

Robotics 2

Dynamic model of robots:Analysis, properties, extensions, uses

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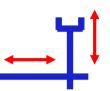
DIPARTIMENTO DI INGEGNERIA INFORMATICA AUTOMATICA E GESTIONALE ANTONIO RUBERTI



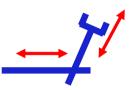
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Analysis of inertial couplings

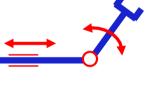
Cartesian robot



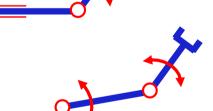
Cartesian "skew" robot



■ PR robot



2R robot



3R articulated robot
 (under simplifying assumptions on the CoMs)

$$M = \begin{pmatrix} m_{11} & 0 \\ 0 & m_{22} \end{pmatrix}$$

$$M = \begin{pmatrix} m_{11} & m_{12} \\ m_{12} & m_{22} \end{pmatrix}$$

$$M = \begin{pmatrix} m_{11} & m_{12}(q_2) \\ m_{12}(q_2) & m_{22} \end{pmatrix}$$

$$M = \begin{pmatrix} m_{11}(q_2) & m_{12}(q_2) \\ m_{12}(q_2) & m_{22} \end{pmatrix}$$

$$M = \begin{pmatrix} m_{11}(q_2, q_3) & 0 & 0 \\ 0 & m_{22}(q_3) & m_{23}(q_3) \\ 0 & m_{23}(q_3) & m_{33} \end{pmatrix}$$

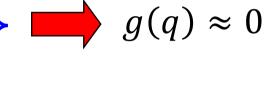




- absence of gravity
 - constant U_q (motion on horizontal plane)
 - applications in remote space
- static balancing
 - distribution of masses (including motors)
- mechanical compensation
 - articulated system of springs
 - closed kinematic chains









STOOM SE

Bounds on dynamic terms

• for an open-chain (serial) manipulator, there always exist positive real constants k_0 to k_7 such that, for any value of q and \dot{q}

$$\begin{aligned} k_0 &\leq \|M(q)\| \leq k_1 + k_2 \|q\| + k_3 \|q\|^2 & \text{inertia matrix} \\ \|S(q,\dot{q})\| &\leq (k_4 + k_5 \|q\|) \|\dot{q}\| & \text{factorization matrix of Coriolis/centrifugal terms} \\ \|g(q)\| &\leq k_6 + k_7 \|q\| & \text{gravity vector} \end{aligned}$$

if the robot has only revolute joints, these simplify to

$$k_0 \le \|M(q)\| \le k_1 \|S(q, \dot{q})\| \le k_4 \|\dot{q}\| \|g(q)\| \le k_6$$

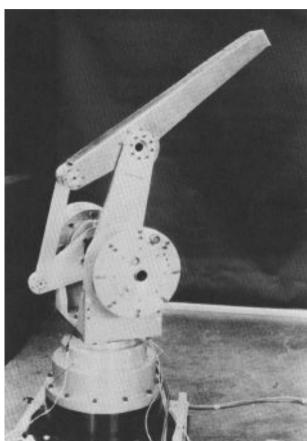
(the same holds true with bounds $q_{i,min} \le q_i \le q_{i,max}$ on prismatic joints)

NOTE: norms are either for vectors or for matrices (induced norms)

Robots with closed kinematic chains - 1







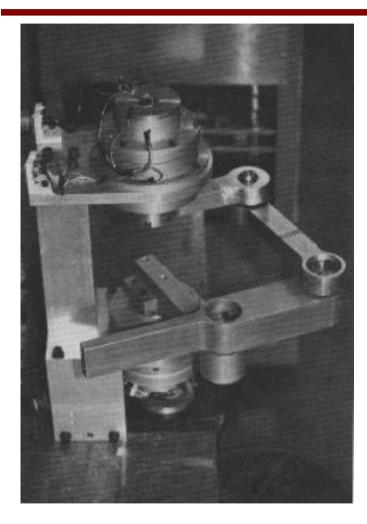


Comau Smart NJ130

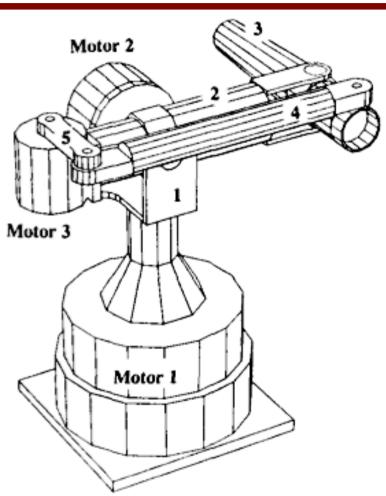
MIT Direct Drive Mark II and Mark III

Robots with closed kinematic chains - 2





MIT Direct Drive Mark IV (planar five-bar linkage)

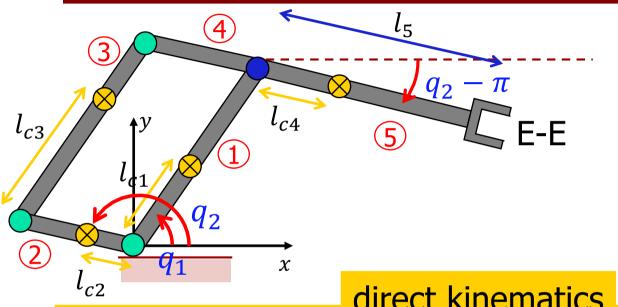


UMinnesota Direct Drive Arm (spatial five-bar linkage)

Robot with parallelogram structure



(planar) kinematics and dynamics



center of mass: arbitrary l_{ci}

parallelogram:

$$l_1 = l_3$$

$$l_2 = l_4$$

direct kinematics

$$p_{EE} = {l_1 c_1 \choose l_1 s_1} + {l_5 \cos(q_2 - \pi) \choose l_5 \sin(q_2 - \pi)} = {l_1 c_1 \choose l_1 s_1} - {l_5 c_2 \choose l_5 s_2}$$

position of center of masses

$$p_{c1} = \begin{pmatrix} l_{c1}c_1 \\ l_{c1}S_1 \end{pmatrix} \quad p_{c2} = \begin{pmatrix} l_{c2}c_2 \\ l_{c2}S_2 \end{pmatrix} \quad p_{c3} = \begin{pmatrix} l_2c_2 \\ l_2S_2 \end{pmatrix} + \begin{pmatrix} l_{c3}c_1 \\ l_{c3}S_1 \end{pmatrix} \quad p_{c4} = \begin{pmatrix} l_1c_1 \\ l_1S_1 \end{pmatrix} - \begin{pmatrix} l_{c4}c_2 \\ l_{c4}S_2 \end{pmatrix}$$



Kinetic energy

linear/angular velocities

$$\begin{aligned} v_{c1} &= \binom{-l_{c1}s_1}{l_{c1}c_1} \dot{q}_1 & v_{c3} &= \binom{-l_{c3}s_1}{l_{c3}c_1} \dot{q}_1 + \binom{-l_2s_2}{l_2c_2} \dot{q}_2 & \omega_1 &= \omega_3 &= \dot{q}_1 \\ v_{c2} &= \binom{-l_{c2}s_2}{l_{c2}c_2} \dot{q}_2 & v_{c4} &= \binom{-l_1s_1}{l_1c_1} \dot{q}_1 + \binom{l_{c4}s_2}{-l_{c4}c_2} \dot{q}_2 & \omega_2 &= \omega_4 &= \dot{q}_2 \end{aligned}$$

Note: a (planar) 2D notation is used here!

$$T_{1} = \frac{1}{2} m_{1} l_{c1}^{2} \dot{q}_{1}^{2} + \frac{1}{2} I_{c1,zz} \dot{q}_{1}^{2} \qquad T_{2} = \frac{1}{2} m_{2} l_{c2}^{2} \dot{q}_{2}^{2} + \frac{1}{2} I_{c2,zz} \dot{q}_{2}^{2}$$

$$T_{3} = \frac{1}{2} m_{3} (l_{2}^{2} \dot{q}_{2}^{2} + l_{c3}^{2} \dot{q}_{1}^{2} + 2 l_{2} l_{c3} c_{2-1} \dot{q}_{1} \dot{q}_{2}) + \frac{1}{2} I_{c3,zz} \dot{q}_{1}^{2}$$

$$T_{4} = \frac{1}{2} m_{4} (l_{1}^{2} \dot{q}_{1}^{2} + l_{c4}^{2} \dot{q}_{2}^{2} - 2 l_{1} l_{c4} c_{2-1} \dot{q}_{1} \dot{q}_{2}) + \frac{1}{2} I_{c4,zz} \dot{q}_{2}^{2}$$

Robot inertia matrix

$$T = \sum_{i=1}^{4} T_i = \frac{1}{2} \dot{q}^T M(q) \dot{q}$$

$$M(q) = \begin{pmatrix} I_{c1,zz} + m_1 l_{c1}^2 + I_{c3,zz} + m_3 l_{c3}^2 + m_4 l_1^2 & \text{symm} \\ (m_3 l_2 l_{c3} - m_4 l_1 l_{c4}) c_{2-1} & I_{c2,zz} + m_2 l_{c2}^2 + I_{c4,zz} + m_4 l_{c4}^2 + m_3 l_2^2 \end{pmatrix}$$

structural condition in mechanical design

$$m_3 l_2 l_{c3} = m_4 l_1 l_{c4} \tag{*}$$



M(q) diagonal and constant \Rightarrow centrifugal and Coriolis terms $\equiv 0$

mechanically DECOUPLED and LINEAR dynamic model (up to the gravity term g(q))



$$\begin{pmatrix} M_{11} & 0 \\ 0 & M_{22} \end{pmatrix} \begin{pmatrix} \ddot{q}_1 \\ \ddot{q}_2 \end{pmatrix} = \begin{pmatrix} u_1 \\ u_2 \end{pmatrix}$$

big advantage for the design of a motion control law!



Potential energy and gravity terms

from the y-components of vectors p_{ci}

$$U_1 = m_1 g_0 l_{c1} s_1 \qquad U_2 = m_2 g_0 l_{c2} s_2$$

$$U_3 = m_3 g_0 (l_2 s_2 + l_{c3} s_1) \quad U_4 = m_4 g_0 (l_1 s_1 - l_{c4} s_2)$$

$$U = \sum_{i=1}^{4} U_i$$

$$g(q) = \left(\frac{\partial U}{\partial q}\right)^T = \begin{pmatrix} g_0(m_1l_{c1} + m_3l_{c3} + m_4l_1)c_1 \\ g_0(m_2l_{c2} + m_3l_2 - m_4l_{c4})c_2 \end{pmatrix} = \begin{pmatrix} g_1(q_1) \\ g_2(q_2) \end{pmatrix}$$
 components are always "decoupled"

in addition, when (*) holds



$$m_{11}\ddot{q}_1 + g_1(q_1) = u_1$$

$$m_{22}\ddot{q}_2 + g_2(q_2) = u_2$$

$$m_{23}\ddot{q}_2 + g_3(q_2) = u_2$$

$$m_{24}\ddot{q}_1 + g_1(q_1) = u_1$$

$$m_{25}\ddot{q}_2 + g_3(q_2) = u_3$$

$$m_{26}\ddot{q}_3 + g_3(q_3) = u_3$$

$$m_{36}\ddot{q}_4 + g_3(q_3) = u_3$$

$$m_{36}\ddot{q}_5 + g_3(q_3) = u_3$$

performing work on q_i

further structural conditions in the mechanical design lead to $g(q) \equiv 0!!$

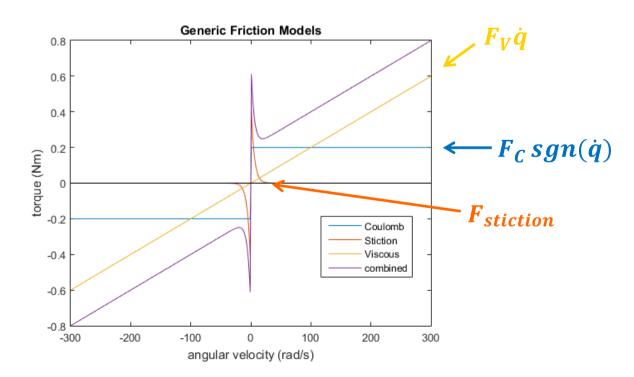


Adding dynamic terms ...

- 1) dissipative phenomena due to friction at the joints/transmissions
 - viscous, Coulomb, stiction, Stribeck, LuGre (dynamic)...
 - local effects at the joints
 - difficult to model in general, except for:

$$u_{V,i} = -F_{V,i} \dot{q}_i$$

$$u_{C,i} = -F_{C,i} \operatorname{sgn}(\dot{q}_i)$$

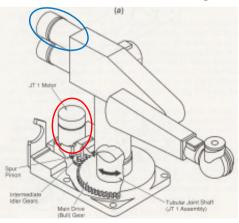


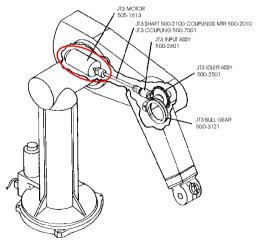


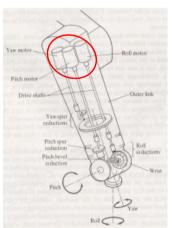
Adding dynamic terms ...

- 2) inclusion of electrical actuators (as additional rigid bodies)
 - motor i mounted on link i 1 (or before), with very few exceptions
 - often with its spinning axis aligned with joint axis i
 - (balanced) mass of motor included in total mass of carrying link
 - (rotor) inertia has to be added to robot kinetic energy
 - transmissions with reduction gears (often, large reduction ratios)
 - in some cases, multiple motors cooperate in moving multiple links: use a transmission coupling matrix Γ (with off-diagonal elements)

Unimation PUMA family





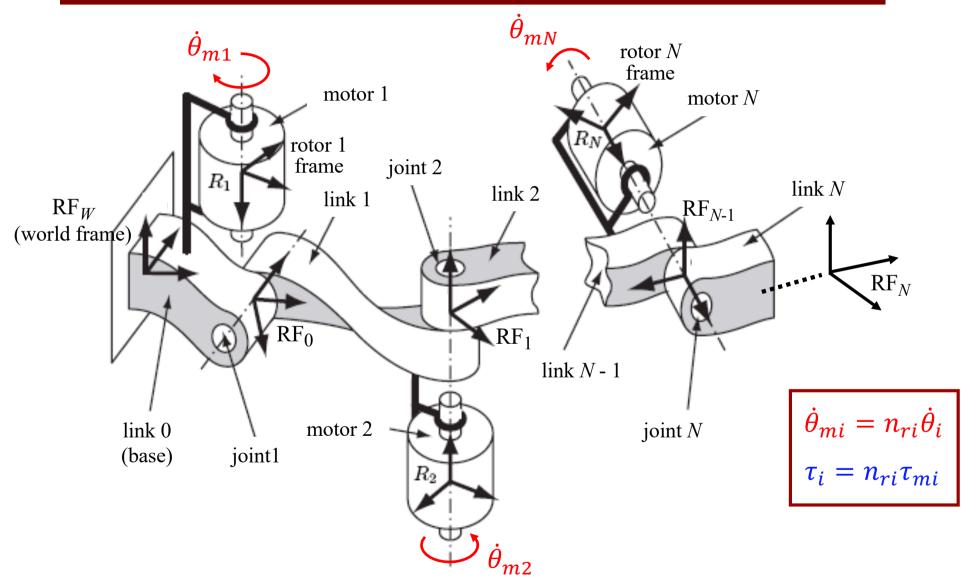




Mitsubishi RV-3S

SALAN YES

Placement of motors along the chain





Resulting dynamic model

 simplifying assumption: in the rotational part of the kinetic energy, only the "spinning" rotor velocity is considered

$$T_{mi} = \frac{1}{2} I_{mi} \dot{\theta}_{mi}^2 = \frac{1}{2} I_{mi} n_{ri}^2 \dot{q}_i^2 = \frac{1}{2} B_{mi} \dot{q}_i^2 \qquad T_m = \sum_{i=1}^N T_{mi} = \frac{1}{2} \dot{q}^T B_m \dot{q}$$
diagonal, > 0

including all added terms, the robot dynamics becomes

$$(M(q) + B_m)\ddot{q} + c(q, \dot{q}) + g(q) + F_V\dot{q} + F_C \operatorname{sgn}(\dot{q}) = \tau$$

$$constant \rightarrow \operatorname{does} \operatorname{NOT} \qquad F_V > 0, F_C > 0$$

$$contribute to c$$

$$motor torques$$

$$(after reduction gears)$$

scaling by the reduction gears, looking from the motor side

diagonal
$$\left(I_m + \operatorname{diag}\left\{\frac{m_{ii}(q)}{n_{ri}^2}\right\}\right) \ddot{\theta}_m + \operatorname{diag}\left\{\frac{1}{n_{ri}}\right\} \left(\sum_{j=1}^N \overline{M}_j(q) \ddot{q}_j + f(q, \dot{q})\right) = \tau_m \quad \text{(before reduction gears)}$$

$$\text{except the terms } m_{jj}$$

Including joint elasticity



- in industrial robots, use of motion transmissions based on
 - belts
 - harmonic drives
 - long shafts

introduces flexibility between actuating motors (input) and driven links (output)

- in research robots, compliance in transmissions is introduced on purpose for safety (human collaboration) and/or energy efficiency
 - actuator relocation by means of (compliant) cables and pulleys
 - harmonic drives and lightweight (but rigid) link design
 - redundant (macro-mini or parallel) actuation, with elastic couplings
- in both cases, flexibility is modeled as concentrated at the joints
- in most cases, assuming small joint deformation (elastic domain)



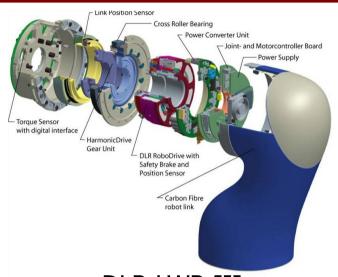




Dexter with cable transmissions

motor elastic load/link spring (stiffness K)

Quanser Flexible Joint (1-dof linear, educational)

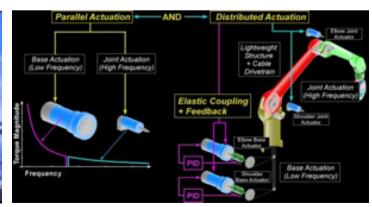


DLR LWR-III with harmonic drives





video



Stanford DECMMA with micro-macro actuation

Dynamic model of robots with elastic joints



- introduce 2N generalized coordinates
 - q = N link positions
- $\dot{\theta} = N$ motor positions (after reduction, $\theta_i = \theta_{mi}/n_{ri}$)
 add motor kinetic energy T_m to that of the links $T_q = \frac{1}{2}\dot{q}^T M(q)\dot{q}$

$$T_{mi} = \frac{1}{2} I_{mi} \dot{\theta}_{mi}^2 = \frac{1}{2} I_{mi} n_{ri}^2 \dot{\theta}_i^2 = \frac{1}{2} B_{mi} \dot{\theta}_i^2 \qquad T_m = \sum_{i=1}^N T_{mi} = \frac{1}{2} \dot{\theta}^T B_m \dot{\theta}$$
 diagonal, > 0

- add elastic potential energy U_e to that due to gravity $U_q(q)$
 - K = matrix of joint stiffness (diagonal, > 0)

$$U_{ei} = \frac{1}{2} K_i \left(q_i - \left(\frac{\theta_{mi}}{n_{ri}} \right) \right)^2 = \frac{1}{2} K_i (q_i - \theta_i)^2 \quad U_e = \sum_{i=1}^N U_{ei} = \frac{1}{2} (q - \theta)^T K (q - \theta)$$

• apply Euler-Lagrange equations w.r.t. (q, θ)

Use of the dynamic model inverse dynamics



- given a desired trajectory $q_d(t)$
 - twice differentiable $(\exists \ddot{q}_d(t))$
 - possibly obtained from a task/Cartesian trajectory $r_d(t)$, by (differential) kinematic inversion

the input torque needed to execute this motion (in free space) is

$$\tau_d = (M(q_d) + B_m)\ddot{q}_d + c(q_d, \dot{q}_d) + g(q_d) + F_V \dot{q}_d + F_C \operatorname{sgn}(\dot{q}_d)$$

- useful also for control (e.g., nominal feedforward)
- however, this way of performing the algebraic computation $(\forall t)$ is not efficient when using the above Lagrangian approach
 - symbolic terms grow much longer, quite rapidly for larger N
 - in real time, numerical computation is based on Newton-Euler method

State equations direct dynamics



Lagrangian dynamic model

$$M(q)\ddot{q} + c(q,\dot{q}) + g(q) = u$$

N differential 2nd order equations

defining the vector of state variables as $x = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} q \\ \dot{q} \end{pmatrix} \in \mathbb{R}^{2N}$

state equations



$$\dot{\mathbf{x}} = \begin{pmatrix} \dot{x}_1 \\ \dot{x}_2 \end{pmatrix} = \begin{pmatrix} x_2 \\ -M^{-1}(x_1)[c(x_1, x_2) + g(x_1)] \end{pmatrix} + \begin{pmatrix} 0 \\ M^{-1}(x_1) \end{pmatrix} u$$

$$= f(x) + G(x)u$$

$$\uparrow \qquad \uparrow$$

$$2N \times 1 \quad 2N \times N$$

2N differential 1st order equations

another choice...
$$\tilde{x} = \begin{pmatrix} q \\ M(q)\dot{q} \end{pmatrix}$$
 generalized momentum $\dot{\tilde{x}} = ...$ (do it as exercise)

$$\tilde{x} = \begin{pmatrix} q \\ M(q)\dot{q} \end{pmatrix}$$

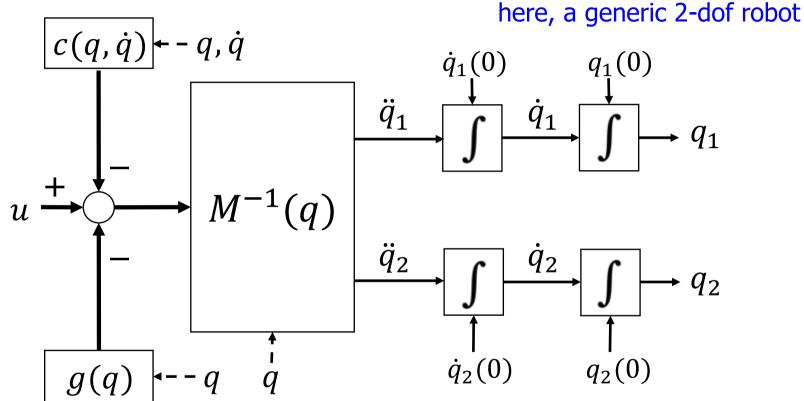
$$\dot{\tilde{x}} = \dots$$
 (do it as exercise)

Dynamic simulation



Simulink block scheme

input torque command (open-loop or in feedback)



including "inv(M)"

- initialization (dynamic coefficients and initial state)
- calls to (user-defined) Matlab functions for the evaluation of model terms
- choice of a numerical integration method (and of its parameters)

e.g., 4th-order Runge-Kutta (ode45)

Approximate linearization



- we can derive a linear dynamic model of the robot, which is valid locally around a given operative condition
 - useful for analysis, design, and gain tuning of linear (or, the linear part of) control laws
 - approximation by Taylor series expansion, up to the first order
 - linearization around a (constant) equilibrium state or along a (nominal, time-varying) equilibrium trajectory
 - usually, we work with (nonlinear) state equations; for mechanical systems, it is more convenient to directly use the 2nd order model
 - same result, but easier derivation

equilibrium state
$$(q, \dot{q}) = (q_e, 0) [\ddot{q} = 0]$$
 \Longrightarrow $g(q_e) = u_e$ equilibrium trajectory $(q, \dot{q}) = (q_d(t), \dot{q}_d(t)) [\ddot{q} = \ddot{q}_d(t)]$ \Longrightarrow $M(q_d)\ddot{q}_d + c(q_d, \dot{q}_d) + g(q_d) = u_d$

Linearization at an equilibrium state



variations around an equilibrium state

$$q = q_e + \Delta q$$
 $\dot{q} = \dot{q}_e + \dot{\Delta q} = \dot{\Delta q}$ $\ddot{q} = \ddot{q}_e + \dot{\Delta q} = \ddot{\Delta q}$ $u = u_e + \Delta u$

 keeping into account the quadratic dependence of c terms on velocity (thus, neglected around the zero velocity)

$$M(q_e)\ddot{\Delta q} + g(q_e) + \frac{\partial g}{\partial q}\bigg|_{q=q_e} \Delta q + o(||\Delta q||, ||\Delta q||) = u_e + \Delta u$$
infinitesimal terms of second or higher order

• in state-space format, with $\Delta x = \begin{pmatrix} \Delta q \\ \dot{\Delta q} \end{pmatrix}$

$$\dot{\Delta x} = \begin{pmatrix} 0 & I \\ -M^{-1}(q_e)G(q_e) & 0 \end{pmatrix} \Delta x + \begin{pmatrix} 0 \\ M^{-1}(q_e) \end{pmatrix} \Delta u = A \Delta x + B \Delta u$$

Linearization along a trajectory



variations around an equilibrium trajectory

$$q = q_d + \Delta q$$
 $\dot{q} = \dot{q}_d + \dot{\Delta q}$ $\ddot{q} = \ddot{q}_d + \dot{\Delta q}$ $u = u_d + \Delta u$

developing to 1st order the terms in the dynamic model ...

$$M(q_d + \Delta q)(\ddot{q}_d + \ddot{\Delta q}) + c(q_d + \Delta q, \dot{q}_d + \dot{\Delta q}) + g(q_d + \Delta q) = u_d + \Delta u$$

$$M(q_d + \Delta q) \cong M(q_d) + \sum_{i=1}^{N} \frac{\partial M_i}{\partial q} \Big|_{q=q_d} e_i^T \Delta q \qquad i\text{-th row of the identity matrix}$$

$$g(q_d + \Delta q) \cong g(q_d) + G(q_d) \Delta q \qquad C_1(q_d, \dot{q}_d)$$

$$c(q_d + \Delta q, \dot{q}_d + \dot{\Delta q}) \cong c(q_d, \dot{q}_d) + \underbrace{\frac{\partial c}{\partial q}}_{\dot{q} = \dot{q}_d} \Delta q + \underbrace{\frac{\partial c}{\partial \dot{q}}}_{\dot{q} = \dot{q}_d} \dot{\Delta q}$$





after simplifications ...

$$M(q_d)\ddot{\Delta q} + C_2(q_d, \dot{q}_d)\dot{\Delta q} + D(q_d, \dot{q}_d, \ddot{q}_d)\Delta q = \Delta u$$
 with
$$D(q_d, \dot{q}_d, \ddot{q}_d) = G(q_d) + C_1(q_d, \dot{q}_d) + \sum_{i=1}^N \frac{\partial M_i}{\partial q} \bigg|_{q=q_d} \ddot{q}_d e_i^T$$

in state-space format

$$\begin{split} \dot{\Delta x} &= \begin{pmatrix} 0 & I \\ -M^{-1}(q_d)D(q_d, \dot{q}_d, \ddot{q}_d) & -M^{-1}(q_d)C_2(q_d, \dot{q}_d) \end{pmatrix} \Delta x \\ &+ \begin{pmatrix} 0 \\ M^{-1}(q_d) \end{pmatrix} \Delta u = A(t) \Delta x + B(t) \Delta u \end{split}$$

a linear, but time-varying system!!

Coordinate transformation



$$q \in \mathbb{R}^{N}$$
 $M(q)\ddot{q} + c(q,\dot{q}) + g(q) = M(q)\ddot{q} + n(q,\dot{q}) = u_q$

if we wish/need to use a new set of generalized coordinates p

$$p \in \mathbb{R}^N$$
 $p = f(q)$ $q = f^{-1}(p)$ by (principle of

by duality (principle of virtual work)

$$\dot{p} = \frac{\partial f}{\partial q} \dot{q} = J(q) \dot{q}$$

$$\dot{q} = J^{-1}(q) \dot{p} \quad u_q = J^T(q) u_p$$

$$\dot{q} = J^{-1}(q)\dot{p}$$

$$u_q = J^T(q)u_p$$

$$\ddot{p} = J(q)\ddot{q} + \dot{J}(q)\ddot{q}$$

$$\ddot{p} = J(q)\ddot{q} + \dot{J}(q)\dot{q} \qquad \qquad \ddot{q} = J^{-1}(q)\big(\ddot{p} - \dot{J}(q)J^{-1}(q)\dot{p}\big)$$

$$M(q)J^{-1}(q)\ddot{p} - M(q)J^{-1}(q)\dot{J}(q)J^{-1}(q)\dot{p} + n(q,\dot{q}) = J^{T}(q)u_{p}$$

(q) pre-multiplying the whole equation...

Robot dynamic model

after coordinate transformation



$$J^{-T}(q)M(q)J^{-1}(q)\ddot{p} + J^{-T}(q)(n(q,\dot{q}) - M(q)J^{-1}(q)\dot{J}(q)J^{-1}(q)\dot{p}) = u_p$$

for actual computation, $q \rightarrow p$ these inner substitutions are not necessary $(q, \dot{q}) \rightarrow (p, \dot{p})$

$$(q,\dot{q}) \to (p,\dot{p})$$

non-conservative generalized forces performing work on p

$$M_p(p)\ddot{p} + c_p(p,\dot{p}) + g_p(p) = u_p$$

$$M_p = J^{-T} M J^{-1}$$
 symmetric,
positive definite
(out of singularities) $g_p = J^{-T} g$

$$c_p = J^{-T} \left(c - M J^{-1} \, \dot{J} \, J^{-1} \dot{p} \right) = J^{-T} c - M_p \, \dot{J} \, J^{-1} \dot{p} \quad \begin{array}{l} \text{quadratic} \\ \text{dependence on } \dot{p} \end{array}$$

$$c_p(p,\dot{p}) = S_p(p,\dot{p})\,\dot{p}$$
 $\dot{M}_p - 2S_p$ skew-symmetric

when p = E-E pose, this is the robot dynamic model in Cartesian coordinates

Q: What if the robot is redundant with respect to the Cartesian task?

SALONYM NE

uniform time scaling of motion

- given a smooth original trajectory $q_d(t)$ of motion for $t \in [0, T]$
 - suppose to rescale time as $t \to r(t)$ (a strictly *increasing* function of t)
 - in the new time scale, the scaled trajectory $q_s(r)$ satisfies

$$q_d(t) = q_s(r(t)) \rightarrow \dot{q}_d(t) = \frac{dq_d}{dt} = \frac{dq_s}{dr} \frac{dr}{dt} = q_s' \dot{r}$$

same path executed (at different instants of time)

$$\ddot{q}_d(t) = \frac{d\dot{q}_d}{dt} = \left(\frac{dq_s'}{dr}\frac{dr}{dt}\right)\dot{r} + q_s'\ddot{r} = q_s''\dot{r}^2 + q_s'\ddot{r}$$

• uniform scaling of the trajectory occurs when r(t) = kt

$$\dot{q}_d(t) = kq_s'(kt) \qquad \ddot{q}_d(t) = k^2 q_s''(kt)$$

Q: what is the new input torque needed to execute the scaled trajectory? (suppose dissipative terms can be neglected)



inverse dynamics under uniform time scaling

the new torque could be recomputed through the inverse dynamics, for every $r = kt \in [0, T'] = [0, kT]$ along the scaled trajectory, as

$$\tau_s(kt) = M(q_s)q_s'' + c(q_s, q_s') + g(q_s)$$

however, being the dynamic model linear in the acceleration and quadratic in the velocity, it is

$$\tau_d(t) = M(q_d)\ddot{q}_d + c(q_d)\dot{q}_d + g(q_d) = M(q_s)k^2q_s'' + c(q_s,kq_s') + g(q_s)$$

$$= k^2(M(q_s)q_s'' + c(q_s,q_s')) + g(q_s) = k^2(\tau_s(kt) - g(q_s)) + g(q_s)$$

thus, saving separately the total torque $\tau_d(t)$ and gravity torque $g_d(t)$ in the inverse dynamics computation along the original trajectory, the new input torque is obtained directly as

$$\tau_S(kt) = \frac{1}{k^2} \left(\tau_d(t) - g(q_d(t)) \right) + g(q_d(t))$$

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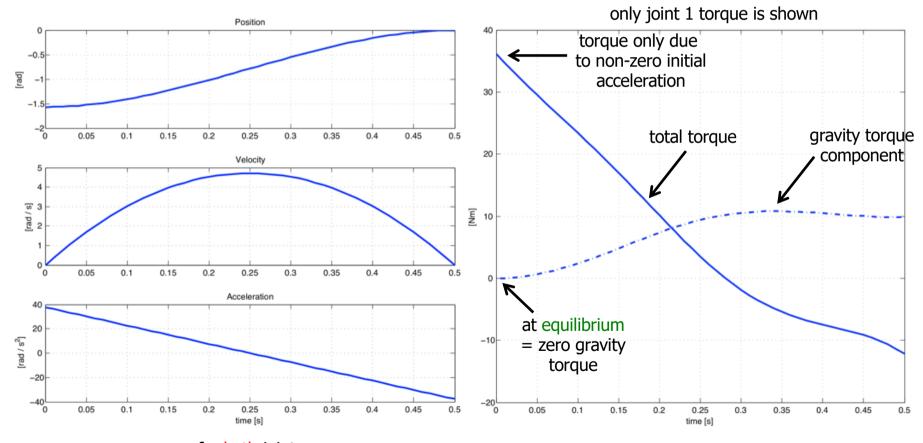
k > 1: slow down ⇒ increase torque

gravity term (only position-dependent): does **NOT** scale!

numerical example

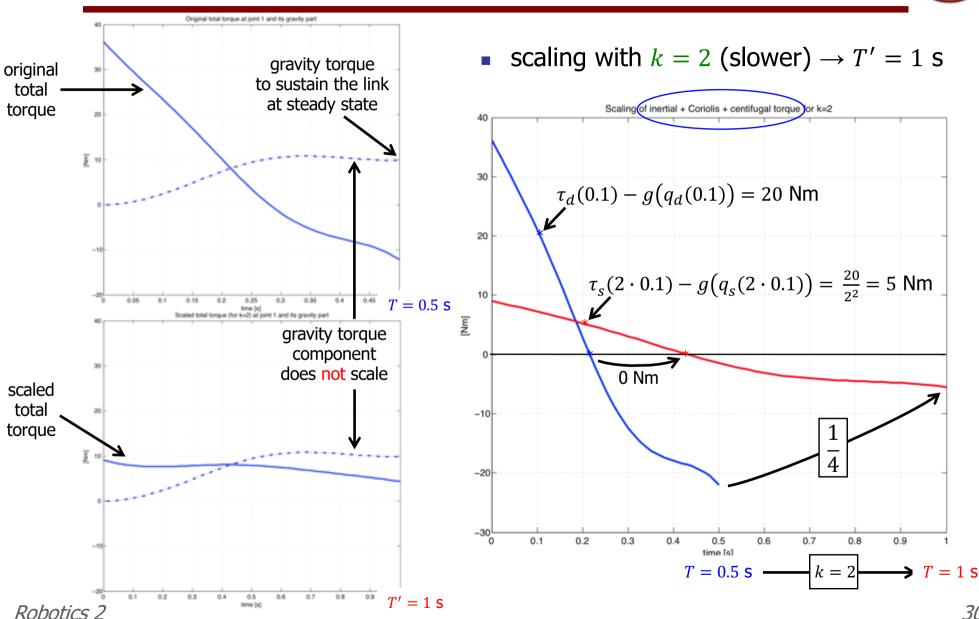


- rest-to-rest motion with cubic polynomials for planar 2R robot under gravity (from downward equilibrium to horizontal link 1 & upward vertical link 2)
- original trajectory lasts T = 0.5 s (but maybe violates the torque limit at joint 1)





numerical example

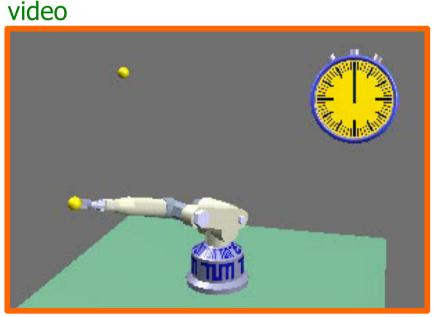


Optimal point-to-point robot motion

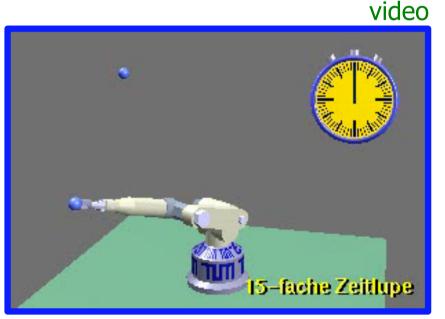
considering the dynamic model



- given the initial and final robot configurations (at rest) and actuator torque bounds, find
 - the minimum-time T_{min} motion
 - the (global/integral) minimum-energy E_{min} motion and the associated command torques needed to execute them
- a complex nonlinear optimization problem solved numerically



$$T_{min} = 1.32 \text{ s}, E = 306$$



$$T = 1.60 \text{ s}, E_{min} = 6.14$$