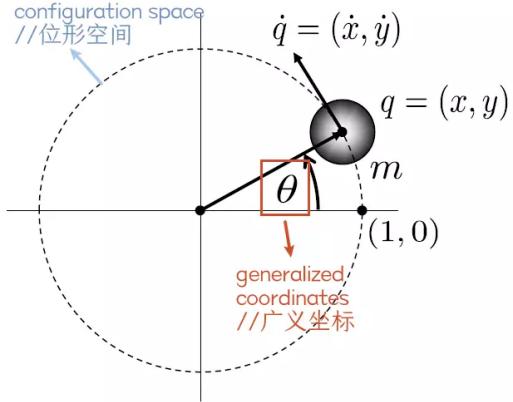


Lecture 03 / 3D Mechanics

Basic concepts / 基本概念

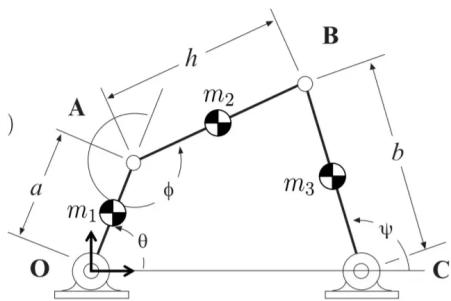
Configuration space, Generalized coordinates / 构型空间与广义坐标



//在表示圆形轨迹时使用笛卡尔坐标的问题：多解与无解点（导数求速度时会出现）

in this case, the configuration space $\mathcal{C} = \{\theta \mid \theta \in [0, 2\pi)\}$

e.g. 四连杆质心位置计算



$$\phi = \arctan \left(\frac{b \sin \psi - a \sin \theta}{g + b \cos \psi - a \cos \theta} \right) - \theta$$

$$\psi(\theta) = \arctan \left(\frac{B}{A} \right) \pm \arccos \left(\frac{C}{\sqrt{A^2 + B^2}} \right)$$

Kinetic energy / 动能

$$T = \frac{1}{2} m \dot{q}^T \dot{q}$$

$$\frac{d}{dt} \left(\frac{\partial T}{\partial \dot{q}} \right) = \frac{d}{dt} (m \dot{q}) = f \quad \left(\text{Given } \dot{q} = \begin{bmatrix} \dot{x} \\ \dot{y} \\ \dot{z} \end{bmatrix}, \frac{\partial T}{\partial \dot{q}} = \begin{bmatrix} \partial T / \partial \dot{x} \\ \partial T / \partial \dot{y} \\ \partial T / \partial \dot{z} \end{bmatrix} \right)$$

//速度(speed)是标量

$$\text{//多个质点: } T = \frac{1}{2} \sum_{i=1}^n m_i \dot{q}_i^T \dot{q}_i$$

注意：刚体质点之间有作用力，不可以简单求导计算单个质点受力

e.g. 四连杆动能计算

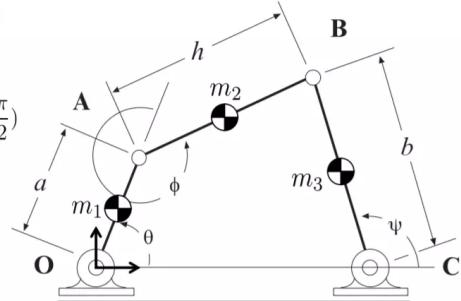
1. 求质心坐标（上个例题）

2. 用位置坐标对时间求导 + 利用速度间关系得到质心处的速度

3. 求系统动能

Assume center of mass of the links are located at the mid-point of the links, compute the kinetic energy of the 4-bar linkage as a function of θ and $\dot{\theta}$ (assume $\phi(\theta), \psi(\theta), \dot{\phi}(\theta, \dot{\theta}), \dot{\psi}(\theta, \dot{\theta})$ are given)

$$\frac{\dot{\psi}}{\dot{\theta}} = \frac{(\mathbf{B} - \mathbf{A})^T J \mathbf{B}}{(\mathbf{B} - \mathbf{A})^T J (\mathbf{B} - \mathbf{C})}, \quad \frac{\dot{\phi}}{\dot{\theta}} = \frac{-(\mathbf{B} - \mathbf{C})^T J \mathbf{B}}{(\mathbf{B} - \mathbf{C})^T J (\mathbf{B} - \mathbf{A})} \quad J = R\left(\frac{\pi}{2}\right)$$



Work function, Potential energy / 功函数与势能

功函数：

If there exists a *work function* $U(q)$ such that:

$$f = \frac{\partial U(q)}{\partial q}$$

we have:

$$\int_{t_0}^{t_f} dW = \int_{t_0}^{t_f} \frac{\partial U}{\partial q} dq = U(q(t_f)) - U(q(t_0))$$

势能：

Putting the kinetic energy and work function together:

$$\int_{t_0}^{t_f} m \ddot{q}^T dq = \int_{t_0}^{t_f} \frac{d}{dt} \left(\frac{1}{2} m \dot{q}^T \dot{q} \right) dt = \int_{t_0}^{t_f} f^T dq = \int_{t_0}^{t_f} \left(\frac{\partial U}{\partial q} \right)^T dq$$



$$T(t_f) - T(t_0) = U(t_f) - U(t_0)$$

or

$$T(t_f) - U(t_f) = T(t_0) - U(t_0)$$

We usually define *potential energy* $V(q) = -U(q)$



$$E(t) \triangleq T(t) + V(t) = \text{constant}$$

The total energy E remains constant. $f = -\partial V / \partial q$ is therefore called a *conservative force*.

//保守力做功与路径无关；典型的非保守力：摩擦力

//弹簧模型：仿真形变计算基础

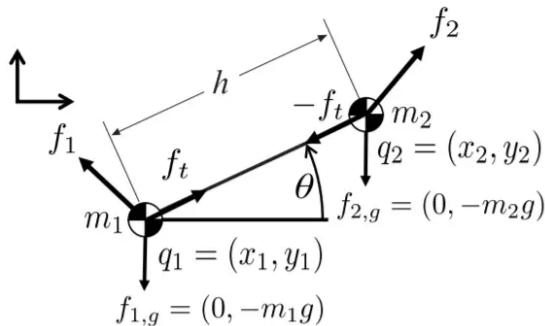
Principle of virtual work / 虚功原理

Virtual displacements, Variation

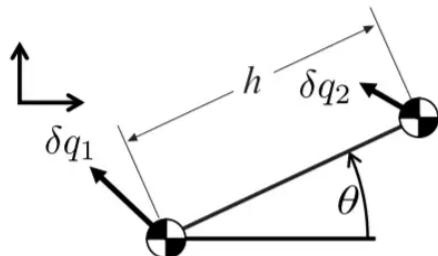
静力平衡:

$$f_1 + f_t + f_{1,g} = 0$$

$$f_2 - f_t + f_{2,g} = 0$$



假设有沿切向的微小位移 (实际上并没有发生) :



$$\delta W \triangleq (f_1 + f_t + f_{1,g})^T \delta q_1 + (f_2 - f_t + f_{2,g})^T \delta q_2 = 0$$

虚功原理(Principle of Virtual Work, or PVW):

向任何方向的微小位移都不做功 = 系统平衡

虚位移(Virtual displacements): $\delta q_1, \delta q_2 \dots$

// δq_1 与 dq_1 : 相似, 但是考虑虚位移时系统时间停止

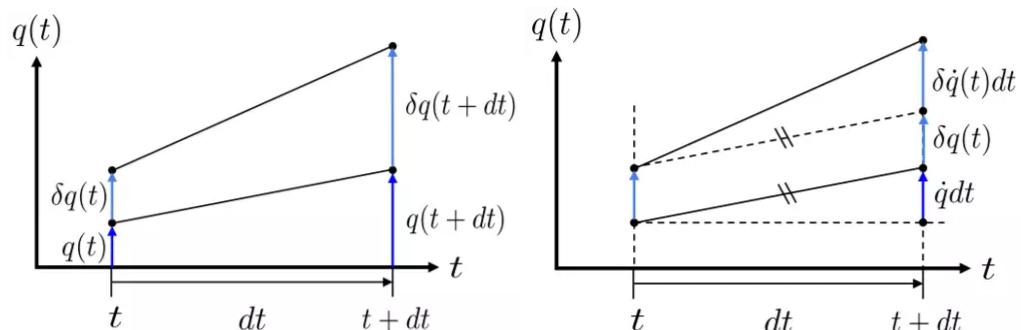
$$(q_1 - q_2)^T (\dot{q}_1 - \dot{q}_2) = 0$$

$$(q_1 - q_2)^T (\delta q_1 - \delta q_2) = 0$$

// δq_1 、 δq_2 的取值可以是满足以上关系任意一组值, 我们可以令 $\delta q_1 = dq_1$!

虚功(Virtual work): δW

虚位移对速度的影响:



给计算带来的好处: 类似偏导, 可以提出来

$$\delta q(t+dt) = \delta q(t) + \delta \dot{q}(t)dt \Rightarrow \frac{d}{dt}(\delta q(t)) = \delta \dot{q}(t)$$

$$d\delta q(t) = \delta dq(t)$$

$$\delta \int_{t_0}^{t_f} q(t)dt = \int_{t_0}^{t_f} \delta q(t)dt$$

使用虚功原理的计算：

1. 移除约束力

Recall we also have:

$$(q_1 - q_2)^T (\delta q_1 - \delta q_2) = 0$$

Since f_t is parallel to the vector $q_1 - q_2$, we have:

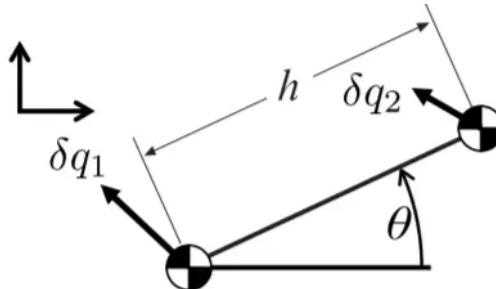
$$f_t^T (\delta q_1 - \delta q_2) = 0$$

$$\delta W = (f_1 + f_{1,g})^T \delta q_1 + (f_2 + f_{2,g})^T \delta q_2 + f_t^T (\delta q_1 - \delta q_2) = 0$$



$$\delta W = (f_1 + f_{1,g})^T \delta q_1 + (f_2 + f_{2,g})^T \delta q_2 = 0$$

2. 虚位移之间的关系



$$\delta q_2 = \delta q_1 + hv\delta\theta, \quad v = \begin{bmatrix} -\sin\theta \\ \cos\theta \end{bmatrix}$$

$$\delta W = (f_1 + f_{1,g} + f_2 + f_{2,g})^T \delta q_1 + h(f_2 + f_{2,g})^T v \delta\theta = 0$$

Since $\delta q_1, \delta\theta$ are linear independent, we have:

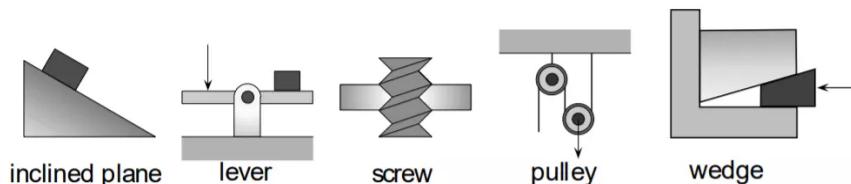
$$f_1 + f_{1,g} + f_2 + f_{2,g} = 0$$

$$(f_2 + f_{2,g})^T v = 0$$

Mechanical advantage

$$f_{in}^T \delta q_{in} - f_{out}^T \delta q_{out} = 0$$

$$f_{in}^T dq_{in}/dt - f_{out}^T dq_{out}/dt = f_{in}^T \dot{q}_{in} - f_{out}^T \dot{q}_{out} = 0 \quad \rightarrow \quad \boxed{\frac{f_{out}}{f_{in}} = \frac{v_{in}}{v_{out}}}$$



Rigid body

$$\boxed{\sum_{i=1}^n f_i^T \delta q_i = 0} \quad \text{where } \delta q_i = \hat{\omega}_{ab} q_i + v_{ab}$$

$$\sum_{i=1}^n (f_i^T \hat{\omega}_{ab} q_i) + v_{ab}^T (\sum_{i=1}^n f_i) = \omega_{ab}^T \sum_{i=1}^n (q_i \times f_i) + v_{ab}^T (\sum_{i=1}^n f_i) = 0$$

$$\sum_{i=1}^n f_i = 0, \quad \sum_{i=1}^n (q_i \times f_i) = 0$$

Define $F_i \triangleq \begin{bmatrix} f_i \\ q_i \times f_i \end{bmatrix} \triangleq \begin{bmatrix} f_i \\ f_i^T J q_i \end{bmatrix}$, called wrenches

// J和叉乘都用于表示法向向量，J更适合矩阵运算



$$\sum_{i=1}^n F_i = 0$$

Differentiation / 微分

微分与逼近 (Approximation)

常数逼近

Definition. A function $f(x)$ is approximated by a **constant** a near x_0 , denoted $f(x) \sim_1 a$, if for any $\epsilon > 0$, there is $\delta > 0$, such that

$$|\Delta x| < \delta \implies |f(x) - a| = |f(x) - a| < \epsilon \cdot 1 = \epsilon.$$

常数逼近与函数连续性

Theorem. Constant approx \iff continuous, and $a = f(x_0)$.

Theorem. Suppose $f(x) \sim_1 a$ and $g(x) \sim_1 b$ near x_0 .

- ▶ If $a > b$, then $f(x) > g(x)$ near x_0 .
- ▶ $f(x) + g(x) \sim_1 a + b$
- ▶ $f(x)g(x) \sim_1 ab$.

线性逼近 (需要平滑)

Definition. Linear approximation $f(x) \sim_{\Delta x} a + b\Delta x$, if for any $\epsilon > 0$, there is $\delta > 0$, such that

$$|\Delta x| < \delta \implies |f(x) - a - b\Delta x| \leq \epsilon |\Delta x|.$$

We say f is **differentiable** at x_0 .

导数：逼近的计算

Definition. The derivative of a function $f(x)$ at x_0 is the coefficient b in the linear approximation $f(x) \sim_{\Delta x} a + b\Delta x$.

$$f'(x_0) = b = \lim_{x \rightarrow x_0} \frac{f(x) - f(x_0)}{x - x_0} = \lim_{\Delta x \rightarrow 0} \frac{f(x_0 + \Delta x) - f(x_0)}{\Delta x}.$$

Example. Find linear approximation of $f(x) = x^3$ at x_0

$$f(x) = (x_0 + \Delta x)^3 = x_0^3 + 3x_0^2 \Delta x + 3x_0 \Delta x^2 + \Delta x^3$$

We have $a = x_0^3 = f(x_0)$, and $b = 3x_0^2$.

Leibniz rule: 逼近的乘法

Suppose $f \sim_{\Delta x} a + b\Delta x$ and $g \sim_{\Delta x} c + d\Delta x$ near x_0 .

Then

$$\begin{aligned} f + g &\sim_{\Delta x} (a + b\Delta x) + (c + d\Delta x) \\ &= (a + c) + (b + d)\Delta x. \end{aligned}$$

We get

$$(f + g)'(x_0) = b + d = f'(x_0) + g'(x_0).$$

Then

$$\begin{aligned} fg &\sim_{\Delta x} (a + b\Delta x)(c + d\Delta x) \\ &= ac + (ad + bc)\Delta x + bd\Delta x^2 \\ &\sim_{\Delta x} ac + (ad + bc)\Delta x. \end{aligned}$$

We get the Leibniz rule

$$(fg)'(x_0) = ad + bc = f(x_0)g'(x_0) + f'(x_0)g(x_0).$$

链式法则：逼近的复合

Suppose $y = f(x) \sim_{\Delta x} a + b\Delta x$ and $g(y) \sim_{\Delta y} c + d\Delta y$ near $y_0 = f(x_0)$.

Then (e.g.: $(1 + 2y) \circ (3 + 4x) = 1 + 2(3 + 4x) = 7 + 8x$)

$$\begin{aligned} g(f(x)) &\sim_{\Delta x} (c + d\Delta y) \circ (a + b\Delta x) \\ &= c + db\Delta x. \end{aligned}$$

We get the chain rule

$$(g \circ f)'(x_0) = db = g'(y_0)f'(x_0) = g'(f(x_0))f'(x_0).$$

*链式法则基于可以微分成立...

*可导一定可微，可微不一定可导

高阶导数：真的难算，只用组合即可

Theorem. If $f^{(n)}(x_0)$ exists, then

$$f(x) \sim_{\Delta x^n} f(x_0) + f'(x_0)\Delta x + \frac{f''(x_0)}{2!}\Delta x^2 + \cdots + \frac{f^{(n)}(x_0)}{n!}\Delta x^n.$$

Basic high order approximations at $x_0 = 0$ ($\Delta x = x$):

$$(1 + x)^p = 1 + px + \frac{p(p-1)}{2!}x^2 + \cdots + \frac{p(p-1)\cdots(p-n+1)}{n!}x^n + o(x^n),$$

$$e^x = 1 + x + \frac{1}{2!}x^2 + \cdots + \frac{1}{n!}x^n + o(x^n),$$

$$\log(1 + x) = x - \frac{1}{2}x^2 + \cdots + (-1)^{n-1}\frac{1}{n}x^n + o(x^n),$$

$$\cos x = 1 - \frac{1}{2!}x^2 + \frac{1}{4!}x^4 - \cdots + (-1)^n \frac{1}{(2n)!}x^{2n} + o(x^{2n+1}),$$

$$\sin x = x - \frac{1}{3!}x^3 + \frac{1}{5!}x^5 - \cdots + (-1)^n \frac{1}{(2n+1)!}x^{2n+1} + o(x^{2n+2}).$$

This is the last time you use high order derivatives!!!

多变量逼近

*偏导数是多变量逼近的系数!

Definition. Linear approximation $f(\vec{x}) \sim_{\Delta \vec{x}} a + b(\Delta \vec{x})$, if \dots ,

$$\|\Delta \vec{x}\| < \delta \implies |f(\vec{x}) - a - b(\Delta \vec{x})| \leq \epsilon \|\Delta \vec{x}\|.$$

We say f is **differentiable** at x_0 , and denote the **(total) derivative**

$$f'(\vec{x}_0) = b, \quad f'(\vec{x}_0)(\vec{v}) = \vec{b} \cdot \vec{v} = \nabla f(\vec{x}_0) \cdot \vec{v}.$$

Calculation:

$$a = f(x_{10}, x_{20}), \quad b_1 = \frac{\partial f}{\partial x_1}(x_{10}, x_{20}), \quad b_2 = \frac{\partial f}{\partial x_2}(x_{10}, x_{20}).$$

线性变换与雅可比行列式:

More generally, consider multifunction of multivariable

$$F(\vec{x}) = (f_1(\vec{x}), \dots, f_m(\vec{x})) : \mathbb{R}^n \rightarrow \mathbb{R}^m$$

A linear map (multifunction) is

$$P(\vec{x}) = \vec{a} + B\Delta \vec{x}, \quad \Delta \vec{x} = \vec{x} - \vec{x}_0, \quad \vec{a} \in \mathbb{R}^m.$$

$B : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is a linear transformation, or an $m \times n$ matrix.

Definition. A map $F(\vec{x}) : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is approximated by a linear map $\vec{a} + B\Delta \vec{x}$ near \vec{x}_0 , and denoted $F(\vec{x}) \sim_{\Delta \vec{x}} \vec{a} + B\Delta \vec{x}$, if for all $\epsilon > 0$, there is $\delta > 0$, such that

$$\|\Delta \vec{x}\| < \delta \implies \|F(\vec{x}) - \vec{a} - B\Delta \vec{x}\| \leq \epsilon \|\Delta \vec{x}\|.$$

We have

$$\vec{a} = F(\vec{x}_0), \quad B = F'(\vec{x}_0) = \begin{pmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \frac{\partial f_1}{\partial x_3} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} & \frac{\partial f_2}{\partial x_3} \end{pmatrix}.$$

多变量的Leibniz rule和链式法则:

- **Leibniz rule:** $F \sim_{\|\Delta \vec{x}\|^n} P$ and $G \sim_{\|\Delta \vec{x}\|^n} Q \implies FG \sim_{\|\Delta \vec{x}\|^n} PQ$, and cutting terms of order $> n$ in PQ .
- **Chain rule:** $F(\vec{x}) \sim_{\|\Delta \vec{x}\|^n} P(\vec{x})$ near \vec{x}_0 , and $G(\vec{y}) \sim_{\|\Delta \vec{y}\|^n} Q(\vec{y})$ near $\vec{y}_0 = f(\vec{x}_0) \implies G(F(\vec{x})) \sim_{\|\Delta \vec{x}\|^n} Q(P(\vec{x}))$, and cutting terms of order $> n$.

Chain rule $(G \circ F)'(\vec{x}_0) = G'(\vec{y}_0)F'(\vec{x}_0)$ means

$$\begin{pmatrix} \frac{\partial g_1(f_1, f_2, f_3)}{\partial x_1} & \frac{\partial g_1(f_1, f_2, f_3)}{\partial x_2} \\ \frac{\partial g_2(f_1, f_2, f_3)}{\partial x_1} & \frac{\partial g_2(f_1, f_2, f_3)}{\partial x_2} \end{pmatrix} = \begin{pmatrix} \frac{\partial g_1}{\partial y_1} & \frac{\partial g_1}{\partial y_2} \\ \frac{\partial g_2}{\partial y_1} & \frac{\partial g_2}{\partial y_2} \end{pmatrix} \begin{pmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \frac{\partial f_1}{\partial x_3} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} & \frac{\partial f_2}{\partial x_3} \end{pmatrix}.$$

线性逼近可以求逆 \rightarrow 函数可逆 (对矩阵也一致)

矩阵的逆映射与求导:

Example.

Derivative of inverse matrix map $F(X) = X^{-1} : \mathbb{R}^{n^2} \rightarrow \mathbb{R}^{n^2}$.

(1) Derivative $F'(I)$ at $X_0 = I$. Denote $H = \Delta X = X - I$.

$$X^{-1} = (I + H)^{-1} = I - H + H^2 - H^3 + \dots$$

Therefore $F'(I)(H) = -H$, $\frac{1}{2!}F''(I)(H) = H^2$, $\frac{1}{3!}F'''(I)(H) = -H^3$

(2) Derivative $F'(A)$ at $X_0 = A$. F is the composition:

$$X_A \xrightarrow{A^{-1} ?} A^{-1} X_I \xrightarrow{?^{-1}} X^{-1} A_I \xrightarrow{?A^{-1}} X^{-1}.$$

The maps $X \rightarrow A^{-1}X$ and $X \rightarrow XA^{-1}$ are linear. Therefore

$$(A^{-1}?)'(A)(H) = A^{-1}H, \quad (?A^{-1})'(I)(H) = HA^{-1}.$$

Then

$$F'(A)(H) = [(?A^{-1})'(I) \circ F'(I) \circ (A^{-1}?)'(A)](H) = -A^{-1}HA^{-1}.$$

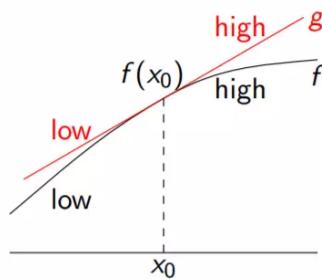
Local Extreme

→ 线性逼近 (Linear Approximation) : 导数判别法背后的原理

Suppose $f(x) \sim_{\Delta x} p(x) = a + b\Delta x$. If $b > 0$, then

high on right and low on left \implies high on right and low on left

\implies not local extreme.



Similarly, $b < 0 \implies$ not local extreme.

Theorem. If f is two-sided differentiable at x_0 , then x_0 is a local

◆ extreme $\implies f'(x_0) = 0$.

◆ 一定要两边都能逼近：论证中需要此条件

→ 二次逼近 (Quadratic Approximation)

Suppose $f(x) \sim_{\Delta x^2} p(x) = a + c\Delta x^2$ (this means $f'(x_0) = 0$).

$c > 0 \implies p(x) > a = p(x_0)$ for $\Delta x \neq 0$ (local minimum of p)

$\implies f(x) > f(x_0)$ for $\Delta x \neq 0$ (local minimum of f).

Theorem. Suppose $f(x) \sim_{\Delta x^2} p(x) = a + c\Delta x^2$ near x_0 .

► If $c > 0$, then x_0 is a local minimum.

► If $c < 0$, then x_0 is a local maximum.

Remark. By Taylor expansion, we have $f'(x_0) = b = 0$ and $f''(x_0) = 2c > 0 \implies$ local minimum.

Theorem. Suppose $f(x) \sim_{\Delta x^n} p(x) = a + c\Delta x^n$ near x_0 .

► If n is odd and $c \neq 0$, then x_0 is not a local extreme.

► If n is even and $c > 0$, then x_0 is a local minimum.

◆ ► If n is even and $c < 0$, then x_0 is a local maximum.

◆ n 次逼近中低阶项变成常数，所以只考虑高阶项

→ 双变量与二次逼近

- ◆ 通过二次型正定/负定判别确定c的正负

Suppose $f(\vec{x}) \sim \|\Delta\vec{x}\|^2$ $p(\vec{x}) = a + c(\Delta\vec{x})$, where c is a quadratic function

$$c(v_1, v_2) = c_{11}v_1^2 + c_{22}v_2^2 + 2c_{12}v_1v_2.$$

Theorem.

$c(\vec{v}) > 0$ for $\vec{v} \neq \vec{0}$ (positive definite) \implies local minimum.

$c(\vec{v}) < 0$ for $\vec{v} \neq \vec{0}$ (negative definite) \implies local maximum.

Remark. In terms of partial derivatives,

$$b_1 = f_{x_1}(\vec{x}_0) = 0, b_2 = f_{x_2}(\vec{x}_0) = 0,$$

$$c_{11} = \frac{1}{2}f_{x_1^2}(\vec{x}_0), c_{22} = \frac{1}{2}f_{x_2^2}(\vec{x}_0), c_{12} = \frac{1}{2}f_{x_1x_2}(\vec{x}_0).$$

Remark. Positive definite $\iff c_{11} > 0$ and $c_{11}c_{22} - c_{12}^2 > 0$

- ◆ $\iff f_{x_1^2} > 0$ and $f_{x_1^2}f_{x_2^2} - f_{x_1x_2}^2 > 0$.

→ 引入约束域

- ◆ 线性逼近转换题目

Problem. Find local extreme $\vec{x}_0 = (x_0, y_0)$ of $f(x, y) = xy$ on the circle $g(x, y) = x^2 + y^2 = 1$.

We have $g(\vec{x}_0) = 1$, and need linear approximation for all data:

$$\begin{aligned} f(\vec{x}) \sim p_f(\vec{x}) &= f(\vec{x}_0) + f_x(\vec{x}_0)\Delta x + f_y(\vec{x}_0)\Delta y \\ &= f(\vec{x}_0) + y_0\Delta x + x_0\Delta y, \\ g(\vec{x}) \sim p_g(\vec{x}) &= g(\vec{x}_0) + 2x_0\Delta x + 2y_0\Delta y. \end{aligned}$$

Approximate Problem. Find local extreme $\vec{x}_0 = (x_0, y_0)$ of $p_f = a + y_0\Delta x + x_0\Delta y$ on $p_g = 1$.

By $g(\vec{x}_0) = 1$, the constraint $p_g = 1$ is $2x_0\Delta x + 2y_0\Delta y = 0$

$$\implies \Delta y = -\frac{x_0}{y_0}\Delta x \implies p_f = a + \frac{y_0^2 - x_0^2}{y_0}\Delta x$$

Consider both $\Delta x > 0$ and < 0 , we need $y_0^2 - x_0^2 = 0$.

We also need $g(\vec{x}_0) = x_0^2 + y_0^2 = 1 \implies x_0 = \pm\frac{1}{\sqrt{2}}, y_0 = \pm\frac{1}{\sqrt{2}}$.

- ◆ 二重约束

Problem. Find local extreme \vec{x}_0 of $f(\vec{x})$ subject to the constraints $g_1(\vec{x}) = c_1$ and $g_2(\vec{x}) = c_2$.

We have $g_1(\vec{x}_0) = c_1$ and $g_2(\vec{x}_0) = c_2$.

We need linear approximation for all data:

$$\begin{aligned} f(\vec{x}) &\sim p_f(\vec{x}) = f(\vec{x}_0) + \nabla f(\vec{x}_0) \cdot \Delta \vec{x}, \\ g_1(\vec{x}) &\sim p_{g_1}(\vec{x}) = g_1(\vec{x}_0) + \nabla g_1(\vec{x}_0) \cdot \Delta \vec{x}, \\ g_2(\vec{x}) &\sim p_{g_2}(\vec{x}) = g_2(\vec{x}_0) + \nabla g_2(\vec{x}_0) \cdot \Delta \vec{x}. \end{aligned}$$

Approximate Problem. Find local extreme \vec{x}_0 of p_f subject to

$$p_{g_1} = c_1 \text{ and } p_{g_2} = c_2.$$

Necessary condition:

$$\nabla g_1(\vec{x}_0) \cdot \Delta \vec{x} = \nabla g_2(\vec{x}_0) \cdot \Delta \vec{x} = 0 \implies \nabla f(\vec{x}_0) \cdot \Delta \vec{x} = 0.$$

This is equivalent to

$$\nabla f(\vec{x}_0) = \lambda_1 \nabla g_1(\vec{x}_0) + \lambda_2 \nabla g_2(\vec{x}_0).$$

Differentiation on Manifold (流形)

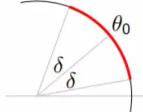
→ 圆上的微分

◆ 用角度 θ 做参数: (存在映射问题)

Function $f(x, y) = xy$ on the circle $S^1 = \{(x, y) : x^2 + y^2 = 1\}$.
Parameterise circle by angle θ

$$x = \cos \theta, \quad y = \sin \theta, \quad f(x, y) = \cos \theta \sin \theta.$$

The part of the circle corresponding to $|\theta - \theta_0| < \delta$ is an interval.



Differentiation at $(x_0, y_0) = (\cos \theta_0, \sin \theta_0)$:

$$\begin{aligned} f(x, y) &= (\cos \theta_0 - \sin \theta_0 \Delta \theta + o(\Delta \theta))(\sin \theta_0 + \cos \theta_0 \Delta \theta + o(\Delta \theta)) \\ &= f(x_0, y_0) + \cos 2\theta_0 \Delta \theta + o(\Delta \theta). \end{aligned}$$

◆ Local extreme $\implies \cos 2\theta_0 = 0 \implies (x_0, y_0) = \frac{1}{\sqrt{2}}(\pm 1, \pm 1)$.

◆ 用 x / y 做参数 (略复杂)

Parameterise by x (only covering the upper half circle)

$$y = \sqrt{1 - x^2}, \quad f(x, y) = x \sqrt{1 - x^2}.$$

The part of the circle corresponding to $|x - x_0| < \delta$ is an interval.



Differentiation at $(x_0, y_0) = (x_0, \sqrt{1 - x_0^2})$:

$$f(x, y) = f(x_0, y_0) + (1 - x_0^2)^{\frac{1}{2}} \left(1 - \frac{x_0^2}{1 - x_0^2} \right) \Delta x + o(\Delta x)$$

$$\text{Local extreme } \implies 1 - \frac{x_0^2}{1 - x_0^2} = 0 \implies (x_0, y_0) = \frac{1}{\sqrt{2}}(\pm 1, 1).$$

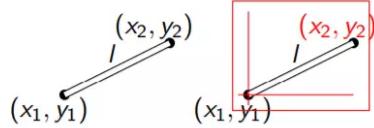
◆ Also need to cover the lower half circle by $y = -\sqrt{1 - x^2}$.

→ 球上的微分

◆ 用 (x, y) 或 (θ, ρ)

→ 机械系统中的例子:

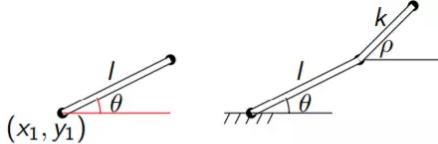
- ◆ Configuration of one link: 表示一根连杆所有的可能位置



Configuration of one link

$$\begin{aligned}\mathcal{C} &= \{(x_1, x_2, x_3, x_4) : (x_1 - x_2)^2 + (y_1 - y_2)^2 - l^2 = 0\} \subset \mathbb{R}^4 \\ &= \{(x_1, y_1, x_2, y_2) : x_2^2 + y_2^2 - l^2 = 0\} = \mathbb{R}^2 \times \mathbb{S}^1 \\ \blacksquare &= \{(x_1, y_1, \theta)\}.\end{aligned}$$

- ◆ Configuration of two links:



Configuration of two links

$$\mathcal{C} = \{(\theta, \rho)\} = \mathbb{S}^1 \times \mathbb{S}^1 = \mathbb{T}^2.$$

- Both are manifolds.
- ◆ 旋转矩阵
All rotations in \mathbb{R}^2

$$SO(2) = \{R_\theta\}, \quad \text{Note: } R_\theta = R_{\theta+2\pi}.$$

All the rotations of \mathbb{R}^3

$$SO(3) = \{R_{\vec{\omega}, \theta}\}, \quad \text{Note: } R_{\vec{\omega}, \theta} = R_{-\vec{\omega}, -\theta} = R_{\vec{\omega}, \theta+2\pi}.$$

Here $\vec{\omega} \in \mathbb{S}^2$ is the axis of rotation, and θ is the rotation angle (as measured by the right hand rule).

- Both are manifolds.
- Presenting Manifold / 流形的表示
 - ◆ Parameterisation / 参数化表示
 - The parameters are **generalised coordinates**.
 - Two problems:
 - ▶ May need several parameterisation pieces to cover the whole manifold.
 - ▶ Two parameters may correspond to the same point (e.g., θ and $\theta + 2\pi$).
 - Not affecting differentiation. Can be solved by defining **atlas of coordinate charts**.
- ◆ Holonomic constraints

$$(x - x_0)^2 + (y - y_0)^2 = R^2 \quad x^2/a^2 + y^2/b^2 + z^2/d^2 = 1 \quad (x - x_0)^2 + (y - y_0)^2 + (z - z_0)^2 = R^2$$

Subset of \mathbb{R}^n satisfying k independent equations

$$f_1(x_1, \dots, x_n) = 0, f_2(x_1, \dots, x_n) = 0, \dots, f_k(x_1, \dots, x_n) = 0.$$

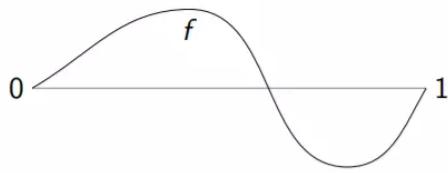
Example

$$SO(n) = \{U : U^T U = I\} \subset \mathbb{R}^{n^2}.$$

Variation / 变分

Calculus of Variation: Shortest path / 最小路径问题

Find shortest path between two points.



The path is given by $y = f(x)$, $x \in [0, 1]$, $f(0) = f(1) = 0$.
The length of the path is a function of f

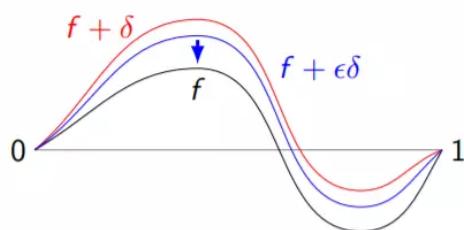
$$\mathcal{L}(f) = \int_0^1 \sqrt{1 + f'^2} dx.$$

By shortest path, we mean

$$\mathcal{L}(g) \geq \mathcal{L}(f) \quad \text{for any function } g \text{ close to } f.$$

- 从数字到函数 (泛函)

Take nearby $g(x) = f(x) + \delta(x)$, where $\delta(x)$ is a small function satisfying $\delta(0) = \delta(1) = 0$.



For each $\delta(x)$, further consider the family of nearby $g(x) = f(x) + \epsilon\delta(x)$. Change the problem to

$$\mathcal{L}(f(x) + \epsilon\delta(x)) \geq \mathcal{L}(f(x)).$$

Then the function $\mathcal{L}(f(x) + \epsilon\delta(x))$ of ϵ minimizes at $\epsilon = 0$.

- 继续用逼近

$$\begin{aligned}
\mathcal{L}(f(x) + \epsilon\delta(x)) &= \int_0^1 \sqrt{1 + (f'(x) + \epsilon\delta'(x))^2} dx \\
&= \int_0^1 \sqrt{1 + f'(x)^2 + 2\epsilon f'(x)\delta'(x) + o(\epsilon)} dx \\
&= \int_0^1 \sqrt{1 + f'(x)^2} \left(1 + \epsilon \frac{2f'(x)\delta'(x)}{1 + f'(x)^2} + o(\epsilon) \right)^{\frac{1}{2}} dx \\
&= \int_0^1 \sqrt{1 + f'(x)^2} \left(1 + \frac{1}{2}\epsilon \frac{2f'(x)\delta'(x)}{1 + f'(x)^2} + o(\epsilon) \right) dx \\
&= \int_0^1 \left(\sqrt{1 + f'(x)^2} + \epsilon \frac{f'(x)\delta'(x)}{\sqrt{1 + f'(x)^2}} + o(\epsilon) \right) dx \\
&= \mathcal{L}(f(x)) + \epsilon \int_0^1 \frac{f'(x)\delta'(x)}{\sqrt{1 + f'(x)^2}} dx + o(\epsilon).
\end{aligned}$$

The red term should vanish.

- 对所有 δ 积分都等于0 $\rightarrow F'(x)=0 \rightarrow F(x)$ 是常数 $\rightarrow f(x)$ 是常数
 $\mathcal{L}(f(x) + \epsilon\delta(x))$ minimises at $\epsilon = 0$.

$$\begin{aligned}
\int_0^1 \frac{f'(x)\delta'(x)}{\sqrt{1 + f'(x)^2}} dx &= \int_0^1 \frac{f'(x)}{\sqrt{1 + f'(x)^2}} d\delta(x) \\
&= \frac{f'(x)}{\sqrt{1 + f'(x)^2}} \delta(x) \Big|_{x=0}^{x=1} - \int_0^1 \frac{d}{dx} \left(\frac{f'(x)}{\sqrt{1 + f'(x)^2}} \right) \delta(x) dx \\
&= - \int_0^1 \frac{d}{dx} \left(\frac{f'(x)}{\sqrt{1 + f'(x)^2}} \right) \delta(x) dx. \quad (\text{by } \delta(0) = \delta(1) = 0)
\end{aligned}$$

The above should vanishes for all $\delta(x)$

$$\begin{aligned}
&\Rightarrow \frac{d}{dx} \left(\frac{f'(x)}{\sqrt{1+f'(x)^2}} \right) = 0 \Rightarrow \frac{f'(x)}{\sqrt{1+f'(x)^2}} = \text{constant} \\
&\Rightarrow f'(x) = \text{constant}, \text{ i.e., straight line.}
\end{aligned}$$

- 一般化推广

Problem. For all $f(a) = A$ and $f(b) = B$, minimise

$$\mathcal{L}(f) = \int_a^b L(t, f, f') dt.$$

把t,f,f'都当成L的变量

This becomes minimising the following at $\epsilon = 0$

$$\begin{aligned}
\mathcal{L}(f + \epsilon\delta) &= \int_a^b L(t, f + \epsilon\delta, f' + \epsilon\delta') dt \\
&= \int_a^b \left[L(t, f, f') + \epsilon \frac{\partial L}{\partial f}(t, f, f') \delta + \epsilon \frac{\partial L}{\partial f'}(t, f, f') \delta' + o(\epsilon) \right] dt \\
&= \mathcal{L}(f) + \epsilon \int_a^b \frac{\partial L}{\partial f}(t, f, f') \delta dt + \epsilon \int_a^b \frac{\partial L}{\partial f'}(t, f, f') d\delta + o(\epsilon) \\
&= \mathcal{L}(f) + \epsilon \int_a^b \frac{\partial L}{\partial f} \delta dt + \epsilon \frac{\partial L}{\partial f'} \delta \Big|_{t=a}^{t=b} - \int_a^b \frac{d}{dt} \left(\frac{\partial L}{\partial f'} \right) \delta dt + o(\epsilon) \\
&= \mathcal{L}(f) + \epsilon \int_a^b \left(\frac{\partial L}{\partial f} - \frac{d}{dt} \left(\frac{\partial L}{\partial f'} \right) \right) \delta dt + o(\epsilon).
\end{aligned}$$

Necessary condition: For all $\delta(t)$, we have

$$\int_a^b \left(\frac{\partial L}{\partial f} - \frac{d}{dt} \left(\frac{\partial L}{\partial f'} \right) \right) \delta(t) dt = 0.$$

This is equivalent to Euler-Lagrange equation

$$\frac{\partial L}{\partial f} - \frac{d}{dt} \left(\frac{\partial L}{\partial f'} \right) = 0.$$

- 多变量推广

Problem. For $f_1(a) = A_1, f_1(b) = B_1, f_2(a) = A_2, f_2(b) = B_2$, minimise

$$\mathcal{L}(f_1, f_2) = \int_a^b L(t, f_1, f_2, f'_1, f'_2) dt.$$

This becomes minimising the following at $\epsilon = 0$

$$\begin{aligned} \mathcal{L}(f_1 + \epsilon \delta_1, f_2 + \epsilon \delta_2) &= \int_a^b L(t, f_1 + \epsilon \delta_1, f_2 + \epsilon \delta_2, f'_1 + \epsilon \delta'_1, f'_2 + \epsilon \delta'_2) dt \\ &= \int_a^b \left[L(t, f_1, f_2, f'_1, f'_2) + \epsilon \frac{\partial L}{\partial f_1}(t, f_1, f_2, f'_1, f'_2) \delta_1 + \epsilon \frac{\partial L}{\partial f_2}(t, f_1, f_2, f'_1, f'_2) \delta_2 \right. \\ &\quad \left. + \epsilon \frac{\partial L}{\partial f'_1}(t, f_1, f_2, f'_1, f'_2) \delta'_1 + \epsilon \frac{\partial L}{\partial f'_2}(t, f_1, f_2, f'_1, f'_2) \delta'_2 + o(\epsilon) \right] dt \\ &= \mathcal{L}(f_1, f_2) + \epsilon \int_a^b \left(\frac{\partial L}{\partial f_1} - \frac{d}{dt} \left(\frac{\partial L}{\partial f'_1} \right) \right) \delta_1 dt \\ &\quad + \epsilon \int_a^b \left(\frac{\partial L}{\partial f_2} - \frac{d}{dt} \left(\frac{\partial L}{\partial f'_2} \right) \right) \delta_2 dt + o(\epsilon). \end{aligned}$$

The necessary condition for (f_1, f_2) to be a local extreme is the Euler-Lagrange equation

$$\frac{\partial L}{\partial f_1} - \frac{d}{dt} \left(\frac{\partial L}{\partial f'_1} \right) = 0, \quad \frac{\partial L}{\partial f_2} - \frac{d}{dt} \left(\frac{\partial L}{\partial f'_2} \right) = 0.$$

Or

$$\frac{\partial L}{\partial(f_1, f_2)} - \frac{d}{dt} \left(\frac{\partial L}{\partial(f'_1, f'_2)} \right) = \vec{0}.$$

- Least action path / 例题计算

Change $(f_1(t), f_2(t))$ to a path $\vec{x}(t) = (x_1(t), \dots, x_n(t))$ in \mathbb{R}^n , connecting $\vec{x}(a) = \vec{A}$ and $\vec{x}(b) = \vec{B}$. Minimising the action

$$\mathcal{L}(\vec{x}(t)) = \int_a^b L(t, \vec{x}(t), \vec{x}'(t)) dt.$$

Euler-Lagrange $\frac{\partial L}{\partial x_i} - \frac{d}{dt} \left(\frac{\partial L}{\partial x'_i} \right) = 0$.

Example. For $L = t(x_1 x'_2 - x_2 x'_1)$, x1, x2都要对t求导

$$\frac{\partial L}{\partial x_1} = t x'_2, \quad \frac{d}{dt} \left(\frac{\partial L}{\partial x'_1} \right) = \frac{d}{dt}(-t x_2) = -x_2 - t x'_2,$$

$$\frac{\partial L}{\partial x_2} = -t x'_1, \quad \frac{d}{dt} \left(\frac{\partial L}{\partial x'_2} \right) = \frac{d}{dt}(t x_1) = x_1 + t x'_1.$$

Euler-Lagrange: $2t x'_2 + x_2 = 0$ and $-2t x'_1 - x_1 = 0$. Solution $(x_1, x_2) = \frac{1}{\sqrt{t}}(c_1, c_2)$, straight line passing through the origin.

- 广义坐标下的变分

Use example $x = r \cos \theta$ and $y = r \sin \theta$.

An expression of L in x, y can then be rewritten in terms of r, θ

$$\begin{aligned} L(t, x, y, x', y') &= L(t, r \cos \theta, r \sin \theta, r' \cos \theta - r\theta' \sin \theta, r' \sin \theta + r\theta' \cos \theta) \\ &= \tilde{L}(t, r, \theta, r', \theta'). \end{aligned}$$

Then

$$\int_a^b L(t, x(t), y(t), x'(t), y'(t)) dt = \int_a^b \tilde{L}(t, r(t), \theta(t), r'(t), \theta'(t)) dt.$$

Minimising left \iff minimising the right. Therefore solution of

$$\frac{\partial L}{\partial x} - \frac{d}{dt} \left(\frac{\partial L}{\partial x'} \right) = 0, \quad \frac{\partial L}{\partial y} - \frac{d}{dt} \left(\frac{\partial L}{\partial y'} \right) = 0,$$

is related to the solution of

$$\frac{\partial \tilde{L}}{\partial r} - \frac{d}{dt} \left(\frac{\partial \tilde{L}}{\partial r'} \right) = 0, \quad \frac{\partial \tilde{L}}{\partial \theta} - \frac{d}{dt} \left(\frac{\partial \tilde{L}}{\partial \theta'} \right) = 0,$$

by $x(t) = r(t) \cos \theta(t)$ and $y(t) = r(t) \sin \theta(t)$.

In general, change of general coordinates $\vec{y} = \phi(\vec{x})$. Then

$\vec{y}' = \phi'(\vec{x})\vec{x}'$, and

$$L(t, \vec{y}, \vec{y}') = L(t, \phi(\vec{x}), \phi'(\vec{x})\vec{x}') = \tilde{L}(t, \vec{x}, \vec{x}').$$

Then solutions of

$$\frac{\partial L}{\partial \vec{y}} - \frac{d}{dt} \left(\frac{\partial L}{\partial \vec{y}'} \right) = 0, \quad \frac{\partial \tilde{L}}{\partial \vec{x}} - \frac{d}{dt} \left(\frac{\partial \tilde{L}}{\partial \vec{x}'} \right) = 0,$$

are related by $\vec{y}(t) = \phi(\vec{x}(t))$.

This means the calculus of variations can be done over manifolds.

Mechanics / 不同观点下的物理公式

Gravity

First Method: Newton's second law of motion

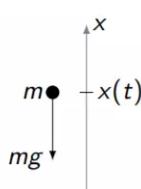
$$mx''(t) = F = -mg.$$

This is $x''(t) = -g$, and we get

$$x(t) = a + vt - gt^2.$$

$a = x(0)$ is the initial height of the mass.

$v = x'(0)$ is the initial velocity of the mass.



Second Method: Conservation of energy

Kinetic energy

$$K = \frac{1}{2}mv^2.$$

Potential energy

$$V = mgx.$$

Conservation of energy: The total energy is constant

$$T = K + V = \frac{1}{2}mv^2 + mgx = \frac{1}{2}mx'^2 + mgx$$

This is the same as the equation by the first method

$$T' = (mx'' + mg)x' = 0.$$

Third Method: Euler–Lagrange

The dynamics is the path $x(t)$ that minimises the action

$$\mathcal{L}(x(t)) = \int_0^b L(x, x') dt, \quad L = K - V = \frac{1}{2}mx'^2 - mgx.$$

We have

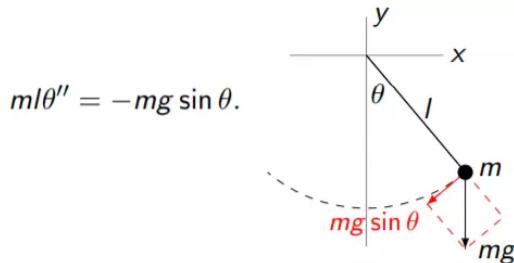
$$\frac{\partial L}{\partial x} = -mg, \quad \frac{\partial L}{\partial x'} = mx', \quad \frac{d}{dt} \left(\frac{\partial L}{\partial x'} \right) = mx''.$$

Euler-Lagrange is the same as the equation by the first method

$$-mg - mx'' = 0.$$

Pendulum / 简单运动 – 单摆

Newton's second law of motion



*需要找到运动方向；难以找到精确解...

Conservation of energy

Kinetic energy: The speed is $v = l\theta'$

$$K = \frac{1}{2}mv^2 = \frac{1}{2}ml^2\theta'^2.$$

Potential energy: Height $y = -l \cos \theta$

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

Conservation of energy: The total energy is constant

$$T = K + V = \frac{1}{2}ml^2\theta'^2 - mgl \cos \theta$$

This is the same as the equation by the first method

$$T' = (ml\theta'' + mg \sin \theta)l\theta' = 0.$$

Euler–Lagrange

The dynamics is the path $\theta(t)$ that minimises the **action**

$$\mathcal{L}(x(t)) = \int_0^b L(x, x') dt, \quad L = K - V = \frac{1}{2}ml^2\theta'^2 + mgl \cos \theta.$$

We have

$$\frac{\partial L}{\partial \theta} = -mgl \sin \theta, \quad \frac{\partial L}{\partial \theta'} = ml^2\theta', \quad \frac{d}{dt} \left(\frac{\partial L}{\partial \theta'} \right) = ml^2\theta''.$$

Euler–Lagrange is the same as the equation by the first method

$$-mgl \sin \theta - ml^2\theta'' = 0.$$

Euler–Lagrange Equation

D'Alembert principle

→ 达朗贝尔原理：用惯性力(Inertial force)将动力问题转换为静力问题

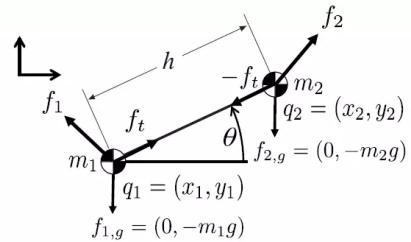
Consider again the example on the right. This time we assume the two mass points are in motion, with $\dot{q}_1, \ddot{q}_1, \dot{q}_2, \ddot{q}_2$.

$$\begin{aligned} f_1 + f_t + f_{1,g} - m_1\ddot{q}_1 &= 0 \\ f_2 - f_t + f_{2,g} - m_2\ddot{q}_2 &= 0 \end{aligned}$$

D'Alembert says, we can treat $-m_1\ddot{q}_1, -m_2\ddot{q}_2$ as forces (which we shall call **inertial forces**), and then we can apply PVW as before:

$$\begin{aligned} \delta W &= (f_1 + f_{1,g} - m_1\ddot{q}_1)^T \delta q_1 + \dots \\ &\quad (f_2 + f_{2,g} - m_2\ddot{q}_2)^T \delta q_2 = 0 \end{aligned}$$

This is known as **d'Alembert's principle**.



Jean d'Alembert
(1717–1783)

From d'Alembert principle to Hamilton's principle

$$\Rightarrow \delta W = 0 \rightarrow \int \delta W = 0$$

→ Details:

- ◆ 动能 $T = m_i \dot{q}_i^T \dot{q}_i / 2$
- ◆ 势能 $V = m_1 g y_1 + m_2 g y_2$
- ◆ 计算过程 (分部积分) :

We consider a general case with n particles:

$$\begin{aligned} \int_{t_0}^{t_f} \delta W dt &= \int_{t_0}^{t_f} \sum_{i=1}^n \left(-\frac{\partial V}{\partial q_i} - m_i \ddot{q}_i \right)^T \delta q_i dt = \int_{t_0}^{t_f} (-\delta V) dt - \int_{t_0}^{t_f} \sum_{i=1}^n m_i \delta q_i^T d\dot{q}_i \\ &= \int_{t_0}^{t_f} (-\delta V) dt - \sum_{i=1}^n m_i \delta q_i^T \dot{q}_i \Big|_{t_0}^{t_f} + \int_{t_0}^{t_f} \sum_{i=1}^n m_i \dot{q}_i^T \frac{d(\delta q_i)}{dt} dt \\ &= \int_{t_0}^{t_f} (-\delta V) dt - \sum_{i=1}^n m_i \delta q_i^T \dot{q}_i \Big|_{t_0}^{t_f} + \int_{t_0}^{t_f} \sum_{i=1}^n m_i \dot{q}_i^T \delta \dot{q}_i dt \\ &= \int_{t_0}^{t_f} (-\delta V) dt - \sum_{i=1}^n m_i \delta q_i^T \dot{q}_i \Big|_{t_0}^{t_f} + \int_{t_0}^{t_f} \sum_{i=1}^n \delta \left(\frac{1}{2} m_i \dot{q}_i^T \dot{q}_i \right) dt \\ &= \int_{t_0}^{t_f} \delta \underbrace{(T - V)}_L dt - \sum_{i=1}^n m_i \dot{q}_i^T \delta q_i \Big|_{t_0}^{t_f} = 0 \end{aligned}$$

- ◆ L is called the **Lagrangian function**

→ 由拉格朗日函数还原系统方程

- ◆ 约束任取的虚位移, 固定轨迹端点位置: $\delta q_i(t_0) = \delta q_i(t_f) = 0, i = 1, \dots, n$
- ◆ Action integral: $A = \int_{t_0}^{t_f} L dt$
- ◆ $\delta A = \int_{t_0}^{t_f} \delta L dt = \delta \int_{t_0}^{t_f} L dt = 0$

→ **Hamilton's principle**

if the action integral A is **stationary** for a particular trajectory, i.e. $\delta A = 0$, then it is the actual trajectory of the system!

From Hamilton's principle to Euler–Lagrange equation

Starting with a given Lagrangian $L(q_1, \dots, q_n, \dot{q}_1, \dots, \dot{q}_n)$ w.r.t. to generalized coordinates $(q_1, \dots, q_n) \in \mathbb{R}^n$, Hamilton's principle leads to:

$$\begin{aligned} \delta A &= \int_{t_0}^{t_f} \delta L(q_1, \dots, q_n, \dot{q}_1, \dots, \dot{q}_n) dt = \int_{t_0}^{t_f} \sum_{i=1}^n \left(\frac{\partial L}{\partial q_i} \delta q_i + \frac{\partial L}{\partial \dot{q}_i} \delta \dot{q}_i \right) dt \\ &= \int_{t_0}^{t_f} \sum_{i=1}^n \frac{\partial L}{\partial q_i} \delta q_i dt + \int_{t_0}^{t_f} \sum_{i=1}^n \frac{\partial L}{\partial \dot{q}_i} d\delta q_i(t) = \int_{t_0}^{t_f} \sum_{i=1}^n \left(\frac{\partial L}{\partial q_i} \delta q_i - \frac{d}{dt} \left(\frac{\partial L}{\partial \dot{q}_i} \right) \delta q_i \right) dt + \frac{\partial L}{\partial \dot{q}_i} \delta q_i \Big|_{t_0}^{t_f} \\ &= \int_{t_0}^{t_f} \sum_{i=1}^n \left(\frac{\partial L}{\partial q_i} - \frac{d}{dt} \left(\frac{\partial L}{\partial \dot{q}_i} \right) \right) \delta q_i dt = 0 \end{aligned}$$

Since δq_i 's are arbitrarily chosen other than the fact that $\delta q_i(t_0) = \delta q_i(t_f) = 0$, we conclude that:

$$\frac{\partial L}{\partial q_i} - \frac{d}{dt} \left(\frac{\partial L}{\partial \dot{q}_i} \right) = 0, \quad i = 1, \dots, n$$

This is called **Euler–Lagrange equations**, a set of second order ordinary differential equations.

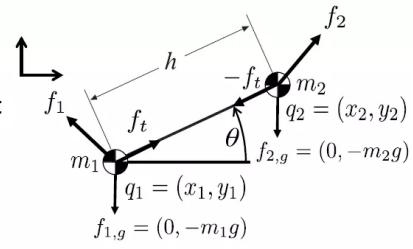
→ 欧拉–拉格朗日方程的优势: 消除约束力, 选取最少坐标描述系统方程

In Hamilton's principle and Euler-Lagrange equation, q_1, \dots, q_n ; $\dot{q}_1, \dots, \dot{q}_n$ can be chosen to be any generalized coordinates, not just Cartesian coordinates!

Example: in the two-particle system, q_1, q_2 are not independent:

$$q_2 = q_1 + h \begin{bmatrix} \cos \theta \\ \sin \theta \end{bmatrix}$$

and its Lagrangian becomes a function of $q_1, \dot{q}_1, \theta, \dot{\theta}$; we have:



$$\int_{t_0}^{t_f} \left(\frac{\partial L}{\partial q_1} - \frac{d}{dt} \left(\frac{\partial L}{\partial \dot{q}_1} \right) \right)^T \delta q_1 + \left(\frac{\partial L}{\partial \theta} - \frac{d}{dt} \left(\frac{\partial L}{\partial \dot{\theta}} \right) \right) \delta \theta dt = 0$$

Since δq_1 (it is a vector and has the component $\delta x_1, \delta y_1$) and $\delta \theta$ are linearly independent:

$$\frac{d}{dt} \left(\frac{\partial L}{\partial \dot{q}_1} \right) - \frac{\partial L}{\partial q_1} = 0, \quad \frac{d}{dt} \left(\frac{\partial L}{\partial \dot{\theta}} \right) - \frac{\partial L}{\partial \theta} = 0$$

For systems with **non-conservative generalized forces** Q_i 's (we will see how to derive Q_i 's in a moment) associated with generalized coordinates q_i 's:

$$\frac{\partial L}{\partial q_i} - \frac{d}{dt} \left(\frac{\partial L}{\partial \dot{q}_i} \right) + Q_i = 0, \quad i = 1, \dots, n \quad \text{or} \quad \frac{d}{dt} \left(\frac{\partial L}{\partial \dot{q}_i} \right) - \frac{\partial L}{\partial q_i} = Q_i, \quad i = 1, \dots, n$$

→ 拉格朗日乘子 (Lagrange multipliers) : 消除多余的虚位移

For systems with holonomic constraints $h_1(q_1, \dots, q_n) = 0, \dots, h_m(q_1, \dots, q_n) = 0$, not all virtual displacements δq_i 's are linear independent; we use **Lagrange multipliers** $\lambda_1, \dots, \lambda_m$ (**they are new unknown variables now!**) to eliminate redundant virtual displacements.

$$\int_{t_0}^{t_f} \sum_{i=1}^n \left(\frac{\partial L}{\partial q_i} - \frac{d}{dt} \left(\frac{\partial L}{\partial \dot{q}_i} \right) + \lambda_1 \frac{\partial h_1}{\partial q_i} + \dots + \lambda_m \frac{\partial h_m}{\partial q_i} \right) \delta q_i dt = 0$$

which is the same as (because $\delta(\lambda_j h_j) = \lambda_j \delta h_j + \delta \lambda_j h_j = \lambda_j \delta h_j$ since $h_j = 0$):

$$\int_{t_0}^{t_f} \delta L + \sum_{j=1}^m \lambda_j \delta h_j dt = \delta \underbrace{\int_{t_0}^{t_f} (L + \sum_{j=1}^m \lambda_j h_j) dt}_{\bar{L}} = 0, \quad h_1 = h_2 = \dots = h_m = 0$$

DM 283 Mechanics for design

We define **augmented Lagrangian** $\bar{L}(q_1, \dots, q_n, \dot{q}_1, \dots, \dot{q}_n, \lambda_1, \dots, \lambda_m) \triangleq L + \lambda_1 h_1 + \dots + \lambda_m h_m$, then

Hamilton's principle still holds for the **augmented Lagrangian** \bar{L} in place of L :

$$\int_{t_0}^{t_f} (\delta \bar{L} + \sum_{j=1}^m \lambda_j \delta h_j) dt = \delta \underbrace{\int_{t_0}^{t_f} (L + \sum_{j=1}^m \lambda_j h_j) dt}_{\bar{L}} = 0, \quad h_1 = h_2 = \dots = h_m = 0$$

Now the Euler-Lagrange equation becomes (along with $h_1 = \dots = h_m = 0$):

$$\frac{\partial L}{\partial q_i} - \frac{d}{dt} \left(\frac{\partial L}{\partial \dot{q}_i} \right) + Q_i + \lambda_1 \frac{\partial h_1}{\partial q_i} + \dots + \lambda_m \frac{\partial h_m}{\partial q_i} = 0, \quad i = 1, \dots, n$$

or

$$\underbrace{\frac{d}{dt} \left(\frac{\partial L}{\partial \dot{q}_i} \right) - \frac{\partial L}{\partial q_i}}_{\substack{\text{All inertial force} \\ \text{and conservative} \\ \text{forces go here}}} = \underbrace{Q_i + \lambda_1 \frac{\partial h_1}{\partial q_i} + \dots + \lambda_m \frac{\partial h_m}{\partial q_i}}_{\substack{\text{Whatever} \\ \text{forces not} \\ \text{covered by } L}} \quad i = 1, \dots, n$$

Constraint forces

Example – Approach 1

- 用广义坐标(Generalized coordinates)表示 L

We now recover Newton's equation for the example with two particles. We have:

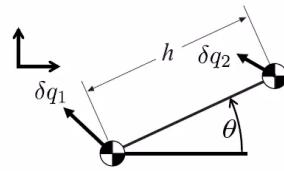
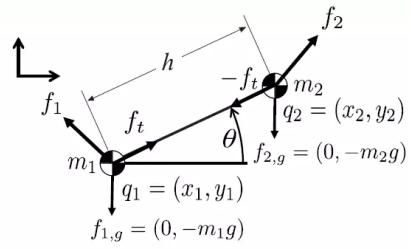
$$L = T - V = \frac{1}{2}m_1\dot{q}_1^T\dot{q}_1 + \frac{1}{2}m_2\dot{q}_2^T\dot{q}_2 - m_1gy_1 - m_2gy_2$$

We first express L as function of the generalized coordinates x_1, y_1, θ :

$$\dot{q}_2 = \begin{bmatrix} \dot{x}_1 - h \sin \theta \dot{\theta} \\ \dot{y}_1 + h \cos \theta \dot{\theta} \end{bmatrix}, \quad y_2 = y_1 + h \sin \theta$$



$$L = \frac{1}{2}m_1(\dot{x}_1^2 + \dot{y}_1^2) + \frac{1}{2}m_2((\dot{x}_1 - h \sin \theta \dot{\theta})^2 + (\dot{y}_1 + h \cos \theta \dot{\theta})^2) - m_1gy_1 - m_2g(y_1 + h \sin \theta)$$



2. 求广义力(Generalized force)

To derive generalized force from f_1 and f_2 , we go back one step in Hamilton's principle:

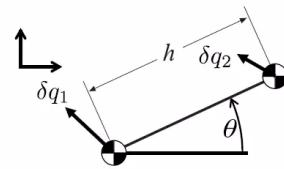
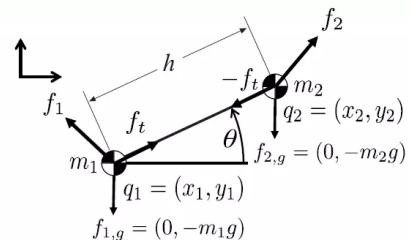
$$\int_{t_0}^{t_f} (\delta L + f_1^T \delta \dot{q}_1 + f_2^T \delta \dot{q}_2) dt = 0$$

We recall that:

$$\delta q_2 = \delta q_1 + h \begin{bmatrix} -\sin \theta \\ \cos \theta \end{bmatrix} \delta \theta$$

Then we

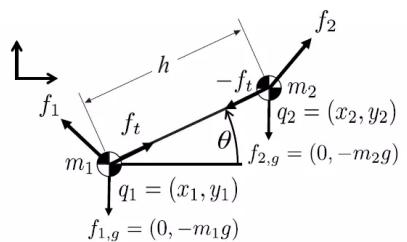
$$\int_{t_0}^{t_f} \left(\frac{\partial L}{\partial q_1} - \frac{d}{dt} \left(\frac{\partial L}{\partial \dot{q}_1} \right) + f_1 + f_2 \right)^T \delta q_1 + \left(\frac{\partial L}{\partial \theta} - \frac{d}{dt} \left(\frac{\partial L}{\partial \dot{\theta}} \right) + f_2^T h \begin{bmatrix} -\sin \theta \\ \cos \theta \end{bmatrix} \right) \delta \theta dt = 0$$



3. 得到方程

The corresponding Euler-Lagrange equations are (they are three equations):

$$\begin{aligned} \frac{d}{dt} \left(\frac{\partial L}{\partial \dot{q}_1} \right) - \frac{\partial L}{\partial q_1} &= f_1 + f_2 \\ \frac{d}{dt} \left(\frac{\partial L}{\partial \dot{\theta}} \right) - \frac{\partial L}{\partial \theta} &= f_2^T h \begin{bmatrix} -\sin \theta \\ \cos \theta \end{bmatrix} \end{aligned}$$



which leads to

$$\begin{aligned} m_1 \ddot{x}_1 + m_2 (\ddot{x}_1 - h \cos \theta \ddot{\theta}^2 - h \sin \theta \ddot{\theta}) &= f_{1,x} + f_{2,x} \\ m_1 \ddot{y}_1 + m_2 (\ddot{y}_1 - h \sin \theta \ddot{\theta}^2 + h \cos \theta \ddot{\theta}) &= f_{1,y} + f_{2,y} - (m_1 + m_2)g \\ m_2 (-\ddot{x}_1 h \sin \theta + \ddot{y}_1 h \cos \theta + h^2 \ddot{\theta}) &= -f_{2,x} h \sin \theta + f_{2,y} h \cos \theta - m_2 g h \cos \theta \end{aligned}$$

We can see that:

$$\begin{aligned} m_1 \ddot{x}_1 + m_2 \underbrace{(\ddot{x}_1 - h \cos \theta \ddot{\theta}^2 - h \sin \theta \ddot{\theta})}_{\ddot{x}_2} &= f_{1,x} + f_{2,x} \\ m_1 \ddot{y}_1 + m_2 \underbrace{(\ddot{y}_1 - h \sin \theta \ddot{\theta}^2 + h \cos \theta \ddot{\theta})}_{\ddot{y}_2} &= f_{1,y} + f_{2,y} - (m_1 + m_2)g \\ \underbrace{m_2 (-\ddot{x}_1 h \sin \theta + \ddot{y}_1 h \cos \theta + h^2 \ddot{\theta})}_{-m_2 \ddot{x}_2 h \sin \theta + m_2 \ddot{y}_2 h \cos \theta} &= -f_{2,x} h \sin \theta + f_{2,y} h \cos \theta - m_2 g h \cos \theta \end{aligned}$$

$\curvearrowleft f_2$

Example – Approach 2

使用拉格朗日乘子法可以恢复约束力！

1.

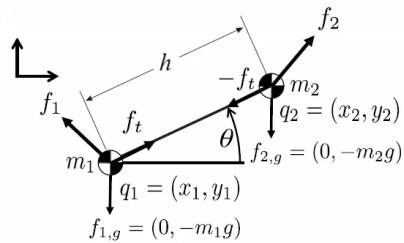
Just use Cartesian coordinates q_1, q_2 , along with the constraint:

$$(q_1 - q_2)^T (q_1 - q_2) - h^2 = 0$$

Now we have to use augmented Lagrangian: 两点间距离

$$\bar{L} = \frac{1}{2}m_1\dot{q}_1^T\dot{q}_1 + \frac{1}{2}m_2\dot{q}_2^T\dot{q}_2 - m_1gy_1 - m_2gy_2 + \dots$$

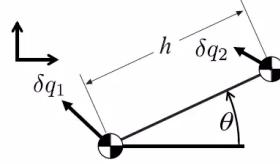
$$\lambda((q_1 - q_2)^T(q_1 - q_2) - h^2)$$



We will get four Euler-Lagrange equations:

$$\frac{d}{dt}\left(\frac{\partial \bar{L}}{\partial \dot{q}_1}\right) - \frac{\partial \bar{L}}{\partial q_1} = f_1$$

$$\frac{d}{dt}\left(\frac{\partial \bar{L}}{\partial \dot{q}_2}\right) - \frac{\partial \bar{L}}{\partial q_2} = f_2$$

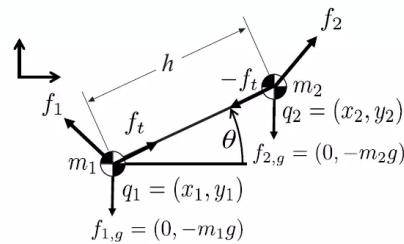


2.

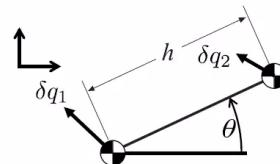
The Euler-Lagrange equations become:

$$m_1\ddot{q}_1 = f_1 + \boxed{\lambda(q_1 - q_2)} - \begin{bmatrix} 0 \\ m_1g \end{bmatrix}$$

$$m_2\ddot{q}_2 = f_2 - \lambda(q_1 - q_2) - \begin{bmatrix} 0 \\ m_2g \end{bmatrix}$$



You can easily see now that $f_t = \lambda(q_1 - q_2)$!



Question: But how do we solve λ ?

$$(q_1 - q_2)^T (q_1 - q_2) - h^2 = 0 \Rightarrow$$

$$(q_1 - q_2)^T (\ddot{q}_1 - \ddot{q}_2) = -(\dot{q}_1 - \dot{q}_2)^T (\dot{q}_1 - \dot{q}_2)$$



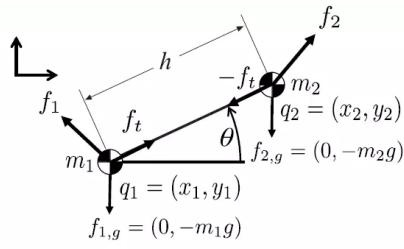
3.

From the E-L equations:

$$\ddot{q}_1 - \ddot{q}_2 = \frac{f_1}{m_1} - \frac{f_2}{m_2} + \lambda\left(\frac{1}{m_1} + \frac{1}{m_2}\right)(q_1 - q_2)$$

Multiply on both sides $(q_1 - q_2)^T$:

$$-(\dot{q}_1 - \dot{q}_2)^T (\ddot{q}_1 - \ddot{q}_2) = (q_1 - q_2)^T \frac{m_2 f_1 - m_1 f_2}{m_1 m_2} + \lambda \frac{h^2(m_1 + m_2)}{m_1 m_2}$$



$$\downarrow \quad \dot{q}_2 - \dot{q}_1 = h \begin{bmatrix} -\sin \theta \\ \cos \theta \end{bmatrix} \dot{\theta}$$

$$\lambda = -\frac{m_1 m_2}{m_1 + m_2} \dot{\theta}^2 + \frac{(m_2 f_{1,x} - m_1 f_{2,x}) \cos \theta + (m_2 f_{1,y} - m_1 f_{2,y}) \sin \theta}{h(m_1 + m_2)}$$

4.

leading to:

$$m_1\ddot{q}_1 = f_1 + \left(-\frac{m_1 m_2}{m_1 + m_2} \dot{\theta}^2 + \frac{(m_2 f_{1,x} - m_1 f_{2,x}) \cos \theta + (m_2 f_{1,y} - m_1 f_{2,y}) \sin \theta}{h(m_1 + m_2)}\right)(q_1 - q_2) + \begin{bmatrix} 0 \\ -m_1 g \end{bmatrix}$$

$$m_2\ddot{q}_2 = f_2 - \left(-\frac{m_1 m_2}{m_1 + m_2} \dot{\theta}^2 + \frac{(m_2 f_{1,x} - m_1 f_{2,x}) \cos \theta + (m_2 f_{1,y} - m_1 f_{2,y}) \sin \theta}{h(m_1 + m_2)}\right)(q_1 - q_2) + \begin{bmatrix} 0 \\ -m_2 g \end{bmatrix}$$

But is it equivalent to our previous derivations? ✓

$$m_1\ddot{x}_1 + m_2(\ddot{x}_1 - h \cos \theta \dot{\theta}^2 - h \sin \theta \ddot{\theta}) = f_{1,x} + f_{2,x}$$

$$m_1\ddot{y}_1 + m_2(\ddot{y}_1 - h \sin \theta \dot{\theta}^2 + h \cos \theta \ddot{\theta}) = f_{1,y} + f_{2,y} - (m_1 + m_2)g$$

$$m_2(-\ddot{x}_1 h \sin \theta + \ddot{y}_1 h \cos \theta + h^2 \ddot{\theta}) = -f_{2,x} h \sin \theta + f_{2,y} h \cos \theta - m_2 g h \cos \theta$$

You are about to find out in the written assignment... ;-)

Newton–Euler Equation

Spatial velocity v.s. Body velocity

关于b-frame: 跟随刚体, 瞬时状态下关于原点坐标系静止 (打点计时器)?
Recall the definition of **spatial velocity** of a rigid body:

$$q_a(t) \mapsto \dot{q}_a(t) = \hat{V}_{ab}(t)q_a(t) = (\dot{g}_{ab}(t)g_{ab}^{-1}(t))q_a = \hat{\omega}_{ab}(t)q_a + v_{ab}(t)$$

Sometimes, it's easier to work with the b-frame (body frame). We define:

$$\dot{q}_b(t) \triangleq g_{ab}^{-1}\dot{q}_a(t) = \underbrace{g_{ab}^{-1}\dot{g}_{ab}}_{\hat{V}_{ab}^b} q_b$$

That is, \dot{q}_b is the velocity of point q w.r.t. to a-frame (ref. frame) but observed in b-frame. Here, the **body velocity** $\hat{V}_{ab}^b = g_{ab}^{-1}\dot{g}_{ab}$ is defined similarly to the **spatial velocity** $\hat{V}_{ab} = \dot{g}_{ab}g_{ab}^{-1}$:

$$\hat{V}_{ab}^b = g_{ab}^{-1}\dot{g}_{ab} = \begin{bmatrix} \hat{\omega}_{ab}^b & v_{ab}^b \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} R_{ab}^T \dot{R}_{ab} & R_{ab}^T \dot{p}_{ab} \\ 0 & 0 \end{bmatrix}$$

$$\hat{V}_{ab}^s \text{(patial)}$$

$$\hat{V}_{ab}^b \text{(ody)}$$

Note: In MLS94, \hat{V}_{ab} is written as \hat{V}_{ab}^s to further distinguish from \hat{V}_{ab}^b . We didn't write it that way until now to avoid confusion. From now on, we shall write \hat{V}_{ab}^s in place of \hat{V}_{ab} .

两者之间的关系与转换 (伴随变换)

$$\text{body velocity } \hat{V}_{ab}^b = g_{ab}^{-1}\dot{g}_{ab}$$

$$\text{spatial velocity } \hat{V}_{ab} = \dot{g}_{ab}g_{ab}^{-1}$$

$$\hat{V}_{ab}^s = \dot{g}_{ab}g_{ab}^{-1} = g_{ab}g_{ab}^{-1}\dot{g}_{ab}g_{ab}^{-1} = g_{ab}\hat{V}_{ab}^b g_{ab}^{-1}$$



$$\begin{aligned} \hat{V}_{ab}^s &= \begin{bmatrix} \hat{\omega}_{ab}^s & v_{ab}^s \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} R_{ab} & p_{ab} \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \hat{\omega}_{ab}^b & v_{ab}^b \\ 0 & 0 \end{bmatrix} \begin{bmatrix} R_{ab}^T & -R_{ab}^T p_{ab} \\ 0 & 1 \end{bmatrix} \\ &= \begin{bmatrix} R_{ab}\hat{\omega}_{ab}^b R_{ab}^T & -R_{ab}\hat{\omega}_{ab}^b R_{ab}^T p_{ab} + R_{ab}v_{ab}^b \\ 0 & 0 \end{bmatrix} \end{aligned}$$



$$V_{ab}^s = \begin{bmatrix} v_{ab}^s \\ \omega_{ab}^s \end{bmatrix} = \begin{bmatrix} R_{ab} & \hat{p}_{ab}R_{ab} \\ 0 & R_{ab} \end{bmatrix} \begin{bmatrix} v_{ab}^b \\ \omega_{ab}^b \end{bmatrix} = \underbrace{Ad_{g_{ab}}}_{\text{Adjoint transformation}} V_{ab}^b$$

熟悉的二维版本:

Planar case:

$$\begin{aligned}
 \hat{V}_{ab}^s &= \begin{bmatrix} J\dot{\theta} & \dot{p}_{ab} - \dot{\theta}Jp_{ab} \\ 0 & 0 \end{bmatrix} & V_{ab}^s &= \begin{bmatrix} v_{ab}^s \\ \dot{\theta} \end{bmatrix} \\
 \hat{V}_{ab}^b &= \begin{bmatrix} J\dot{\theta} & R_{ab}^T \dot{p}_{ab} \\ 0 & 0 \end{bmatrix} & &= \begin{bmatrix} R_{ab} & -Jp_{ab} \\ 0 & 1 \end{bmatrix} \begin{bmatrix} v_{ab}^b \\ \dot{\theta} \end{bmatrix} \\
 v_{ab}^s &= \dot{p}_{ab} - \dot{\theta}Jp_{ab}, \omega_{ab}^s = \dot{\theta} & &= Ad_{g_{ab}} V_{ab}^b \\
 v_{ab}^b &= R_{ab}^T \dot{p}_{ab}, \omega_{ab}^b = \dot{\theta} & & \\
 J &= R\left(\frac{\pi}{2}\right) & &
 \end{aligned}$$

Newton–Euler equation

Generalized inertia matrix and momentum

*求导: 动能 \rightarrow 动量 \rightarrow 力

The **kinetic energy** of a rigid in motion:

$$\begin{aligned}
 T &= \frac{1}{2} \int_V \rho \dot{q}_a^T \dot{q}_a dV \\
 &= \frac{1}{2} \int_V \rho (\hat{\omega}_{ab}^s q_a + v_{ab}^s)^T (\hat{\omega}_{ab}^s q_a + v_{ab}^s) dV \\
 &= \frac{1}{2} [v_{ab}^{s T} \quad \omega_{ab}^{s T}] \underbrace{\begin{bmatrix} mI_{3 \times 3} & -\int_V \rho \hat{q}_a dV \\ \int_V \rho \hat{q}_a dV & -\int_V \rho \hat{q}_a^2 dV \end{bmatrix}}_{M^s} \begin{bmatrix} v_{ab}^s \\ \omega_{ab}^s \end{bmatrix} \\
 &= \frac{1}{2} V_{ab}^{s T} [M^s V_{ab}^s], \quad m = \int_V \rho dV
 \end{aligned}$$

广义惯量矩阵与广义动量

M^s is called the **generalized inertia matrix** (superscript s means computed in spatial frame a).

We define **generalized momentum** by $M^s V_{ab}^s$.

$$M^s = \begin{bmatrix} mI_{3 \times 3} & -\int_V \rho \hat{q}_a dV \\ \int_V \rho \hat{q}_a dV & -\int_V \rho \hat{q}_a^2 dV \end{bmatrix}$$

Newton–Euler equation

牛顿–欧拉法可以直接得到约束力!

The **Newton-Euler equation** is given by:

$$\frac{d}{dt} (M^s V_{ab}^s) = M^s \dot{V}_{ab}^s + \dot{M}^s V_{ab}^s = F^s = \begin{bmatrix} f^s \\ \tau^s \end{bmatrix}$$

(F^s denotes wrench in a -frame)

which leads to:

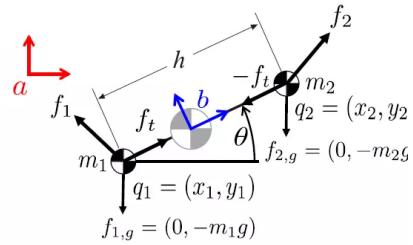
$$M^s \dot{V}_{ab}^s - \begin{bmatrix} \hat{\omega}_{ab}^s & \hat{v}_{ab}^s \\ 0 & \hat{\omega}_{ab}^s \end{bmatrix}^T M^s V_{ab}^s = F^s \quad (\text{3-dimension})$$

$$M^s \dot{V}_{ab}^s - \begin{bmatrix} J\dot{\theta} & -Jv_{ab}^s \\ 0 & 0 \end{bmatrix}^T M^s V_{ab}^s = F^s \quad (\text{2-dimension})$$

Example 1

Recall the planar spatial velocity ($J = R(\pi/2)$):

$$\hat{V}_{ab}^s = \begin{bmatrix} J\dot{\theta} & v_{ab}^s \\ 0 & 0 \end{bmatrix} \quad \rightarrow \quad \dot{q}_{1a} = J\dot{\theta}q_{1a} + v_{ab}^s \\ \dot{q}_{2a} = J\dot{\theta}q_{2a} + v_{ab}^s$$



which leads to:

$$T = \frac{1}{2}m_1(J\dot{\theta}q_{1a} + v_{ab}^s)^T(J\dot{\theta}q_{1a} + v_{ab}^s) + \frac{1}{2}m_2(J\dot{\theta}q_{2a} + v_{ab}^s)^T(J\dot{\theta}q_{2a} + v_{ab}^s) \\ = \frac{1}{2}m_1 q_{1a}^T q_{1a} \dot{\theta}^2 + \frac{1}{2}m_2 q_{2a}^T q_{2a} \dot{\theta}^2 + \frac{1}{2}(m_1 + m_2)v_{ab}^{s T} v_{ab}^s + m_1 v_{ab}^{s T} J q_{1a} \dot{\theta} + m_2 v_{ab}^{s T} J q_{2a} \dot{\theta} \\ = \frac{1}{2} \underbrace{\begin{bmatrix} v_{ab}^s & \dot{\theta} \end{bmatrix}}_{V_{ab}^s T} \underbrace{\left[\begin{array}{cc} (m_1 + m_2)I & m_1 J q_{1a} + m_2 J q_{2a} \\ (m_1 J q_{1a} + m_2 J q_{2a})^T & m_1 q_{1a}^T q_{1a} + m_2 q_{2a}^T q_{2a} \end{array} \right]}_{\text{这是对称阵! } \rightarrow M^s} \underbrace{\begin{bmatrix} v_{ab}^s \\ \dot{\theta} \end{bmatrix}}_{V_{ab}^s}$$

Recall the center of mass c is given by:

$$c_a = \frac{m_1 q_{1a} + m_2 q_{2a}}{m_1 + m_2}$$

Also denote $m_1 + m_2$ by m for simplicity; then:

$$T = \frac{1}{2}V_{ab}^{s T} \underbrace{\left[\begin{array}{cc} mI & mJc_a \\ (mJc_a)^T & m_1 q_{1a}^T q_{1a} + m_2 q_{2a}^T q_{2a} \end{array} \right]}_{M^s} V_{ab}^s$$

耦合项
惯量相关

The N-E equation becomes:

$$M^s \dot{V}_{ab}^s + \dot{M}^s V_{ab}^s = F_1 + F_2 + F_{1,g} + F_{2,g}$$

Where $F_1, F_2, F_{1,g}, F_{2,g}$ are the wrenches corresponding to $f_1, f_2, f_{1,g}, f_{2,g}$.

The N-E equation becomes:

$$\begin{bmatrix} mI & mJc_a \\ (mJc_a)^T & m_1 q_{1a}^T q_{1a} + m_2 q_{2a}^T q_{2a} \end{bmatrix} \begin{bmatrix} \dot{v}_{ab}^s \\ \dot{\theta} \end{bmatrix} + \dots \\ \begin{bmatrix} mI & mJ\dot{c}_a \\ (mJ\dot{c}_a)^T & 2m_1 q_{1a}^T \dot{q}_{1a} + 2m_2 q_{2a}^T \dot{q}_{2a} \end{bmatrix} \begin{bmatrix} v_{ab}^s \\ \dot{\theta} \end{bmatrix} = \dots \\ \begin{bmatrix} f_1 \\ f_1^T J q_{1a} \end{bmatrix} + \begin{bmatrix} f_2 \\ f_2^T J q_{2a} \end{bmatrix} + \begin{bmatrix} f_{1,g} \\ f_{1,g}^T J q_{1a} \end{bmatrix} + \begin{bmatrix} f_{2,g} \\ f_{2,g}^T J q_{2a} \end{bmatrix}$$

Remember we say:

$$\dot{M}^s V_{ab}^s = - \begin{bmatrix} J\dot{\theta} & -Jv_{ab}^s \\ 0 & 0 \end{bmatrix}^T M^s V_{ab}^s \quad \dot{q}_{1a} = \dot{\theta} T q_{1a} + V_{ab}^{s,v} \\ X^T J X = 0_v$$

Inertia Matrix

The generalized inertia matrix M^s is not constant; it is easier to work with b -frame in this case.

$$\begin{aligned} T &= \frac{1}{2} \int_V \rho \dot{q}_a^T \dot{q}_a dV = \frac{1}{2} \int_V \rho \dot{q}_b^T R_{ab}^T R_{ab} \dot{q}_b dV = \frac{1}{2} \int_V \rho \dot{q}_b^T \dot{q}_b dV \\ &= \frac{1}{2} \int_V \rho (\hat{\omega}_{ab}^b q_b + v_{ab}^b)^T (\hat{\omega}_{ab}^b q_b + v_{ab}^b) dV \\ &= \frac{1}{2} m v_{ab}^{b^T} v_{ab}^b + \frac{1}{2} \omega_{ab}^b {}^T \underbrace{\int_V -\rho \hat{q}_b^2 dV}_{\mathcal{I}^b} \omega_{ab}^b - v_{ab}^{b^T} \int_V \rho \hat{q}_b dV \omega_{ab}^b \end{aligned}$$

To proceed, we first define the [center of mass](#):

$$c_a = \frac{1}{m} \int_V \rho q_a dV, \quad m = \int_V \rho dV$$



If we attach b -frame to the center of mass, we have $c_b = 0$:

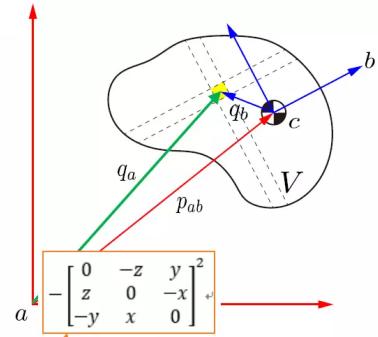
$$m \hat{c}_b = \int_V \rho \hat{q}_b dV = 0$$



$$T = \frac{1}{2} \begin{bmatrix} v_{ab}^b \\ \omega_{ab}^b \end{bmatrix} {}^T \underbrace{\begin{bmatrix} m I_{3 \times 3} & \\ & \mathcal{I}^b \end{bmatrix}}_{M^b} \begin{bmatrix} v_{ab}^b \\ \omega_{ab}^b \end{bmatrix} = \frac{1}{2} V_{ab}^{b^T} M^b V_{ab}^b$$

Since $V_{ab}^s = Ad_{g_{ab}} V_{ab}^b$, we have:

$$M^s = Ad_{g_{ab}}^{-T} M^b Ad_{g_{ab}}^{-1} \quad \mathcal{I}^b = \int_V -\rho \hat{q}_b^2 dV = \int_V \rho \begin{bmatrix} y^2 + z^2 & -xy & -xz \\ -xy & z^2 + x^2 & -yz \\ -xz & -yz & x^2 + y^2 \end{bmatrix} dx dy dz$$



This is the usual inertia matrix you see in your physics class.