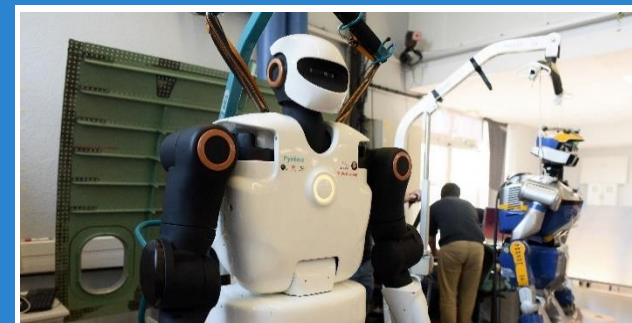


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Pinocchio

Fast forward & inverse dynamics

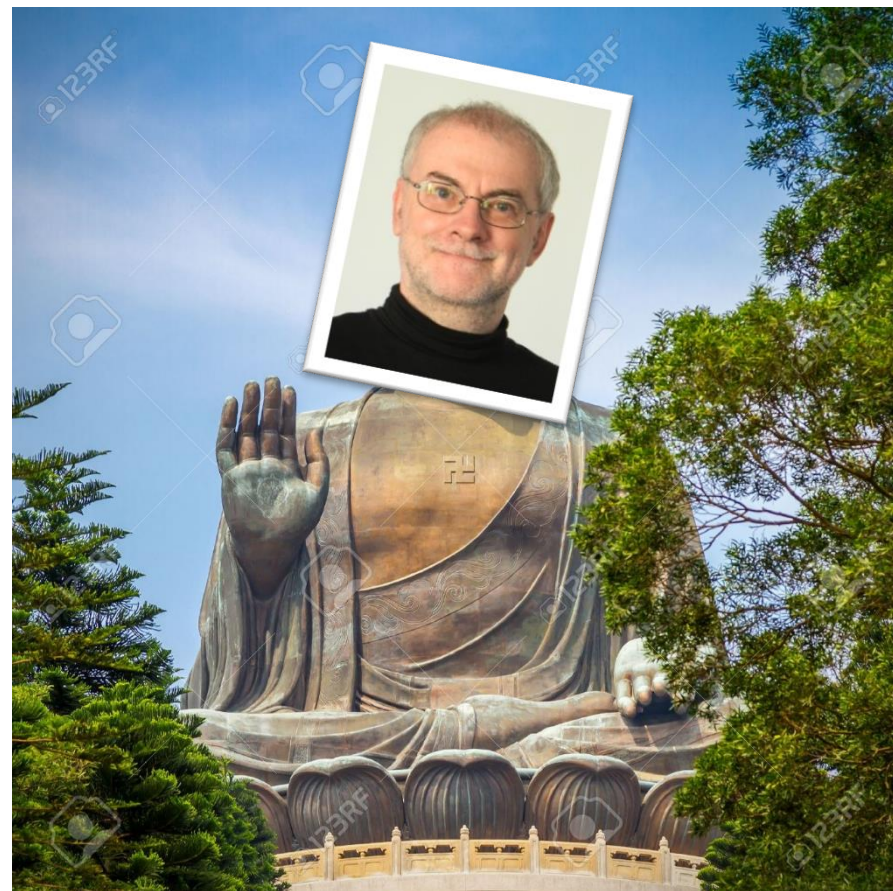


Nicolas Mansard
(CNRS)





Justin Carpentier (INRIA)



Roy Featherstone (IIT)

- ❑ Web site
 - ❑ <https://stack-of-tasks.github.io/pinocchio>
- ❑ Doxygen
 - ❑ Documentation tab on github.io
- ❑ Tutorials:
 - ❑ Practical exercises in the documentation

- ❑ Also use the ? In Python

- ❑ GitHub project
 - ❑ <https://github.com/stack-of-tasks/pinocchio>
- ❑ Post issues for contributing
- ❑ We are looking for doc-devs!
 - ❑ Feedback some material as a thank-you note
 - ❑ In the doc: “examples” is waiting for you

❑ C++ Library

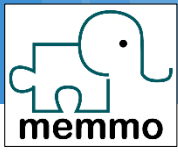
- ❑ Fast, careful implementation
- ❑ Using curiously recursive template pattern (CRTP)
- ❑ You likely don't want to develop code there
- ❑ Using it is not so complex (think Eigen)

❑ Python bindings

- ❑ A 1-to-0.99 map from C++ API to Python API
- ❑ Start by developing in Python
- ❑ Beware of the lack of accuracy ... speed is ok

- ❑ Pinocchio is a modeling library
 - ❑ Not an application
 - ❑ Not a solver
 - ❑ Some key features directly available

- ❑ You don't want the solver inside Pinocchio
 - ❑ Inverse dynamics: TSID
 - ❑ Planning and contact planning: HPP
 - ❑ Optimal control: Crocodyl
 - ❑ Optimal estimation, reinforcement learning, inverse kinematics, contact simulation ...



List of features

- ❑ URDF parser
- ❑ Forward kinematics and Jacobians
- ❑ Mass, center of mass and gen.inertia matrix
- ❑ Forward and inverse dynamics
- ❑ Model display (with Gepetto-viewer)
- ❑ Collision detection and distances (with HPP-FCL)
- ❑ Derivatives of kinematics and dynamics
- ❑ Type templatization and code generation



- ❑ Pinocchio for
 - ❑ Computing the inertia matrix, jacobians, kinematics

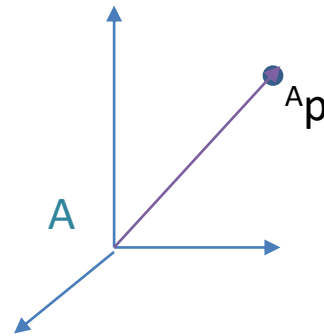
- ❑ Formulation of tasks
- ❑ Contact models
- ❑ QP resolution

- ❑ Pinocchio for
 - ❑ Geometry, collision (hpp-fcl)
 - ❑ Projectors with inverse kinematics
 - ❑ Balance constraint with dynamics
- ❑ Pinocchio encapsulated in hpp-Pinocchio
- ❑ Stochastic exploration algorithm (RRT)
- ❑ Contact checking
- ❑ Re-arrangement algorithms

- ❑ Pinocchio for
 - ❑ Kinematics and dynamics
 - ❑ And their derivatives
 - ❑ Display with Gepetto-viewer
- ❑ DDP optimizer
- ❑ Task/cost formulation



This is a point



This is not a point

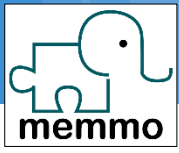
This is the representation of a point



Representing the physical world

- ❑ Pinocchio is a model
 - ❑ Of course, models are wrong
- ❑ **The way you represent geometry matters**
- ❑ Example of $SO(3)$
 - ❑ r is a map from $E(3)$ to $E(3)$
 - ❑ R is a orthonormal positive matrix
 - ❑ w is a 3D vector
 - ❑ q is a quaternion represented as a 4D vector
 - ❑ Roll-Pitch-Yaw & other Euler angles should not be used



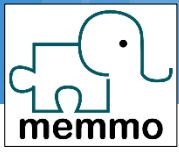




Pinocchio bases

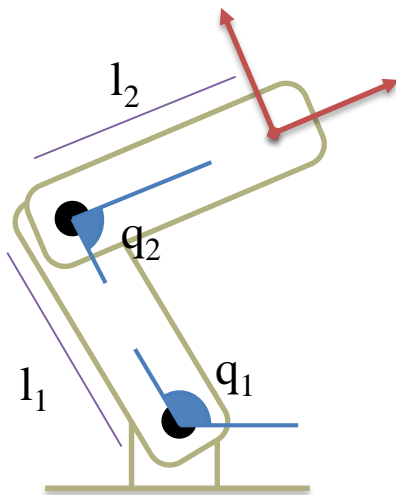
- ❑ Urdf model
- ❑ Kinematic tree
- ❑ Forward kinematics
- ❑ Display
- ❑ Spatial algebra

- ❑ Inside robot model:
 - ❑ joints: joint types and indices
 - ❑ names: joint names
 - ❑ jointPlacements: constant placement wrt parent
 - ❑ parents: hierarchy of joints representing the tree
- ❑ No bodies
 - ❑ masses and geoms are attached as tree decorations
- ❑ First joint represent the universe
 - ❑ If $nq == 7$ then $\text{len}(\text{rmodel.joints}) == 8$



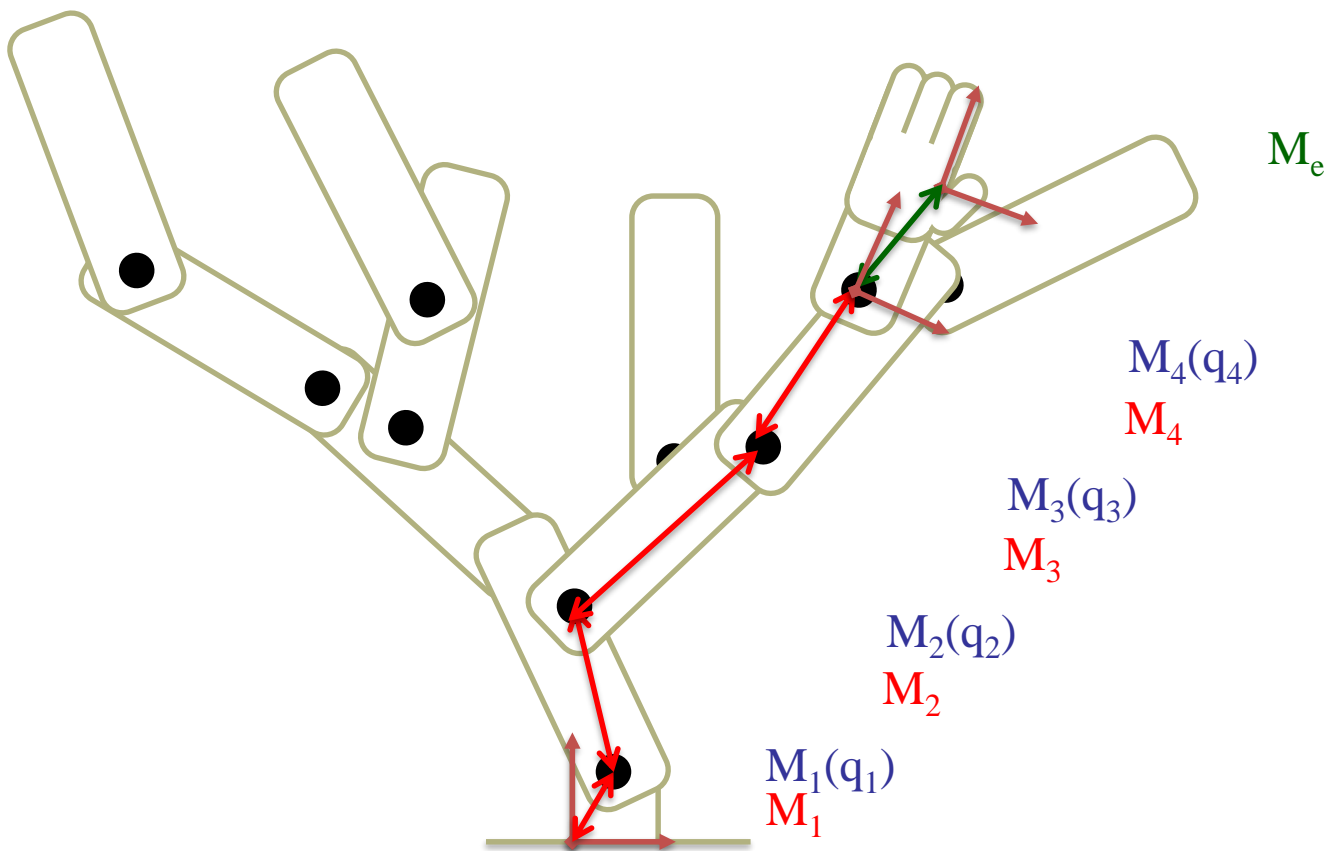
Kinematic tree





$$M(q) = \begin{bmatrix} l_1 \cos(q_1) + l_2 \cos(q_1 + q_2) \\ l_1 \sin(q_1) + l_2 \sin(q_1 + q_2) \end{bmatrix}$$

- The geometric model is a tree of joints and bodies



$$M(q) = \mathbf{M}_1 \oplus \mathbf{M}_1(q_1) \oplus \mathbf{M}_2 \oplus \dots \oplus \mathbf{M}_4 \oplus \mathbf{M}_4(q_4) \oplus \mathbf{M}_e$$



□ Direct geometry

$h: q \rightarrow h(q), \quad C^1$ continuous function

□ Direct kinematics

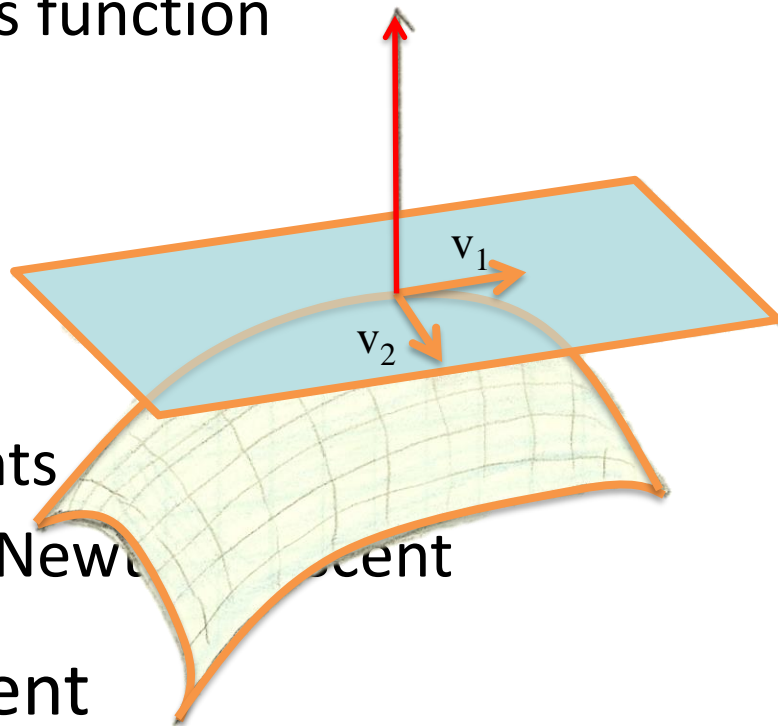
$v: q, \dot{q} \rightarrow v(q, \dot{q}) = J(q) \dot{q}$

□ Inverse geometry

- Ill defined, singular points
- Numerical inversion by Newton-Raphson

□ Integration of the descent

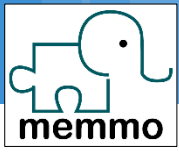
- Robot trajectory
- Quadratic problem at each step



- ❑ Gepetto-viewer is a display server
 - ❑ Python can create a client to this server
- ❑ Gepetto-viewer does not know the kinematic tree
 - ❑ Pinocchio must place the bodies
 - ❑ RobotWrapper is doing that for you (not in C++)

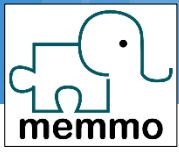
- M : placement in SE3
- v : “spatial” velocity of SE3
 - $\dot{M} = v \times M$
- α : “spatial” acceleration in SE3
 - $v \in M^6 = \mathfrak{se}(3)$
 - $\alpha \in M^6 = \mathfrak{se}(3)$
 - $\alpha = \dot{v}$
- ϕ : “spatial” force in SE3
 - Power $P = \langle \phi | v \rangle = {}^A\phi^T {}^A v \in \mathbb{R}$
 - $\eta \in F^6$: momentum
- Y : “spatial” inertia in SE3
 - $\eta = Y v$
 - $\phi = Y \alpha$





Placement





Displacements





Velocities



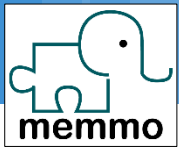


Acceleration



$${}^A \frac{d}{dt} \mathbf{v} = \frac{d}{dt} {}^A \mathbf{v} + {}^{A} \mathbf{v}_A \times {}^A \mathbf{v}$$

$${}^A \frac{d}{dt} \boldsymbol{\phi} = \frac{d}{dt} {}^A \boldsymbol{\phi} + {}^{A} \mathbf{v}_A \times {}^A \boldsymbol{\phi}$$

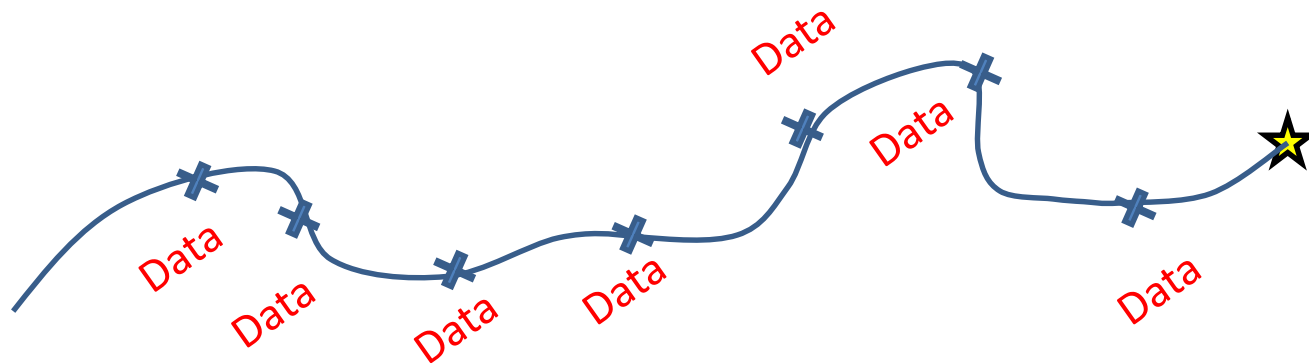


Inertias





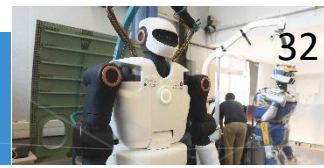
- ❑ Pinocchio.Model should be constant
 - ❑ Kinematic tree, joint model, masses, placements ...
 - ❑ Plain names used here
- ❑ Pinocchio.Data is modified by the algorithms
 - ❑ oMi , v , a
 - ❑ J , $Jcom$
 - ❑ M
 - ❑ τ , nle
- ❑ 1 Model, several Data



$$\min_{X,U} l_T(x_T) + \sum_{t=0}^{T-1} l(x_t, u_t)$$

$$\text{s.t. } x_{t+1} = f(x_t, u_t)$$

← 1 model





Forward kinematics

- ❑ `pinocchio.forwardKinematics(rmodel,rdata,
q,vq,aq)`
- ❑ Compute all the joint placements in `data.oMi`
- ❑ `M = data.oMi[-1]` : last placement
- ❑ `R = M.rotation`
- ❑ `p = M.translation`

- ❑ Pinocchio works with NumPY.Matrix
 - ❑ R is a matrix
 - ❑ p is a 1d matrix: `p.shape == (3,1)`
 - ❑ You can multiply $R * p$

- ❑ NumPy works better with Array
 - ❑ `np.zeros([3,3])` is an array
 - ❑ You cannot multiply array
 - ❑ Use `np.dot` or obtain a coefficient-wise multiplication

- ❑ SciPy works with array too

```
from scipy.optimize import fmin_slsqp  
fmin_slsqp?
```

```
fmin_slsqp(x0 = np.zeros(7),  
           func= costFunction,  
           f_eqcons = constraintFunction,  
           callback = callbackFunction)
```


- ❑ Make the optimization problem a class:
 - ❑ Problem parameters in the `__init__`
 - ❑ Cost method taking `x` as input
 - ❑ Constraint and callback method if need be

```
class OptimProblem:
```

```
    def __init__(self,rmodel):
```

```
        # Put your parameters here
```

```
        self.rmodel = rmodel
```

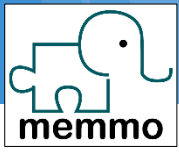
```
        self.rdata = self.rmodel.createData()
```

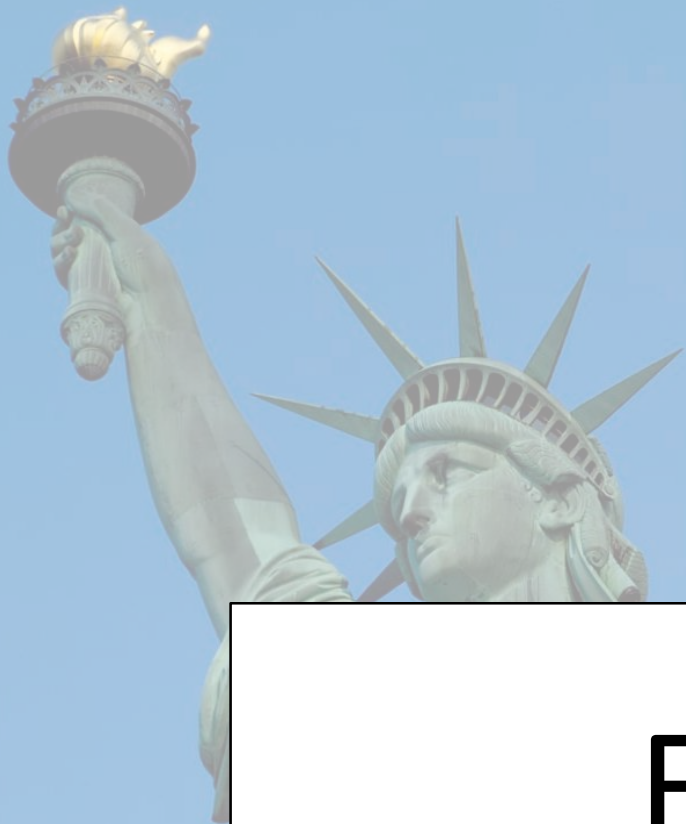
```
    def cost(self,x): return sum( x**2 )
```

```
    def callback(self,x): print(self.cost(x))
```

```
pbm = OptimProblem(robot.model)
```

```
fmin_slsqp(x0=x0,func=pbm.cost,callback=pbm.callback)
```





Frames &+

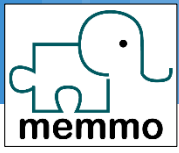
□ Joint frames

- Skeleton of the kinematic chain
- Computed by forward kinematics in `rdata.oMi`

□ “Operational” frames

- Added as decoration to the tree
- Placed with respect to a joint parent
- Stored in `rmodel.frames`
- Computed by `updateFramePlacements` in `rdata.oMf`

- ❑ Parsed from urdf
- ❑ In `rmodel.lowerPositionLimits` and `rmodel.upperPositionLimits`
- ❑ Beware, infinity by default





Log and difference

- ❑ Difference of positions
 - ❑ residuals = $p - p^*$

- ❑ Difference of rotations
 - ❑ residuals = $\log_3(R^T R^*)$

- ❑ Difference of placements
 - ❑ residuals = $\log_6(M^{-1} M^*)$

- ❑ Revolute joint
 - ❑ q of dimension one, $v_q = \dot{q}$
- ❑ Free flyer

$$q_{\text{next}} = \text{pinocchio.integrate}(q, v_q) \in Q$$

$$q_{\text{next}} = q \oplus v_q$$

$$\Delta q = v_q = \text{pinocchio.difference}(q_1, q_2) \in T_{q_1}Q$$

$$\Delta q = q_2 (-) q_1$$

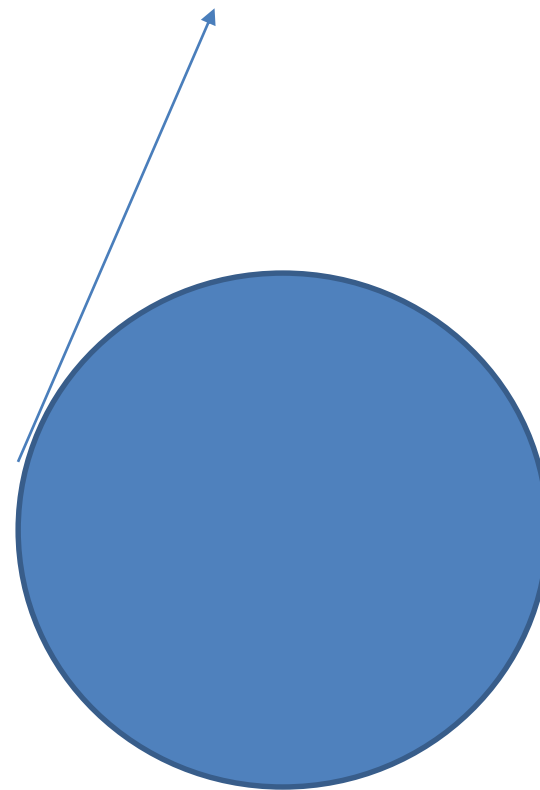
□ On a Matrix Lie Group

$$q \oplus v_q = \text{Matrix}(q) \exp(\text{skew}(v_q)) = Q \text{Exp}(v_q)$$

$$q_2 (-) q_1 = \log(Q_2^{-1} Q_1)$$

- $q = (x, y, z, \underline{q}, \dots)$ with \underline{q} unitary
- What is the result with a solver ?

```
def constraint_q(self, x):  
    return norm(x[3:7])-1
```



- We represent q
 - as the displacement v_q
 - from a reference configuration q_0

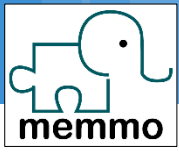
$$q = q_0 \oplus v_q$$

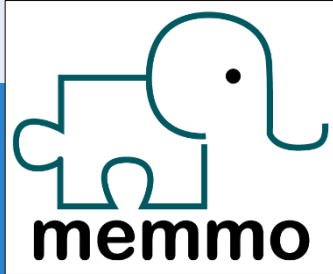


Random configuration

`pinocchio.randomConfiguration(rmodel)`





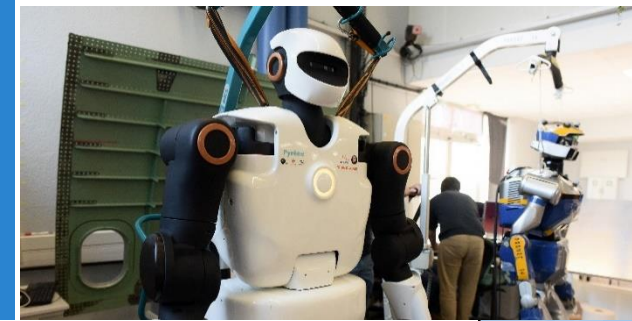


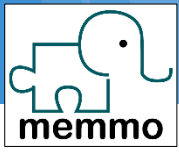
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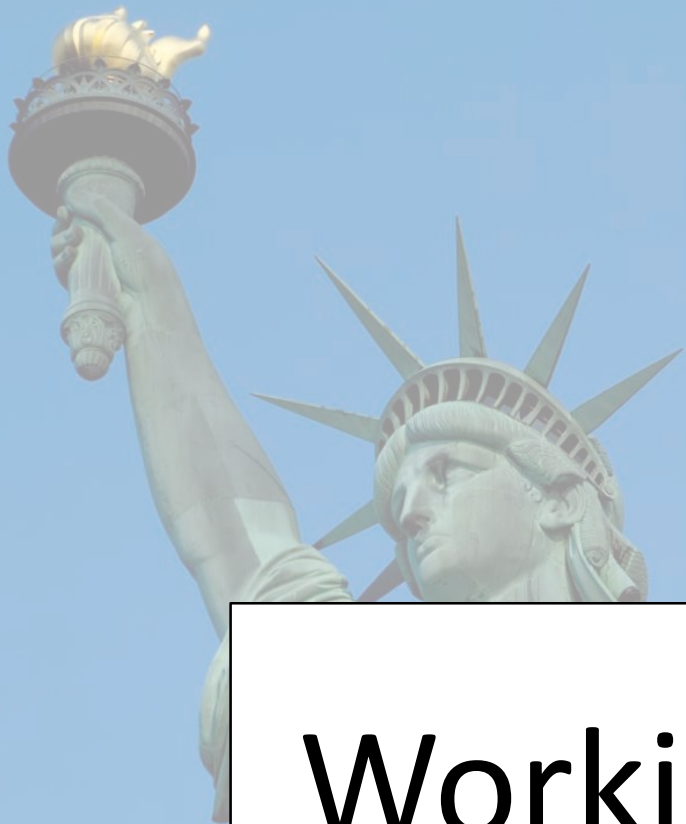
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Part II

Differencial kinematics







Working in manifolds

□ Function f :

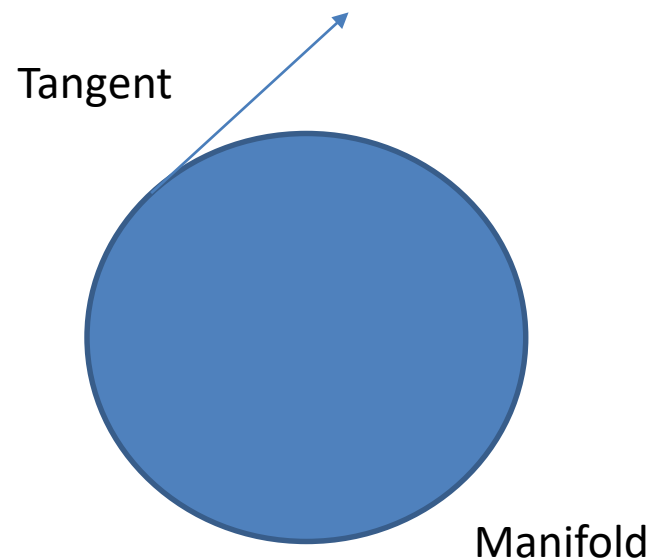
- From manifold to manifold
- $M: q \in Q \rightarrow M(q) \in SE3$

□ Derivative F_x

- From tangent to tangent
- $M_q: v_q \in TQ \rightarrow v \in M^6$

□ $v(q, v_q) = J(q) v_q$

- J : from vector to vector

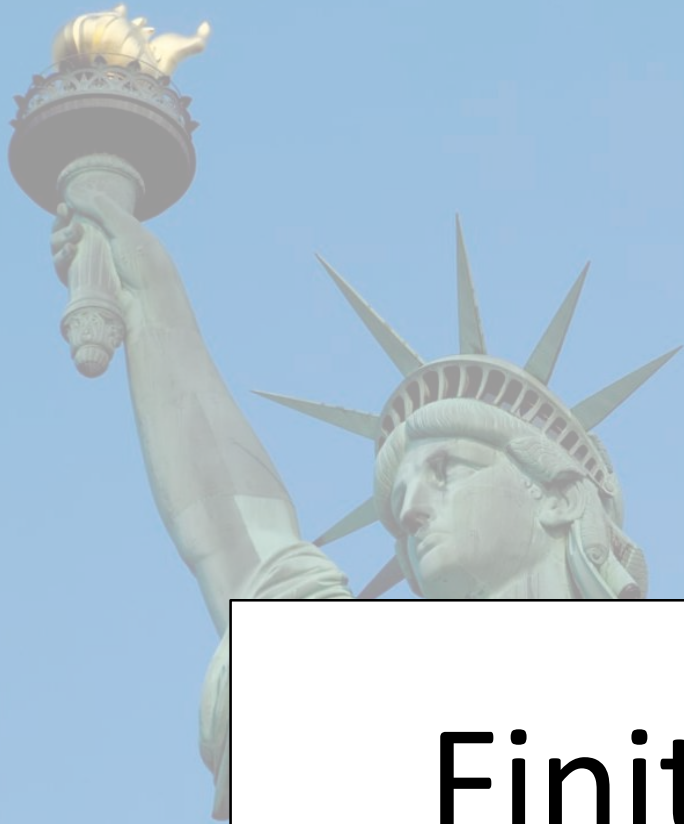


- You should know in which tangent space you work
 - Typically at the local point, or at the origin

$$v(q, v_q) = J(q) v_q$$

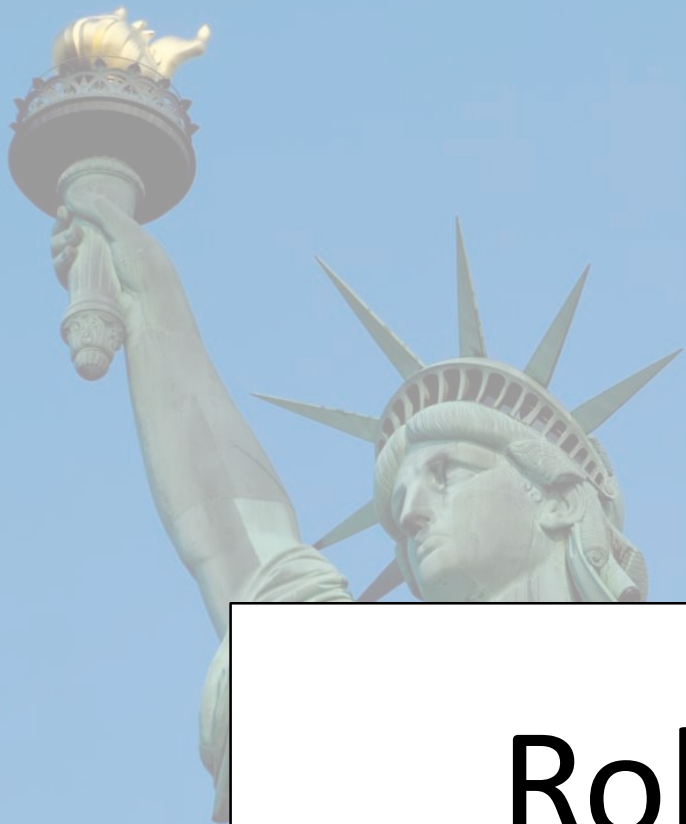
- In **Pinocchio**,
the velocity are often represented locally
 - Velocity of the free flyer in the frame of the hip





Finite differences





Robot jacobian



- ❑ Computed by two steps:
 - ❑ `computeJointJacobians(rmodel,rdata,q)`
 - ❑ `getJointJacobian(rmodel,rdata,IDX,LOCAL/GLOBAL)`
- ❑ From local to global



Frame jacobian

- Just add the additional displacement

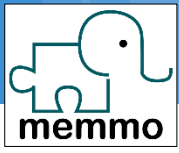
- 4 steps

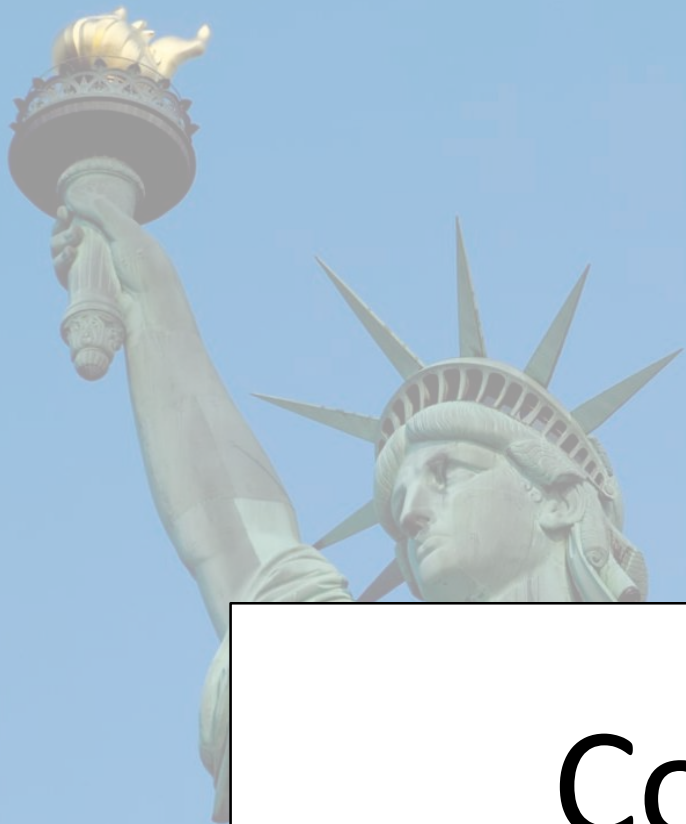
 - ComputeJointJacobian

 - updateFramePlacements

 - getFrameJacobian







Cost jacobian

$$\text{Cost}(q) = \log(M(q))$$

$$\text{Cost} = \log \circ M$$

$$\text{Cost}_q = \log_M M_q$$

M_q ?

\log_M



Log jacobian

- ❑ Computed in pinocchio
- ❑ Pinocchio.Jlog



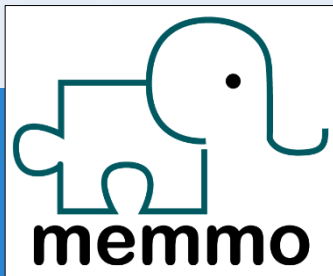


Free-flyer reparam

- ❑ Recall $q = q_0 \oplus v_q = r(v_q)$
- ❑ $c(v_q) = \log(M(r(v_q)))$

- ❑ Chain rule ...
 - ❑ $r(v) = \text{integrate}(q_0, v)$
 - ❑ $R_v = d\text{Integrate}_{dv}(q_0, v)$
 - ❑ Not implemented yet in Pinocchio
 - ❑ But it is the inverse of $d\text{Diffence}$ which is implemented

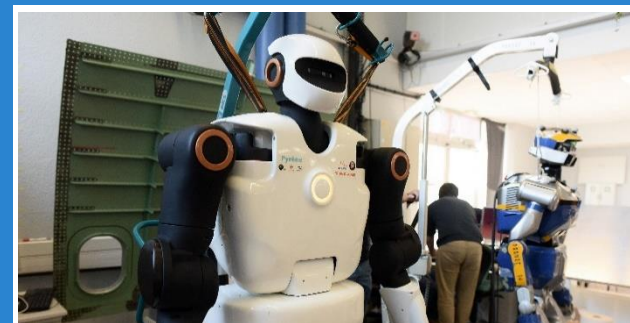




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Part III Dynamics



- Dynamic equation of the robot

$$M(q)\dot{v}_q + c(q, v_q) + g(q) = \tau_q$$



- Dynamic equation of the robot

$$M(q)\dot{v}_q + c(q, v_q) + g(q) = \tau_q$$

- Actuation of the robot

- Fixed manipulator: $\tau_q = \tau_m$

- Floating robot: $\tau_q = \begin{bmatrix} 0 \\ \tau_m \end{bmatrix} = S^T \tau_m$

- Robot in contact: : $\tau_q = S^T \tau_m + J^T \phi$





An intuition of M?





RNEA algorithm





Other algorithms

- CRBA

- ABA

- ComputeAllTerms

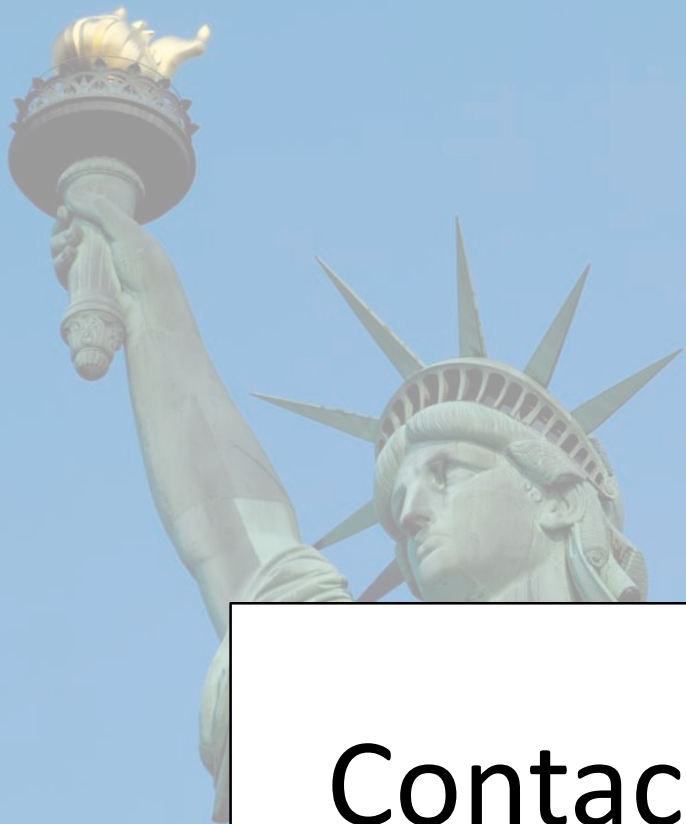




RNEA with forces







Contact inverse dynamics

$$\min_{\tau, \varphi} \|M\dot{v}_q + b(q, v_q) - \tau - J^T \varphi\|$$

- ❑ OptimProblem class
- ❑ With a x2var function that makes the dispatch
- ❑ It is a linear problem: we should not use NLP
- ❑ See TSID tomorrow