

Trajectory representation as polynomials and splines

Robotics and Computer vision (RoVi)

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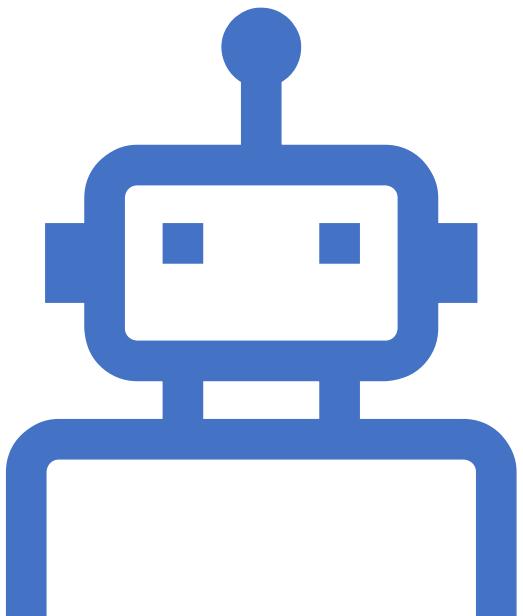
SDU Robotics

The Maersk Mc-Kinney Moller Institute
University of Southern Denmark

Based on M. Mihelj, T. Bajd. et al. Robtics – 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, CRC Press, 2011



Agenda

- Limits on joint positions, velocities and accelerations,
- Cubic polynomial interpolation and splines
- Dynamic systems for trajectory planning

To start with...



and the score

Viewing Division 735: NIST ...

Results – Manufacturing Track

Team	Subtask	Threading Fasteners	Insertions	Wire Routing	Belt Drive	Time Bonus	Subtotal	Total
SDU Robotics	Disassembly	54	27	8	3	18	110	427
	Assembly	72	54	40	43	108	317	
CASIA&Wenjing College Yantai University	Disassembly	0	24	0	0	0	24	109
	Assembly	0	27	30	28	0	85	
New Dexterity Research Group	Disassembly	12	27	2	3	0	44	44
	Assembly	0	0	0	0	0	0	
Robotic Materials	Disassembly	0	15	0	0	0	15	27
	Assembly	0	12	0	0	0	12	
JAKS	Disassembly	0	12	0	0	0	12	18
	Assembly	0	6	0	0	0	6	
Munich School of Robotics and Machine	Disassembly	Did not compete						
	Assembly							

Handling limits on joint positions, velocities and accelerations

Limits on joint positions and velocities

- we consider the problem:

$$J(q)\Delta q = \Delta u$$

- The positional joint limits can be written as:

$$q_{min} \leq q + \Delta q \leq q_{max}$$

- Velocities can be treated similarly:

$$\dot{q}_{min}\Delta t_i \leq \Delta q \leq \dot{q}_{max}\Delta t_i \quad \Delta t_i = t_{i+1} - t_i$$

$$\dot{q}_{min}\Delta t_i + q^i \leq q + \Delta q \leq \dot{q}_{max}\Delta t_i + q^i$$

Limits on accelerations

$$G_{\min}(q^i, q^{i-1}, t_i) = \max\left\{q_{\min}, \dot{q}_{\min}\Delta t_i + q^i, \frac{1}{2}\ddot{q}_{\min}\Delta t_i[\Delta t_{i-1} + \Delta t_i] + q^i + \frac{\Delta t_i}{\Delta t_{i-1}}(q^i - q^{i-1})\right\}$$

$$G_{\max}(q^i, q^{i-1}, t_i) = \min\left\{q_{\max}, \dot{q}_{\max}\Delta t_i + q^i, \frac{1}{2}\ddot{q}_{\max}\Delta t_i[\Delta t_{i-1} + \Delta t_i] + q^i + \frac{\Delta t_i}{\Delta t_{i-1}}(q^i - q^{i-1})\right\}$$

- we can finally write the problem as:

$$\min \| J(\mathbf{q})\Delta\mathbf{q} - \Delta\mathbf{u} \|$$

$$G_{\min}(q^i, q^{i-1}, t_i) \leq \mathbf{q} + \Delta\mathbf{q} \leq G_{\max}(q^i, q^{i-1}, t_i)$$

Pros and cons of this approach

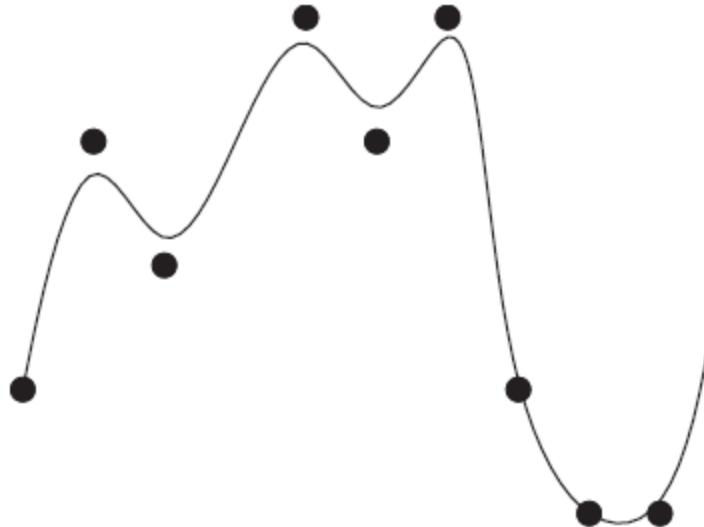
- Advantages:
 - Suitable for time-critical trajectories (e.g. visual servoing),
 - good handling of positional limits,
 - always computes the feasible solution closest to the desired tool movement.
- Disadvantages:
 - a bit complicated to program (Quadratic Programming Problem - QPP),
 - tool path may deviate in space.

From Points, via- points to trajectories

From Points, via-points to trajectories



Interpolation



Approximation

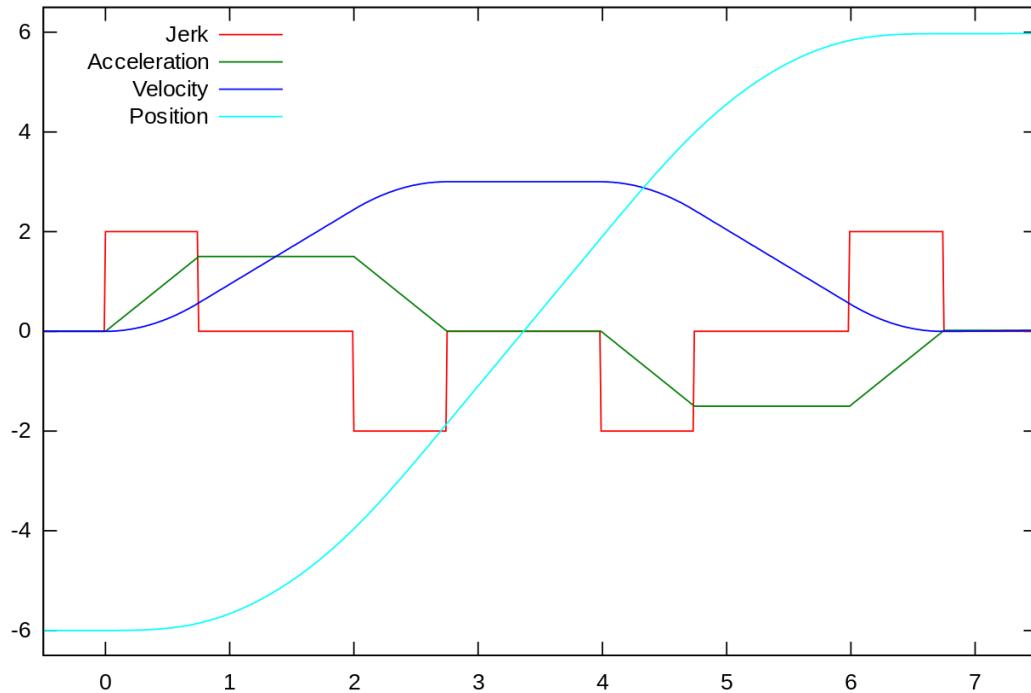
- Definition: Interpolation Constructing new data points within the range of a discrete set of known data points (exact fitting).
- Definition: Approximation Inexact fitting of a discrete set of known data points.

Interpolation between points

- In the last lecture we introduced:
 - linear interpolation between points,
 - linear interpolation with parabolic blends,
- In comparison, velocities and accelerations are better approximated with the parabolic blends.
- Still this method does not give a completely smooth trajectory!
- For this we must look at the derivative of the acceleration at blend points.

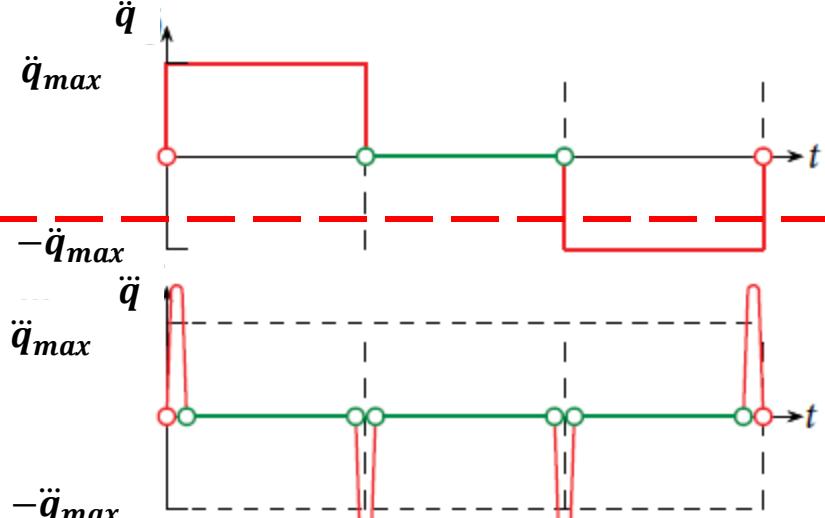
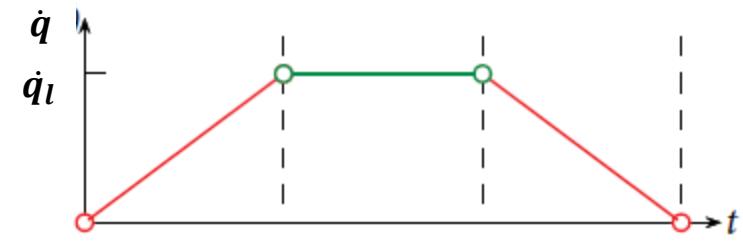
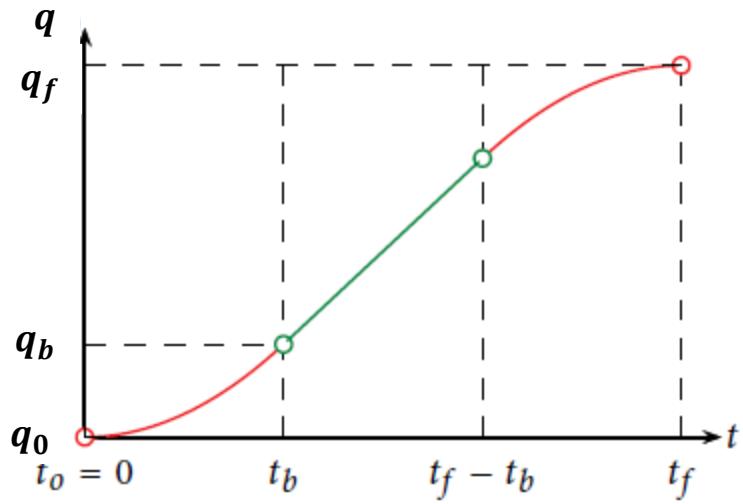
Definition of jerk for motion planning

- Jerk in motion control is the rate at which an object's acceleration changes with respect to time.
- It is a vector quantity (having both magnitude and direction). and expressed in m/s^3 (SI units).
- Jerk can be calculated as:
$$\vec{J} = \frac{da(t)}{dt} = \frac{d^2v(t)}{dt} = \frac{d^3q(t)}{dt}$$
- jerk can be calculated for joint and task space trajectories.



<https://commons.wikimedia.org/w/index.php?curid=46643265>

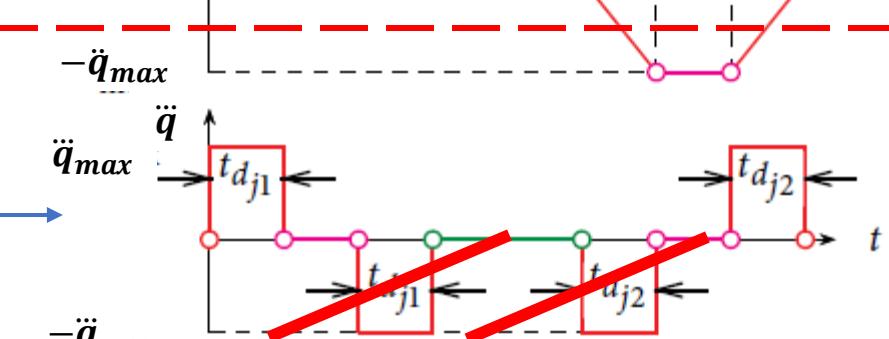
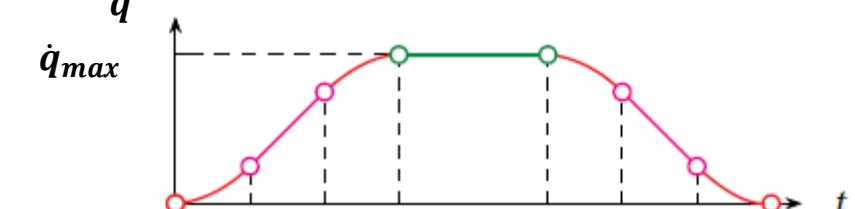
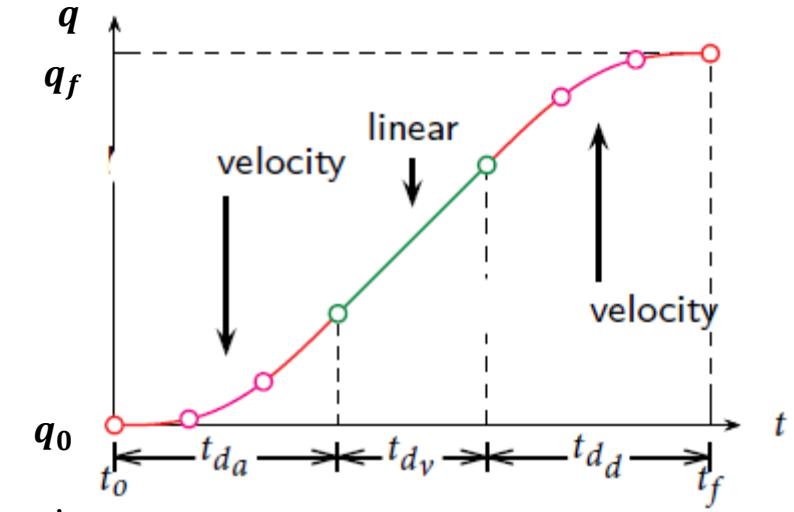
How does this show?



Introduction of trapezoidal acc. blend

= S blends [L. Biagiotti and C. Melchiorri. *Trajectory Planning for Automatic Machines and Robots*. Springer Verlag, 2008]

We are looking for
= polynomial interpolation



Cubic Polynomial Interpolation

Cubic Polynomial Interpolation

- Cubic polynomial interpolation can be applied in task and joint space.
- Dense here means that we may assume that the displacements between adjacent frames are small.
- In such a case, linear interpolation is generally not suitable because the segments are too small (velocity is not continuous between segments).
- Here, we recommend to use cubic interpolations or splines.

Cubic Polynomial Interpolation

- Cubic interpolation can be used in:
 - joint space, using n-tuple $X_i \equiv q^i$,
 - tool space, using 7-tuple $X_i \equiv (p^i, Q^i)$ for $i = 0, 1, \dots, N$.
- Orientations should be represented as quaternions **(don't forget to normalize at the end!!)**.
- Between two frames (X_{i-1} and X_i) we obtain a curve, defined as:

$$X_{i-1,i}(t) = a_i t^3 + b_i t^2 c_i t + d_i$$

- where parameters a_i, b_i, c_i, d_i have to be determined.
- depending on the number of curves we have $4N$ parameters to determine.

Cubic Polynomial Interpolation

- The curves must satisfy the following constraints

$$X_{i-1,i}(t_{i-1}) = a_i t_{i-1}^3 + b_i t_{i-1}^2 + c_i t_{i-1} + d_i = X_{i-1} \quad i = 1, \dots, N$$

$$X_i(t_i) = a_i t_i^3 + b_i t_i^2 + c_i t_i + d_i = X_i \quad i = 1, \dots, N$$

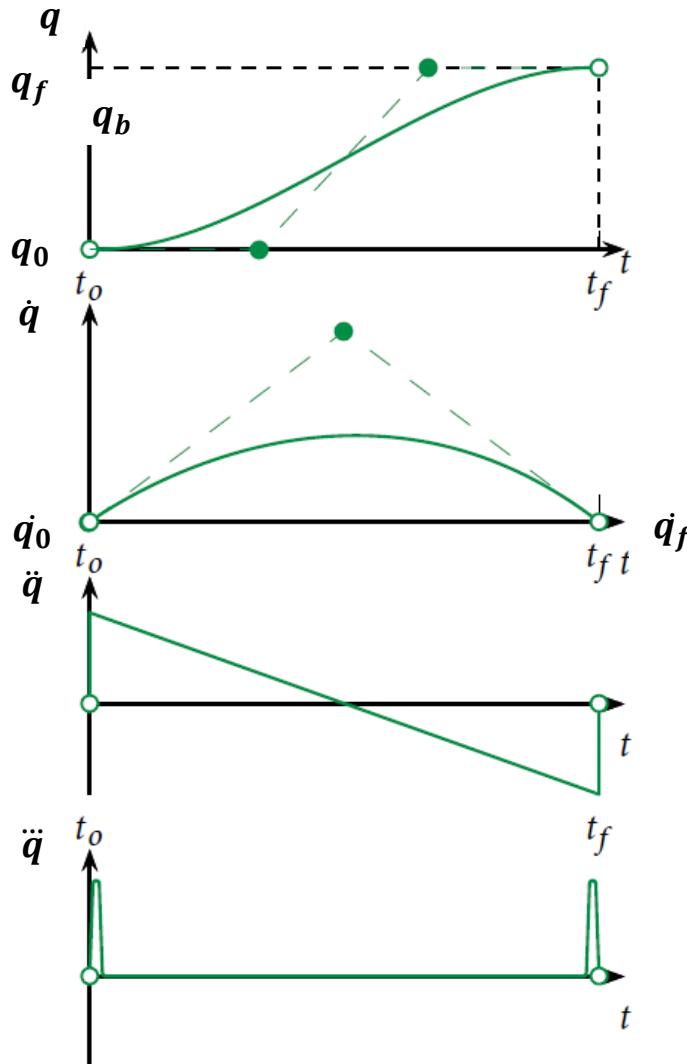
- which gives a $2N$ linear equations with unknowns,
- the remaining $2N$ equations can be given in several ways.
- for example if velocities at passage points are given we get

Cubic Polynomial Interpolation

$$\begin{aligned}\dot{X}_{i-1}(t_{i-1}) &= 3a_i t_{i-1}^2 + 2b_i t_{i-1} + c_i = \dot{X}_{i-1} \quad i = 1, \dots, N \\ \dot{X}_i(t_i) &= 3a_i t_i^2 + 2b_i t_i + c_i = \dot{X}_i \quad i = 1, \dots, N\end{aligned}$$

- which gives 4 linear equations with 4 unknowns for each sub-path.

Let's take a closer look at this problem...



- Let's consider a cub. interp. between two positions in joint space.
- where 4 parameters a_0, a_1, a_2, a_3 are to be determined by boundary conditions.
- Property: bounded acceleration, jerk impulse at both ends.

$$q(t) = a_0 + a_1(t - t_0) + a_2(t - t_0)^2 + a_3(t - t_0)^3$$

$$t \in [t_0, t_f], t_d \triangleq t_f - t_0$$

$$\begin{cases} q(t_0) = a_0 = q_0 \\ \dot{q}(t_0) = a_1 = \dot{q}_0 \\ q(t_f) = a_0 + a_1 t_d + a_2 t_d^2 + a_3 t_d^3 = q_f \\ \dot{q}(t_f) = a_1 + 2a_2 t_d + 3a_3 t_d^2 = \dot{q}_f \end{cases} \Rightarrow \begin{cases} a_0 = q_0 \\ a_1 = \dot{q}_0 \\ a_2 = \frac{3h - (2\dot{q}_0 + \dot{q}_f)t_d}{t_d^2} \\ a_3 = \frac{-2h + (\dot{q}_0 + \dot{q}_f)t_d}{t_d^3} \end{cases}$$

Multi point cubic interpolation

- Given $q_0, q_f, \dot{q}_0, \dot{q}_f$ at t_0, t_f and via points $[q_k]_1^m$ and $[t_k]_1^m$ where m is the number of sections between the points and k is the number of the specific point, we solve:

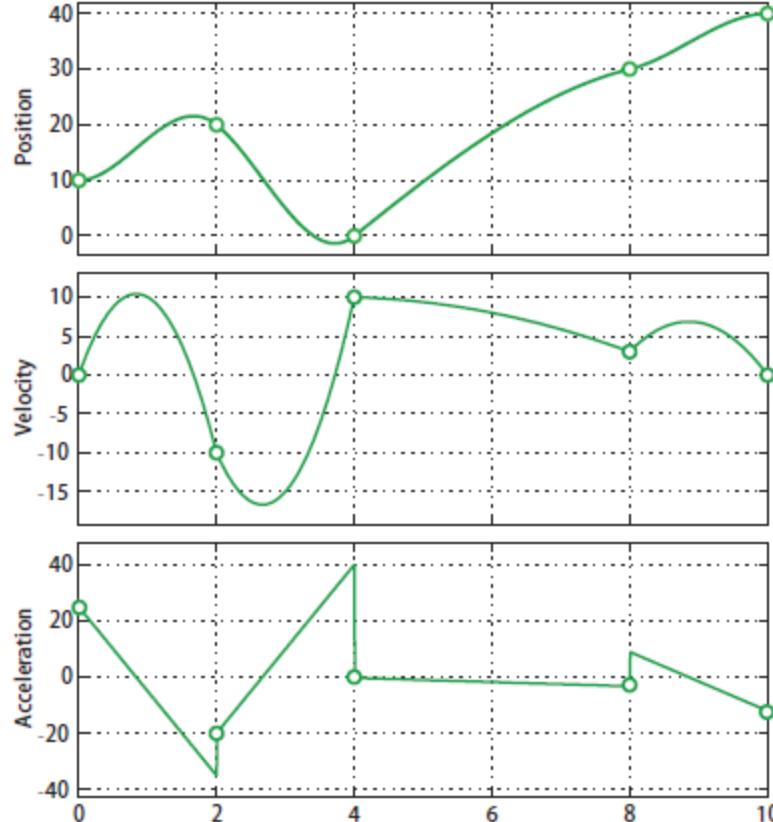
$$a_{0k} + a_{1k}(t - t_k) + a_{2k}(t - t_k)^2 + a_{3k}(t - t_k)^3$$

for all unknowns $\{a_{0k}, a_{1k}, a_{2k}, a_{3k}\}_0^m$.

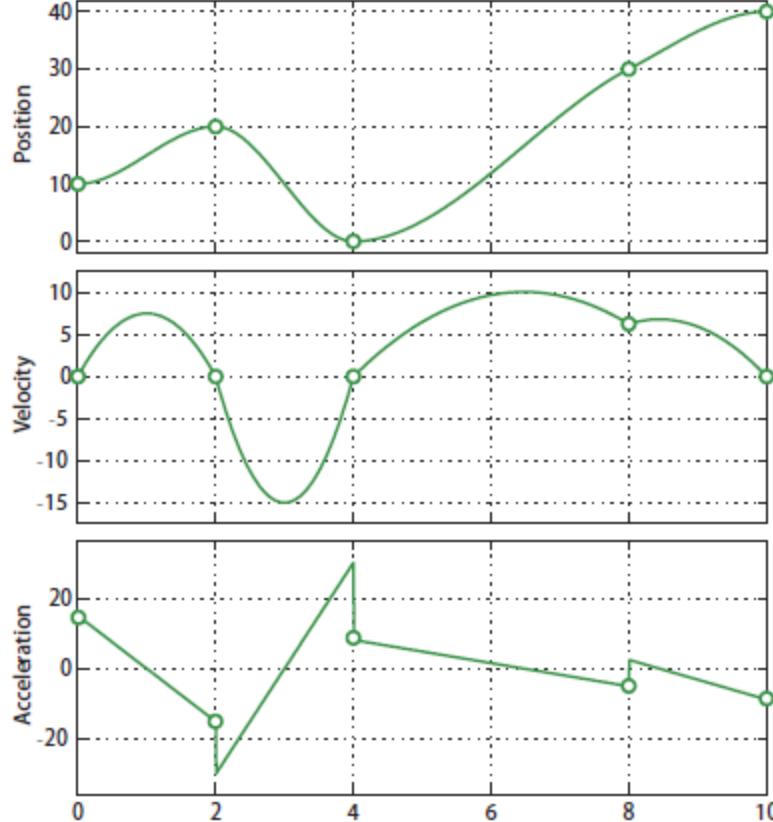
- accelerations at via points can be assigned by the user $[\ddot{q}_k]_1^m$ and $t_d = t_f - t_0, h = q_f - q_0$, we can compute the coefficients.

$$\begin{cases} a_{0k} = q_0 \\ a_{2k} = \frac{3h - (2\dot{q}_0 + \dot{q}_f)t_d}{t_d^2} \\ a_{3k} = \frac{-2h + (2\dot{q}_0 + \dot{q}_f)t_d}{t_d^3}, k = 0, 1, \dots, m \end{cases}$$

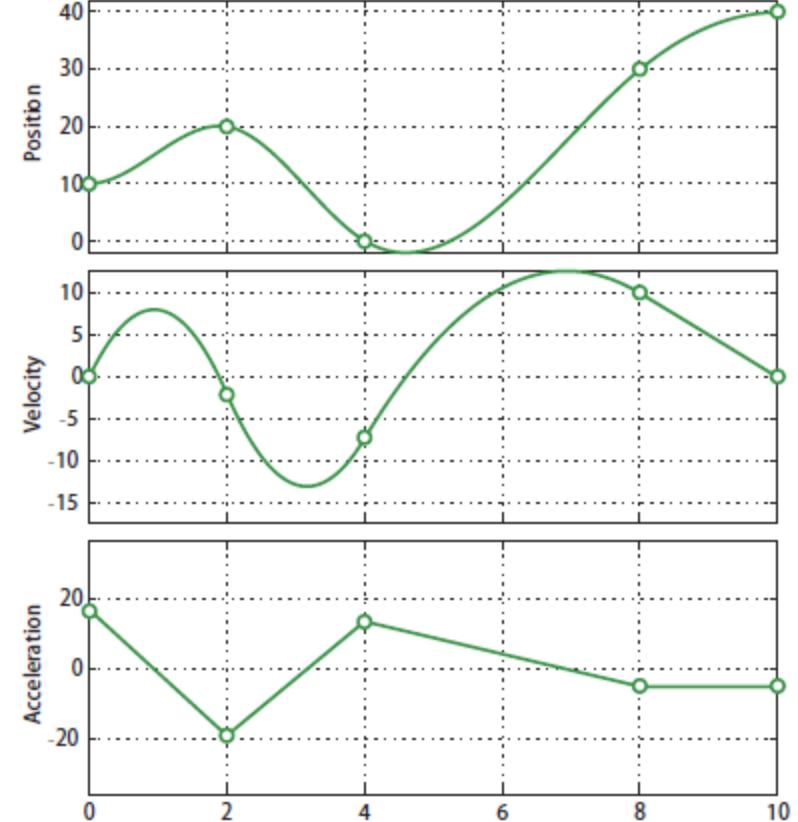
With via points Example:



Via points velocities are arbitrary assigned – large discontinuous acc.



Via points velocities assigned – large discontinuous acc.



Higher order polynomial – continuous acc.

Higher order polynomial – Quintic polynomial

$$\theta(t) = a_0 + a_1(t - t_0) + a_2(t - t_0)^2 + a_3(t - t_0)^3 + a_4(t - t_0)^4 + a_5(t - t_0)^5, t \in [t_0, t_f]$$

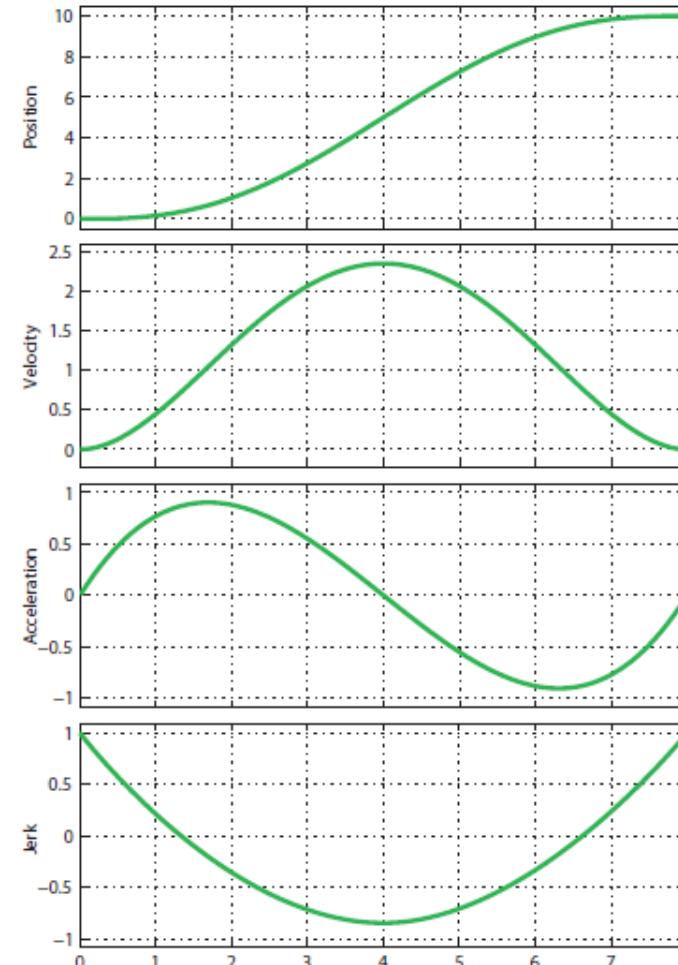
- With 6 unknown's coefficients $a_i, i = 0, \dots, 5$.
- **Properties:**
- Smooth and bounded jerk
- Acc. continuity in composite curves.

Boundary conditions:

$$\theta(t_o) = \theta_o, \quad \theta(t_f) = \theta_f$$

$$\dot{\theta}(t_o) = \dot{\theta}_o, \quad \dot{\theta}(t_f) = \dot{\theta}_f$$

$$\ddot{\theta}(t_o) = \ddot{\theta}_o, \quad \ddot{\theta}(t_f) = \ddot{\theta}_f$$



We calculate the coefficients:

We define $t_d = t_f - t_0$, $h = q_f - q_0$, then

$$a_0 = q_0$$

$$a_1 = \dot{q}_0$$

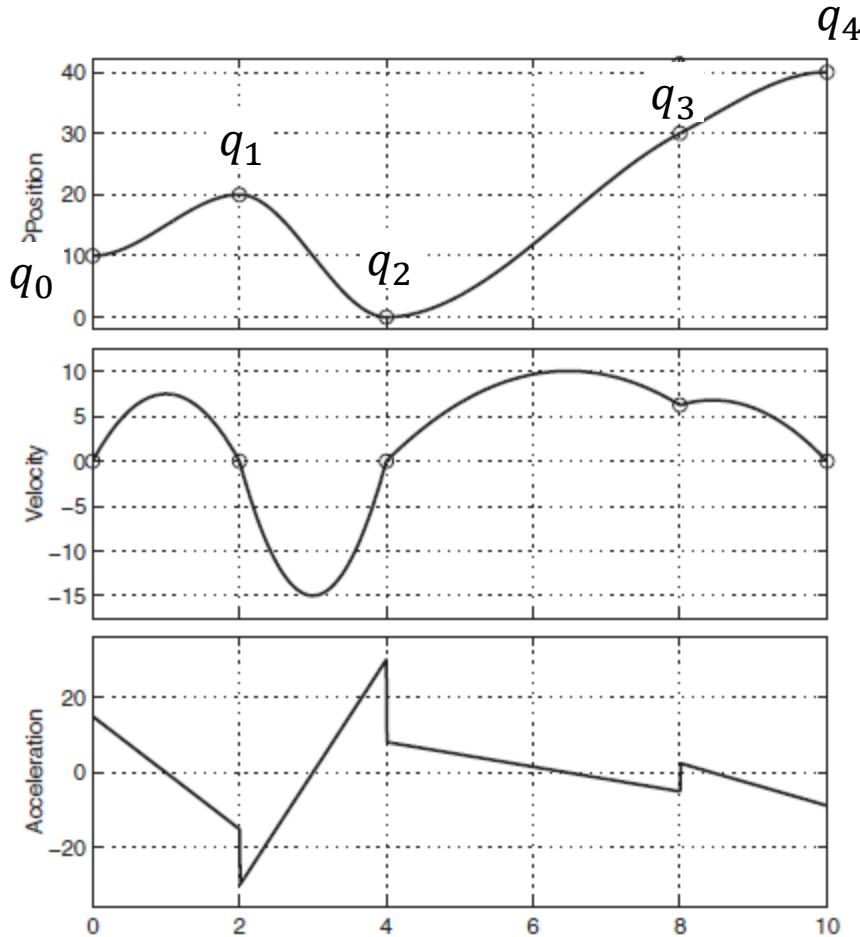
$$a_2 = \frac{1}{2} \ddot{q}_0$$

$$a_3 = \frac{1}{2t_d^3} [20h - (8\dot{q}_f + 12\dot{q}_0)t_d - (3\ddot{q}_0 - \ddot{q}_f)t_d^2]$$

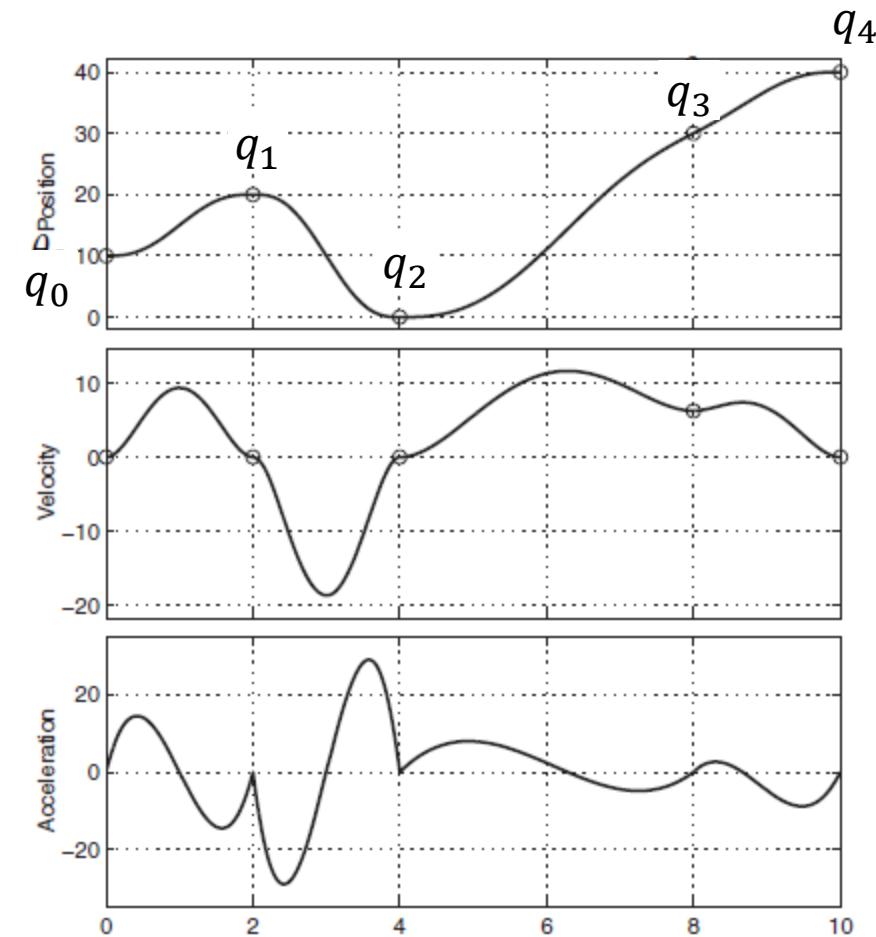
$$a_4 = \frac{1}{2t_d^4} [-30h - (14\dot{q}_f + 16\dot{q}_0)t_d - (3\ddot{q}_0 - 2\ddot{q}_f)t_d^2]$$

$$a_5 = \frac{1}{2t_d^2} [12h - 6(\dot{q}_f + \dot{q}_0)t_d - (\ddot{q}_f - \ddot{q}_0)t_d^2]$$

Comparison



Cubic polynomials: acceleration discontinuity.



Quintic polynomials: continuity in acceleration.

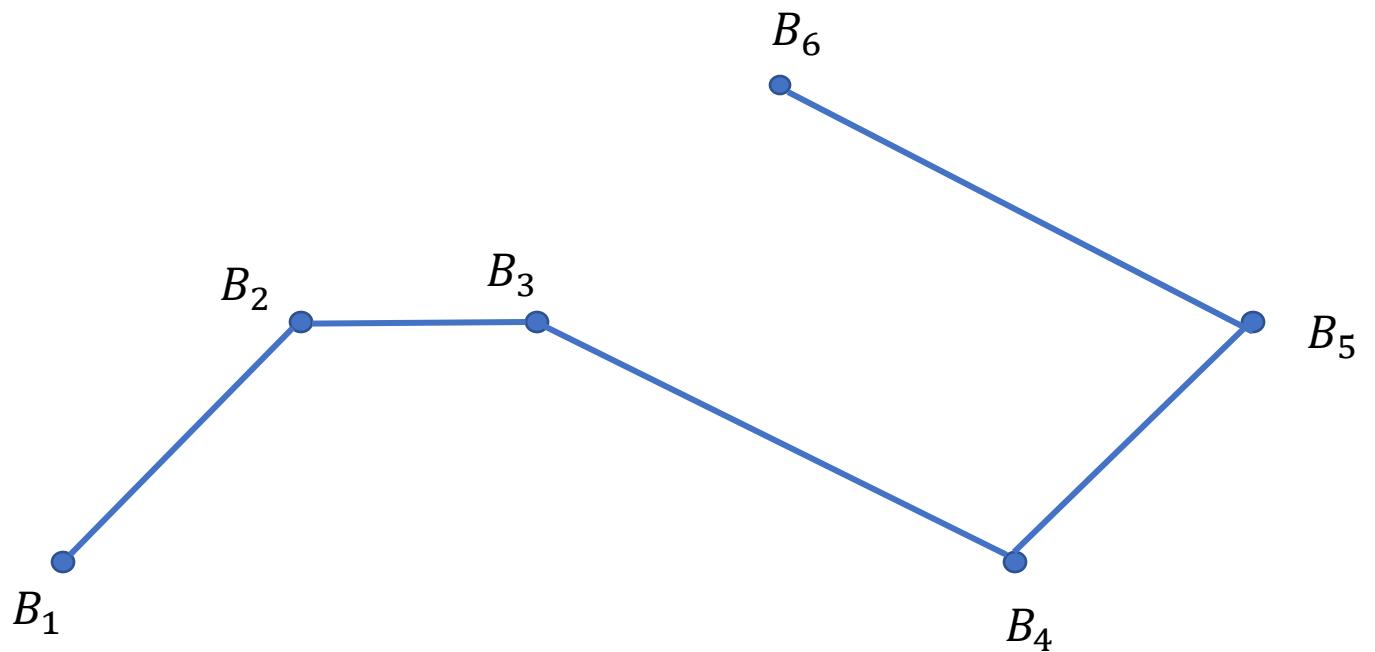
Splines

Splines

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

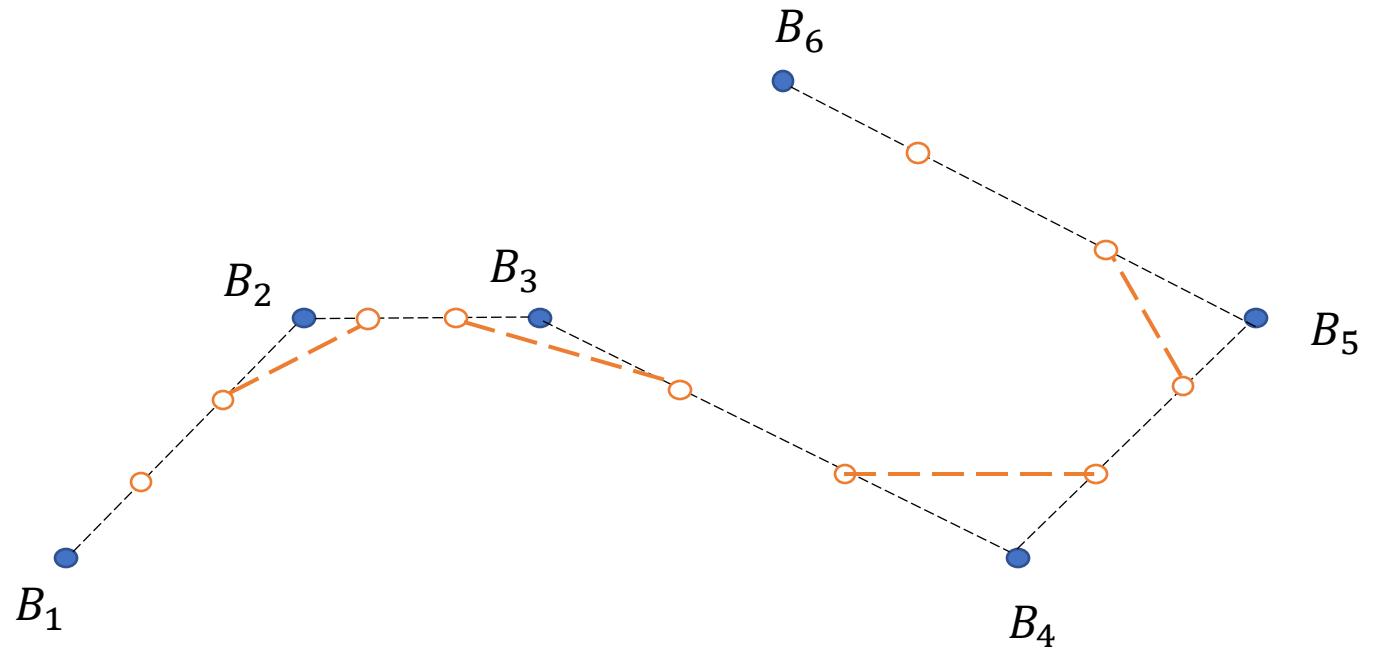
Splines - by hand

- Lets consider a couple of points.
- we can connect them with a line (do linear interpolation).
- maybe not optimal.



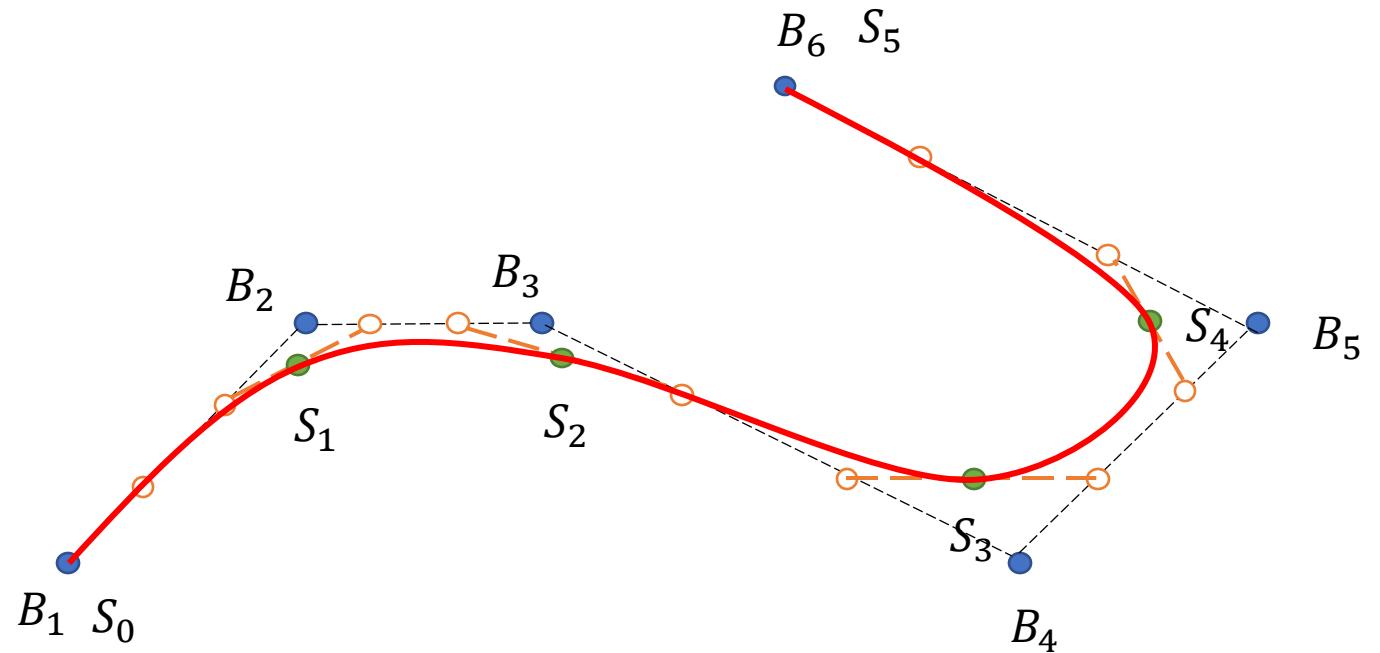
Splines – by hand

- We can divide the space between the points into thirds,
- and connect them between each other.



Splines – by hand

- Split the connecting line into half,
- which gives us the anchor points of the spline,
- A cubic spline is defined via fitting a Bezier Curve to the anchor points.



Cubic splines

- If the velocities are prescribed only at the first and last frame then the following set of equations are used:

$$\dot{X}_{0,1}(t_0) = 3a_1 t_0^2 + 2b_1 t_0 + c_1 = \dot{X}_0$$

$$\dot{X}_{N-1,N}(t_N) = 3a_N t_N^2 + 2b_N t_N + c_N = \dot{X}_N$$

$$3a_i t_i^2 + 2b_i t_i + c_i = 3a_{i+1} t_i^2 + 2b_{i+1} t_i + c_{i+1} \quad i = 1, \dots, N$$

$$6a_i t_i + 2b_i = 6a_{i+1} t_i + 2b_{i+1} \quad i = 1, \dots, N$$

- thus we obtain a set of $4N$ coupled linear equations that can be solved by standard numerical methods.

Dynamic Movement Primitives

Dynamic movement primitives (DMPs)

- DMPs are defined at the acceleration level:

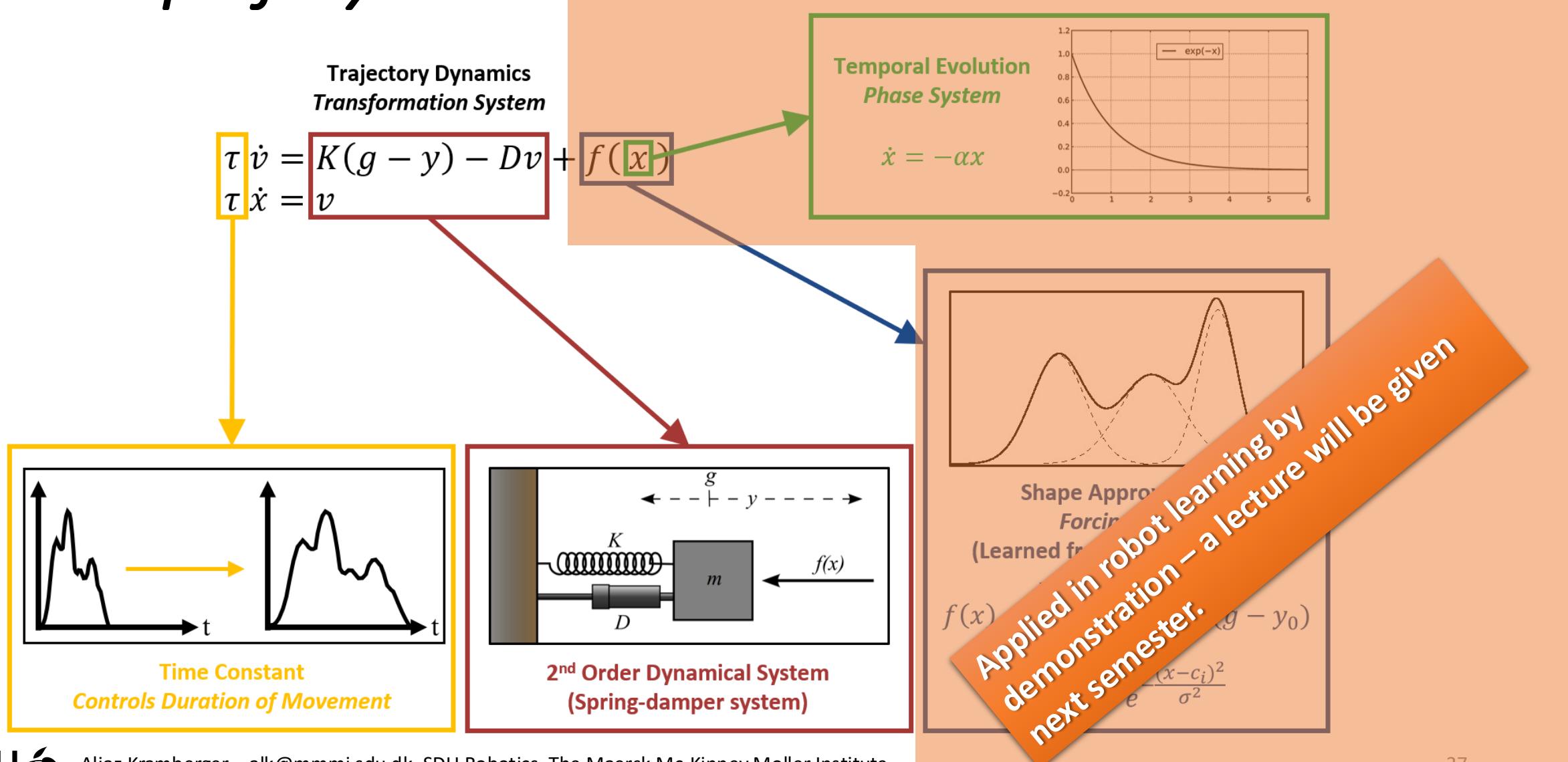
$$\begin{aligned}\tau \dot{z} &= K(g - y_0) - Dz + f(x) \\ \tau \dot{y} &= z\end{aligned}$$

- Guarantees continuity and smoothness for robot control,
- preserves the physical meaning of $f(x)$ as a force,
- g represents the goal of the desired movement,
- y_0 represents the start of the desired movement,
- K, D positive gains, configured in such a way that the dynamics has a stable point at the defined g .

Benefits

- Not explicit time dependent,
- demonstrated/ recorded movement can be reproduced,
- the movement can be altered online,
- the movement can be defined with a couple of simple parameters,
- the movement can be stopped and resumed again,
- the system is stable, and can handle disturbances,
- can be implemented on various systems.

Recap of Dynamic Movement Primitives



DMP: Cartesian-space Orientation

Position

- Three independent regular DMPs.
- Expressed here in vector form.

$$\tau \dot{z} = \alpha_z (\beta_z (g_p - p) - z)$$

$$\tau \dot{p} = z$$

Orientation

- Represented as unit quaternions.
- Singularity free representation.
- Similar properties as positional DMPs.

$$\tau \dot{\eta} = \alpha_z (\beta_z (2 \log(g_q * \bar{q})) - \eta)$$

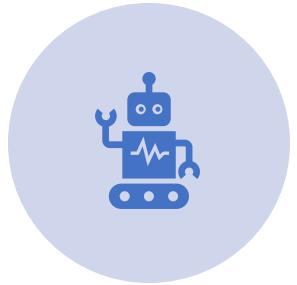
$$\tau \dot{q} = \frac{1}{2} \eta * q$$

Shared Phase

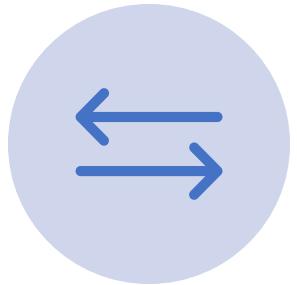
- Synchronizes position and orientation

$$\tau \dot{x} = -\alpha_x x$$

Take home message



Position and velocity limits can be directly incorporated in the inverse kinematics'.



With a higher order polynomial, better approximation and acceleration , jerk conditions are obtained.



DMP's can be utilized for constructing point to point movements – properties.



Splines, provide continuous acceleration and velocities, with the consequence of precision inaccurate.