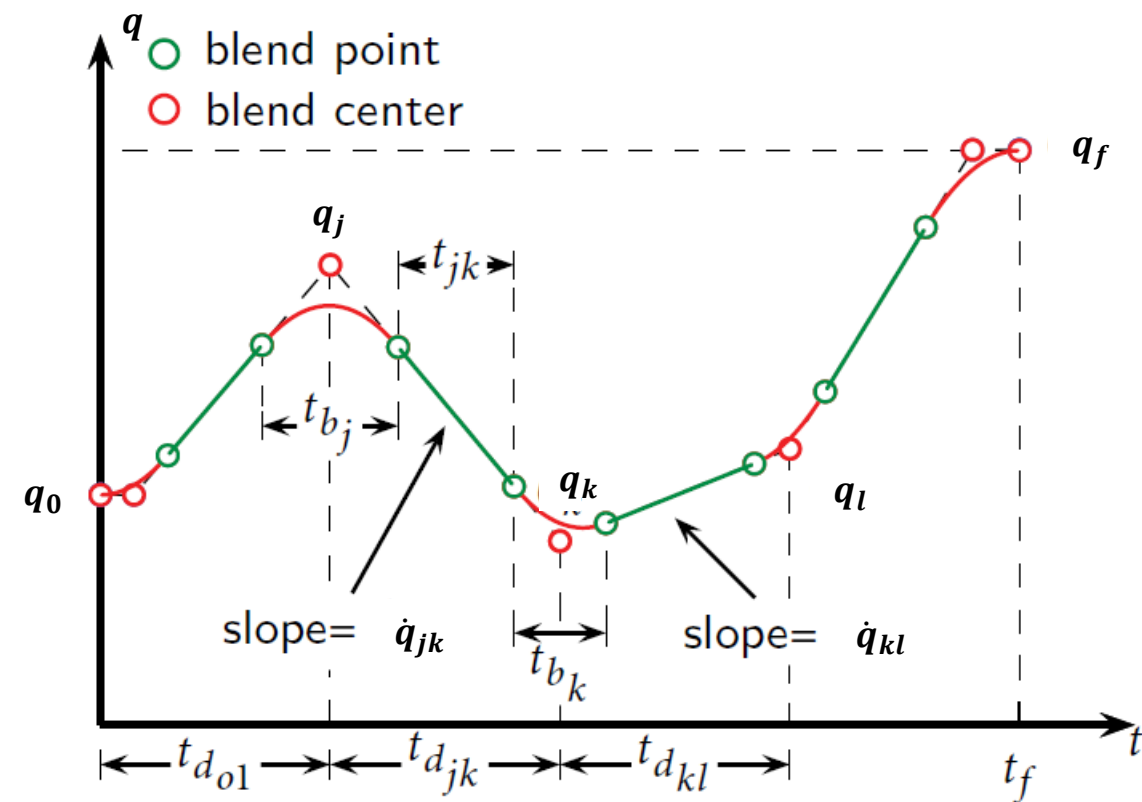
Parabolic blend for a path with via points



- Given: $t_0, q_0, t_f, q_f, \ddot{q}_b$, via points $\{q_i\}_1^m$ at time $\{t_i\}_1^m$ (time duration $t_{d_{jk}} \triangleq t_k - t_j$).
- Find: $q(t)$ interpolating q_0, q_f and approximating $\{q_i\}_1^m$.
- Solution: *Linear Parabolic Blend* with via points

Parabolic blend for a path with via points

- Solution: LPB with via points



for via points $j, k, l = 1, \dots, m$:

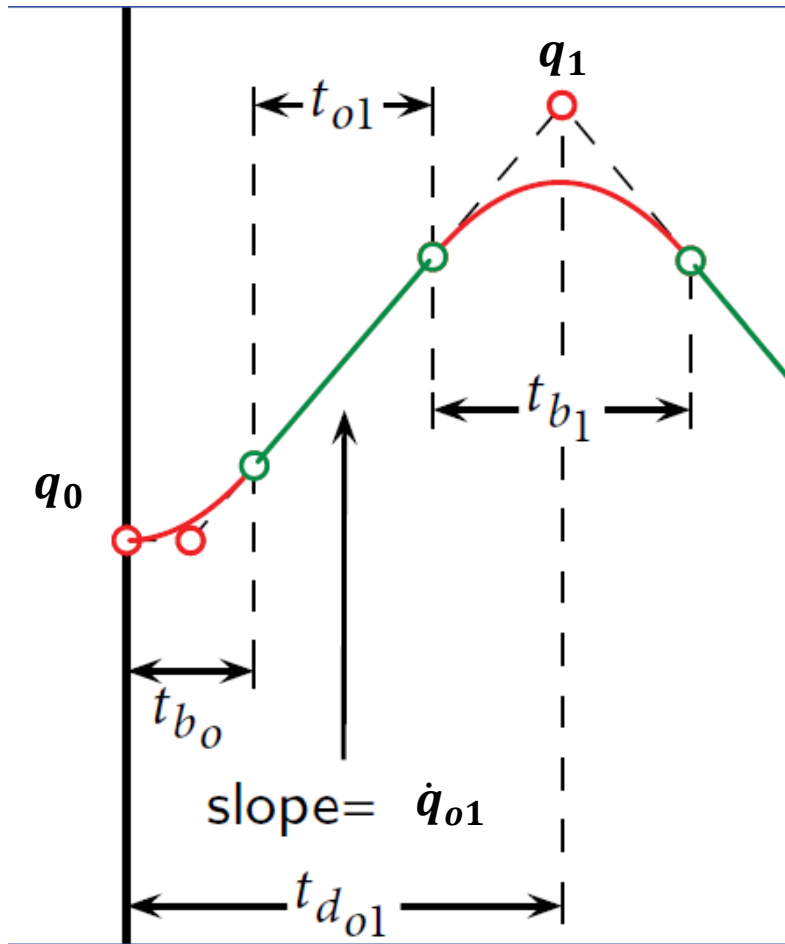
$$\dot{q}_{jk} = \frac{q_k - q_j}{t_{d_{jk}}} \text{ (lin. vel),}$$

$$\ddot{q}_k = \text{Sgn}(\dot{q}_{kl} - \dot{q}_{jk}) |\ddot{q}_b| \text{ (blend acc.),}$$

$$t_{b_k} = \frac{\dot{q}_{kl} - \dot{q}_{jk}}{\ddot{q}_k} \text{ (blend dur.)}$$

$$t_{jk} = t_{d_{jk}} - \frac{1}{2}t_{b_j} - \frac{1}{2}t_{b_k} \text{ (lin. dur.)}$$

Let's divide it into two segments



- First segment:

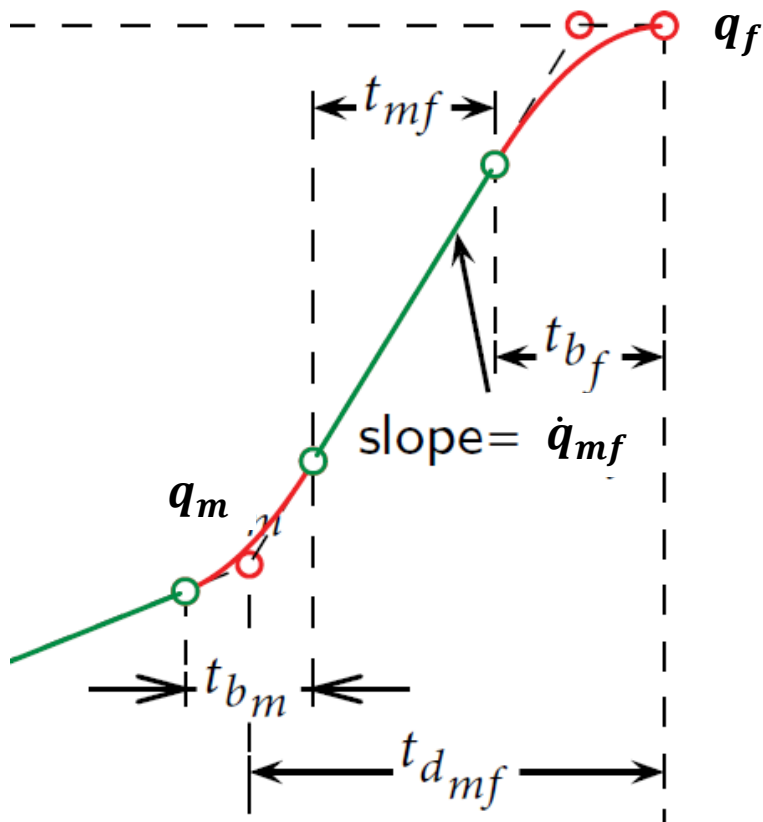
$$\ddot{q}_o = \text{Sgn}(q_1 - q_o) |\ddot{q}_b| \text{ (Blend acc.)}$$

$$t_{b_o} = t_{d_{o1}} - \sqrt{t_{d_{o1}}^2 - \frac{2(q_1 - q_o)}{\ddot{q}_o}} \text{ (Blend dur.)}$$

$$\dot{q}_{o1} = \frac{q_1 - q_o}{t_{d_{o1}} - \frac{1}{2}t_{b_o}} \text{ (lin. vel.)}$$

$$t_{o1} = t_{d_{o1}} - t_{b_o} - \frac{1}{2}t_{b1} \text{ (lin. dur.)}$$

Let's divide it into two segments



• Last segment:

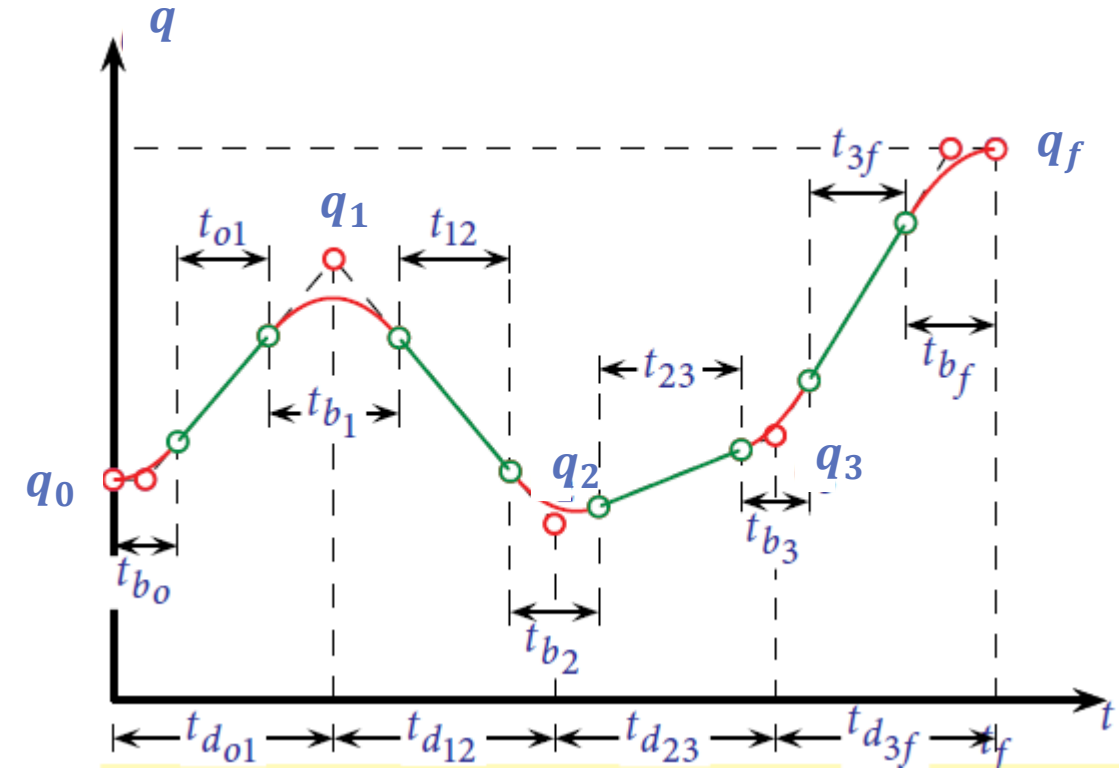
$$\ddot{q}_f = \text{Sgn}(q_m - q_f) |\ddot{q}_b| \text{ (Blend acc.)}$$

$$t_{bf} = t_{dmf} - \sqrt{t_{dmf}^2 - \frac{2(q_f - q_m)}{\ddot{q}_f}} \text{ (Blend dur.)}$$

$$\dot{q}_{mf} = \frac{q_f - q_m}{t_{dmf} - \frac{1}{2}t_{bf}} \text{ (lin. vel.)}$$

$$t_{mf} = t_{dmf} - t_{bf} - \frac{1}{2}t_{bm} \text{ (lin. dur.)}$$

Example with 3 via points



• Given:

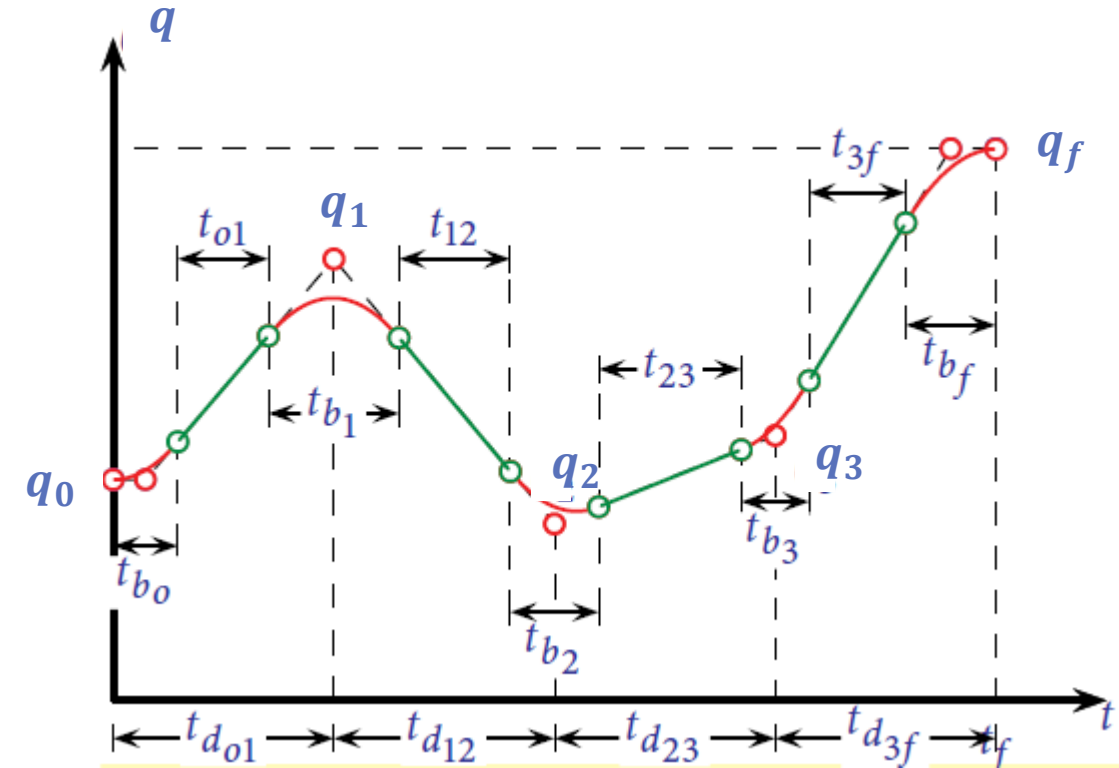
$q_o = 1$ $q_f = 2.5$ $\ddot{q}_b = 4$ $q_1 = 2$ $q_2 = 0.8$ $q_3 = 1.2$
we calculate using the equations from the previous slide:

$$\ddot{q}_0 = 4,$$

$$t_{b_o} = 1 - \sqrt{1^2 - \frac{2(2-1)}{4}} = 0.29$$

$$\dot{q}_{o1} = \frac{2-1}{1 - \frac{1}{2}0.29} = 1.17$$

Example with 3 via points



we calculate:

$$\dot{q}_{12} = \frac{0.8-2}{1} = -1.2,$$

$$\ddot{q}_1 = -4,$$

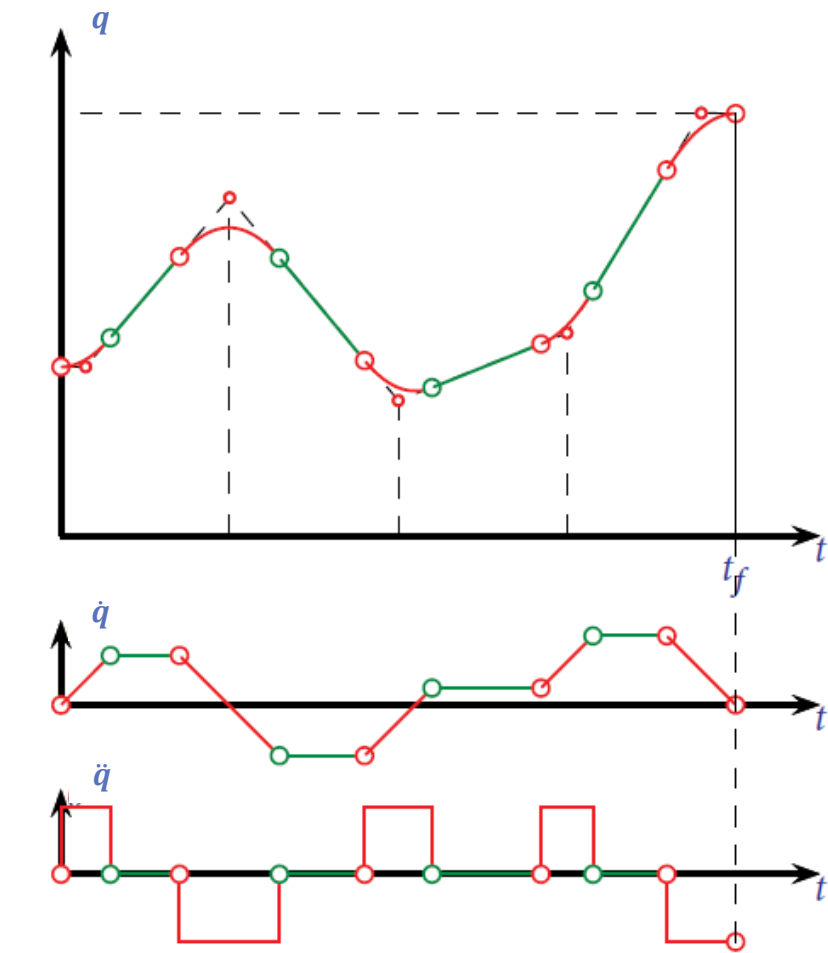
$$t_{b_1} = \frac{-1.2-1.17}{-4} = 0.59, t_{01} = 1 - 0.29 - \frac{1}{2} 0.59 = 0.41$$

$$\dot{q}_{23} = \frac{1.2-0.8}{1} = 0.4,$$

$$\ddot{q}_2 = 4, t_{b_2} = \frac{0.4+1.2}{4} = 0.4,$$

$$t_{12} = 1 - \frac{1}{2} 0.59 - \frac{1}{2} 0.4 = 0.51$$

Example with 3 via points



we calculate:

$$\ddot{q}_f = -4,$$

$$t_{bf} = 1 - \sqrt{1^2 + \frac{2(2.5-1.2)}{-4}} = 0.41$$

$$\ddot{q}_3 = 4,$$

$$t_{b3} = \frac{1.63-0.4}{4} = 0.31$$

$$t_{3f} = 1 - \frac{1}{2} 0.31 - 0.41 = 0.44$$

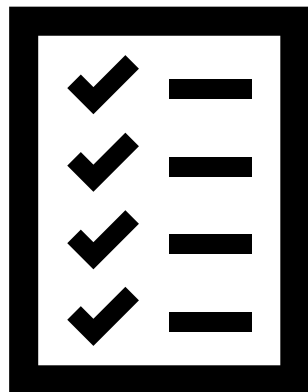
$$t_{23} = 1 - \frac{1}{2} 0.4 - \frac{1}{2} 0.31 = 0.65$$

Parabolic blend in Task space

- Parabolic blend in task space, can for the positional part be calculated in the same way as for joint space parabolic blends.
- Orientation can be computed with piece wise computing slerp trajectories in quaternion space (more info *Dantam N, Stilman M. Spherical parabolic blends for robot workspace trajectories. In 2014 IEEE/RSJ International Conference on Intelligent Robots and Systems 2014 Sep 14 (pp. 3624-3629)*).
- Followed by solving the inverse kinematics for the generated task space path.

Example

Python example



Construct a linear path using python functions (**ctrj**) between the initial joint configuration of the UR robot (`self.q0`), and an offset tool space pose given as:

- `sm.SE3.Trans(0.2, 0.2, 0.35)`
 - with 100 samples
 - extend the generated trajectory with another linear interpolation starting at the current pose and ending at:
 - `sm.SE3.Trans(0,0,-0.4)`
 - with 50 samples.
 - Calculate a joint interpolated trajectory (**jtraj**) starting at the current configuration and ending at a user defined configuration (using sliders)
 - with 20 samples.
1. Plot the position part and the orientation part of the trajectoire.
 2. Plot the velocities and accelerations.