

Trajectory Planning for Robots: The Challenges of Industrial Considerations

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Introduction market, applications, and trends

- 1.5 million industrial robots in the world (2013, IFR)
; used in large industries (automotive, display)



Hyundai Motors, Czech Republic



LG Display, S. Korea

- Multi-purpose and efficient (6DOF, easy to program, fast with good repeatability)
-> provides various applications (different from other automated machines)
- Trends: more intelligence (vision & force sensors), safety (collision, human-robot interaction)
competitive market (cost reduction)
 - light weight design (less stiff)
 - the high motion performance is dependent on control methods (the main theme)

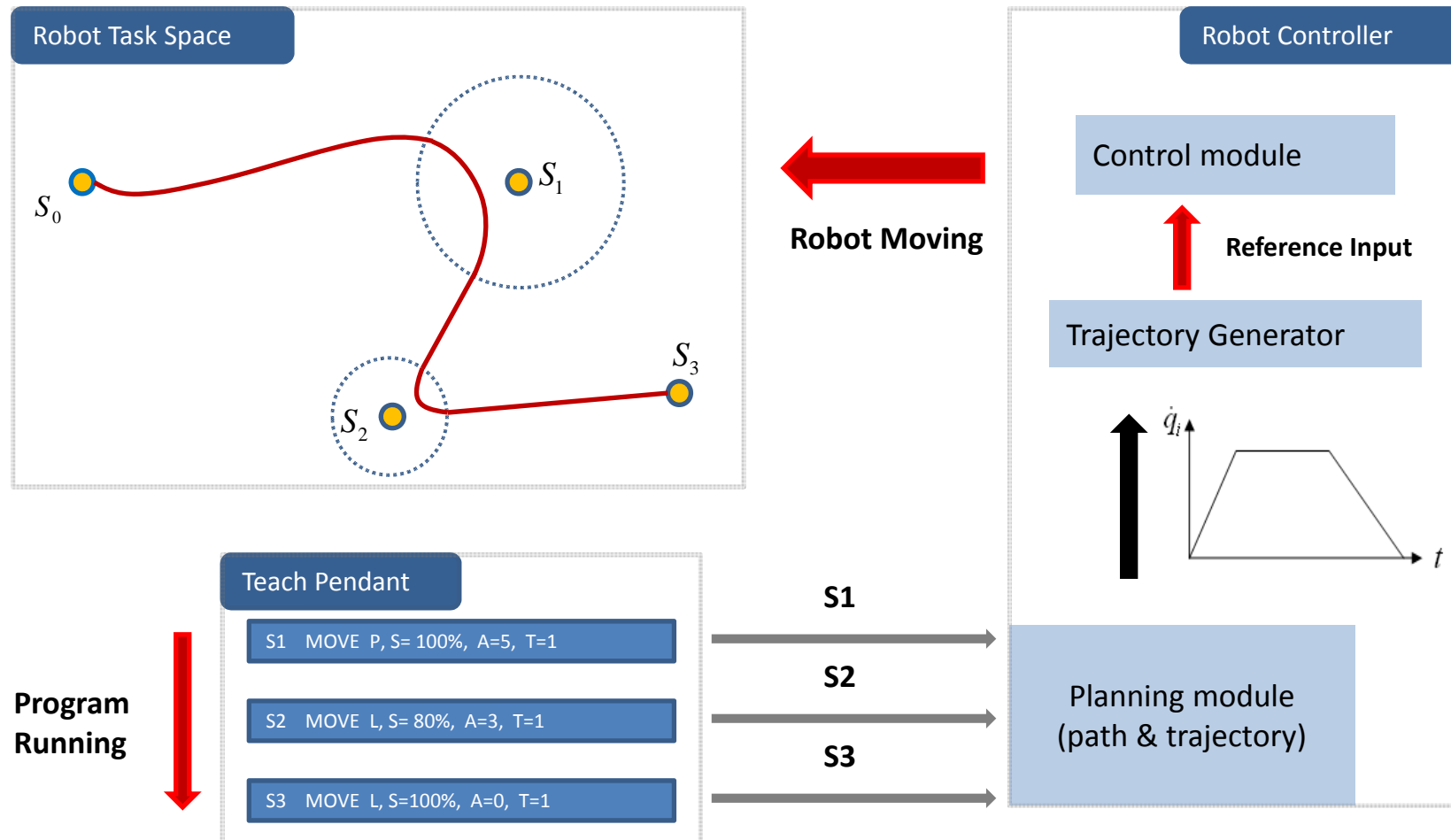
Purpose

- To introduce a typical motion planning method for industrial robots
- To review trajectory planning methods proposed in the literature (related to motion performance of industrial robots)
- To review some common but the most important constraints for generic industrial robot controllers
- To explain why some of well-known methods are not used in practice

Assumptions

- 6DOF industrial robots
- A generic robot controller (architecture, servo control loop)
- A good trajectory:
 - ; simple (implementation, maintenance)
 - ; efficient (small CPU burden)
 - ; being able to address important constraints
 - ; maximizes motion performance (high speed and high accuracy)

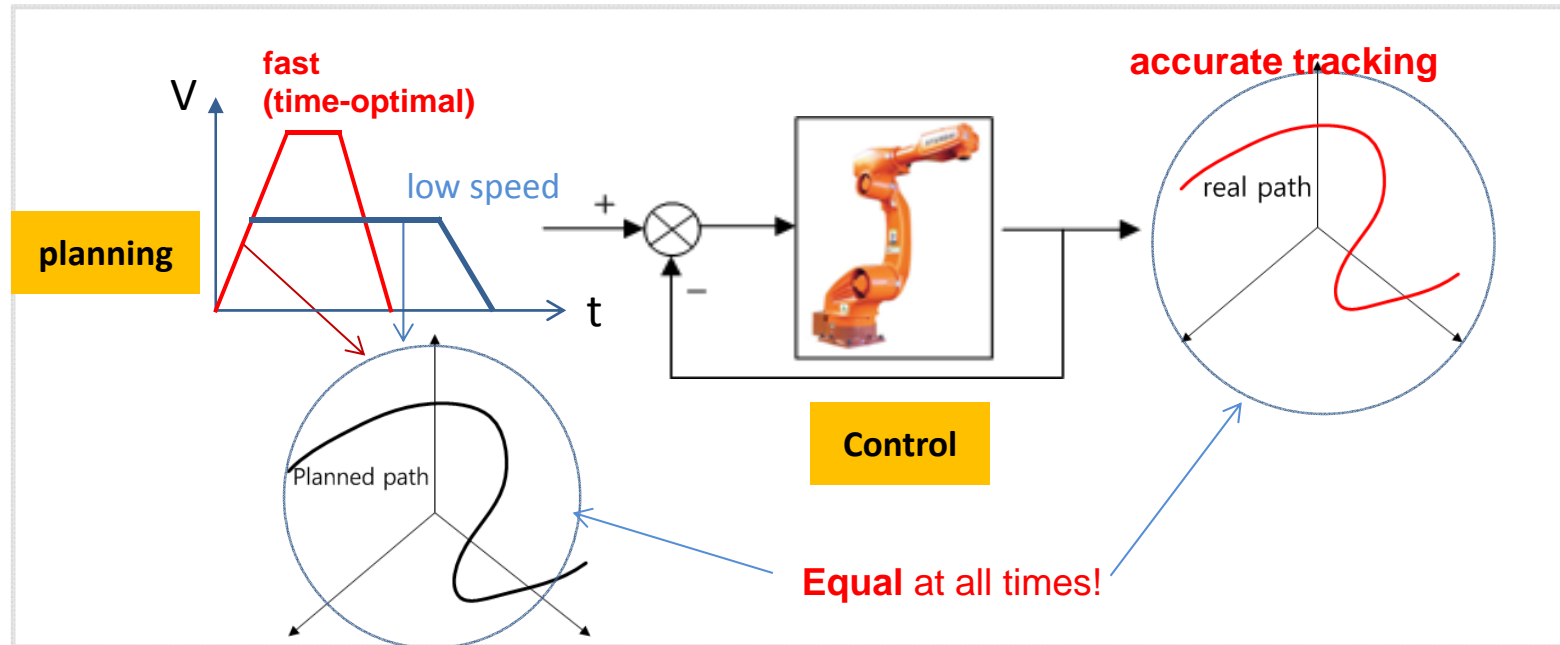
Introduction motion generation



HHI (Hyundai Heavy Industries Co., Ltd.) robot system

Introduction An ideal motion control for industrial robots

Path-Invariant Time-Optimal Motion



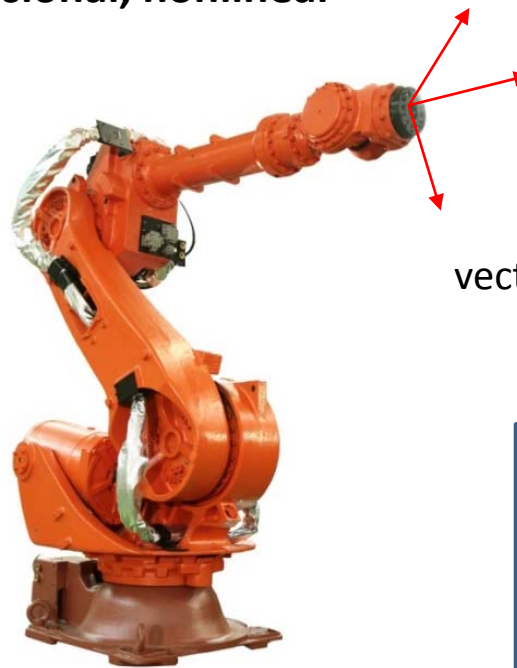
Requirements for trajectories

- *time-optimal*
- *generate a path independent of trajectories*
- *band-limited*

Challenges kinematics

Multi-dimensional, nonlinear

joint space
 \mathbb{R}^6

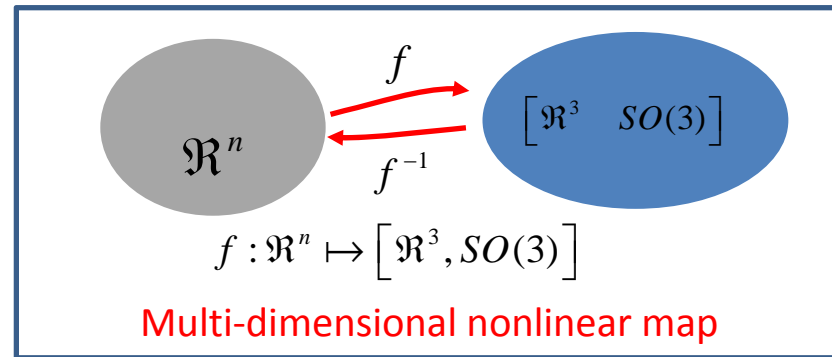
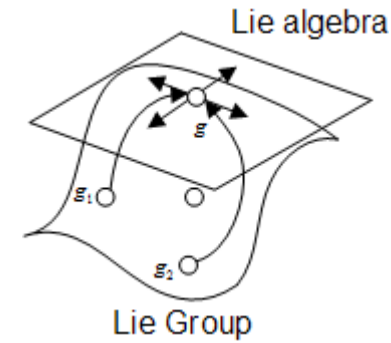


HHI's HS165 (6DOF, payload: 165kg)

Task space
 $[\mathbb{R}^3 \ SO(3)]$

vector space

curved space (Lie Group)
(not a vector space)



- Need to address both positions and orientations
- Orientations: much more complex than positions

Kinematic constraints: path, maximum speed/acceleration/jerk

Challenges dynamics

Multi-dimensional, nonlinear, highly coupled



Equation of motion

$$\tau = \mathbf{M}(\mathbf{q})\ddot{\mathbf{q}} + \mathbf{B}(\mathbf{q}, \dot{\mathbf{q}})\dot{\mathbf{q}} + \mathbf{N}(\mathbf{q}) \in \mathcal{R}^{12}$$

Inertia

- position-dependent
- coupled

Coriolis-Centrifugal and friction

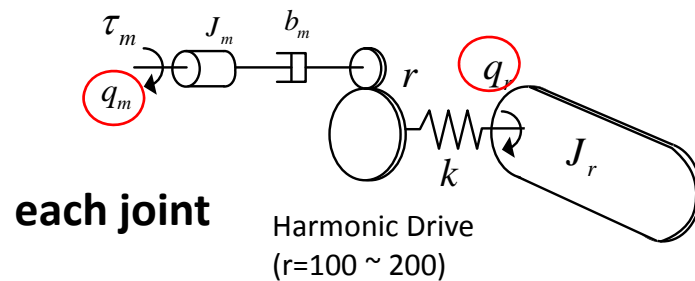
- Pos. & vel. dependent
- coupled

Joint compliance, gravity

- position dependent
- coupled

Natural frequencies at steady-state: 3 ~ 25 Hz (1~2nd mode)
(very low, position dependent)

Joint (link) positions: **not measurable!**



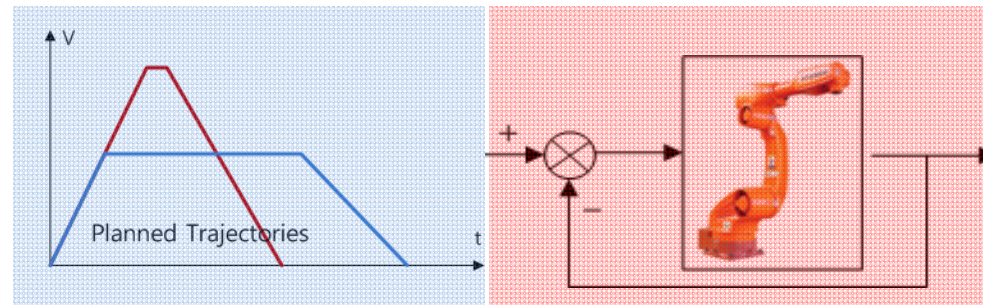
Constraints

- maximum torques (hard constraint)
- not exciting natural modes

Traditional Approach

Complex geometry + complex dynamics (too complicated)

→ Divide & Conquer (Planning / Control)



Planning Task (goals)

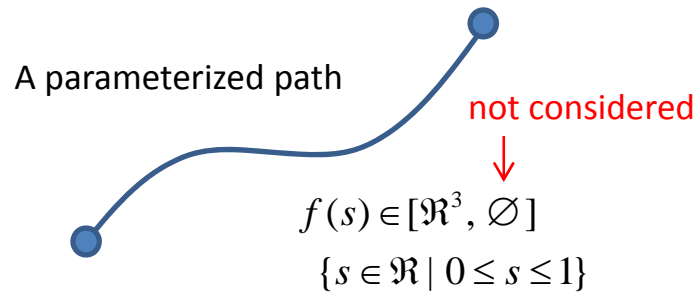
- **Increase speed**
; time-optimal trajectory planning
- **minimize vibrations**
; smooth motion trajectory planning

Control Task (goals)

- **Track accurately, respond quickly,**
- **Obtain stable motions**
; various control methods
(linear, nonlinear, SISO, MIMO)

Well-Known Methods Time-optimal trajectories

Time-Optimal Trajectory Planning



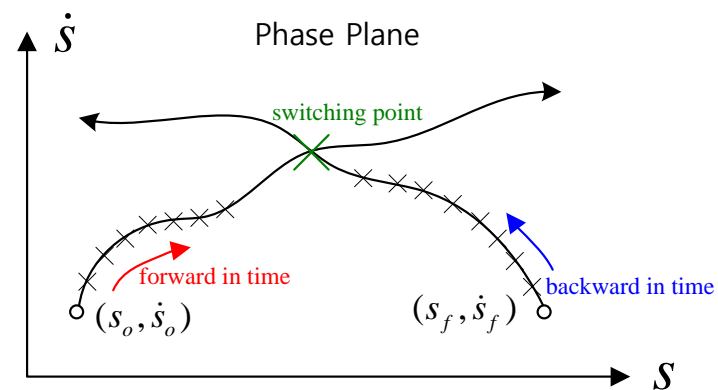
EOM $\tau = \mathbf{m}(s)\ddot{s} + \mathbf{c}(s, \dot{s}) \in \mathbb{R}^n$

Constraints

$$-\left|\tau_{\max}\right| \leq \mathbf{m}(s)\ddot{s} + \mathbf{c}(s, \dot{s}) \leq \left|\tau_{\max}\right|$$

$$\rightarrow \ddot{s}_{\max, acc} = \min \left[\frac{\left|\tau_{\max, i}\right| - c_i(s, \dot{s})}{m_i(s)} \right]$$

Find the time-optimal trajectory



Pros: problem simplified by the path parameter
very close to true time-optimal trajectories

Cons: high CPU burden, infinite jerk (in general),
dynamic model never perfect

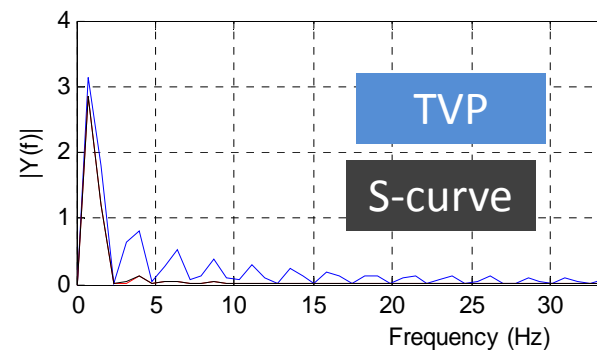
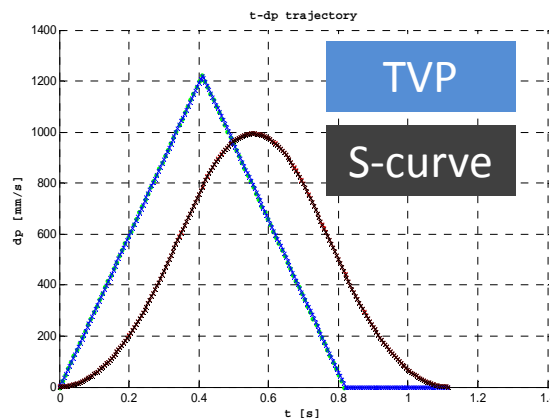
Shin and McKay (1985)

Bobrow *et al.* (1985)

Constantinescu and Croft (2000)

Well-Known Methods Smooth-Trajectories

- Argument
 - ; time-optimal: too complicated, high frequencies (advanced control methods required)
 - ; smooth trajectories can obtain faster settling times due to small residual vibrations.



Pros: small oscillation, no special controllers needed

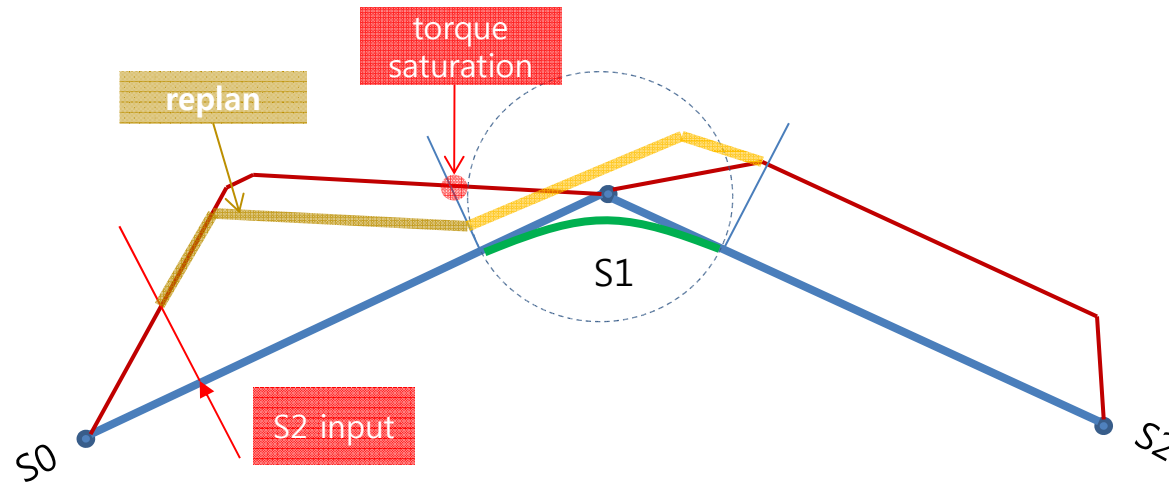
Cons: slow (no dynamics)

**too many cases for non-zero boundary conditions
cannot completely suppress high freq.
natural freq. must be known**

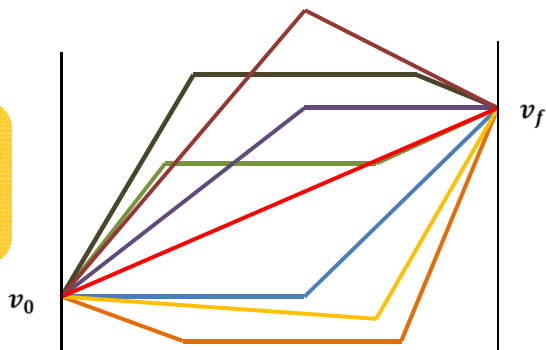
**Meckl and Woods (1985)
Singer (1990)
Macfalan and Croft (2003)
Kroger and Walh (2010)**

Challenges online planning

Online trajectory planning: A robot's trajectory must be planned (replanned) during motion for any (stationary) target given kinematic and dynamic constraints.



Trapezoid Velocity

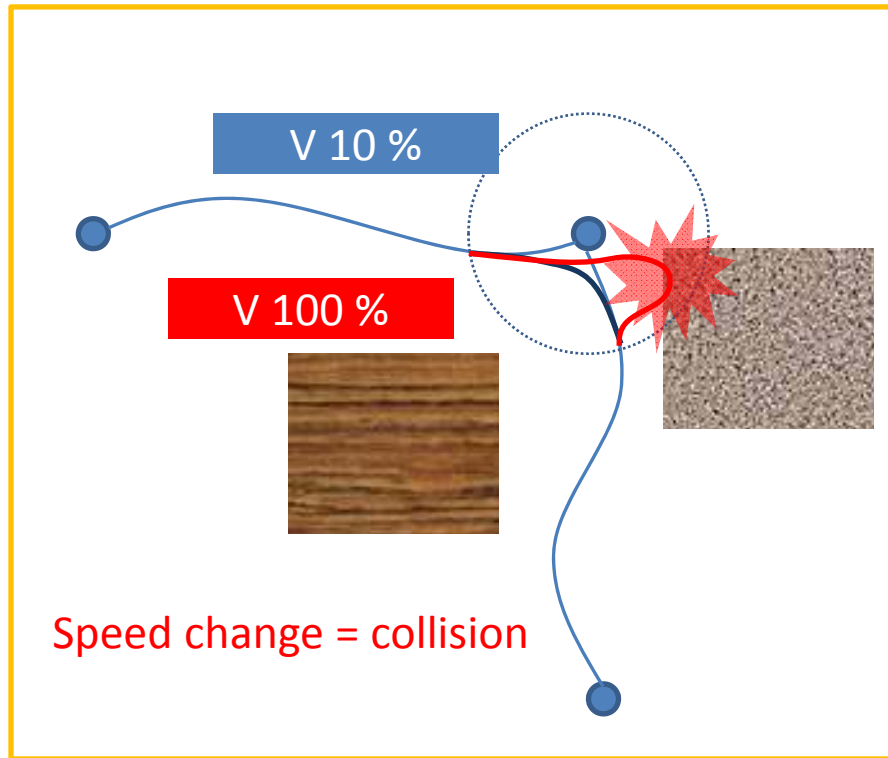


Non-zero boundary conditions

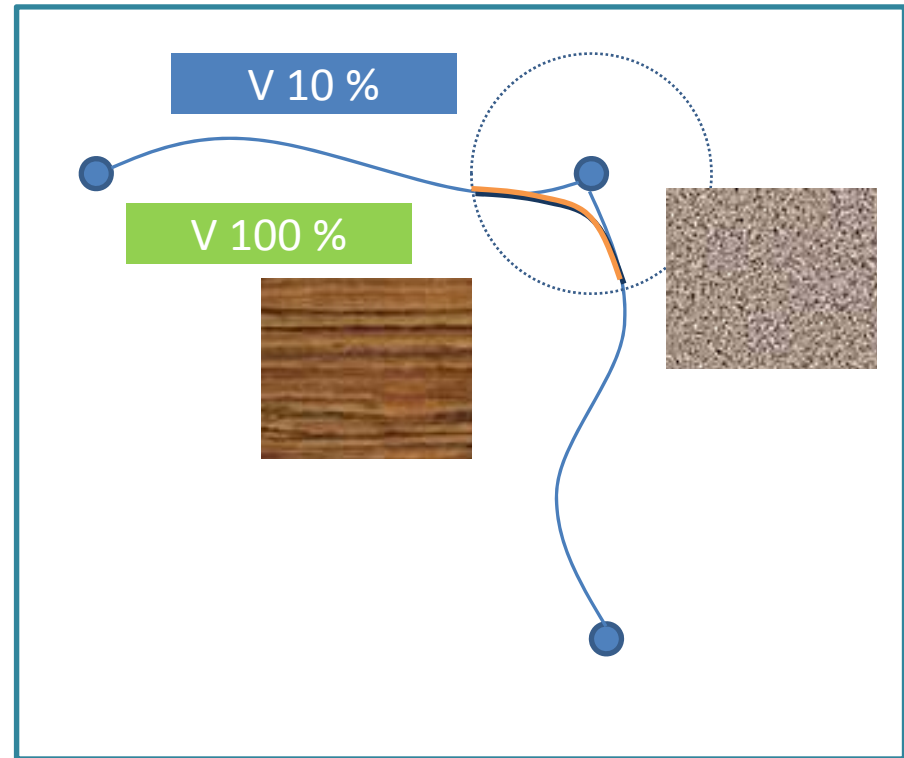
S-curves or higher polynomials?

- too many cases to consider
- prone to generate s/w errors
- > difficult to implement and maintain

Challenges path and speed



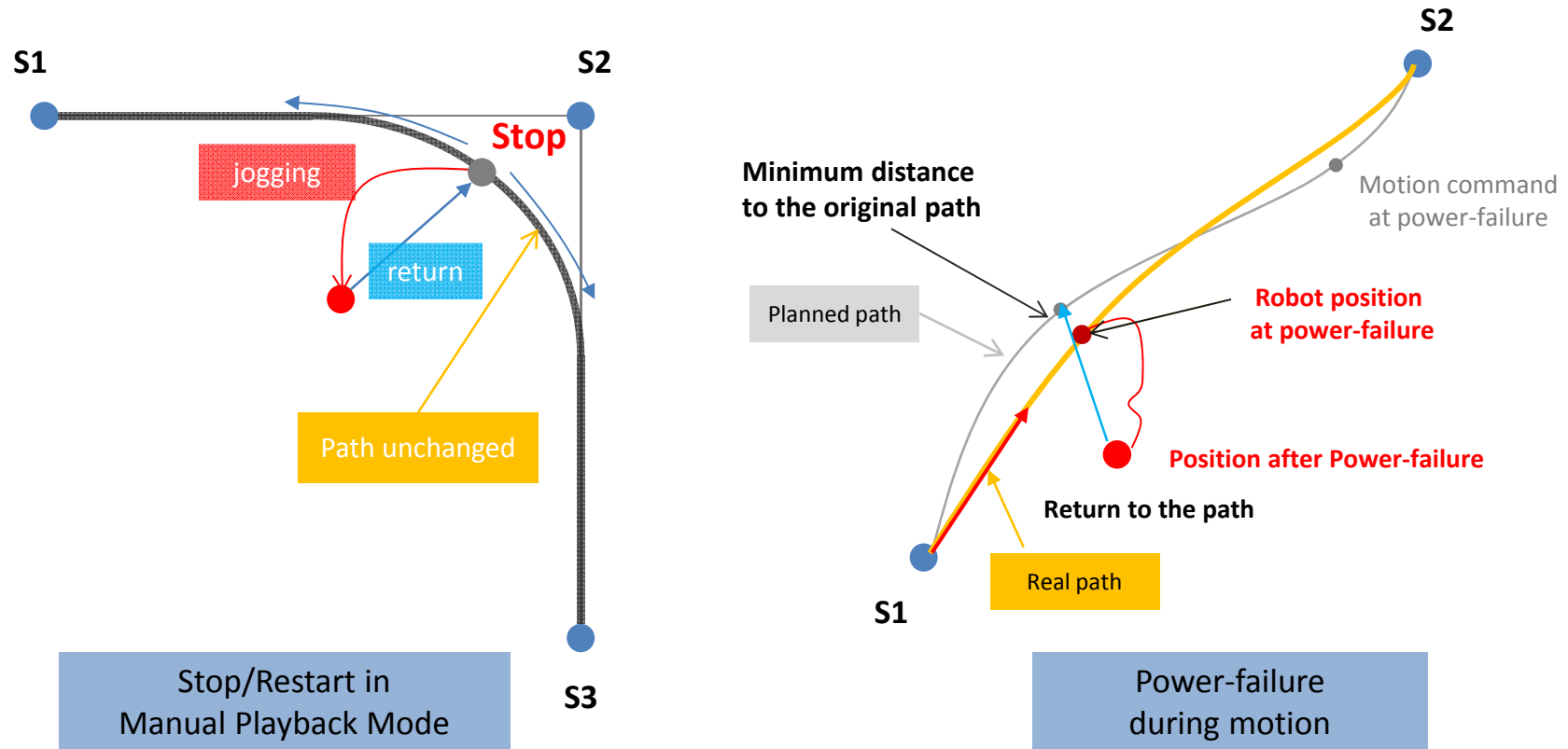
bad



good

Path must be the same regardless of speed changes.

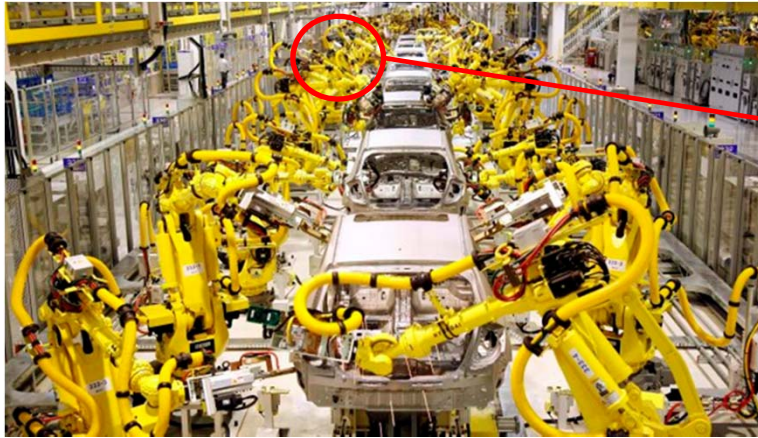
Challenges path invariance for all operation conditions



A programmed path must not be changed under any circumstances unless such a request is made by the user.

Path Recovery

Imagine you are in charge of this factory!
And there is a chance of black-out!



No
power
failure



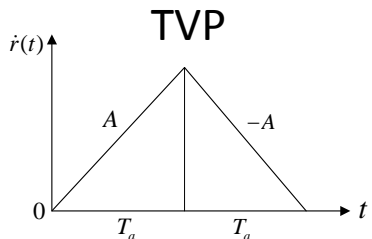
Power failure



Path Recovery

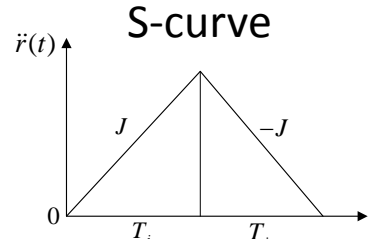
Trajectories Vs. Residual Vibrations

Frequency Analysis via Fourier Transform



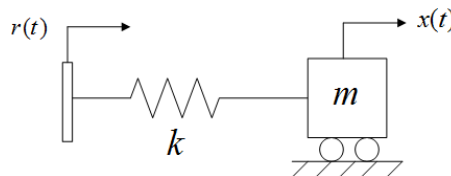
TVP

$$R_{TVP}(j\omega) = \frac{A}{-\omega^2} (1 - e^{-jT_a\omega})$$



S-curve

$$R_{S-curve}(j\omega) = \frac{J}{-j\omega^3} (1 - e^{-jT_j\omega})$$



$m\ddot{x} + k(x - r) = 0$

$$\frac{X(j\omega)}{R(j\omega)} = H(j\omega) = \frac{\omega_n^2}{\omega_n^2 - \omega^2}$$

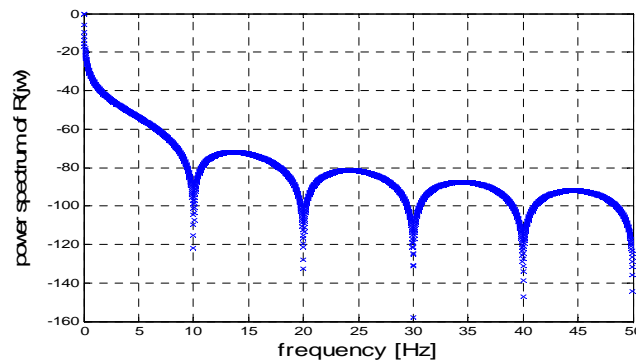
$$X_{TVP}(j\omega) = H(j\omega) R_{TVP}(j\omega)$$

$$= \frac{A}{-\omega^2} \left(\frac{\omega_n^2}{\omega_n^2 - \omega^2} \right) (1 - e^{-jT_a\omega})$$

$$X_{S-curve}(j\omega) = H(j\omega) R_{S-curve}(j\omega)$$

$$= \frac{J}{-j\omega^3} \left(\frac{\omega_n^2}{\omega_n^2 - \omega^2} \right) (1 - e^{-jT_j\omega})$$

FRF of TVP



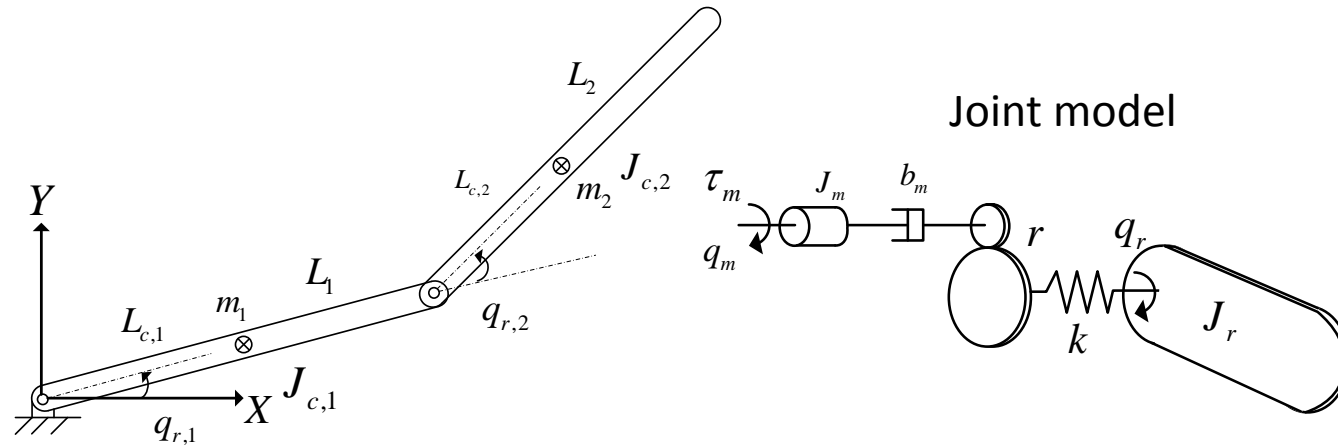
- Big A, Big J -> large Oscillation
- TVP Larger vibration than S-curve
- Vibration can be 0 at some T_a, T_j
- S-curves still have high freq.

Simulation dynamic model

Robot: 2DOF industrial Robot with flexible joints

Initial position: $q_{r0} = [90, -90]$ [deg]

Natural frequencies: 7 Hz, 21 Hz at q_{r0}

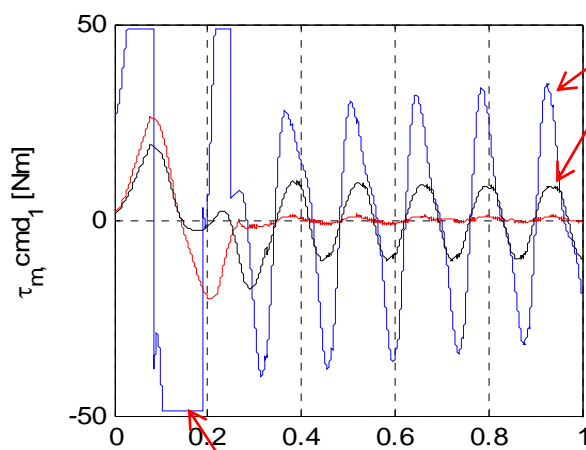
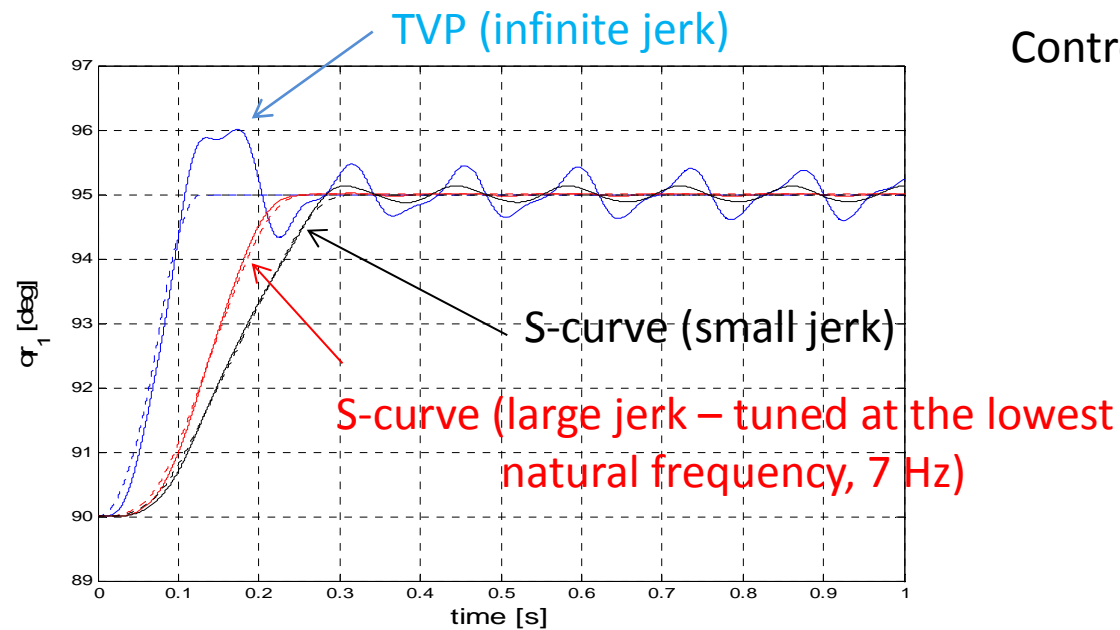


Parameter	Joint 1	Joint 2	Parameter	Joint 1	Joint 2
Mass [kg]	80	170	k [Nm/rad]	1010700	505350
Jr [kg.m ²]	9.97	13.7	bm [Nm s/rad]	0.0281	0.0088
L [m]	0.87	1.05	Max. motor torque [Nm]	49	35.86
Lc [m]	0.38	0.2	r (gear ratio)	201	145

Simulation trajectories Vs. vibrations

Controller: Decentralized PD controllers

First joint position



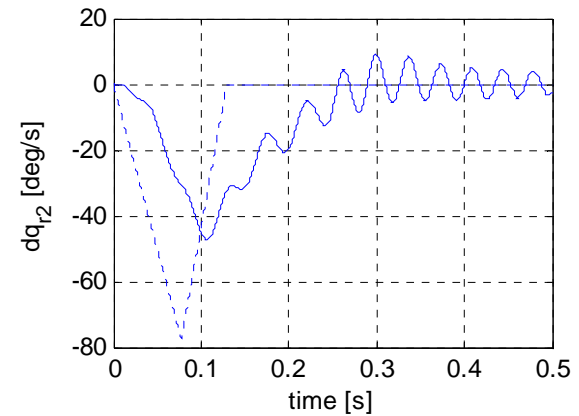
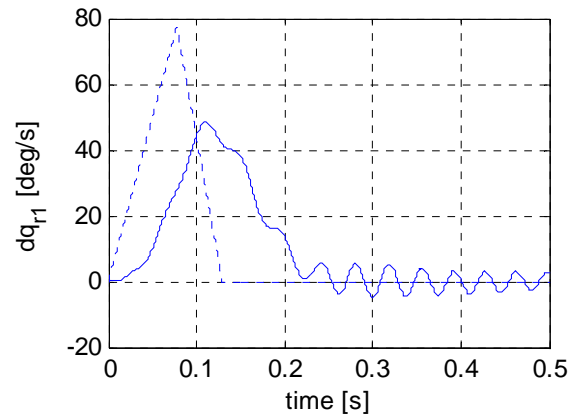
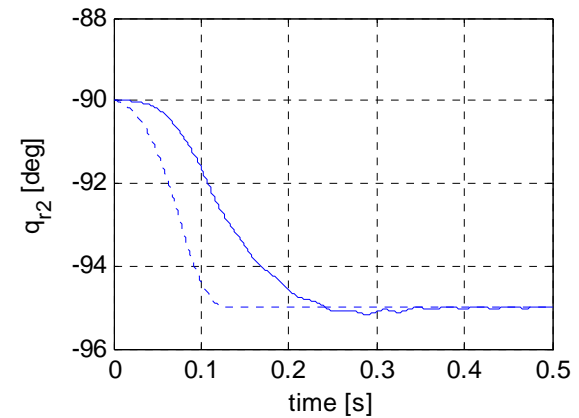
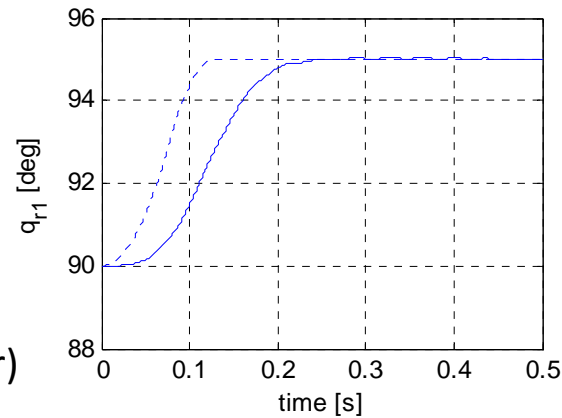
Oscillation at 7 Hz (1st mode)

- Control performance is affected by trajectories.
- Trajectories must be band-limited.
- Such limitation of the band cannot be determined arbitrarily.

Torque saturation

Simulation trajectories Vs. controller

Trajectory: TVP
Controller: decentralized
& linear
full-state feedback
(linear state-observer)
fixed gain by pole-placement



- Large tracking errors
- Large oscillations
- > Decentralized and linear control methods do not work well.
- > A more advanced control method is needed.

Conclusions

- Trajectory planning for industrial robots is challenging.
; online planning, safety, efficiency, complex kinematics and dynamics
- Traditionally, a good trajectory is regarded as either time-optimal or smooth.
- Time-optimal trajectories
; high CPU burden and high frequency components
- Smooth trajectories
; too slow, still complex (high order polynomials), vague jerk limitation
- What is the best trajectory for industrial robots?
; not clear but TVP is still widely used!
; it depends on servo controller.