

Multi-joint kinematics and dynamics

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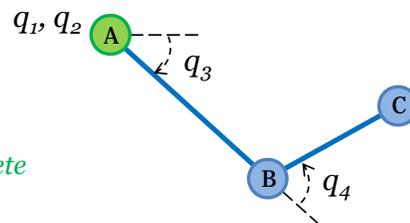
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Kinematics in generalized vs. Cartesian coordinates²

$$\begin{array}{ll} \text{generalized} & q = (q_1, q_2, q_3, q_4)^T \\ \text{Cartesian} & x = (A_x, A_y, B_x, B_y, C_x, C_y)^T \end{array}$$

dim(q) equals the number of degrees of freedom (DOFs)
dim(x) > dim(q), i.e. Cartesian coordinates are over-complete



forward kinematics:

always well-defined mapping from q to $x = h(q)$

$$A_x = q_1$$

$$A_y = q_2$$

$$B_x = q_1 + |AB| \cos(q_3)$$

$$B_y = q_2 + |AB| \sin(q_3)$$

$$C_x = q_1 + |AB| \cos(q_3) + |BC| \cos(q_3 + q_4)$$

$$C_y = q_2 + |AB| \sin(q_3) + |BC| \sin(q_3 + q_4)$$

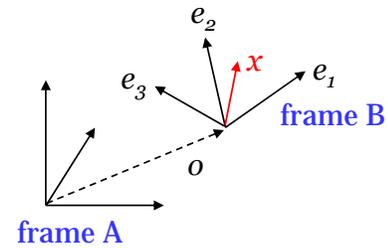
inverse kinematics:

usually not well-defined, but one can resolve redundancy via optimization:
of all q satisfying $x = h(q)$ for given x , pick the one closest to a preferred q^*

$$q = \arg \min_{\hat{q}: x=h(\hat{q})} \|\hat{q} - q^*\|$$

Attach a spatial frame to each body - usually at the center of mass, or at the center of the joint connecting it to its parent body. Rotations can be expressed as

$${}^A_B R = \begin{pmatrix} | & | & | \\ e_1 & e_2 & e_3 \\ | & | & | \end{pmatrix} \quad R^T = R^{-1}, \det(R) = 1$$



Transformation between frames: ${}^A x = {}^A_B O + {}^A_B R {}^B x$

Addition and multiplication can be combined using homogeneous coordinates:

$${}^A_B \hat{R} = \begin{pmatrix} {}^A_B R & {}^A_B O \\ \mathbf{0} & 1 \end{pmatrix} \quad \hat{x} = \begin{pmatrix} x \\ 1 \end{pmatrix} : \quad {}^A \hat{x} = {}^A_B \hat{R} {}^B \hat{x} \quad {}^A \hat{R} = {}^A_B \hat{R} {}^B \hat{R}$$

The spatial relations between body frames depend on the joint parameters q . Let $q_{(n)}$ be the parameters specifying the joint between body n and its parent. Then the corresponding transformation is parameterized as ${}^{\text{parent}(n)}_n \hat{R}(q_{(n)})$

Computing the forward kinematics involves a forward recursion:

$${}^{\text{world}}_n \hat{R} = {}^{\text{world}}_{\text{parent}(n)} \hat{R} {}^{\text{parent}(n)}_n \hat{R}(q_{(n)})$$

*Numerically, forward kinematics is more accurate using **quaternions** instead of 3x3 matrices.*

Rigid-body dynamics

Velocity

velocity (ω, \mathbf{v}_p) at P
 is equivalent to (ω, \mathbf{v}_o) at O
 where $\mathbf{v}_o = \mathbf{v}_p + \vec{OP} \times \omega$

Force

general force $(\mathbf{f}, \mathbf{n}_p)$ at P
 is equivalent to $(\mathbf{f}, \mathbf{n}_o)$ at O
 where $\mathbf{n}_o = \mathbf{n}_p + \vec{OP} \times \mathbf{f}$

Rigid Body Inertia

mass: m
 CoM: C
 inertia at CoM: \mathbf{I}_C

spatial inertia tensor: $\hat{\mathbf{I}}_O = \begin{bmatrix} \mathbf{I}_O & m \tilde{\mathbf{c}} \\ m \tilde{\mathbf{c}}^T & m \mathbf{1} \end{bmatrix}$
 where $\mathbf{I}_O = \mathbf{I}_C - m \tilde{\mathbf{c}} \tilde{\mathbf{c}}$

$\tilde{\mathbf{r}} = \begin{bmatrix} 0 & -r_z & r_y \\ r_z & 0 & -r_x \\ -r_y & r_x & 0 \end{bmatrix} \quad \tilde{\mathbf{r}} \mathbf{v} = \mathbf{r} \times \mathbf{v}$

Equation of Motion (Newton-Euler)

$$\mathbf{f} = \frac{d}{dt}(\mathbf{I}\mathbf{v}) = \mathbf{I}\mathbf{a} + \mathbf{v} \times \mathbf{I}\mathbf{v}$$

\mathbf{f} = net force acting on a rigid body
 \mathbf{I} = inertia of rigid body
 \mathbf{v} = velocity of rigid body
 $\mathbf{I}\mathbf{v}$ = momentum of rigid body
 \mathbf{a} = acceleration of rigid body

see Featherstone's slides on *Spatial Vector Algebra*

Newtonian mechanics with implicit constraints

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Newton's second law for a scalar point mass is $m\ddot{x} = f$

For a set of n point masses in 3D we have

$$\begin{pmatrix} m_1 I_3 & & \\ & \ddots & \\ & & m_n I_3 \end{pmatrix} \begin{pmatrix} \ddot{x}_{1,1} \\ \vdots \\ \ddot{x}_{n,3} \end{pmatrix} = \begin{pmatrix} f_{1,1} \\ \vdots \\ f_{n,3} \end{pmatrix}$$

which in vector notation is $D\ddot{x} = f$

Now consider a set of m positional **equality constraints** defined *implicitly* as $\phi(x) = 0$. They could specify that some masses belong to the same rigid body, or that some rigid bodies are constrained by joints, etc. The constraints eliminate m DOFs and create a $3n-m$ dimensional configuration manifold parameterized by q .

The constraint forces can only act within the null space, which is spanned by the rows of the Jacobian matrix $J = \frac{\partial \phi}{\partial x}$. Thus $f_{tot} = f + J^T \lambda$ for some m -dimensional vector λ , found by taking into account the differentiated constraints:

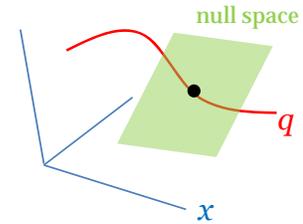
$$\dot{\phi} = J\dot{x}, \quad \ddot{\phi} = J\ddot{x} + \dot{J}\dot{x} = 0, \quad \text{where } \dot{J} = \sum_i \frac{\partial J}{\partial x_i} \dot{x}_i$$

The constrained dynamics $D\ddot{x} = f_{tot}$ are the solution to the linear in \ddot{x}, λ equation

$$\begin{pmatrix} D & -J^T \\ -J & 0 \end{pmatrix} \begin{pmatrix} \ddot{x} \\ \lambda \end{pmatrix} = \begin{pmatrix} f \\ \dot{J}\dot{x} \end{pmatrix}$$

The constrained dynamics are

$$D\ddot{x} = f - J^T (JD^{-1}J^T)^{-1} (\dot{J}\dot{x} + JD^{-1}f)$$



Constrained inertia and the Gauss principle

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When the system is stationary, the constrained dynamics simplify to $\ddot{x} = A f$ where A is the inverse of the constrained inertia matrix:

$$A = D^{-1} - D^{-1} J^T (JD^{-1} J^T)^{-1} JD^{-1}$$

There is no acceleration in the null space: $J\ddot{x} = J A f = 0$, which follows from

$$J A = JD^{-1} - JD^{-1} J^T (JD^{-1} J^T)^{-1} JD^{-1} = JD^{-1} - JD^{-1} = 0$$

A is singular, with $\text{rank}(A) = \dim(q)$.

Using the matrix inversion lemma, we can represent A as

$$A = \lim_{\varepsilon \rightarrow \infty} (D + \varepsilon J^T J)^{-1}$$

Thus the constrained inertia is " $D + \infty J^T J$ ", and is infinite in the null space.

The same results can be obtained from the more general **Gauss principle**: the constrained acceleration \ddot{x} is the solution to the minimization problem

$$\ddot{x} = \arg \min_a (a - \ddot{x}_0)^T D (a - \ddot{x}_0) \quad \text{s.t. } J a = b$$

\ddot{x}_0 is the unconstrained acceleration; J, b can encode general constraints.

Explicit constraints

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The implicitly-constrained dynamics $D\ddot{x} = f - J_\phi^T (J_\phi D^{-1} J_\phi^T)^{-1} (J_\phi \dot{x} + J_\phi D^{-1} f)$ are expressed in over-complete Cartesian coordinates (x), which is often undesirable. Instead it is better to express the dynamics in generalized (q) coordinates. This is done through explicit constraints given by the forward kinematics function $x = h(q)$

Differentiating the constraints twice yields $\ddot{x} = J(q)\ddot{q} + \dot{J}(q)\dot{q}$

The dynamics are $D\ddot{x} = f + f_c$ where f_c are the constraint forces. Since the columns of J span the tangent space to the manifold, $J(q)^T f_c = 0$

Assembling these equations, we obtain a system which is linear in \ddot{x}, \ddot{q}, f_c

$$\begin{pmatrix} D & -I & 0 \\ I & 0 & -J \\ 0 & J^T & 0 \end{pmatrix} \begin{pmatrix} \ddot{x} \\ f_c \\ \ddot{q} \end{pmatrix} = \begin{pmatrix} f \\ \dot{J}\dot{q} \\ 0 \end{pmatrix}$$

The constrained dynamics are

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

where

$$M = J^T D J$$
$$c = J^T D \dot{J} \dot{q}$$
$$\tau = J^T f$$

Coordinate transformations

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Consider any set of coordinates x , related to q as $x = h(q)$

Velocities in the two coordinate systems relate as $\dot{x} = J(q)\dot{q}$

Let f and τ denote the same force expressed in x and q coordinates respectively.

Power is coordinate-independent:

$$\dot{q}^T \tau = \dot{x}^T f = \dot{q}^T J^T f$$

Since this holds for any velocity, forces in the two coordinate systems relate as

$$\tau = J(q)^T f$$

Let D and M denote the same inertia expressed in x and q coordinates respectively.

Kinetic energy is coordinate-independent:

$$\dot{q}^T M \dot{q} = \dot{x}^T D \dot{x} = \dot{q}^T J^T D J \dot{q}$$

Since this holds for any velocity, inertias in the two coordinate systems relate as

$$M(q) = J(q)^T D(x) J(q)$$

Equality constraints are handled as in the case of point-mass dynamics:
we solve the linear in \ddot{q}, λ equation

$$\begin{aligned} M(q)\ddot{q} + c(q, \dot{q}) &= \tau + J(q)^T \lambda \\ J(q)\ddot{q} + \dot{J}(q)\dot{q} &= 0 \end{aligned}$$

Here the constraints are $\phi(q) = 0$ and the Jacobian is $J(q) = \frac{\partial \phi}{\partial q}$

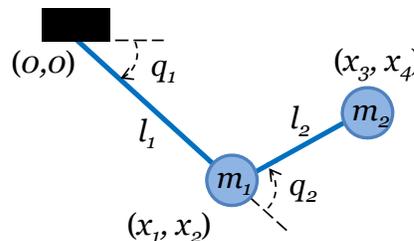
Equality constraints are often used to create kinematic loops (e.g. holding hands)

In simulations, the constraints can be violated numerically due to integration errors.
Thus it is necessary to introduce constraint stabilization (resembling PD control).

Example: 2-link arm

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$$D = \begin{pmatrix} m_1 & & & \\ & m_1 & & \\ & & m_2 & \\ & & & m_2 \end{pmatrix}$$



Implicit constraints:

$$\mathbf{0} = \phi(x) = \begin{pmatrix} x_1^2 + x_2^2 - l_1^2 \\ (x_3 - x_1)^2 + (x_4 - x_2)^2 - l_2^2 \end{pmatrix} \quad J_\phi(x) = 2 \begin{pmatrix} x_1 & x_2 & 0 & 0 \\ x_1 - x_3 & x_2 - x_4 & x_3 - x_1 & x_4 - x_2 \end{pmatrix}$$

Explicit constraints:

$$x = h(q) = \begin{pmatrix} l_1 \cos(q_1) \\ l_1 \sin(q_1) \\ l_1 \cos(q_1) + l_2 \cos(q_1 + q_2) \\ l_1 \sin(q_1) + l_2 \sin(q_1 + q_2) \end{pmatrix} \quad J_h(x) = \begin{pmatrix} -l_1 \sin(q_1) & 0 \\ l_1 \cos(q_1) & 0 \\ -l_1 \sin(q_1) - l_2 \sin(q_1 + q_2) & -l_2 \sin(q_1 + q_2) \\ l_1 \cos(q_1) + l_2 \cos(q_1 + q_2) & l_2 \cos(q_1 + q_2) \end{pmatrix}$$

Computing $M = J^T D J$ and $c = J^T D \dot{J} \dot{q}$ directly is inefficient.

Instead one can use faster algorithms exploiting the structure of kinematic trees.

Let s_i be the 6D motion vector of the (1-dof) joint connecting body i to its parent.

Composite Rigid Body algorithm for computing the inertia matrix $M(q)$

(1) backward recursion:

$$D_i^{\text{comp}} = D_i + \sum_{j \in \text{children}(i)} D_j^{\text{comp}}$$

(2) set:

$$M_{ij} = \begin{cases} s_j^T D_i^{\text{comp}} s_i & \text{if } i \in \text{descendants}(j) \\ s_j^T D_j^{\text{comp}} s_i & \text{if } j \in \text{descendants}(i) \\ 0 & \text{otherwise} \end{cases}$$

Recursive Newton-Euler algorithm for computing the inverse dynamics $(q, \dot{q}, \ddot{q}) \rightarrow \tau$

(1) forward recursion:

$$\dot{x}_i = \dot{x}_{\text{parent}(i)} + s_i \dot{q}_i$$

$$\ddot{x}_i = \ddot{x}_{\text{parent}(i)} + \dot{s}_i \dot{q}_i + s_i \ddot{q}_i$$

(2) backward recursion:

$$f_i = D_i \ddot{x}_i + \dot{x}_i \times D_i \dot{x}_i + \sum_{j \in \text{children}(i)} f_j$$

(3) set:

$$\tau_i = s_i^T f_i$$

running this algorithm with $\ddot{q} = 0$ yields $-c(q, \dot{q})$

Once M and c are computed, we can compute $\ddot{q} = M^{-1}(\tau - c)$ and integrate.

Dynamics in generalized coordinates

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

where $c_k(q, \dot{q}) = \sum_{ij} \Gamma_{ij,k}(q) \dot{q}_i \dot{q}_j$

$$\Gamma_{ij,k}(q) = \frac{1}{2} \left(\frac{\partial M_{ik}(q)}{\partial q_j} + \frac{\partial M_{jk}(q)}{\partial q_i} - \frac{\partial M_{ij}(q)}{\partial q_k} \right)$$

M inertia matrix

c Coriolis and centrifugal forces

g gravitational forces

τ applied/control forces

Γ Christoffel symbols

This can be derived from the **Euler-Lagrange equation**:

$$\frac{d}{dt} \frac{\partial L(q, \dot{q})}{\partial \dot{q}} - \frac{\partial L(q, \dot{q})}{\partial q} = \tau$$

where the *Lagrangian* is the kinetic energy minus the potential energy:

$$L(q, \dot{q}) = K(q, \dot{q}) - P(q)$$

$$K(q, \dot{q}) = \frac{1}{2} \dot{q}^T M(q) \dot{q}$$

$$P(q) = \sum_n 9.81 m_n h_n(q), \quad g(q) = -\frac{\partial P(q)}{\partial q}$$

If M does not depend on q , then $c=0$ and we have Newton's second law: $M \ddot{q} = \tau + g$

The same dynamics can be obtained from the equivalent Hamiltonian formulation, based on the *Hamiltonian* $H = K + P$ instead of the *Lagrangian* $L = K - P$. Now the state is represented in terms of q and the generalized momentum $p = M(q)\dot{q}$

H and L are related by the *Legendre transformation* $H = \dot{q}^T p - L$

Kinetic energy in the new coordinates is $K(q, p) = \frac{1}{2} p^T M(q)^{-1} p = \frac{1}{2} \dot{q}^T M(q) \dot{q} = K(q, \dot{q})$

Hamilton's equations are:
$$\dot{p} = -\frac{\partial H(q, p)}{\partial q} + \tau$$
$$\dot{q} = \frac{\partial H(q, p)}{\partial p}$$

The rate of change of the Hamiltonian (i.e. the total energy) equals power:

$$\frac{d}{dt} H(q, p) = \frac{\partial H}{\partial q^T} \dot{q} + \frac{\partial H}{\partial p^T} \dot{p} = \frac{\partial H}{\partial q^T} \frac{\partial H}{\partial p} - \frac{\partial H}{\partial p^T} \frac{\partial H}{\partial q} + \frac{\partial H}{\partial p^T} \tau = \dot{q}^T \tau$$

In the absence of external forces, the Hamiltonian is conserved.

Manifolds and metrics

Q is a differentiable manifold and $T_q Q$ the tangent space at point q . $T_q^* Q$ denotes the co-tangent (or dual) space.

A *metric* defines a dot-product on the tangent space:

$$\langle u, v \rangle_q = u^T M(q) v = \sum_{ij} M_{ij} u^i v^j \equiv M_{ij} u^i v^j \text{ (Einstein)}$$

The manifold is *Riemannian* if $M(q)$ is s.p.d. for all q .

The dot-product on the co-tangent space is defined by the inverse of M :

$$\langle u^*, v^* \rangle_q = u^{*T} M(q)^{-1} v^* = M^{ij} u_i v_j \quad \text{where} \quad (M^{ij}) \equiv (M_{ij})^{-1}, \quad u = (u^i), \quad u^* = (u_i)$$

The metric provides the mapping between the two spaces:

$$u^* = M u, \quad u = M^{-1} u^*; \quad \text{in coordinates, } u_i = M_{ij} u^j, \quad u^i = M^{ij} u_j$$

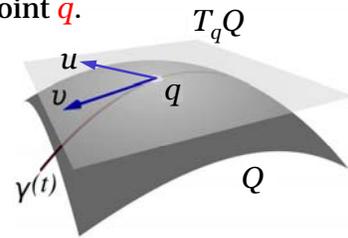
Tangent and co-tangent vectors are multiplied directly: $u^T v^* = u^i v_i = u^T M v$

Application to multi-joint dynamics:

The configuration space of a multi-joint system is a Riemannian manifold with metric given by the joint-space inertia matrix $M(q)$. The tangent vectors are velocities \dot{q} .

The co-tangent vectors are forces f and momenta $p = M(q)\dot{q}$.

$p^T \dot{q}$ is kinetic energy; $f^T \dot{q}$ is power.



The tangent basis vectors are associated with partial derivatives: $e_i = \partial/\partial q_i$

The co-tangent basis vectors are associated with differential forms: $\varepsilon_i = dq_i$

If $f(q)$ is scalar and $v = v^i e_i$ is a tangent vector, then vf is the directional derivative:

$$vf = v^i e_i f = v^i \frac{\partial f}{\partial q_i} = v^T \text{grad}(f)$$

A *connection* specifies how nearby coordinate frames “connect”, i.e. how the basis vectors change over the manifold. The usual vector directional derivative is replaced with the *covariant derivative*, defined in coordinates by the *Christoffel symbols* $\Gamma_{ij}^k(q)$

$$\nabla_{e_i} e_j = \Gamma_{ij}^k e_k$$

For general vectors $u = u^i e_i, v = v^i e_i$ the covariant derivative is $\nabla_v u = \left(v^i \frac{\partial u^k}{\partial q_i} + \Gamma_{ij}^k u^i v^j \right) e_k$

A connection is *flat* when $\Gamma_{ij}^k = 0$. In that case we recover the regular derivative.

For a Riemannian manifold with metric $M(q)$, there exists a unique metric-preserving torsion-free connection (the *Levi-Civita connection*) with Christoffel symbols:

$$\Gamma_{ij}^k = M^{ks} \Gamma_{ij,s} \quad \Gamma_{ij,s} = \frac{1}{2} \left(\frac{\partial M_{is}}{\partial q_j} + \frac{\partial M_{js}}{\partial q_i} - \frac{\partial M_{ij}}{\partial q_s} \right)$$

A *geodesic* is a curve $\gamma(t)$ such that $\nabla_{\dot{\gamma}} \dot{\gamma} = 0$, i.e. $\ddot{\gamma}^k + \Gamma_{ij}^k \dot{\gamma}^i \dot{\gamma}^j = 0$ for all k . This is called the *geodesic equation*.

Unforced motions as geodesics

The unforced motions of a multi-joint system satisfy $M(q)\ddot{q} + c(q, \dot{q}) = 0$

which we can rewrite (using the fact that M is s.p.d.) as $\ddot{q} + M(q)^{-1} c(q, \dot{q}) = 0$

Recalling the expression for c , this can be written in component form as

$$\ddot{q}_k + (M^{-1}c)_k = \ddot{q}_k + \sum_{ij} \Gamma_{ij}^k \dot{q}_i \dot{q}_j = 0$$

where $\Gamma_{ij}^k = \sum_s M_{ks}^{-1} \Gamma_{ij,s}$ and $\Gamma_{ij,s} = \frac{1}{2} \left(\frac{\partial M_{is}}{\partial q_j} + \frac{\partial M_{js}}{\partial q_i} - \frac{\partial M_{ij}}{\partial q_s} \right)$

Thus we have recovered the geodesic equation $\nabla_{\dot{q}} \dot{q} = 0$

The Levi-Civita connection for the Riemannian metric defined by the inertia matrix is called the **mechanical connection**. Its geodesics are the unforced motions.

With external forces and gravity, the dynamics become $M(q) \nabla_{\dot{q}} \dot{q} = \tau + g$

This is equivalent to Newton's second law,

with the covariant derivative in place of the regular derivative: $\frac{d}{dt} \dot{q} \rightarrow \nabla_{\dot{q}} \dot{q}$

Simulation software

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A number of physics engines are available, mostly free. All of them can now simulate complex systems (e.g. humanoids) faster than real time on a laptop.

Dealing with contacts (especially hard contacts) turns out to be harder than dealing with smooth dynamics, and is needed for stable and realistic simulations.

MuJoCo (our engine) is the only one optimized for control rather than simulation. It uses parallelism to compute gradients and Hessians of trajectory costs.

OpenSim is the only one that combines muscle simulation with multi-joint dynamics.

	coordinates	contacts	free
MATLAB Robotics Toolbox	q	n/a	yes + source
SD/Fast	q	n/a	no
OpenSim	q	spring	yes + source
NVidia PhysX	x	hard	yes
Intel Havoc	x	hard	yes
Open Dynamics Engine	x	hard	yes + source
MuJoCo	q	hard	yes + source

Speed comparison (no contacts)

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Number of forward dynamics evaluations per second,
single thread, 3GHz Intel processor

All models have around 30 DOFs

	MuJoCo	SF/FAST	MuJoCo + factor(M)
isolated	140,800	113,600	122,000
chain	100,000	125,000	36,800
hand	96,200	85,500	68,500
humanoid	116,300	125,000	67,600

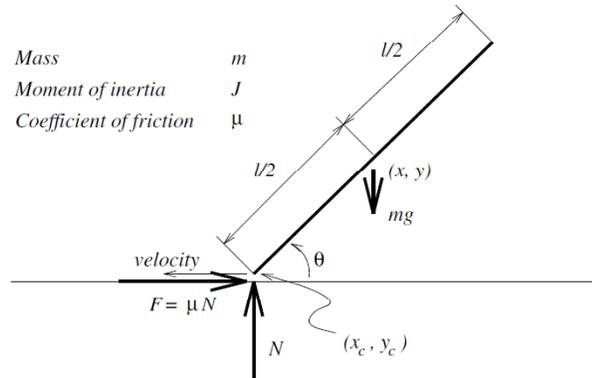
Continuous-time equations of motion
may not have a feasible solution!

Example: Painleve's problem

$$\begin{aligned} m\ddot{x} &= F, \\ m\ddot{y} &= N - mg, \\ J\ddot{\theta} &= (l/2)[+F \sin \theta - N \cos \theta] \end{aligned}$$

$$\ddot{y}_c = \ddot{y} + (l/2) \sin \theta \ddot{\theta} + (l/2) \cos \theta \dot{\theta}^2$$

$$\ddot{y}_c = \left[\frac{1}{m} - \frac{l^2}{4J} \cos \theta (\mu \sin \theta - \cos \theta) \right] N + (l/2) \cos \theta \dot{\theta}^2 - g$$



complementarity condition: $0 \leq \ddot{y}_c \perp N \geq 0$

When J/ml^2 is sufficiently small, the contact force generates a counter-clockwise torque, causing the rod to rotate *into the table*.

The general problem is that “rigid” bodies in contact no longer remain rigid, so all computationally-efficient models are approximations and fail in some cases.

see ICRA 2010 slides

(at the end of the PDF)

Motivation:

- complementary-based formulations yield non-convex NP-hard optimization problems
- a convex optimization problem can be obtained by relaxing complementarity
- this also yields an invertible contact model, which is very useful for trajectory optimization

Discrete-time dynamics in contact coordinates: $\mathbf{v}_{t+1} = \mathbf{c}_t + A_t \mathbf{f}_t$

\mathbf{v} – velocity after contact interaction

$$(A, \mathbf{c}) \xrightarrow{\text{forward}} (\mathbf{f}, \mathbf{v})$$

\mathbf{f} – contact force/impulse

$$(A, \mathbf{v}) \xrightarrow{\text{inverse}} (\mathbf{f}, \mathbf{c})$$

A – inverse inertia in contact space

\mathbf{c} – velocity in the absence of contact interaction (i.e. when $\mathbf{f}=\mathbf{0}$)

Forward dynamics:

the velocity minimizes the kinetic energy $\min_{\mathbf{f}, \mathbf{v}} \frac{1}{2} \mathbf{v}^\top A^{-1} \mathbf{v}$ subject to constraints:

- the contact force must be inside the friction cone $\mu_i^2 f_{i,3}^2 - f_{i,1}^2 - f_{i,2}^2 \geq 0$
- the normal force must be non-negative $N \mathbf{f} \geq 0$
- the normal velocity must avoid penetration $N \mathbf{v} - \mathbf{v}_{\min} \geq 0$

The objective is convex (quadratic) and the constraint set is convex (conic and linear inequalities).

E. Todorov, to appear in ICRA 2011

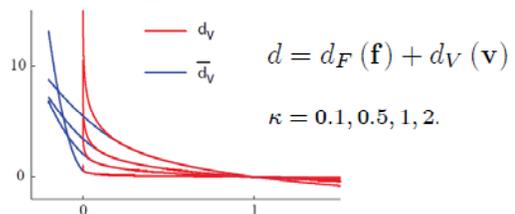
Interior-point minimization

Replace the inequality constraints $s_j(\mathbf{f}) \geq 0$ with the (modified) log-barrier function

$$d(\mathbf{f}) = -\kappa \sum_{j=1}^{2n_i + 3n_c} \log s_j(\mathbf{f})$$

The kinetic energy cost is

$$\ell(\mathbf{f}) = \frac{1}{2} \mathbf{f}^\top (A + R) \mathbf{f} + \mathbf{f}^\top \mathbf{c}$$



Forward dynamics then comes down to unconstrained minimization of the composite cost

$$\ell(\mathbf{f}) + d(\mathbf{f})$$

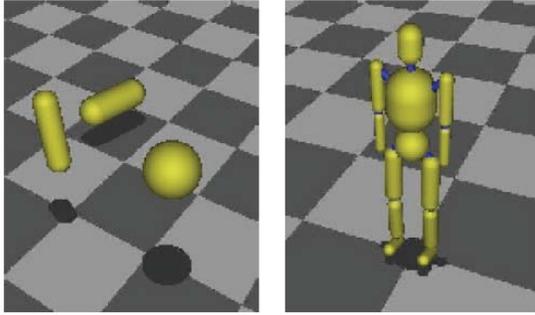
Inverse dynamics

The (now unknown) \mathbf{f} must satisfy $\nabla_{\mathbf{f}} \ell + \nabla_{\mathbf{f}} d = 0$

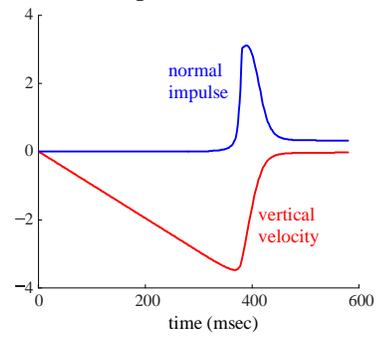
which can be expanded as $R \mathbf{f} + \mathbf{v} + \nabla_{\mathbf{f}} d_F + A \nabla_{\mathbf{v}} d_V = 0$

The solution to this nonlinear equation is the unique minimizer of the convex cost

$$\frac{1}{2} \mathbf{f}^\top R \mathbf{f} + d_F(\mathbf{f}) + \mathbf{f}^\top (\mathbf{v} + A \nabla_{\mathbf{v}} d_V(\mathbf{v}))$$



ball-drop test



numerical performance for inverse dynamics

# contacts	inverse evaluations per second
2	241,774
5	54,190
10	12,991
20	2,361

