

# EE5110/5062/6110: Special Topics in Automation & Control — Autonomous systems Motion planning (III)

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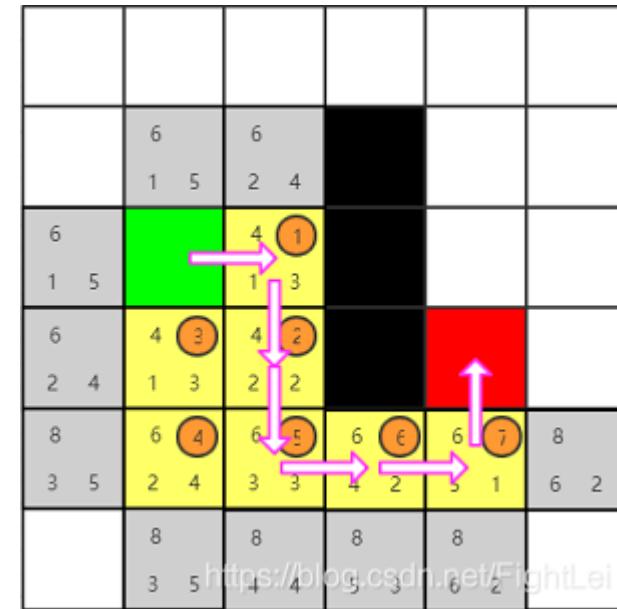
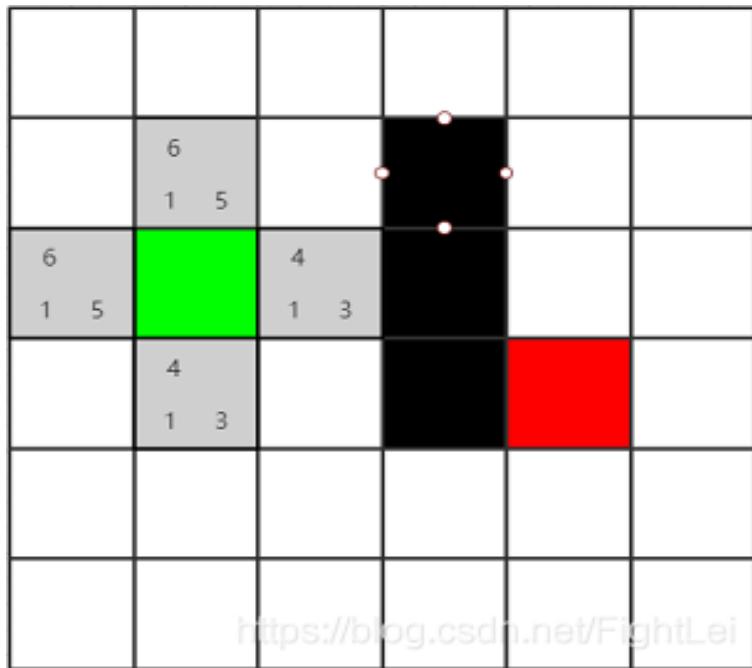


# Review:

- A\* finds the shortest path between starting point A and ending point B without colliding with static obstacles

## An example in MATLAB

# Rule:right->down->left->up



Can A\* handle moving obstacles?

# Learning Objective

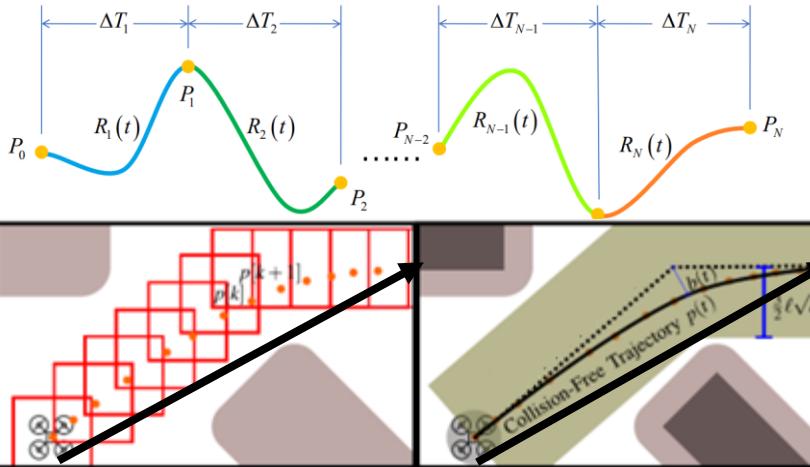
- Understand the trajectory generation—polynomial method
- Can integrate this trajectory with controller.

# Topics To Be Covered

- 1. Concept of trajectory generation**
- 2. Several typical trajectories**
- 3. \*Trajectory represented by polynomial**

# 1. Background

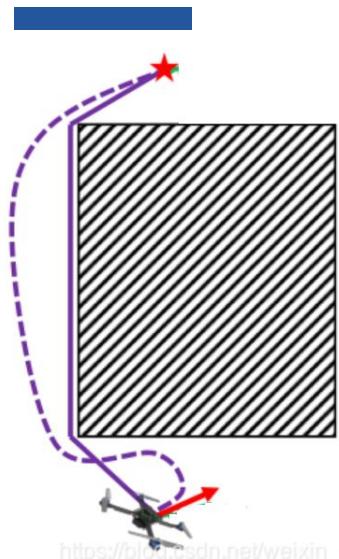
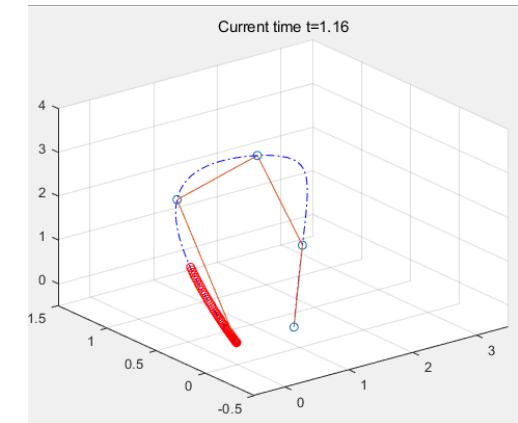
Can we move from point A to point B directly even it is free?



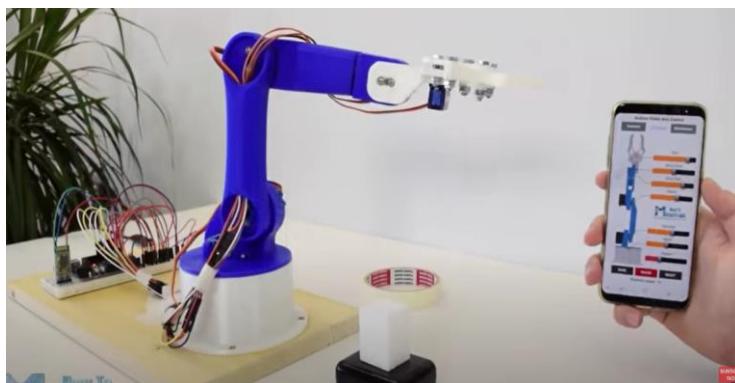
- A\*
- D\*

Fusion work of  
path-finding algorithm

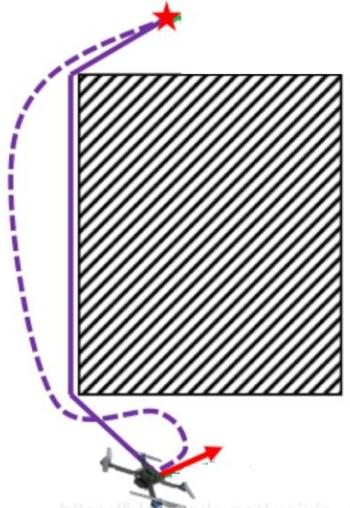
Optimization of path:  
trajectory(smooth)



<https://bit.ly/2dn0n7z>



Following trajectory



# Path vs. Trajectory

[https://blockcsdn.net/weixin\\_449/](https://blockcsdn.net/weixin_449/)

**Path:** is an ordered locus of points in the space (either joint or operational), which the manipulator should follow.

It is a pure geometric description of motion.

**Trajectory:** a path on which timing law is specified, e.g., velocities and accelerations in each point.

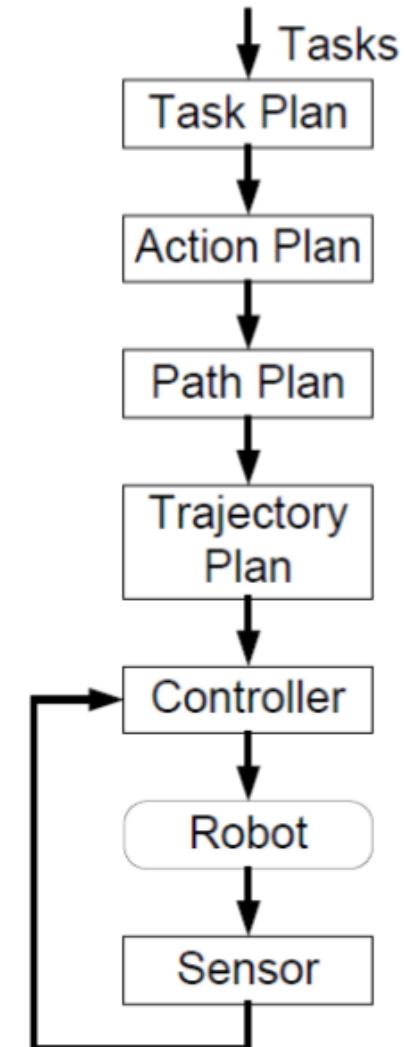
# Robot Motion Planning

## Path planning:

- Geometric path
- Issues: optimal path.

## Trajectory:

- Interpolate or approximate the desired path by a class of polynomial functions.
- Generate a sequence of time-based “control set points” for the control of manipulator from the initial configuration to its destination.

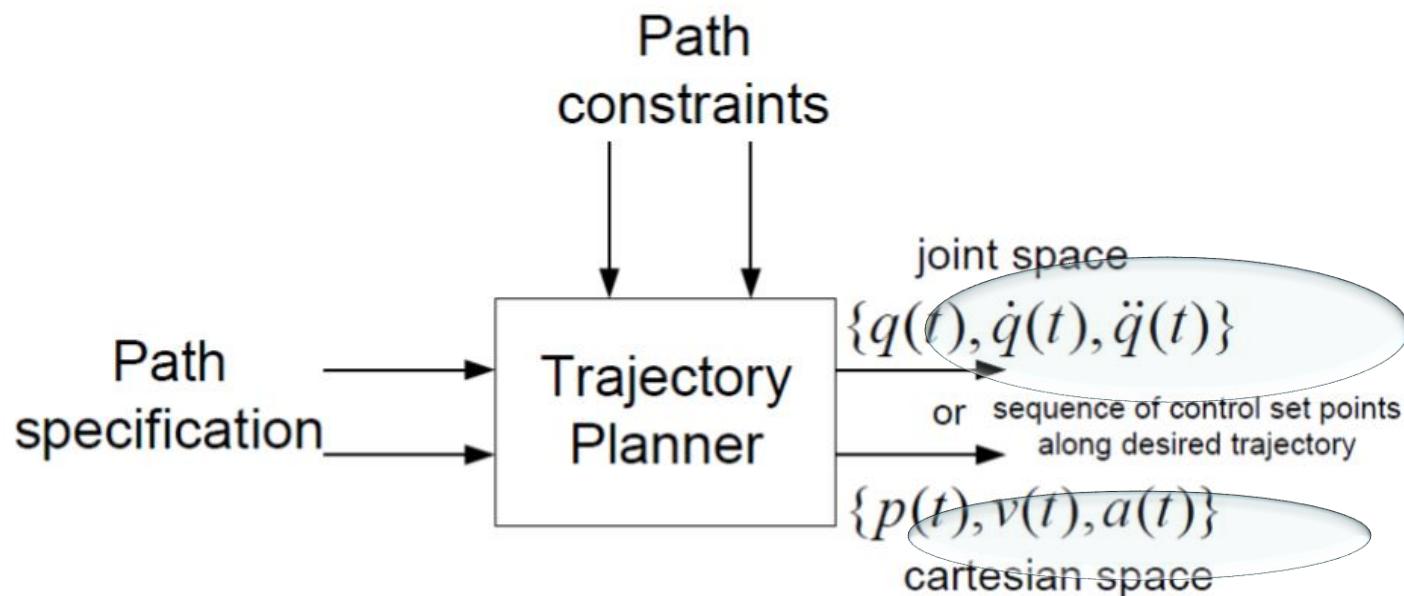


Roughly, a trajectory is a specification of the robot position (configuration) as a function of time.

## Trajectory Generation Problem

- Trajectory:  $\theta(s(t))$  or  $\theta(t)$
- Velocity and acceleration along a trajectory  $\theta(s(t))$ :  
$$\dot{\theta} = \frac{d\theta}{ds} \dot{s}; \quad \ddot{\theta} = \frac{d\theta}{ds} \ddot{s} + \frac{d^2\theta}{ds^2} \dot{s}^2$$
- Jerk and snap along a trajectory  $\theta(s(t))$ :
- $\ddot{\theta}$  and  $\dddot{\theta}$
- Trajectory Generation: Construct a trajectory (path + time scaling) so that the robot reaches a sequence of points in a given time
- Trajectory should be **sufficiently smooth** and respect limits on joint variables, velocities, accelerations, or torques

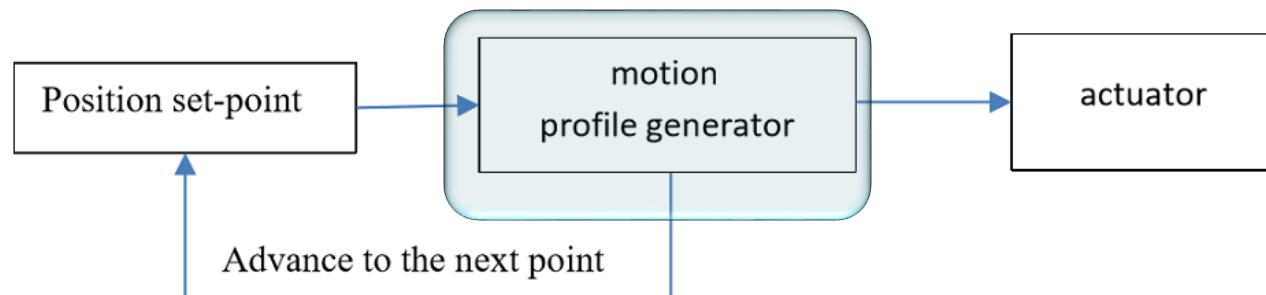
# Trajectory Planning



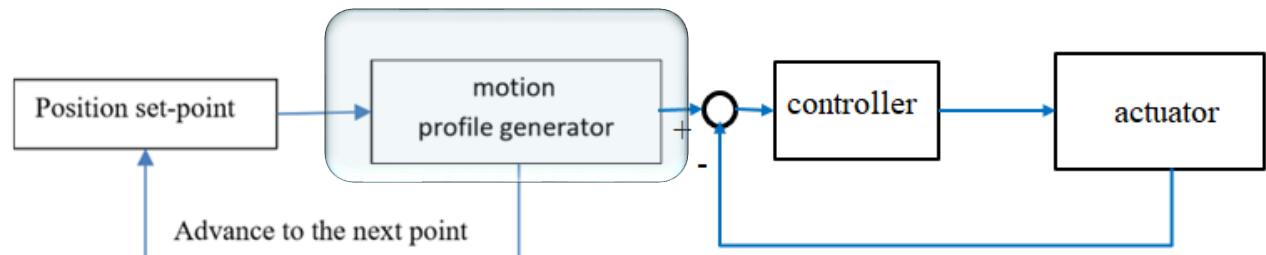
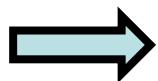
- Path Profile
- Velocity Profile
- Acceleration Profile

# Implementation of the motion profiles generator

Feedforward



Desired



## 2. Several typical motion profiles

The type of motion profile required for an application depends on the purpose of the movement.

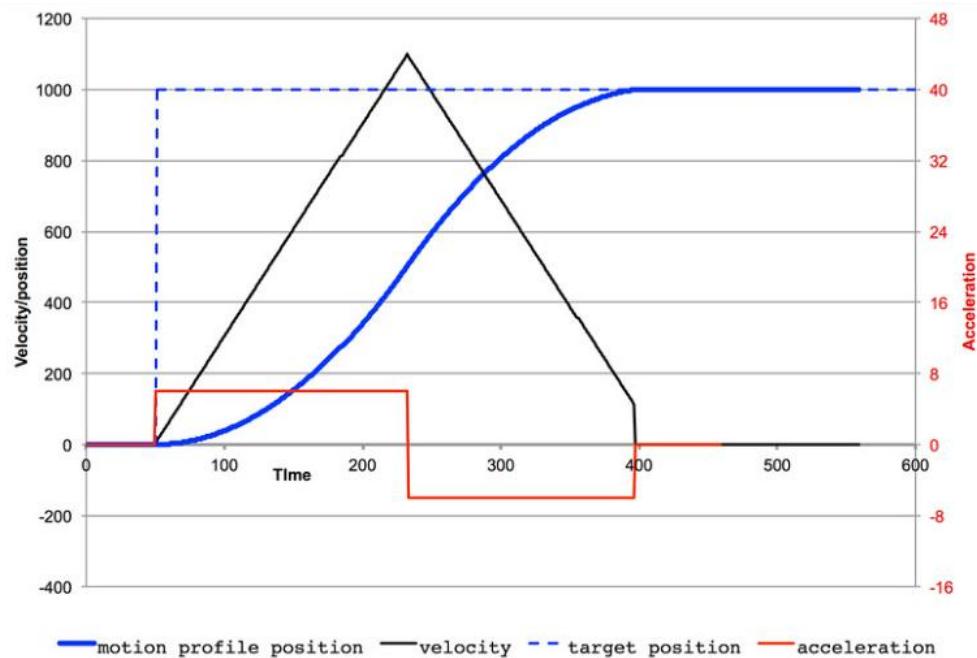
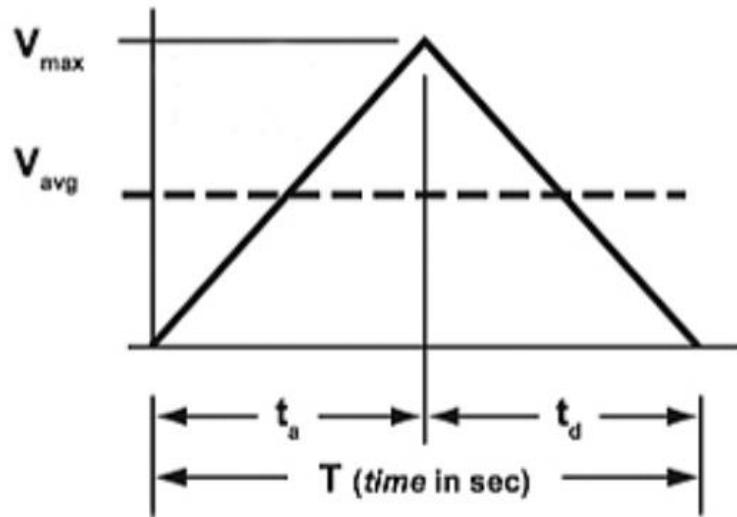
We consider

- **Triangular profile**
- **Trapezoidal motion profiles:**
- **S-curve**

## 2.1.Triangular motion profiles: For quick, point-to-point movements

The basic premise of a triangular move profile is to accelerate to a maximum velocity, then immediately decelerate, with acceleration and deceleration being equal in terms of both time and distance. In other words, if you want to move from here to there in the fastest time, you would use a triangular move profile.

Determining any of the move variables—time, velocity, acceleration, or distance—is based on the geometry of the triangle. Below are some common equations for triangular move profiles.



Not very smooth, but allows for easy incorporation of joint velocity and acceleration limits.

## Triangular motion profiles:

$t \in [0, t_a]$ : constant acceleration phase  $\ddot{s} = a$

$$v = at$$

$$p = \frac{v_0 + v}{2} t = \frac{a}{2} t^2$$

$t \in (t_a, t_{a+d}]$ : constant deceleration phase  $\ddot{s} = -a$

**Objective: find position profile, velocity profile and calculate acceleration.**

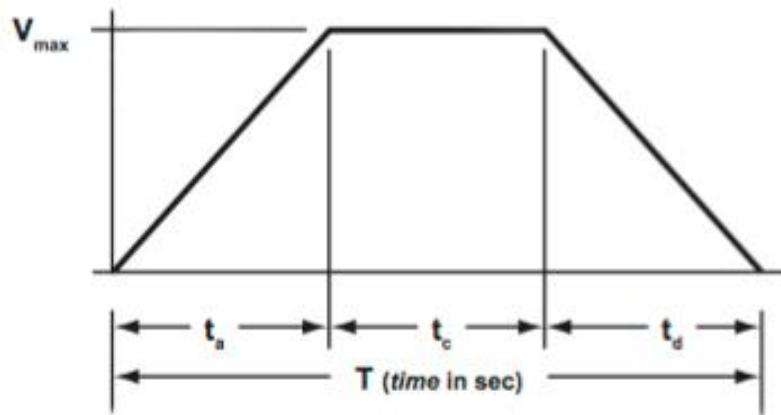
**Not very smooth, but allows for easy incorporation of joint velocity and acceleration limits.**

## Limitations:

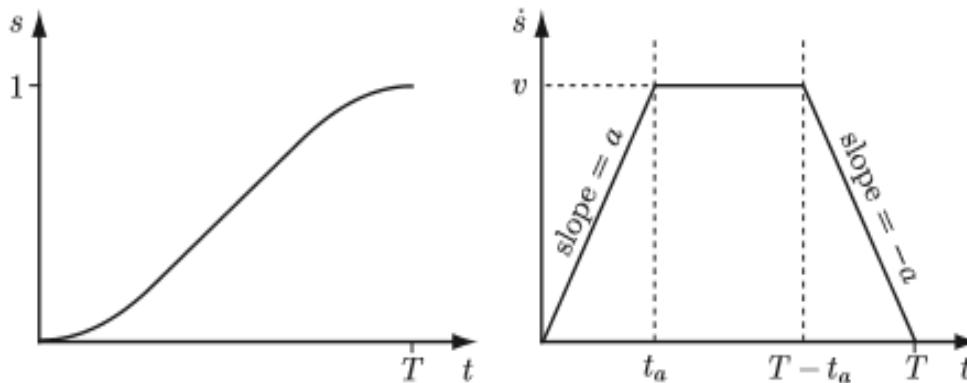
- 1) overall fastest option for getting from one point to the other,**
- 2) Not smooth**

## 2.2Trapezoidal move profile

A trapezoidal move profile is used when the application needs to accelerate to a maximum velocity and then travel at that velocity for a specified time or distance. Some common processes that use trapezoidal move profiles are machining, dispensing, and painting.



The simplest form of trapezoidal move profile, and the one used in the examples below, is the 1/3, 1/3, 1/3 profile. In this case, 1/3 of the time is used for accelerating, 1/3 is used for constant velocity, and 1/3 is used for decelerating. But if you understand the geometry of the move profile, which is essentially two triangles and a rectangle, you can calculate the necessary parameters regardless of whether the time segments are equal or not.



- Trapezoidal time scaling:
  - $t \in [0, t_a]$ : constant acceleration phase  $\ddot{s} = a$        $v = at$        $p = \frac{a}{2} t^2$
  - $t \in (t_a, T - 2t_a]$ : constant velocity phase  $\dot{s} = v$        $p = vt + \frac{a}{2} t_a^2$ .
  - $t \in (T - 2t_a, T]$ : constant deceleration phase  $\ddot{s} = -a$
- Not very smooth, but allows for easy incorporation of joint velocity and acceleration limits.

## Advantages:

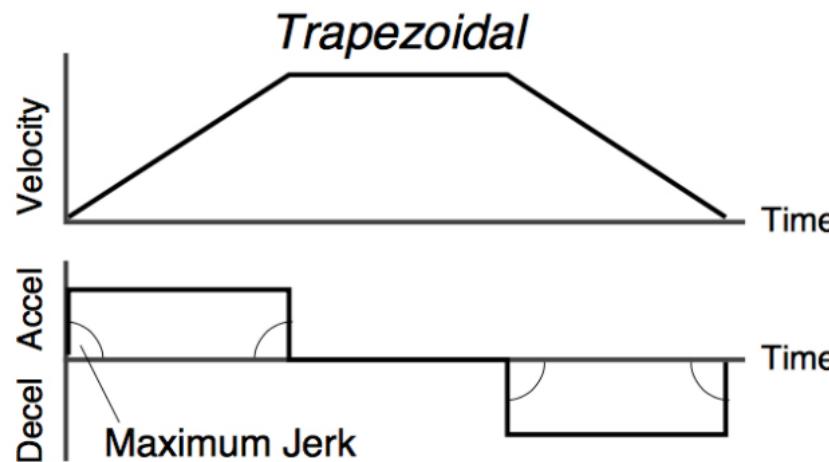
**For applications that benefit from a period of constant velocity, a trapezoidal move profile offers the advantage of having a lower maximum velocity than a triangular profile.**

## Limitations:

**Trapezoidal profile causes discontinuities in acceleration at  $t=0, ta, ta+tc$  and  $T$ .**

## 2.3. Adding curves produces smoother motion

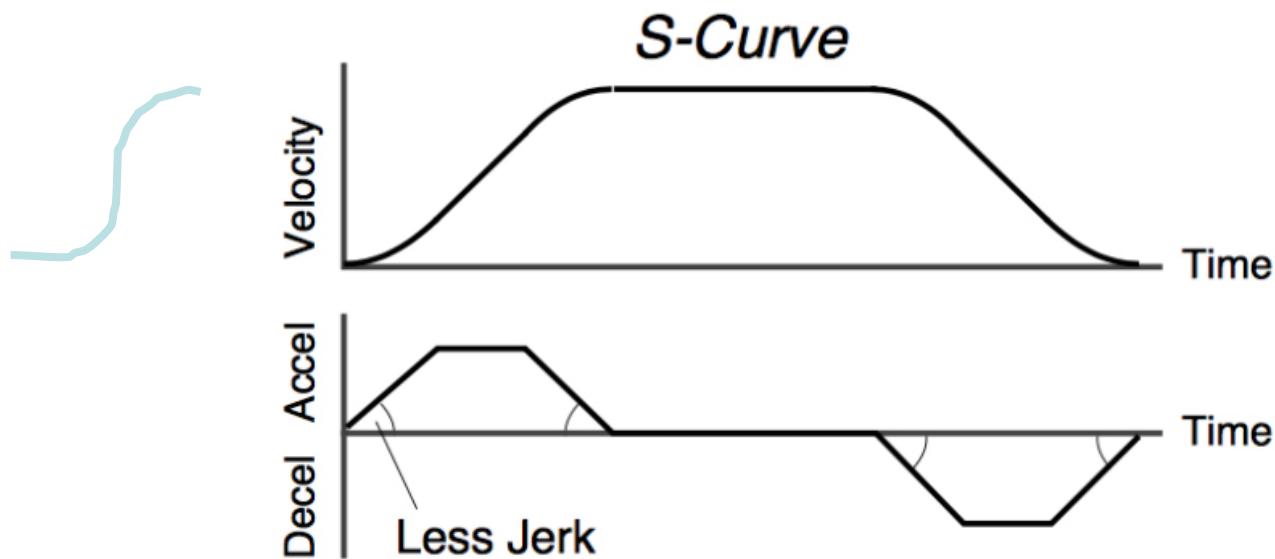
Despite underpinning the majority of motion control applications, neither triangular nor trapezoidal move profiles are ideal for motion systems due to a phenomenon known as “jerk.” **Jerk** is the rate of change of acceleration, and for trapezoidal and triangular move profiles, the initial acceleration and final deceleration occur instantly, meaning jerk is (theoretically) infinite



Standard triangular and trapezoidal (shown above) motion profiles require instant acceleration, which leads to (theoretically) infinite jerk.

Jerk is especially problematic for systems that require smooth, accurate motion because it causes vibrations that can reduce positioning accuracy and extend **settling time**.

To reduce jerk, the beginnings and ends of the acceleration and deceleration phases of the move are smoothed into an “S” shape. This limits the rate of change of acceleration and deceleration (jerk) and produces smoother motion and more accurate positioning.



Smoothing the beginning and end of the acceleration and deceleration phases of motion – known as an S-curve motion profile – allows acceleration to increase and decrease over time, which reduces jerk.

# An improved version is the S-curve. It consists of 7 phases.

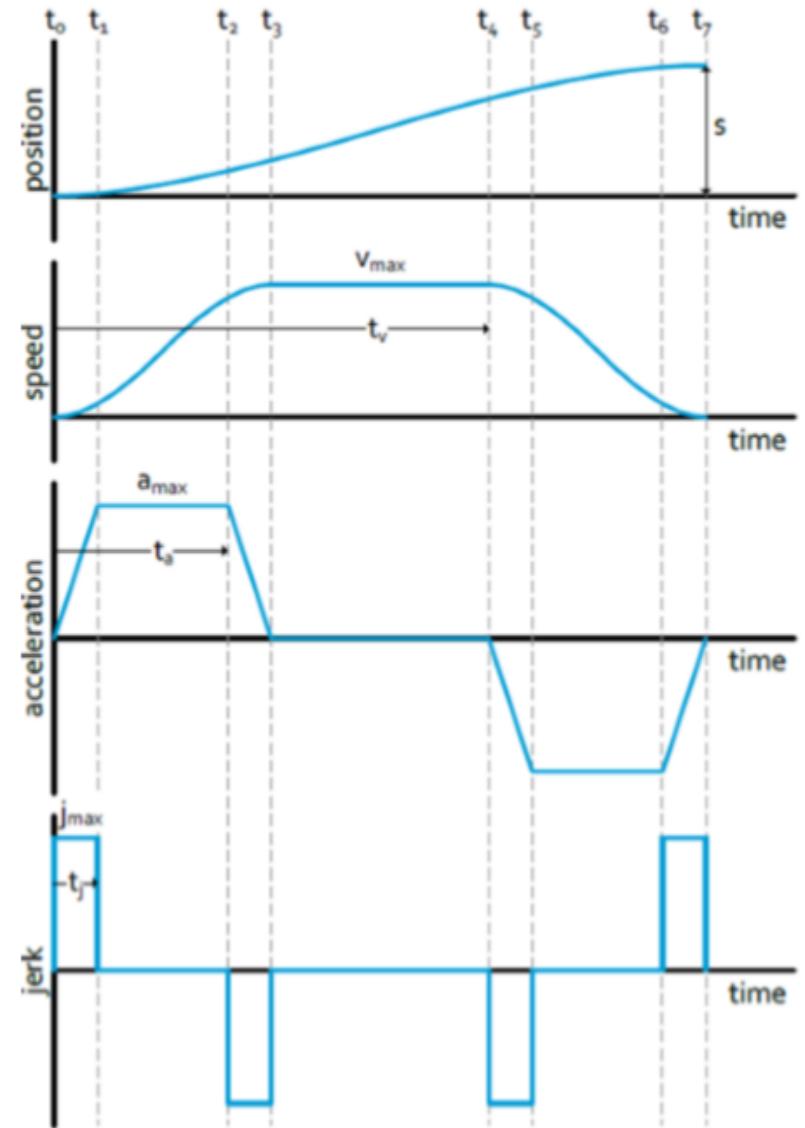
- (1) constant jerk  $d^3s/dt^3 = J$ ;
- (2) constant acceleration  $s = a$ ;
- (3) constant negative jerk  $-J$ ;
- (4) coasting at constant  $v$ ;
- (5) constant negative jerk;
- (6) constant deceleration;
- (7) constant positive jerk

S-curve is divided into seven segments. We use time segments to represent the curve:  $t_1, t_2, t_3, t_4, t_5, t_6, t_7$ .

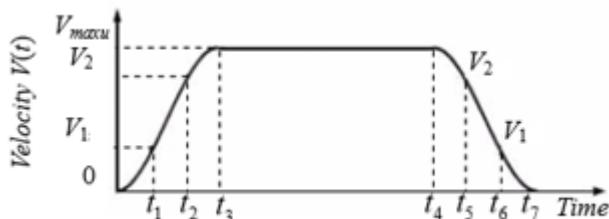
We have to give several parameters before giving the S-curve.

- 1) Displacement  $X_{max}$
- 2) Maximum speed  $v_{max}$
- 3) Maximum acceleration  $a_{max}$
- 4) Maximum jerk  $J_{max}$
- 5) Seven time periods

Objective: plot position X, speed V and acceleration A profiles of each time segment.



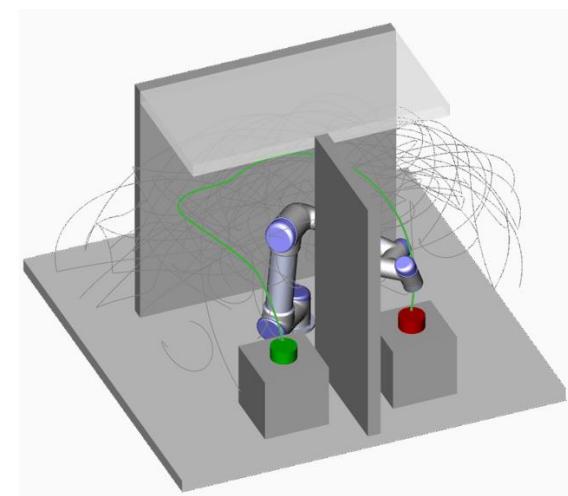
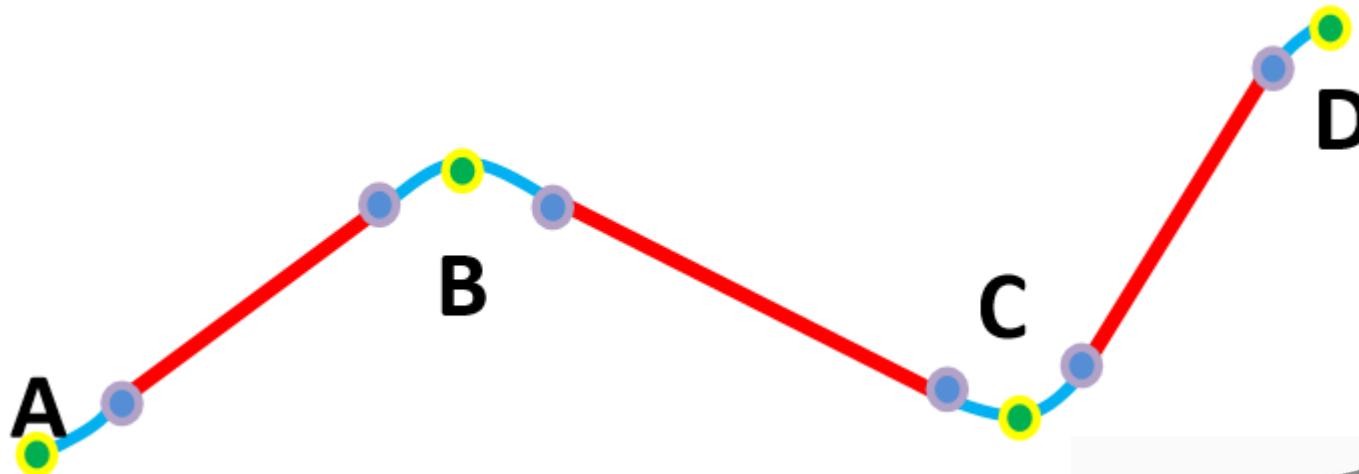
$$v_3 = v_{max}$$



### Motion parameters

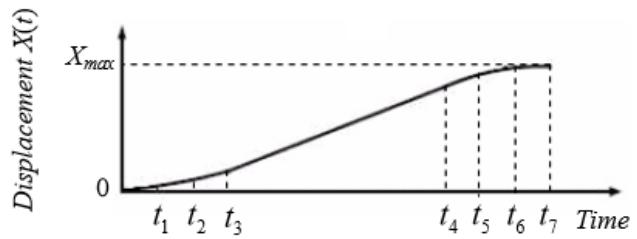
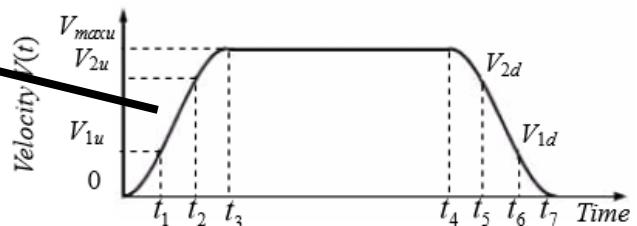
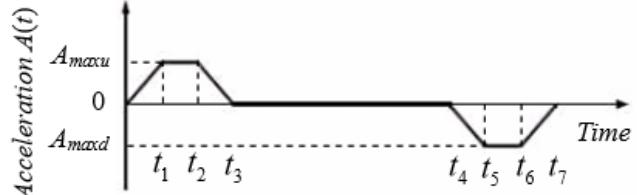
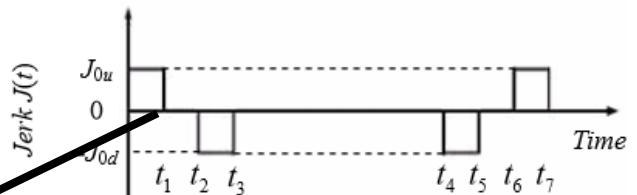
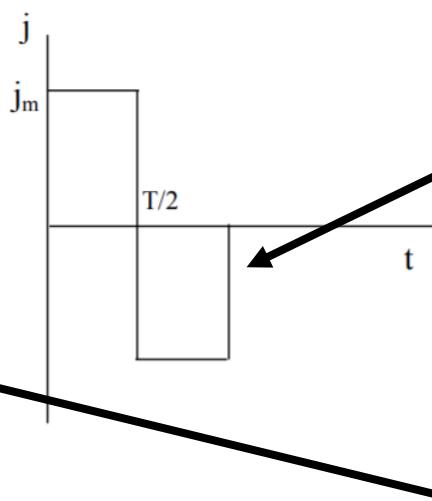
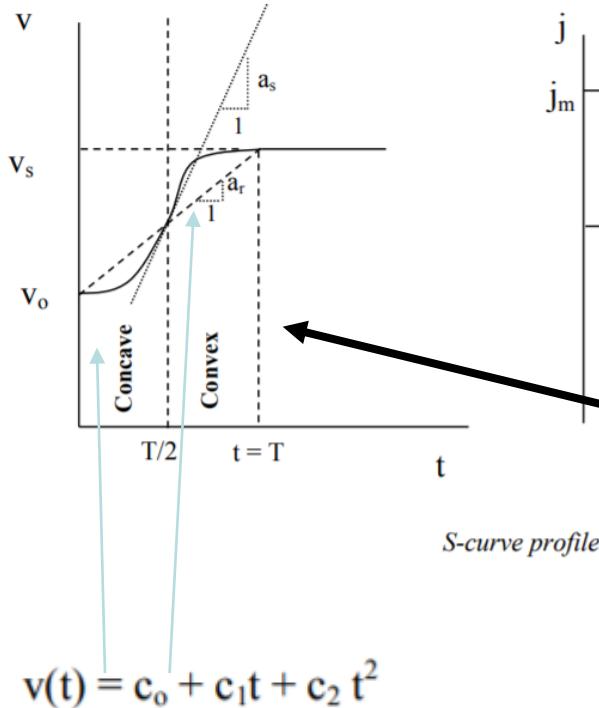
jerk	acceleration	velocity	position
$t_0 \dots t_1$	$j_{max}$	$j_{max} \cdot (t - t_0)$	$\frac{1}{2} \cdot j_{max} \cdot (t - t_0)^2$
$t_1 \dots t_2$	0	$a_1 = a_2$	$v_1 + a_1 \cdot (t - t_1)$
$t_2 \dots t_3$	$-j_{max}$	$a_2 - j_{max} \cdot (t - t_2)$	$v_2 + a_2 \cdot (t - t_2) + \frac{1}{2} \cdot -j_{max} \cdot (t - t_2)^2$
$t_3 \dots t_4$	0	0	$p_2 + v_2 \cdot (t - t_2) + \frac{1}{2} \cdot a_2 \cdot (t - t_2)^2 + \frac{1}{6} \cdot -j_{max} \cdot (t - t_2)^3$
$t_4 \dots t_5$	$-j_{max}$	$-j_{max} \cdot (t - t_4)$	$v_3 = v_4$
$t_5 \dots t_6$	0	$a_5 = a_6$	$p_4 + v_4 \cdot (t - t_4) + \frac{1}{6} \cdot -j_{max} \cdot (t - t_4)^3$
$t_6 \dots t_7$	$j_{max}$	$a_6 + j_{max} \cdot (t - t_6)$	$v_5 - a_{max} \cdot (t - t_5)$
$t_1 = t_j$		$t_2 = t_a$	$v_6 + a_6 \cdot (t - t_6) + \frac{1}{2} \cdot j_{max} \cdot (t - t_6)^2$
$t_3 = t_a + t_j$		$t_4 = t_v$	$p_5 + v_5 \cdot (t - t_5) + \frac{1}{2} \cdot a_5 \cdot (t - t_5)^2$
$t_5 = t_v + t_j$		$t_6 = t_v + t_a$	$p_6 + v_6 \cdot (t - t_6) + \frac{1}{2} \cdot a_6 \cdot (t - t_6)^2 + \frac{1}{6} \cdot j_{max} \cdot (t - t_6)^3$
$t_7 = t_v + t_j + t_a$			

S-curve can generate a trajectory profile as follows



**Limitations:** Complex expression;  
cannot handle other constraints,  
for example, snap and environment

### 3. Trajectory represented by polynomial



One polynomial or multiple polynomials?

### 3.1 Trajectory represented by one polynomial

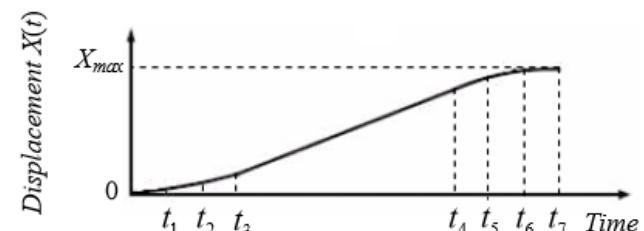
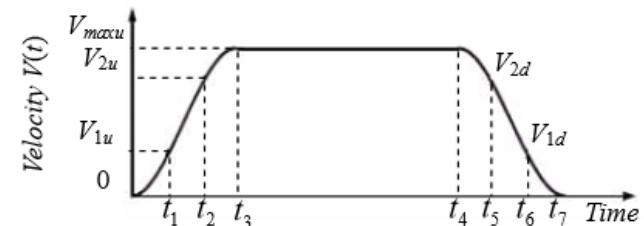
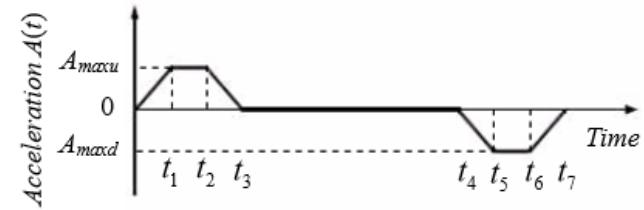
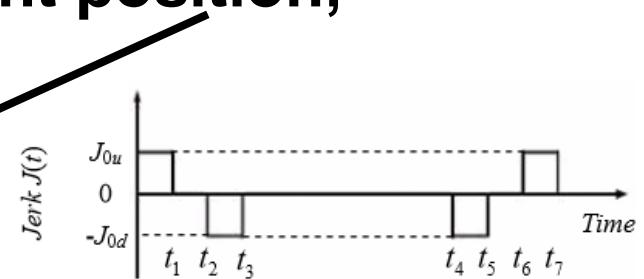
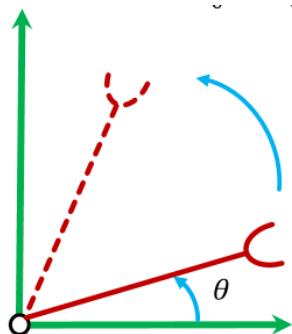
Can we use one polynomial to represent position, velocity and acceleration profiles?

$$\theta(t) = a_0 + a_1 t + a_2 t^2 + a_3 t^3$$

Joint velocity and acceleration along this path are

$$\dot{\theta}(t) = a_1 + 2a_2 t + 3a_3 t^2$$

$$\ddot{\theta}(t) = 2a_2 + 6a_3 t$$



## Question: how do we get coefficients?

$$\theta(t) = a_0 + a_1 t + a_2 t^2 + a_3 t^3$$

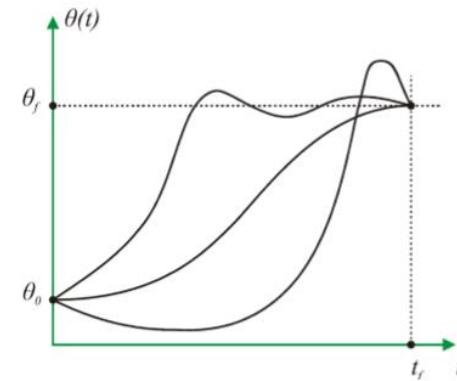
$$\dot{\theta}(t) = a_1 + 2a_2 t + 3a_3 t^2$$

Initial and terminal conditions

- Initial value  $\theta(0) = \theta_0$ ,
- Final value  $\theta(T) = \theta_f$ ,
- Initial velocity  $\dot{\theta}(0) = 0$ ,
- Final velocity  $\dot{\theta}(T) = 0$ .

Boundary conditions:

	Position	Velocity
$t = 0$	$a$	0
$t = T$	$b$	0



These **four constraints** can be satisfied by a polynomial of at least third degree (**four coefficients**)

## Example:

A single-link robot with a rotary joint is motionless at  $\theta = 15$  degrees. It is desired to move the joint in a smooth manner to  $\theta = 75$  degrees in 3 seconds. Find the coefficients of a cubic that accomplishes this motion and brings the manipulator to rest at the goal. Plot the position, velocity, and acceleration of the joint as a function of time.

$$\theta(0) = 15.0, \quad T=3, \quad \theta(3) = 75$$

$$\dot{\theta}(0) = a_1 = 0$$

$$\ddot{\theta}(0) = a_2 + 2a_3 \cdot 3 + 3a_4 \cdot 3^2 = 0 \Rightarrow 9a_2 + 27a_3 = 60$$

$$\dot{\theta}(3) = a_1 + 2a_2 \cdot 3 + 3a_3 \cdot 3^2 = 0 \Rightarrow a_2 = -\frac{9}{2}a_3$$

$$a_0 = 15.0, \quad a_1 = 0.0, \quad a_2 = 20.0, \quad a_3 = -4.44.$$

$$\theta(t) = 15.0 + 20.0t^2 - 4.44t^3,$$

$$\dot{\theta}(t) = 40.0t - 13.33t^2,$$

$$\ddot{\theta}(t) = 40.0 - 26.66t.$$

Boundary conditions:

	Position	Velocity
$t = 0$	$a$	0
$t = T$	$b$	0

$$\begin{bmatrix} a \\ b \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 1 \\ 0 & T^3 & T^2 & T \\ 0 & 0 & 1 & 0 \\ 0 & 3T^2 & 2T & 1 \end{bmatrix} \begin{bmatrix} a_3 \\ a_2 \\ a_1 \\ a_0 \end{bmatrix}$$

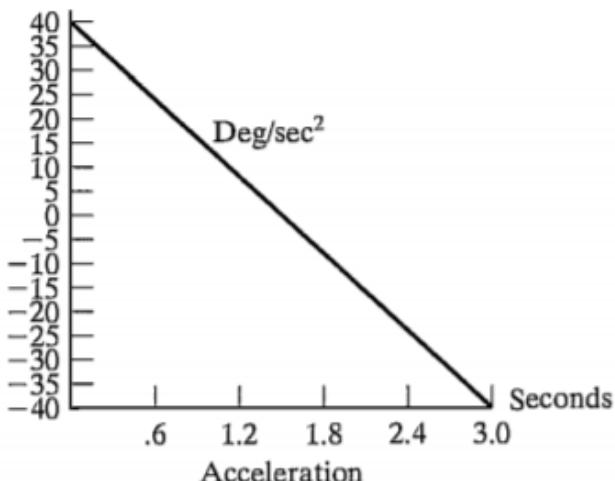
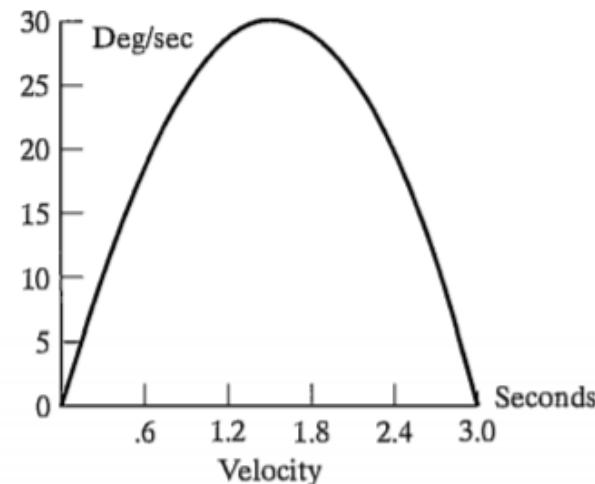
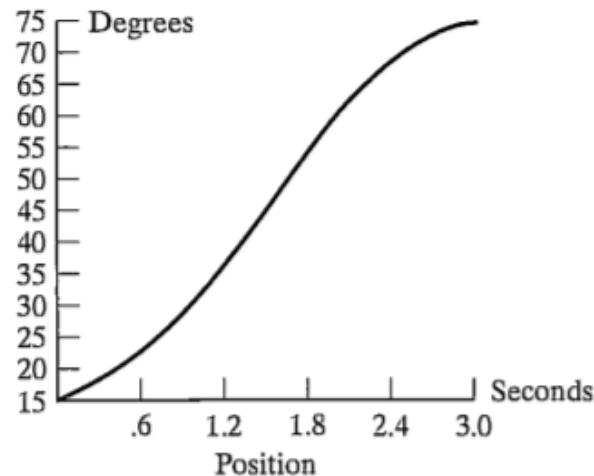


MATLAB ?

$$\theta(t) = 15.0 + 20.0t^2 - 4.44t^3,$$

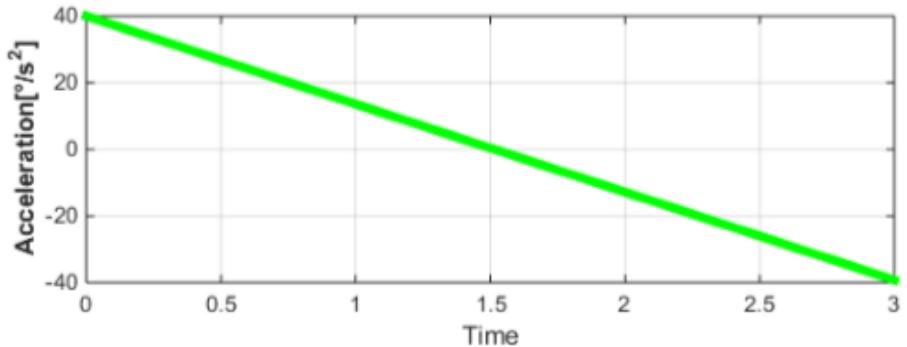
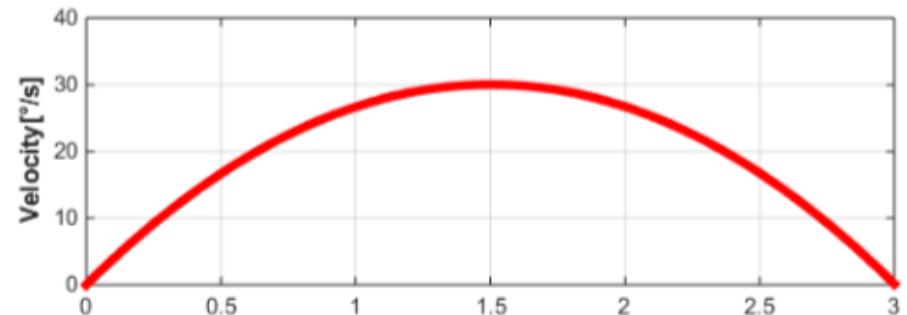
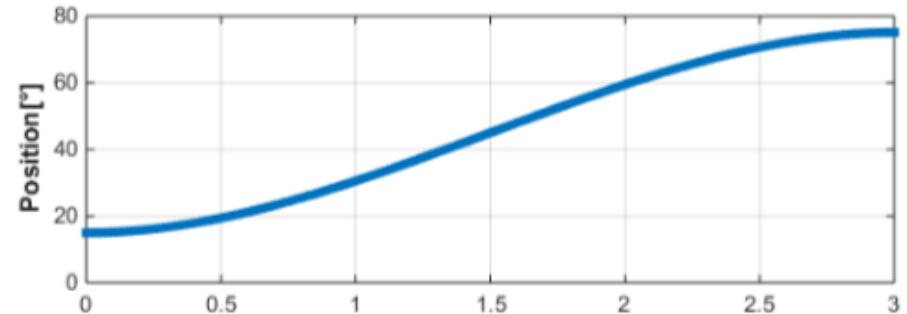
$$\dot{\theta}(t) = 40.0t - 13.33t^2,$$

$$\ddot{\theta}(t) = 40.0 - 26.66t.$$



# MATLAB program

```
2 -      t = 0:0.01:3;  
3 -  
4 -  
5 -      theta_t = 15 + 20.*t.*t - 4.44.*t.*t.*t;  
6 -      theta_d_t = 40.*t - 13.33.*t.*t;  
7 -      theta_2d_t = 40 - 26.6.*t;  
8 -  
9 -      subplot(3,1,1);  
10 -     plot(t,theta_t); grid on;  
11 -     ylabel('Position[°]');  
12 -  
13 -      subplot(3,1,2);  
14 -     plot(t,theta_d_t); grid on;  
15 -     ylabel('Velocity[°/s]');  
16 -  
17 -      subplot(3,1,3);  
18 -     plot(t,theta_2d_t); grid on;  
19 -     ylabel('Acceleration[°/s^2]');  
20 -  
21 -
```



# Summary:

For a third-order polynomial, we have four coefficients. Thus, we need to have four boundary conditions, i.e., four equations for getting the values of coefficients.

$$\theta(t) = a_0 + a_1 t + a_2 t^2 + a_3 t^3$$

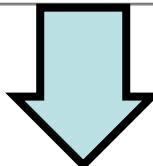
$$\dot{\theta}(t) = a_1 + 2a_2 t + 3a_3 t^2$$

Boundary conditions:



	Position	Velocity
$t = 0$	$a$	0
$t = T$	$b$	0

List equations



$$\begin{bmatrix} a \\ b \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 & 1 \\ T^3 & T^2 & T & 1 \\ 0 & 0 & 1 & 0 \\ 3T^2 & 2T & 1 & 0 \end{bmatrix} \begin{bmatrix} a_3 \\ a_2 \\ a_1 \\ a_0 \end{bmatrix}$$

?

**Question:** can we use a polynomial more than 3<sup>rd</sup> order degree?

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

Now we have six coefficients.

But the four boundary constraints are not enough for obtaining all coefficients. In this case, we need to have more boundary conditions.

Can we use a simpler one, for example, first order ?

**Higher order polynomial model**  $x(t) = p_0 + p_1t + p_2t^2 + \dots + p_nt^n = \sum_{i=0}^n p_i t^i$

Position  $x(t) = p_0 + p_1t + p_2t^2 + p_3t^3 + p_4t^4 + p_5t^5 = \sum_{i=0}^5 p_i t^i$

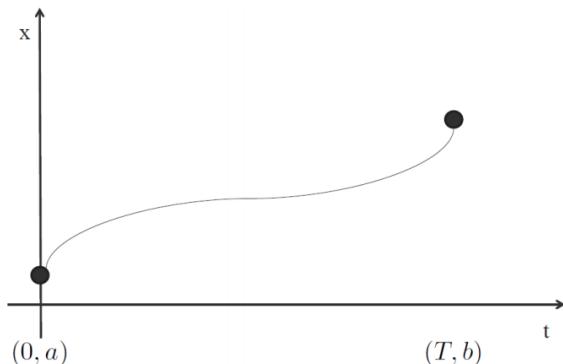
Speed  $\dot{x}(t) = p_1 + 2p_2t + 3p_3t^2 + 4p_4t^3 + 5p_5t^4$

How many parameters?

Acceleration  $\ddot{x}(t) = 2p_2 + 6p_3t + 12p_4t^2 + 20p_5t^3$

### Boundary condition

	Position	Velocity	Acceleration
$t = 0$	$a$	0	0
$t = T$	$b$	0	0



$$\begin{bmatrix} a \\ b \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 1 \\ T^5 & T^4 & T^3 & T^2 & T & 1 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 5T^4 & 4T^3 & 3T^2 & 2T & 1 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 20T^3 & 12T^2 & 6T & 2 & 0 & 0 \end{bmatrix} \begin{bmatrix} p_5 \\ p_4 \\ p_3 \\ p_2 \\ p_1 \\ p_0 \end{bmatrix}$$

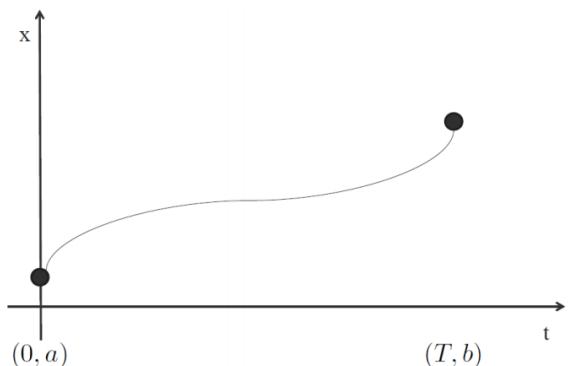
**Six coefficients have six boundary conditions. We have six linear equations.**

## Higher order polynomial model

$$x(t) = p_0 + p_1 t + p_2 t^2 + \cdots + p_n t^n = \sum_{i=0}^n p_i t^i \quad (n = 5)$$

### Boundary condition

	<i>Position</i>	<i>Velocity</i>	<i>Acceleration</i>
$t = 0$	$a$	$v_0$	$a_0$
$t = T$	$b$	$v_T$	$a_T$



$$\begin{bmatrix} a \\ b \\ v_0 \\ v_T \\ a_0 \\ a_T \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 1 \\ T^5 & T^4 & T^3 & T^2 & T & 1 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 5T^4 & 4T^3 & 3T^2 & 2T & 1 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 20T^3 & 12T^2 & 6T & 2 & 0 & 0 \end{bmatrix} \begin{bmatrix} p_5 \\ p_4 \\ p_3 \\ p_2 \\ p_1 \\ p_0 \end{bmatrix}$$

## Summary:

**For 5<sup>th</sup> order polynomial, we have boundary conditions**

1. Initial Position
2. Final Position
3. Initial Velocity
4. Final Velocity
5. Initial Acceleration
6. Final Acceleration

**Continuous, smooth and comfortable  
How about 7<sup>th</sup> order polynomial?**

# Higher order polynomial model

$$x(t) = p_0 + p_1 t + p_2 t^2 + \cdots + p_n t^n = \sum_{i=0}^n p_i t^i$$

**where there are n+1 coefficients. Define coefficient vector**

$$\mathbf{p} = [p_0, p_1, \dots, p_n]^T$$

**Then, position trajectory can be expressed as**

$$\mathbf{p}(t) = [1, t, t^2, \dots, t^n] \cdot \mathbf{p}$$

**We also obtain velocity, acceleration , jerk and snap trajectories**

$$v(t) = \mathbf{p}'(t) = [0, 1, 2t, 3t^2, 4t^3, \dots, nt^{n-1}] \cdot \mathbf{p}$$

$$a(t) = \mathbf{p}''(t) = [0, 0, 2, 6t, 12t^2, \dots, n(n-1)t^{n-2}] \cdot \mathbf{p}$$

$$jerk(t) = \mathbf{p}^{(3)}(t) = [0, 0, 0, 6, 24t, \dots, \frac{n!}{(n-3!)} t^{n-3}] \cdot \mathbf{p}$$

$$snap(t) = \mathbf{p}^{(4)}(t) = [0, 0, 0, 0, 24, \dots, \frac{n!}{(n-4!)} t^{n-4}] \cdot \mathbf{p}$$

# More coefficients, more boundary conditions.

For example, for  $n+1$  coefficients, we need to have  $n+1$  boundary conditions, i.e., equation constraints

	position	speed	acceleration	jerk	snap	.....
t=0	a	$v_0$	$a_0$	$j_0$	$s_0$	.....
t=T	b	$v_T$	$a_T$	$j_T$	$s_T$	.....

Remember to give the time first!!!

Total distance/average speed

# Calculate parameters and conditions, depending on the polynomial you selected

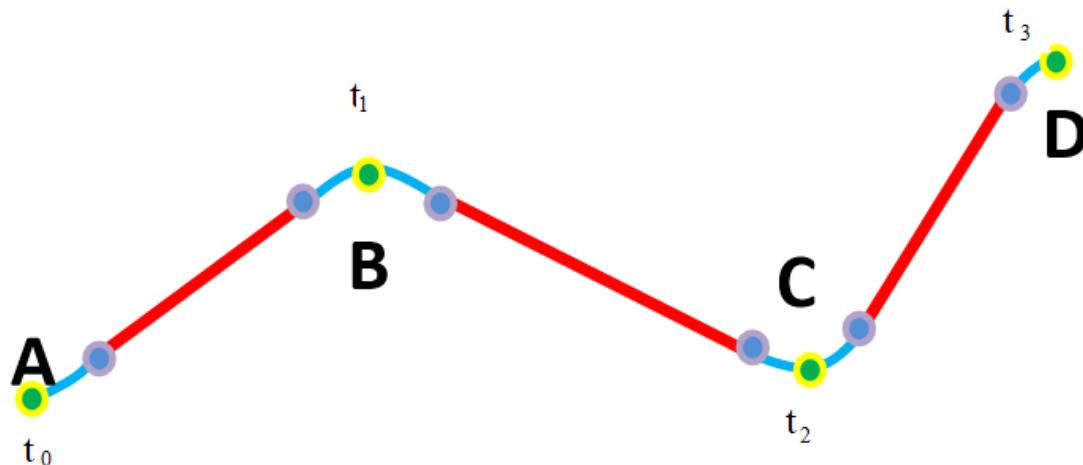
- Initial and terminal points: position, velocity, acceleration, and jerk, i.e., pvaj

4(initial)+4(terminal)

## 3.2 Multiple polynomials

From A->B, we can use one polynomial. How about A→B→C→D? Can we use one polynomial to express it?

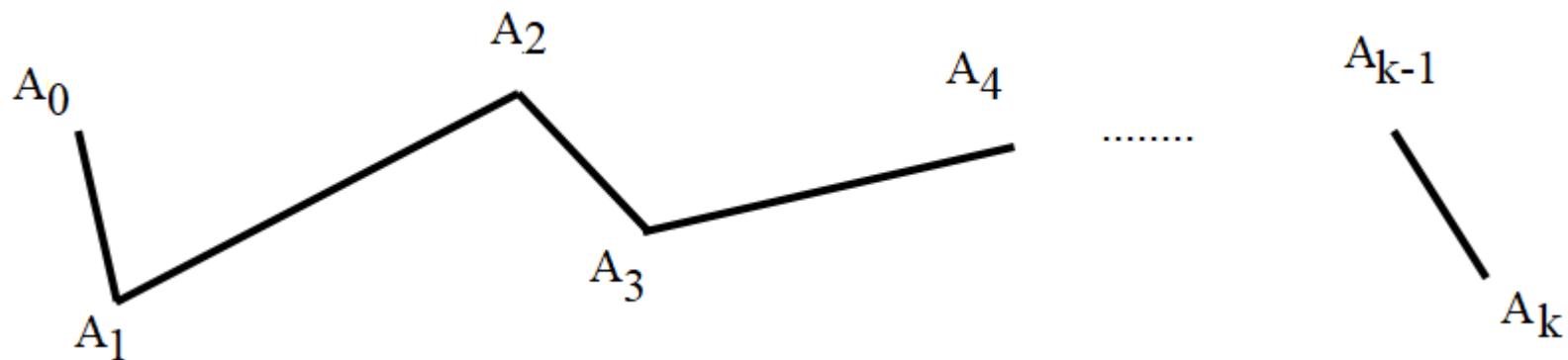
Usually, we use multiple polynomials



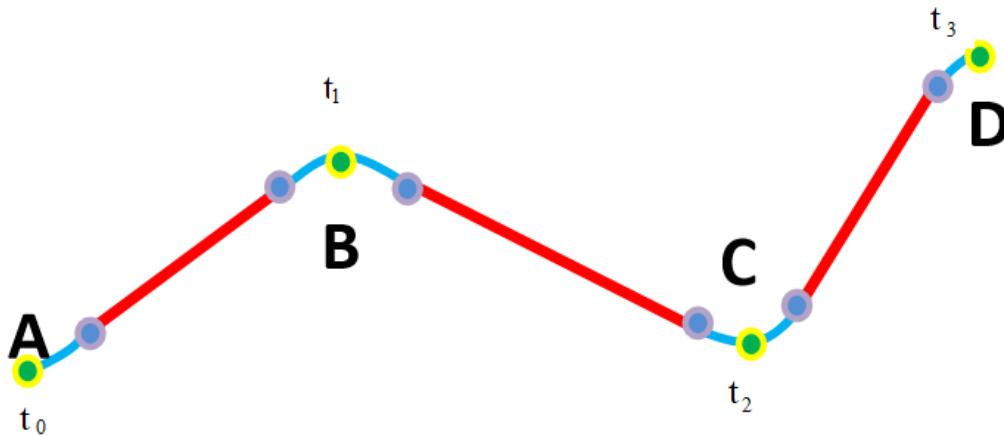
$$p(t) = \begin{cases} [1, t, t^2, \dots, t^n] \cdot p_1 & t_0 \leq t < t_1 \\ [1, t, t^2, \dots, t^n] \cdot p_2 & t_1 \leq t < t_2 \\ [1, t, t^2, \dots, t^n] \cdot p_3 & t_2 \leq t < t_3 \end{cases}$$

# For k segments, we have k polynomials

$$p(t) = \begin{cases} [1, t, t^2, \dots, t^n] \cdot p_1 & t_0 \leq t < t_1 \\ [1, t, t^2, \dots, t^n] \cdot p_2 & t_1 \leq t < t_2 \\ \dots \\ [1, t, t^2, \dots, t^n] \cdot p_k & t_{k-1} \leq t < t_k \end{cases}$$



# You may argue with me. Why we need to learn the multiple polynomials?



**A→B    B→C, C→D, use the same  
boundary condition   v=0, a=0.**

Sometimes, this may take longer time to  
reach target or exceed the limits.

**1) Time intervals: we know the time interval at each segment.**

**2) Boundary conditions**

- **Initial values**
- **Terminal values**
- **For the relationship between two adjacent segments, the continuity of position, velocity, acceleration, and jerk can also be formulated as equality constraint equation, i.e.,  $p v a j$  must be equal**

# For simplicity, all polynomials should have the same order.

Initial values  
Terminal values

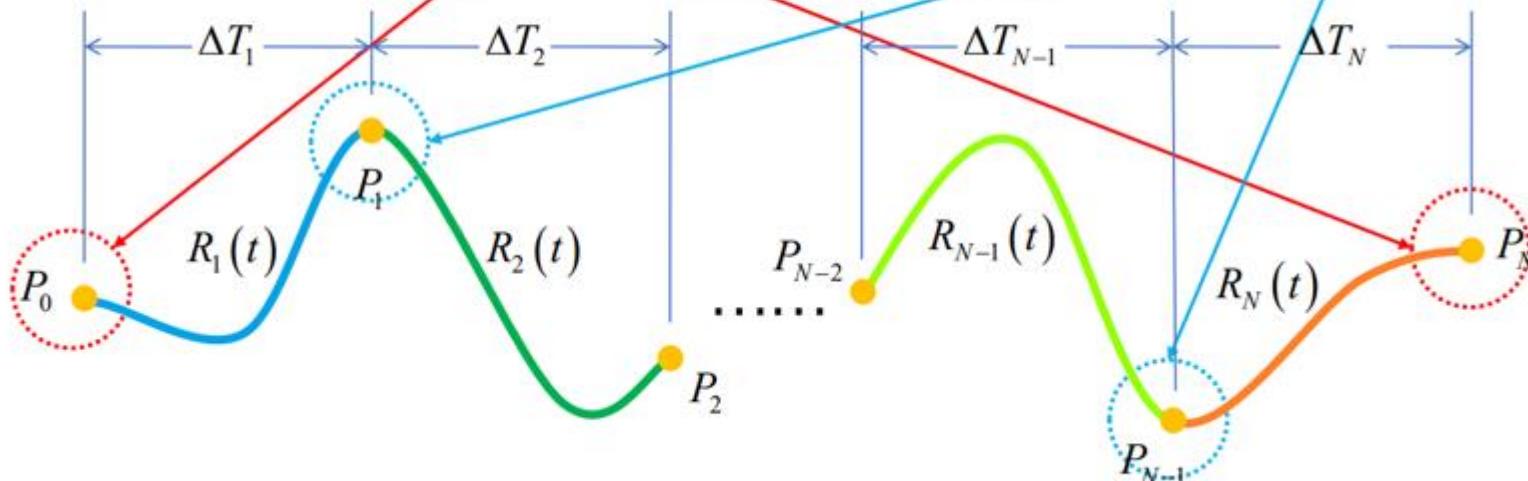
For the relationship between 2 adjacent segments, the continuity of position, velocity and acceleration can also be formulated as equality constraint equation.

- Derivative constraints:

$$\begin{cases} f_j^{(k)}(T_0) = x_{0,j}^{(k)} \\ f_j^{(k)}(T_j) = x_{T,j}^{(k)} \end{cases}$$

- Continuity constraints:

$$f_j^{(k)}(T_j) = f_{j+1}^{(k)}(T_j)$$



For example, continuity constraints:  $[1, t_1, t_1^2, \dots, t_1^n] \cdot p_1 = [1, t_1, t_1^2, \dots, t_1^n] \cdot p_2 \quad t = t_1$

$\Rightarrow [1, t_1, t_1^2, \dots, t_1^n] \cdot p_1 - [1, t_1, t_1^2, \dots, t_1^n] \cdot p_2 = 0 \quad t = t_1$

- **Derivative constraints**

$$\text{position} : [1, t_0, t_0^2, \dots, t_0^n, \underbrace{0\dots0}_{(k-1)(n+1)}]p = p_0$$

$$\text{velocity} : [0, 1, 2t_0, \dots, nt_0^{n-1}, \underbrace{0\dots0}_{(k-1)(n+1)}]p = v_0$$

$$\text{acceleration} : [0, 0, 2, \dots, n(n-1)t_0^{n-2}, \underbrace{0\dots0}_{(k-1)(n+1)}]p = a_0$$

$$\text{Final position} : [1, t_T, t_T^2, \dots, t_T^n, \underbrace{0\dots0}_{(k-1)(n+1)}]p = p_T$$

$$\text{Final velocity} : [0, 1, 2t_T, \dots, nt_T^{n-1}, \underbrace{0\dots0}_{(k-1)(n+1)}]p = v_T$$

$$\text{Final acceleration} : [0, 0, 2, \dots, n(n-1)t_T^{n-2}, \underbrace{0\dots0}_{(k-1)(n+1)}]p = a_T$$

- **Continuity constraints**

$$[\underbrace{0\dots0}_{(i-1)(n+1)}, 1, t_i, t_i^2, \dots, t_i^n, -1, -t_i, -t_i^2, \dots, -t_i^n, \underbrace{0\dots0}_{(k-i-1)(n+1)}]p = 0$$

## Combining equation constraints

$$\left[ \begin{array}{c}
 1, t_0, t_0^2, \dots, t_0^n, 0_{(k-1)(n-1)} \\
 0, 1, 2t_i, \dots, nt_i^{n-1}, 0_{(k-1)(n-1)} \\
 0, 0, 2, \dots, n(n-1)t_i^{n-2}, 0_{(k-1)(n-1)} \\
 \vdots \\
 0_{1(i-1)(n+1)}, 1, t_i, t_i^2, \dots, t_i^n, 0_{(k-i)(n+1)} \\
 \vdots \\
 0_{(k-1)(n+1)}, 1, t_k, t_k^2, \dots, t_k^n \\
 \vdots \\
 0_{(k-1)(n+1)}, 1, t_k, t_k^2, \dots, t_k^n \\
 0_{(k-1)(n+1)}, 0, 1, 2t_k, \dots, nt_k^{n-1} \\
 0_{(k-1)(n+1)}, 0, 0, 2, \dots, n(n-1)t_k^{n-2} \\
 0_{(i-1)(n+1)}, 1, t_i, t_i^2, \dots, t_i^n, -1, -t_i, -t_i^2, \dots, -t_i^n, 0_{(k-i-1)(n+1)} \\
 0_{(i-1)(n+1)}, 0, 1, 2t_i, \dots, nt_i^{n-1}, 0, -1, -2t_i, \dots, -nt_i^{n-1}, 0_{(k-i-1)(n+1)} \\
 \\ 
 0_{(i-1)(n+1)+2}, 2, \dots, \frac{n!}{(n-2)!} t_i^{n-2}, 0, 0, -2, \dots, -\frac{n!}{(n-2)!} t_i^{n-2}, 0_{(k-i-1)(n+1)}
 \end{array} \right] p = \begin{bmatrix} p_0 \\ v_0 \\ a_0 \\ \vdots \\ p_i \\ \vdots \\ p_k \\ v_k \\ a_k \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

Derivative constraints

Continuity constraints

$\downarrow$

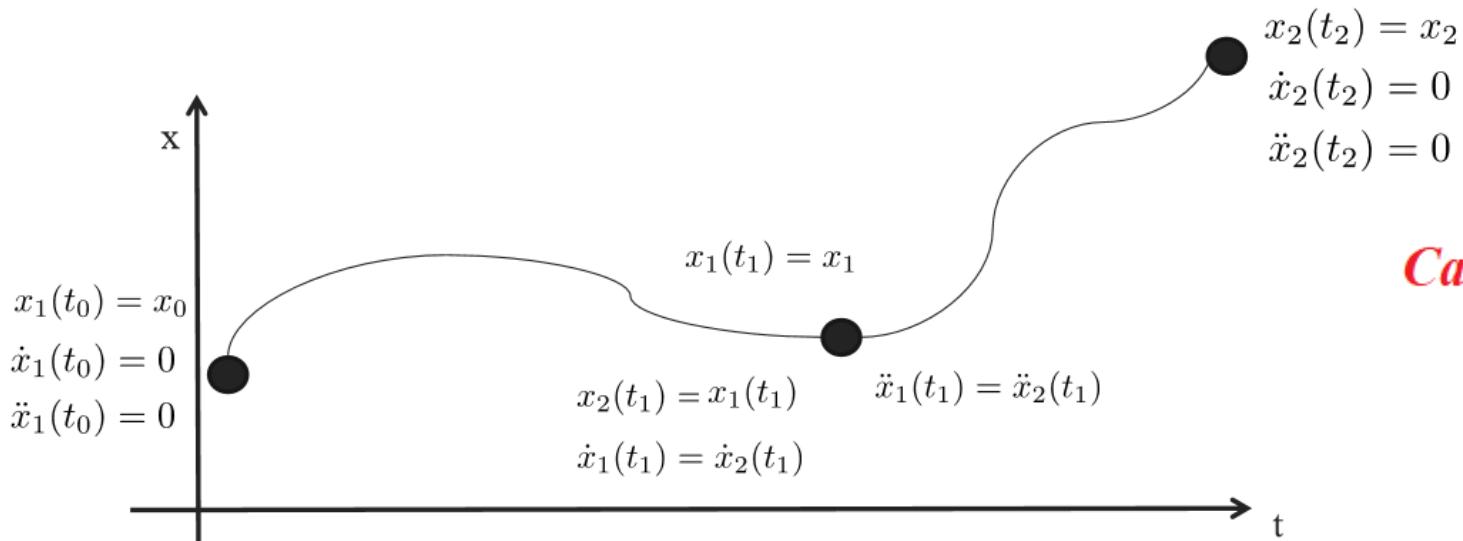
$A_{eq}p = b_{eq}$

## Estimate number of equation constraints

- Equation constraints=3 (initial pva) + $k-1$  (middle p) +3(terminal pva) +3( $k-1$ ) (middle continuity pva ) = $4k+2$ , where k represents polynomials (k segments)
- This implies that we can solve  $4k+2$  parameters.

$$k=2 \rightarrow \text{equation constraints} = 3(\text{initial}) + 3(\text{terminal}) + 1(k-1) + 3 \times 1 = 10$$

*Two segments*

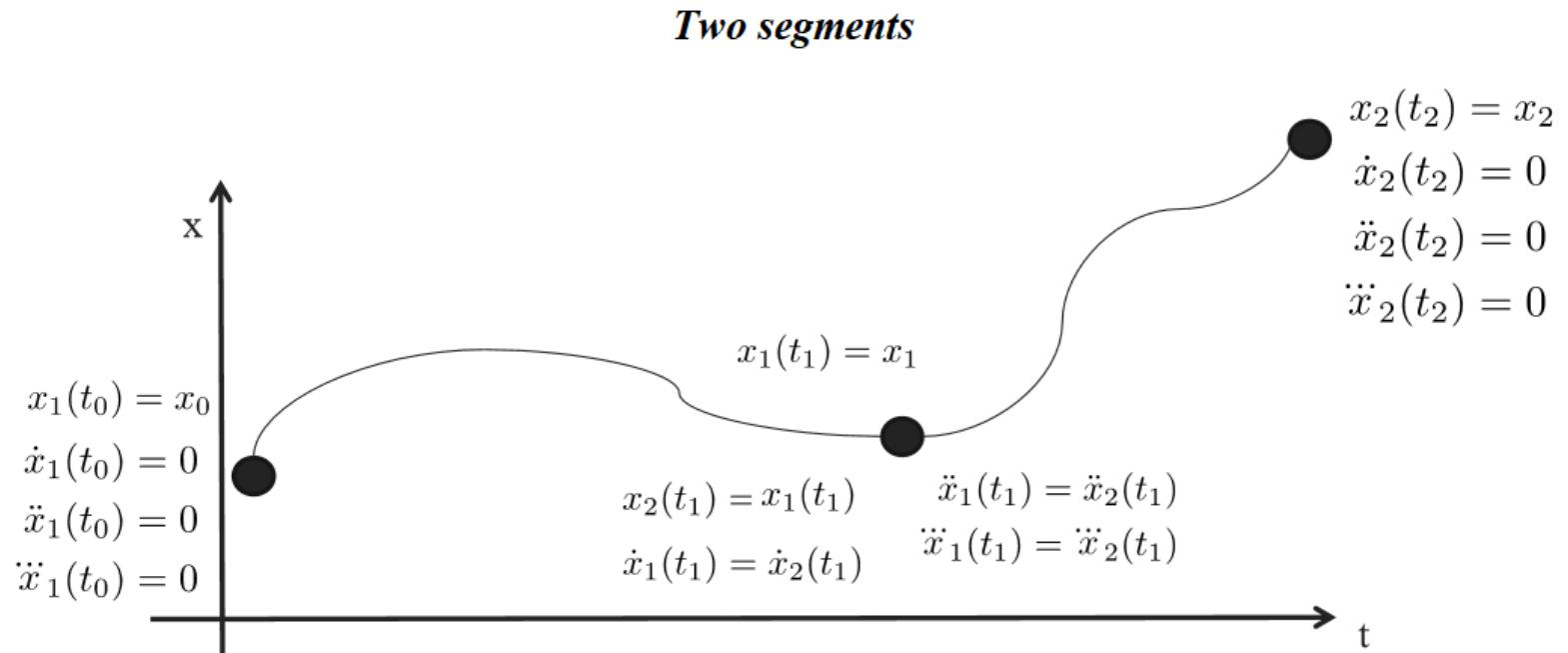


*Can we express like*

$$\begin{aligned}x_1(t_1) &= x_1 \\x_2(t_1) &= x_1\end{aligned}$$

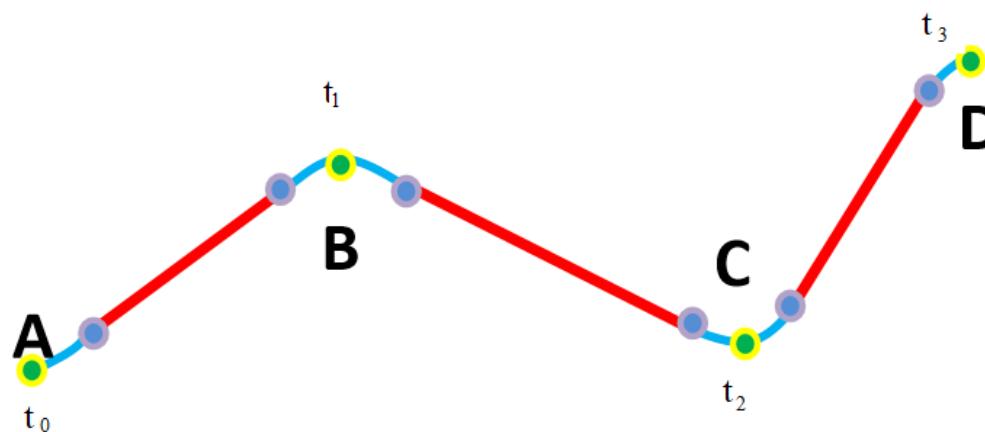
# If we consider pvaj, how many number of equation constraints?

- equation constraints=4+4+1+4=13



# Several issues in polynomial:

- Time is allocated according to the path distance. It is obtained by `totDist/avgSpd`
- Order number (we should have the same order number in all polynomials)
- Boundary conditions



# Limitations:

- **Boundary conditions are set to zero. It is too conservative. Sometimes this causes the velocity or acceleration exceeding the limits.**
- **You have to check if the velocity and acceleration or jerk exceed the limits.**

# 3.3 Trajectory optimization

- More constraints,  
position and velocity

$$x_{min} \leq x \leq x_{max}$$

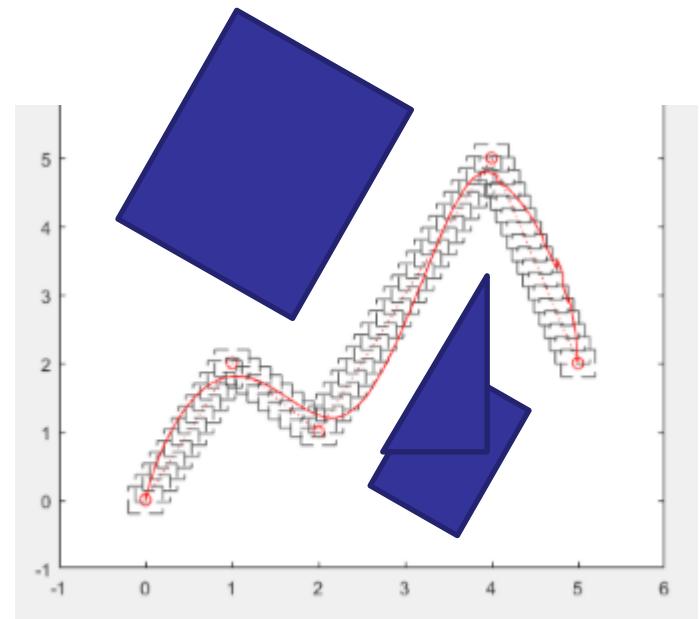
$$v_{min} \leq v \leq v_{max}$$

- Objective optimizations

*minimum snap* :  $\min f(p) = \min \int (p^{(4)}(t))^2 dt$

*minimum jerk* :  $\min f(p) = \min \int (p^{(3)}(t))^2 dt$

*minimum acce* :  $\min f(p) = \min \int (p^{(2)}(t))^2 dt$



Minimum snap result with corridor

?

# Key point:

The problem is formulated as following

$$\min f(p) \quad \longleftarrow \quad \text{Acceleration , jerk or snap}$$

$$s. t. \quad A_{eq}p = b_{eq}, \quad \longleftarrow \quad \text{Equation constraints}$$

$$A_{ieq}p \leq b_{ieq} \quad \longleftarrow \quad \text{Inequality constraints}$$

This is quadratic programming (QP)

Boundary conditions  Equation constraints

$$x_{min} \leq x \leq x_{max}$$

$$v_{min} \leq v \leq v_{max}$$

 Inequality constraints

# In this module,

- We will not concentrate on optimization method. This is beyond this module.
- The main purpose of QP is to use it in trajectory generation.

# A QP example

This example demonstrates a specific QP problem:

$$\begin{aligned} \min f(x) &= 3x_1^2 + x_2^2 + 2x_1x_2 + x_1 + 6x_2 + 2 \\ \text{s.t. } &2x_1 + 3x_2 \geq 4 \\ &x_1, x_2 \geq 0 \end{aligned}$$

Rewrite the constraints.

$$\begin{aligned} g_1 &= -2x_1 - 3x_2 + 4 \leq 0 \\ g_2 &= -x_1 \leq 0 \\ g_3 &= -x_2 \leq 0 \end{aligned}$$

# For example, we consider to minimizing jerk, where 5<sup>th</sup> order polynomial

$$s(t) \in [0, t_f] \longrightarrow s(t) = \alpha_0 + \alpha_1 t + \alpha_2 t^2 + \dots + \alpha_n t^n + \dots$$

$$\text{Jerk} = \ddot{s}(t)$$

$$\text{Total (squared) Jerk} = \int_0^{t_f} \ddot{s}(t)^2 dt$$

Minimize this!

$$s(t) = \sum_{n=0}^{\infty} \alpha_n t^n$$

Using THIS form of s

And incorporating THIS realization

$$\frac{d^m s}{dt^m} = 0 \quad \forall m \geq 6$$

$$\alpha_6, \alpha_7, \dots = 0$$

## MINIMUM 1D JERK TRAJECTORIES

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

# Minimum Snap (usually, 7<sup>th</sup> order )

$$\begin{aligned}
 & \min \int_0^T (p^{(4)}(t))^2 dt \\
 &= \min \sum_{i=1}^k \int_{t_{i-1}}^{t_i} (p^{(4)}(t))^2 dt \\
 &= \min \sum_{i=1}^k \int_{t_{i-1}}^{t_i} \left( [0, 0, 0, 0, 24, \dots, \frac{n!}{(n-4)!} t^{n-4}] \cdot p \right)^T \left[ 0, 0, 0, 0, 24, \dots, \right. \\
 &\quad \left. \frac{n!}{(n-4)!} t^{n-4} \right] \cdot p dt \\
 &= \min \sum_{i=1}^k p^T \int_{t_{i-1}}^{t_i} \left[ 0, 0, 0, 0, 24, \dots, \frac{n!}{(n-4)!} t^{n-4} \right]^T \left[ 0, 0, 0, 0, 24, \dots, \right. \\
 &\quad \left. \frac{n!}{(n-4)!} t^{n-4} \right] dt p
 \end{aligned}$$

# Minimum Snap

$$\min \int_0^T (p^{(4)}(t))^2 dt = \min \sum_{i=1}^k p^T Q_i p$$

**where**

$$\begin{aligned}
 Q_i &= \int_{t_{i-1}}^{t_i} [0, 0, 0, 0, 24, \dots, \frac{n!}{(n-4!)} t^{n-4}]^T [0, 0, 0, 0, 24, \dots, \frac{n!}{(n-4!)} t^{n-4}] dt \\
 &= \begin{bmatrix} 0_{4 \times 4} & 0_{4 \times (n-3)} \\ 0_{(n-3) \times 4} & \frac{r!}{(r-4)!} \frac{c!}{(c-4)!} \frac{1}{(r-4)+(c-4)+1} (t_i^{(r+c-7)} - t_{i-1}^{(r+c-7)}) \end{bmatrix}
 \end{aligned}$$

$$Q = \begin{bmatrix} Q_1 & & & \\ & Q_2 & & \\ & & \ddots & \\ & & & Q_k \end{bmatrix}$$

# Minimum Snap

$$\min \int_0^T (p^{(4)}(t))^2 dt = \min p^T Q p$$

$$Q = \begin{bmatrix} Q_1 & & & \\ & Q_2 & & \\ & & \ddots & \\ & & & Q_k \end{bmatrix}$$

## Equation Constraints:

$$A_{eq}p = b_{eq}$$

# For inequality constraints, it is more complex.

We give one example

$$A_{ieq} p \leq b_{ieq}$$

position constraint

$$\mathbf{x}_{min} \leq [1, t_1, t_1^2, \dots, t_1^n, \underbrace{0 \dots 0}_{(k-1)(n+1)}] p \leq \mathbf{x}_{max}$$

$$[1, t_1, t_1^2, \dots, t_1^n, \underbrace{0 \dots 0}_{(k-1)(n+1)}] p \leq \mathbf{x}_{max}$$



$$\begin{bmatrix} 1, & t_1, & t_1^2, \dots, & t_1^n, & 0 \dots 0 \\ -1, & -t_1, & -t_1^2, \dots, & -t_1^n, & 0 \dots 0 \end{bmatrix} p \leq \begin{bmatrix} \mathbf{x}_{max} \\ -\mathbf{x}_{min} \end{bmatrix}$$

$$- [1, t_1, t_1^2, \dots, t_1^n, \underbrace{0 \dots 0}_{(k-1)(n+1)}] p \leq -\mathbf{x}_{min}$$

# Example:

**A vehicle has obtained five Waypoints from A\*,**

Waypts ( $x, y$ ) = [ (0, 0) ; (1, 2) ; (2, -1) ; (4, 8) ; (5, 2) ]. It is desire to move from the initial point (0,0) to the goal (5,2) in 5s. Please use minimum snap to find the polynomial and plot position, velocity and acceleration profiles. Boundary conditions

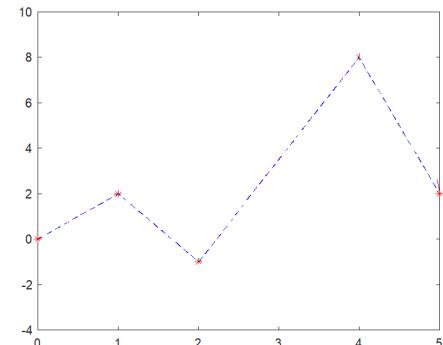
Initial:  $v_0 = [0, 0]; a_0 = [0, 0];$

Terminal:  $v_1 = [0, 0]; a_1 = [0, 0];$

We use 7th order polynomial and the following optimization

$$\min \int_0^T (p^{(4)}(t))^2 dt$$

$$A_{eq}p = b_{eq}$$



We use 7th order polynomials for each axis for each segment.

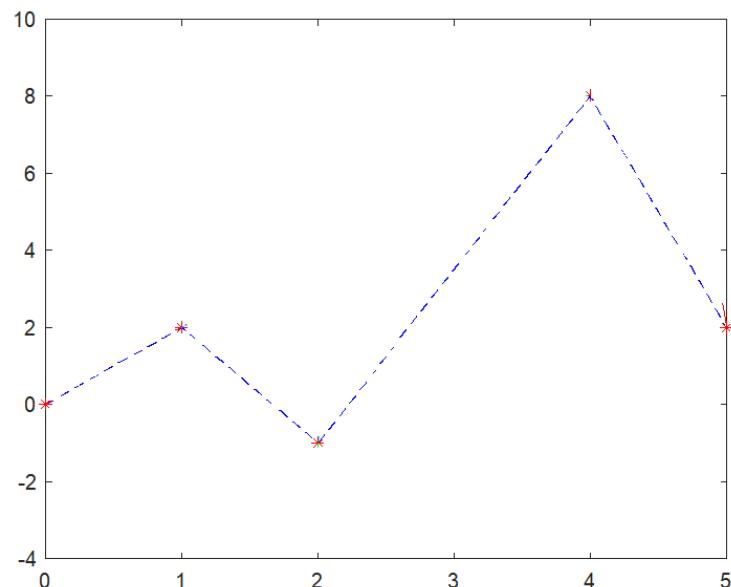
$$x(t) = a_0 + a_1 t + a_2 t^2 + a_3 t^3 + a_4 t^4 + a_5 t^5 + a_6 t^6 + a_7 t^7$$

$$y(t) = b_0 + b_1 t + b_2 t^2 + b_3 t^3 + b_4 t^4 + b_5 t^5 + b_6 t^6 + b_7 t^7$$

To minimize the following cost

$$\begin{aligned} & \min \int_0^T (p^{(4)}(t))^2 dt \\ &= \min \sum_{i=1}^4 \int_{t_{i-1}}^{t_i} \left( [0, 0, 0, 0, 24, \dots, \frac{7!}{(7-4)!} t^{7-4}] \cdot p \right)^T [0, 0, 0, 0, 24, \dots, \frac{7!}{(7-4)!} t^{7-4}] \cdot p dt \end{aligned}$$

$$A_{eq}p = b_{eq}$$



# Time intervals:

- 0 0.5401 1.3039 3.5308 5.0000

# Equation constraints:

6 boundary conditions,i.e.,pva

# Quadratic programming.

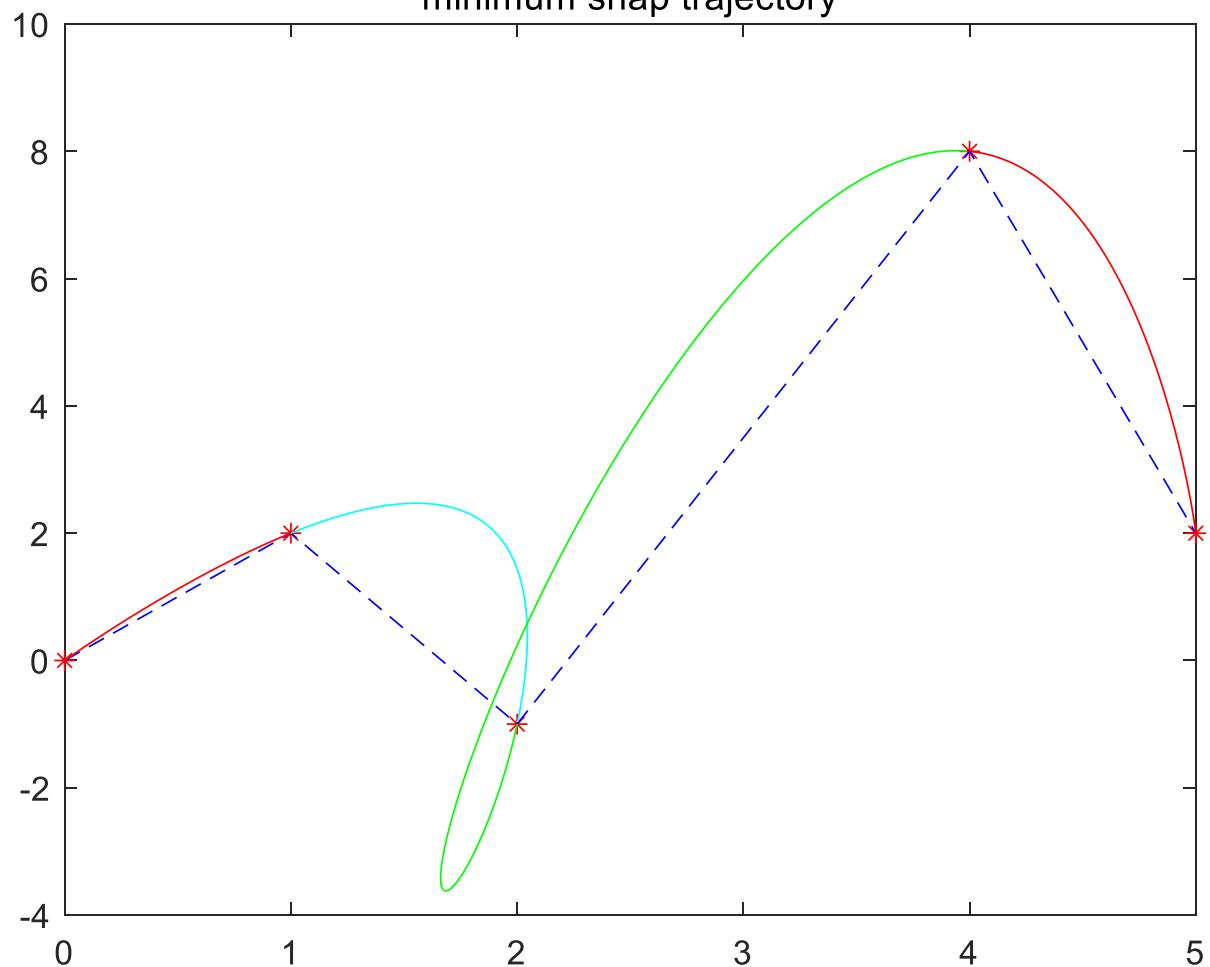
We use **MATLAB quadprog**

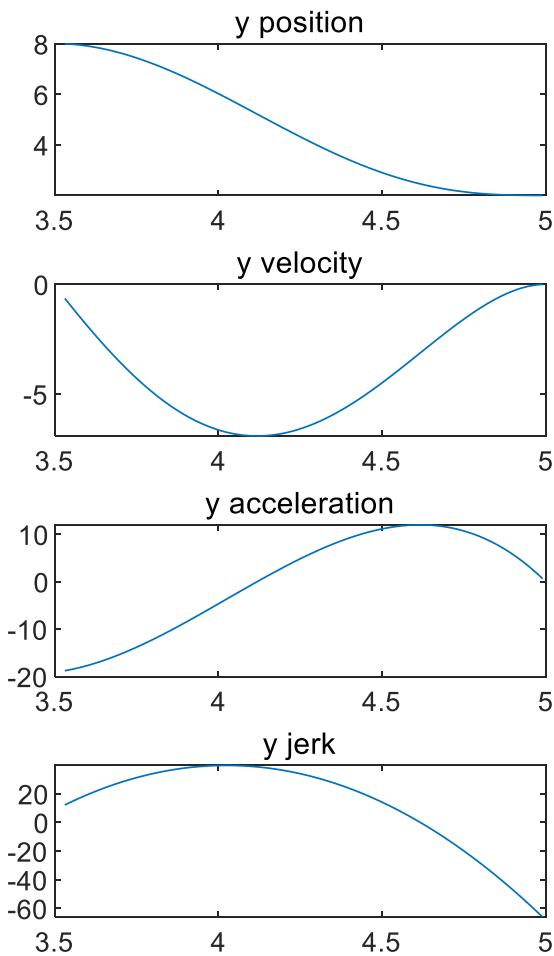
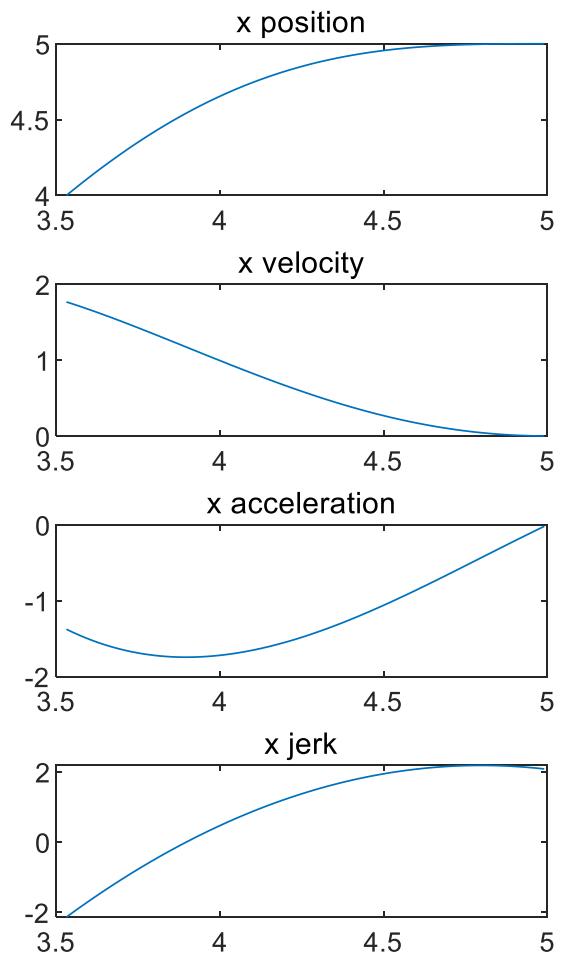
$$\min_{x} \frac{1}{2} x^T H x + f^T x \text{ such that } \begin{cases} A \cdot x \leq b, \\ Aeq \cdot x = beq, \end{cases}$$

$f=0$ ,  $A=[ ]$ ,  $b=[ ]$ , since we don't have inequality constraints

```
x = quadprog(H,f,A,b,Aeq,beq)
```

## minimum snap trajectory





# Comparison between polynomial method 3.2 and 3.3

- Method 3.2 has only equation constraints
- Method 3.3 includes 3.2's equation constraints and it adds cost function and inequality constraints

# How do we select constraints ?

## In general,

- Max Velocity (wrt car capabilities and speed limit)
- Min Velocity
- Max Acceleration
- Min Acceleration
- Steering Angle
- Cannot exceed max speed of vehicle
- Minimum velocity → usually shouldn't be negative (no backing up)
- Lateral acceleration needs to be checked to avoid rollover / skid
- Longitudinal acceleration needs to be checked with powertrain capabilities
- Corresponds to max braking force
- Cannot exceed max / min steering

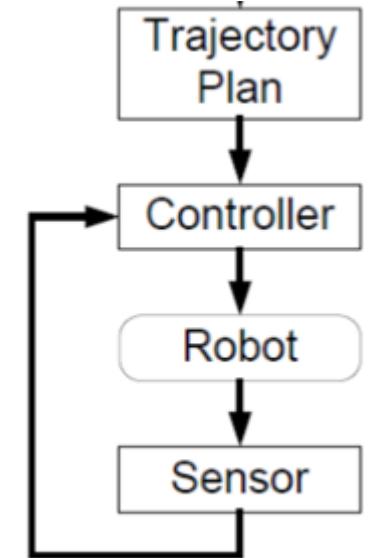
# How do we use trajectories?

- We have trajectories

$$p(t_d), \dot{p}_d(t), \ddot{p}_d(t)$$

- Control=feedforward+feedback

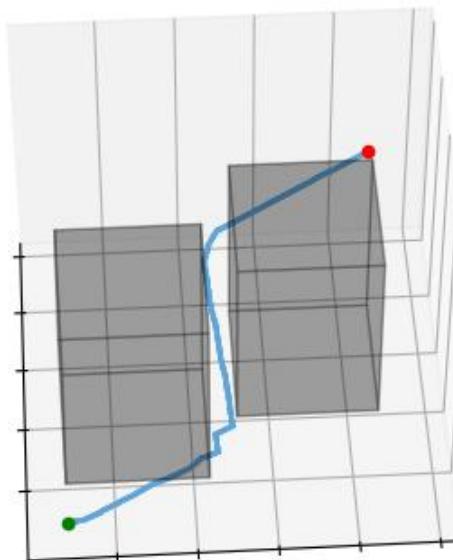
$$\ddot{p}_d(t) + k_p(p_d - p_{\text{meas}}) + k_d(\dot{p}_d - v_{\text{meas}})$$



# One example



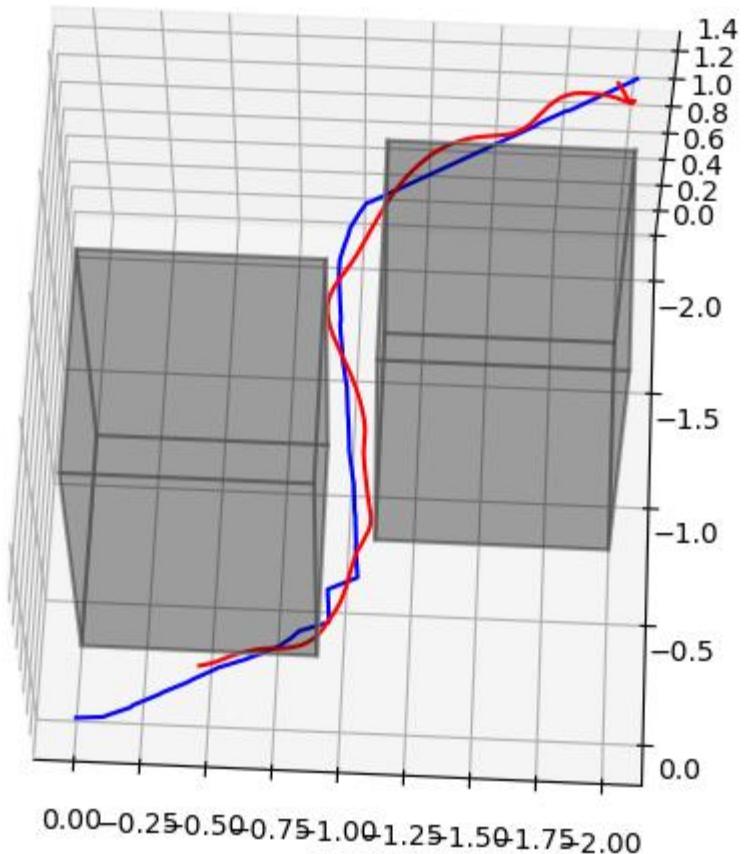
Lab environment  
 $2m * 2m * 1.5m$

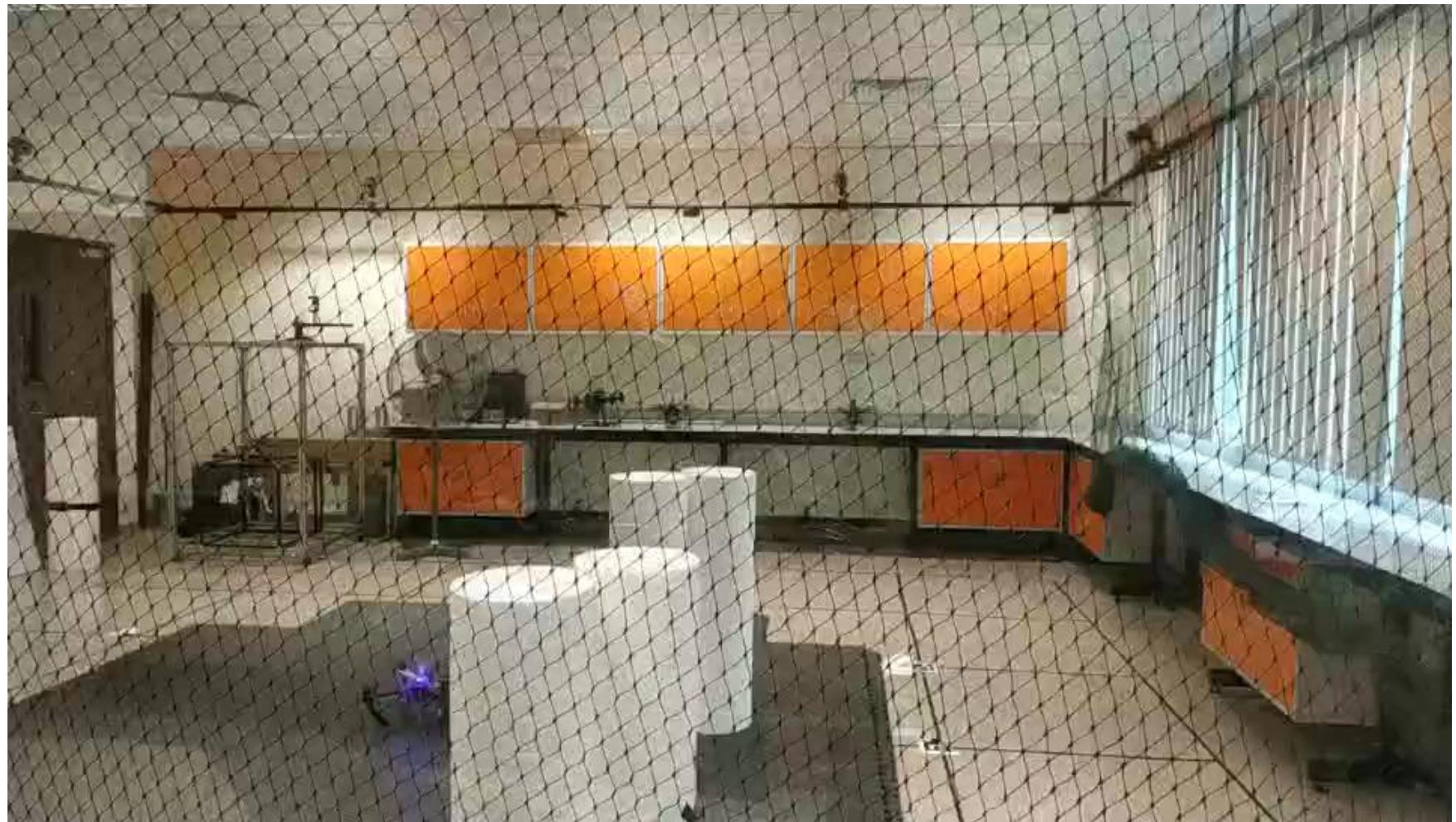


Map size:  $256 * 256 * 256$   
Level: 3

# Desired and actual trajectories

Blue:desired Red:actual





# Summary

- We use polynomial for generating trajectories, position, velocity and acceleration for two points.
- For multiple points, we can use multiple polynomials to generate trajectories.
- Several methods have been discussed. It is hard to say which method is better. This depends on your applications.

# Limitations:

- Computational load is high, especially in multiple waypoints.

# Motion planning

Element	Method	Key point	Remark
C -space	Grid map	Vehicle → point Obstacles with adding vehicle's radius	
Path planning	A*	$F=G+H$ , H --heuristic function	
	Sampling-based search (PRM)	Generate random sampling points Remove points or links colliding with obstacles A* search	
Trajectory generation	Polynomial trajectory without optimization	Define n order → n+1 parameters Define boundary conditions such that we have n+1 equations	
	Polynomial trajectory optimization	Define n order → n+1 parameters Define boundary and continuity conditions → equations Define constraint conditions → inequality constraints QP	

# References

- Choset, K.M. Lynch, S. Hutchinson, G. Kantor, W. Burgard, L.E. Kavraki, and S. Thrun. Principles of Robot Motion: Theory, Algorithms, and Implementations. MIT Press, Boston, MA, 2005.
- Jean-Claude Latombe, Robot Motion Planning
- John J. Craig, Introduction to Robotics

# Thank you!