Motor control

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Motor control

is about the processes of bringing about the physical movement of an arm (robot or human)

📕 this entails

- the mechanical dynamics of an arm
- control principles



Resources

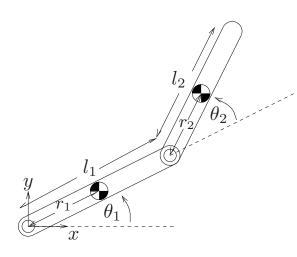
- R M Murray, Z Li, S S. Sastry: A mathematical introduction to robotic manipulation. CRC Press, 1994
- K M Lynch, F C Park: Modern Robotics: Mechanics, Planning, and Control. Cambridge University Press, 2017
- online version of both available...

Newton's law

- for a mass, m, described by a variable, x, in an inertial frame: $m\ddot{x} = f(x, t)$ where f is a force
- in non-inertial frames, e.g. rotating or accelerating frames:
 - 🧲 centripetal forces
 - Coriolis forces

Rigid bodies: constraints

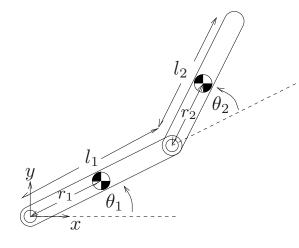
constraints reduce the effective numbers of degrees of freedom...



$$F_i = m_i \ddot{r}_i \qquad r_i \in \mathbb{R}^3, i = 1, \dots, n.$$
$$g_j(r_1, \dots, r_n) = 0 \qquad j = 1, \dots, k.$$

Rigid bodies: constraints

generalized coordinates capture the remaining, free degrees of freedom



$$r_i = f_i(q_1, \dots, q_m)$$
$$i = 1, \dots, n$$

$$g_j(r_1, \dots, r_n) = 0$$

$$j = 1, \dots, k.$$

Lagrangian mechanics

The Lagrangian framework makes it possible to capture dynamics in generalized coordinates that reflect constraints $L(q, \dot{q}) = T(q, \dot{q}) - V(q),$

Lagrange function L = kineticpotential energy

[Murray, Sastry, Li, 94]

Lagrangian mechanics

Least action principle: The integral of L over time=action is minimal $\delta A = \delta \int L(q, \dot{q}, t) dt = 0$

[Murray, Sastry, Li, 94]

Euler-Lagrange equation

$$\delta A = \int (\frac{\partial L}{\partial q} \delta q + \frac{\partial L}{\partial \dot{q}} \delta \dot{q}) dt = 0$$

with $\delta \dot{q} = d\delta q/dt$

and with partial integration

$$\delta A = \left[\frac{\partial L}{\partial \dot{q}}\delta q\right] + \int \left(\frac{\partial L}{\partial q} - \frac{d}{dt}\frac{\partial L}{\partial \dot{q}}\right)\delta q \, dt = 0$$

first term vanishes: no variation at start/end points

Euler-Lagrange equation

$$=>\frac{d}{dt}\frac{\partial L}{\partial \dot{q}} - \frac{\partial L}{\partial q} = 0$$

Image: Image: second state and secon

$$\frac{d}{dt}\frac{\partial L}{\partial \dot{q}} - \frac{\partial L}{\partial q} = \gamma$$

in component form:

$$\frac{d}{dt}\frac{\partial L}{\partial \dot{q}_i} - \frac{\partial L}{\partial q_i} = \Upsilon_i \qquad i = 1, \dots, m,$$

Example: pendulum

d generalized coordinates: $heta, \phi$

$$T = \frac{1}{2}ml^2 \|\dot{r}\|^2 = \frac{1}{2}ml^2 \left(\dot{\theta}^2 + (1 - \cos^2\theta)\dot{\phi}^2\right)$$

$$V = -mgl\cos\theta,$$

ma

$$L(q, \dot{q}) = \frac{1}{2}ml^2 \left(\dot{\theta}^2 + (1 - \cos^2 \theta)\dot{\phi}^2\right) + mgl\cos\theta$$
position relative to base
$$r(\theta, \phi) = \begin{bmatrix} l\sin\theta\cos\phi\\ l\sin\theta\sin\phi\\ -l\cos\theta \end{bmatrix}$$

Example: pendulum

$$\frac{d}{dt}\frac{\partial L}{\partial \dot{\theta}} = \frac{d}{dt}\left(ml^2\dot{\theta}\right) = ml^2\ddot{\theta}$$
$$\frac{\partial L}{\partial \theta} = ml^2\sin\theta\cos\theta\,\dot{\phi}^2 - mgl\sin\theta$$
$$\frac{d}{dt}\frac{\partial L}{\partial \dot{\phi}} = \frac{d}{dt}\left(ml^2\sin^2\theta\,\dot{\phi}\right) = ml^2\sin^2\theta\,\ddot{\phi} + 2ml^2\sin\theta\cos\theta\,\dot{\theta}\dot{\phi}$$
$$\frac{\partial L}{\partial \phi} = 0$$

$$\begin{bmatrix} ml^2 & 0 \\ 0 & ml^2 \sin^2 \theta \end{bmatrix} \begin{bmatrix} \ddot{\theta} \\ \ddot{\phi} \end{bmatrix} + \begin{bmatrix} -ml^2 \sin \theta \cos \theta \, \dot{\phi}^2 \\ 2ml^2 \sin \theta \cos \theta \, \dot{\theta} \dot{\phi} \end{bmatrix} + \begin{bmatrix} mgl \sin \theta \\ 0 \end{bmatrix} = 0.$$

inertial centrifugal gravitational (Coriolis)

Example: two-link planar robot

generalized coordinates: θ_1, θ_2

where $s_i = \sin(\theta_i), c_i = \cos(\theta_i)$

 $\begin{bmatrix} \alpha + 2\beta c_2 & \delta + \beta c_2 \\ \delta + \beta c_2 & \delta \end{bmatrix} \begin{bmatrix} \ddot{\theta}_1 \\ \ddot{\theta}_2 \end{bmatrix} + \begin{bmatrix} -\beta s_2 \dot{\theta}_2 & -\beta s_2 (\dot{\theta}_1 + \dot{\theta}_2) \\ \beta s_2 \dot{\theta}_1 & 0 \end{bmatrix} \begin{bmatrix} \dot{\theta}_1 \\ \dot{\theta}_2 \end{bmatrix} = \begin{bmatrix} \tau_1 \\ \tau_2 \end{bmatrix}$ inertial

centrifugal/Coriolis active torques

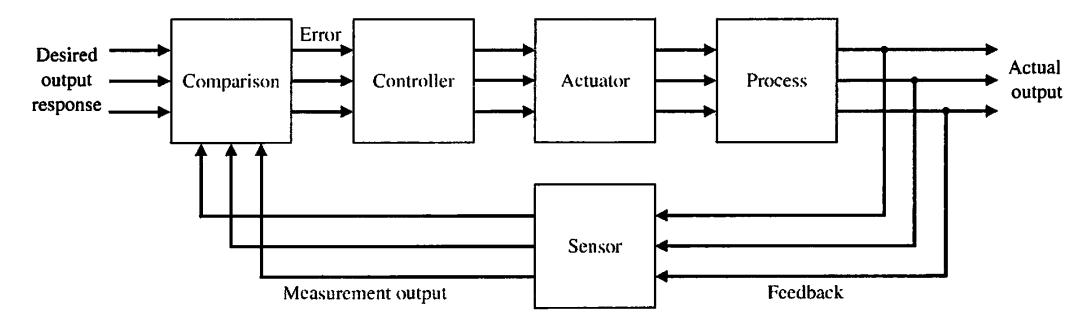
Open-chain manipulator

$$M(\theta)\ddot{\theta} + C(\theta,\dot{\theta})\dot{\theta} + N(\theta,\dot{\theta}) = \tau$$

inertial centrifugal/ gravitational active Coriolis torques

Control systems

robotic motion as a special case of control



[Dorf, Bishop, 2011]

Control systems

$$\dot{x} = f(t, x, u) \qquad \qquad y = \eta(t, x, u)$$

state of process/actuator x

loutput, y

control signal, u

[Dorf, Bishop, 2011]

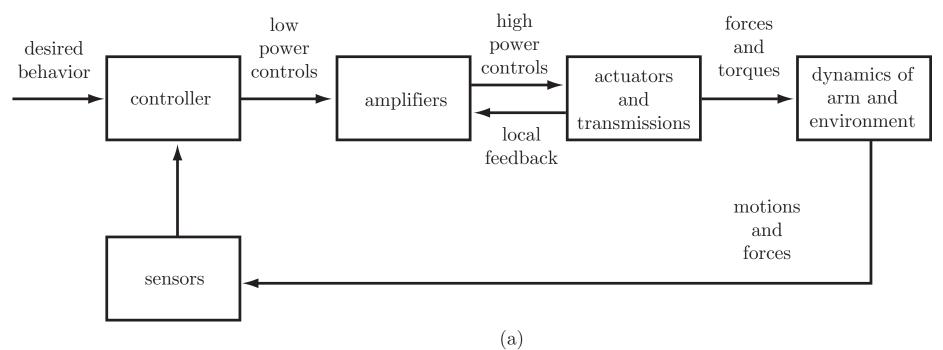
Control systems

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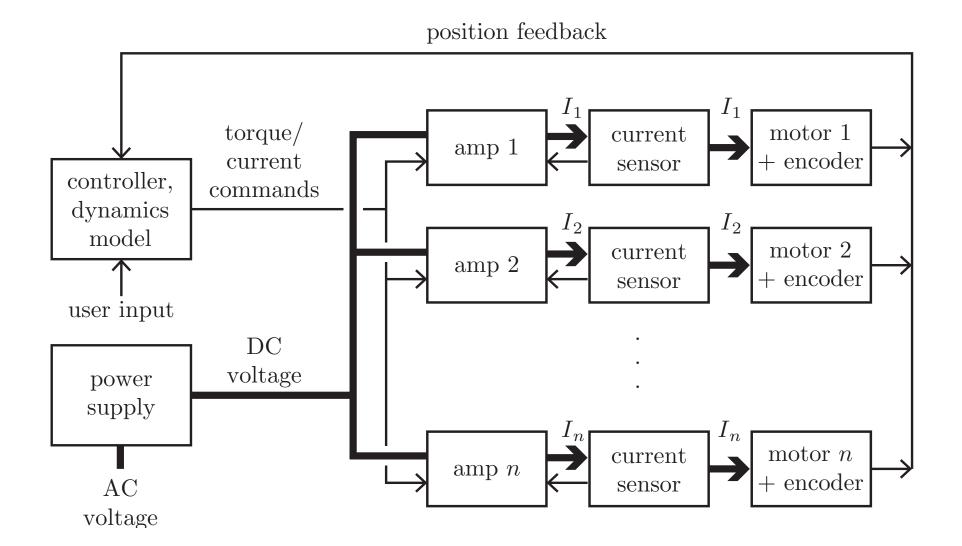
control law: u as a function of y (or \hat{y}), desired response, y_d

disturbances modeled stochastically

[Dorf, Bishop, 2011]



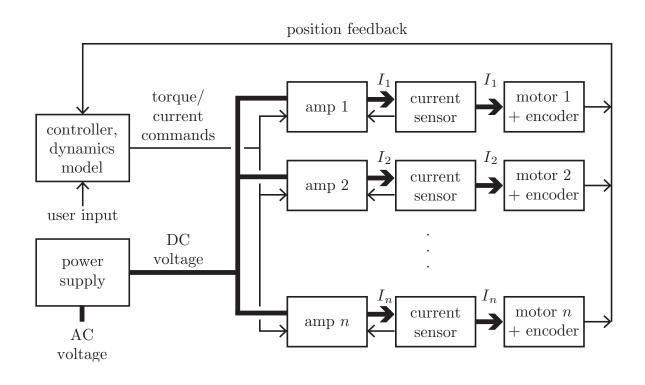
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actuators enable commanding a torque by commanding a current... in good approximation

[Lunch, Park, 2017]

=> control signal: torque



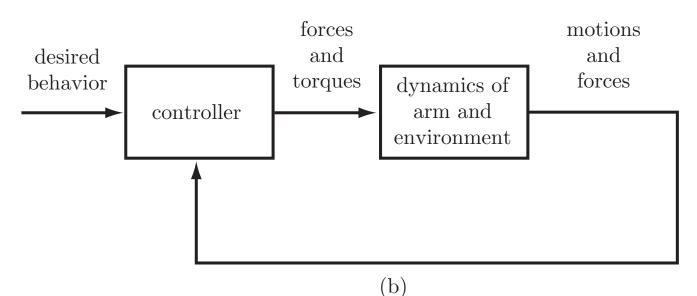
$$\mathbf{A}\dot{x} = f(t, x, u)$$

state variable x(t) = output: kinematic state of robot

[Lunch, Park, 2017]

desired trajectory: $x_d(t)$ (from motion planning)

control signal: u = torques

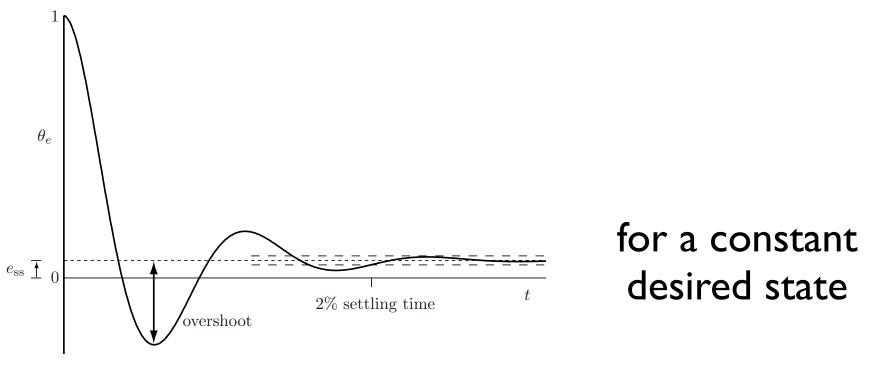


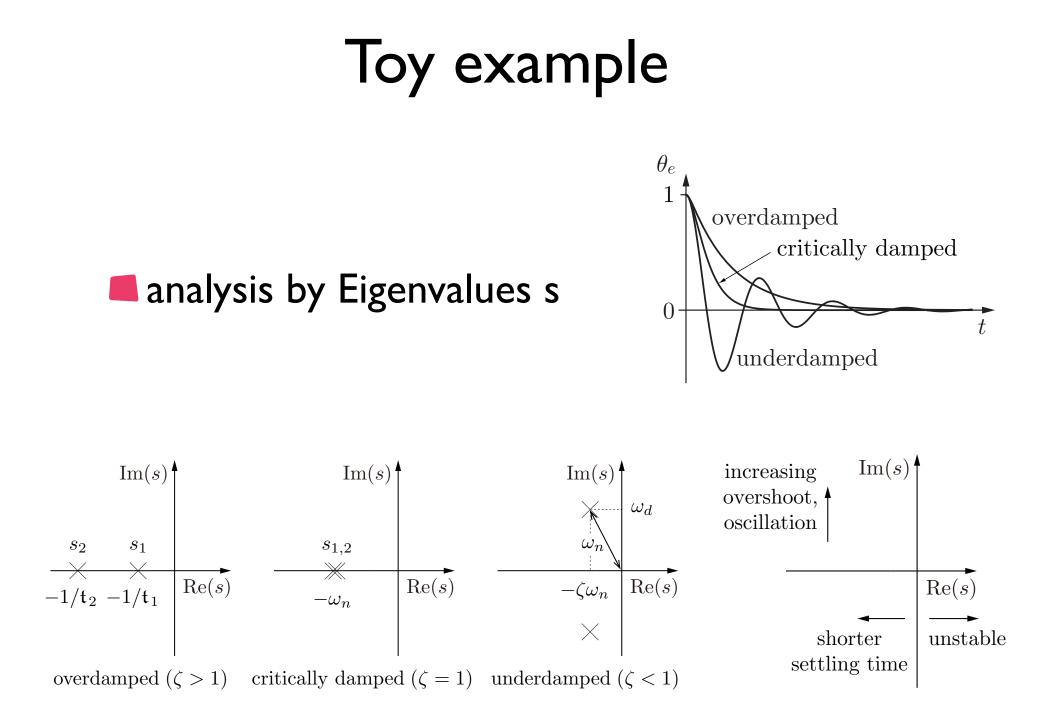
theoretical core of robotic control theory:

- devise control laws that lead to stable control
- (approximate these numerically on hardware and computers)

task: generate joint torques that produce a desired motion... $\theta_d(t)$

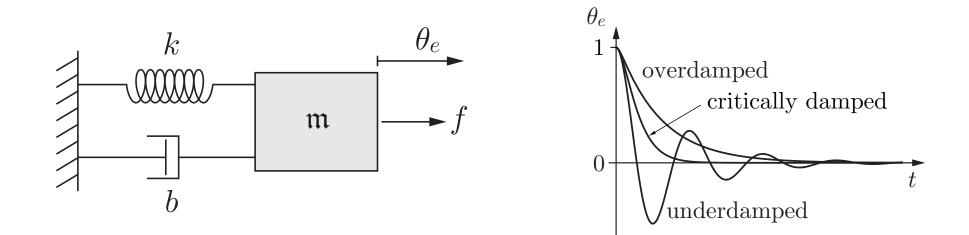
=> make error: $e(t) = \theta(t) - \theta_d(t)$ small



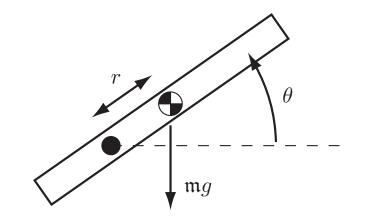


Toy example

linear mass spring model $m\ddot{e}(t) + b\dot{e}(t) + ke(t) = 0$



$= \tau = M\ddot{\theta} + mgr\cos(\theta) + b\dot{\theta}$



$$\tau = M\ddot{\theta} + mgr\cos(\theta) + b\dot{\theta}$$

feedback PID controller

$$\tau = K_p \theta_e + K_d \dot{\theta}_e + K_i \int \theta(t') dt'$$

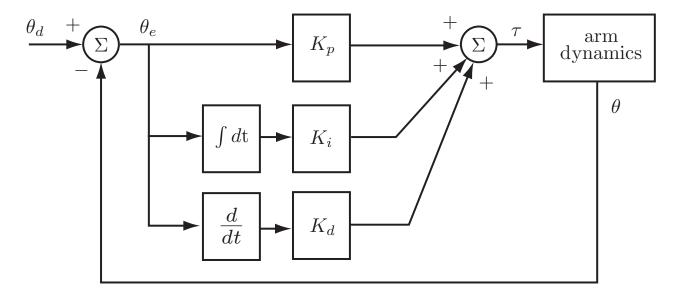
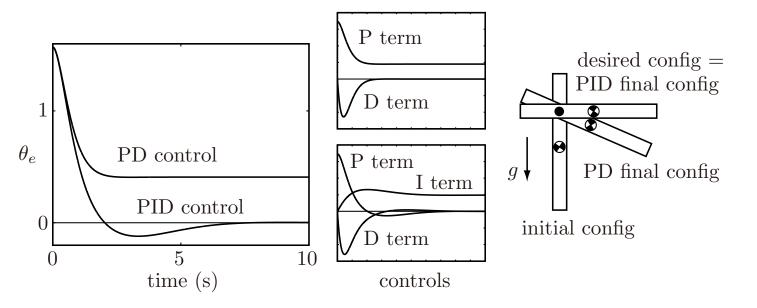


Figure 11.12: Block diagram of a PID controller.

$$\tau = M\ddot{\theta} + mgr\cos(\theta) + b\dot{\theta}$$

feedback PID controller

$$\tau = K_p \theta_e + K_d \dot{\theta}_e + K_i \int \theta(t') dt'$$



 $= \tau = M\ddot{\theta} + mgr\cos(\theta) + b\dot{\theta} = M\ddot{\theta} + h(\theta, \dot{\theta})$

feedforward controller

has model of the dynamics:

 $\blacksquare \tau = \tilde{M} \ddot{\theta} + \tilde{h}(\theta, \dot{\theta})$

compute forward torque

$$\boldsymbol{\tau}(t) = \tilde{M}(\theta_d(t)) \dot{\theta}_d(t) + \tilde{h}(\theta_d, \dot{\theta}_d)$$

if model exact: $\ddot{\theta} \approx \ddot{\theta}_d$

feedforward controller

if model wrong..

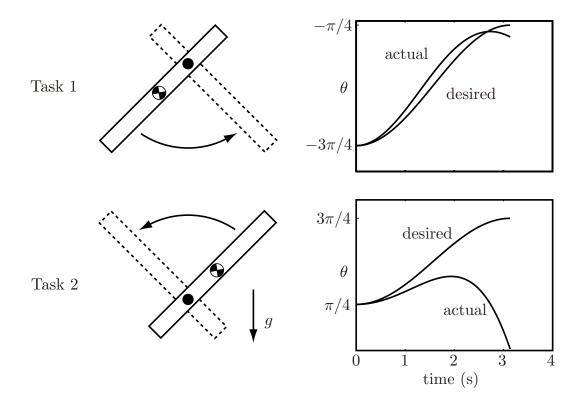


Figure 11.17: Results of feedforward control with an incorrect model: $\tilde{r} = 0.08$ m, but r = 0.1 m. The desired trajectory in Task 1 is $\theta_d(t) = -\pi/2 - (\pi/4)\cos(t)$ for $0 \le t \le \pi$. The desired trajectory for Task 2 is $\theta_d(t) = \pi/2 - (\pi/4)\cos(t), 0 \le t \le \pi$.

combined feedforward and feedback PID controller ...

$$\mathbf{T} = \tilde{M}(\theta) \left(\ddot{\theta}_d + K_p \theta_e + K_d \dot{\theta}_e + K_i \int \theta(t') dt' \right) + \tilde{h}(\theta, \dot{\theta})$$

inverse dynamics or computed torque controller

Control of multi-joint arm

generate joint torques that produce a desired motion... θ_d

$$\blacksquare \operatorname{error} \theta_e = \theta - \theta_d$$

PD control
$$\tau = K_p \theta_e + K_e \dot{\theta}_d + K_i \int \theta_e(t') dt'$$

=> controlling joints independently

$$M(\theta)\ddot{\theta} + C(\theta,\dot{\theta})\dot{\theta} + N(\theta,\dot{\theta}) = \tau$$

Control of multi-joint arm

there are many more sophisticated models that compensate for interaction torques/ inertial coupling... e.g. computed torque control (inverse dynamics)

$$\tau = \underbrace{M(\theta)\ddot{\theta}_d + C\dot{\theta} + N}_{\tau_{\rm ff}} + \underbrace{M(\theta)\left(-K_v\dot{e} - K_p e\right)}_{\tau_{\rm fb}}.$$

$$M(\theta)\ddot{\theta} + C(\theta,\dot{\theta})\dot{\theta} + N(\theta,\dot{\theta}) = \tau$$

$$\Rightarrow (\ddot{\theta} - \ddot{\theta}_d) = \ddot{e} = -K_v \dot{e} - K_p e$$

Control of multi-joint arm

- computed torque control (inverse dynamics)
- but: computational effort can be considerable... simplification.. only compensate for gravity...

$$\tau = K_p \theta_e + K_e \dot{\theta}_d + K_i \int \theta_e(t') dt' + \tilde{N}(\theta)$$

$$M(\theta)\ddot{\theta} + C(\theta,\dot{\theta})\dot{\theta} + N(\theta,\dot{\theta}) = \tau$$

Problem: contact forces

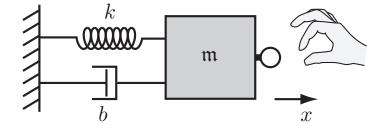
- as soon as the robot arm makes contact, a host of problems arise from the contact forces and their effect on the arm and controller...
- need compliance... resisting to a welldefined degree
- => impedance control... research frontier

Impedance

- to control movement well.. need a very stiff arm and "stiff" controller (high gain K_x)
- to control force/limit force (e.g. for interaction with surfaces or humans) you need a relatively soft arm and soft controller
- design system to give hand, x, a desired impedance: m, b, k in

$$\blacksquare m\ddot{x} + b\dot{x} + kx = f$$

where f is force applied..



Operational space formulation Euler-Langrage in end-effector space $\Lambda(x)\ddot{x} + \mu(x, \dot{x}) + p(x) = F$ with F forces acting on the end-effector equivalent dynamics in joint space $A(q)\ddot{q} + b(q, \dot{q}) + g(q) = \Gamma$ with joint torques $\Gamma = J^T(q)F$

[Khatib, 1987]

Impedance control

Hogan 1985... $\tau = J^{T}(\theta) \left(\tilde{\Lambda}(\theta) \ddot{x} + \tilde{\eta}(\theta, \dot{x}) - (M \ddot{x} + B \dot{x} + K x) \right)$

Link to movement planning

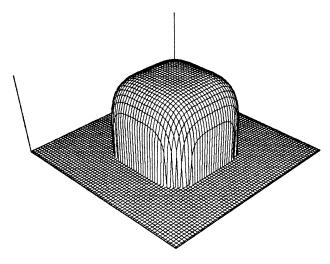
- where does "desired trajectory" come from?
- typically from end-effector level movement planning
 - then add an inverse kinematic...
 - which can be problematic
- alternative: planning and control in endeffector space

Operational space formulation

in end-effector space add constraints as contributions to the "virtual forces"

$$F_{\mathbf{x}_d}^* = -\operatorname{grad}[U_{\mathbf{x}_d}(\mathbf{x})],$$

$$F_O^* = -\operatorname{grad}[U_O(\mathbf{x})].$$



 $\Lambda(x)\ddot{x} + \mu(x, \dot{x}) + p(x) = F$

[Khatib, 1986,1987]

Optimal control

given a plant
$$\dot{x} = f(x, u)$$

find a control signal u(t)

that moves the state from an final position $x_i(0)$ to a terminal position $x_f(t_f)$ within the time t_f

a (difficult) planning problem!

minimize a cost function to find such a signal

How does the human (or other animal) movement system generate movement?

mechanics:... biomechanics

actuator: muscle

control? feedback loops