### Lecture VI: Constraints and Controllers

Parts Based on Erin Catto's Box2D Tutorial

#### **Motion Constraints**

- In practice, no rigid body is free to move around 'on its own'.
- Movement is constrained:
  - wheels on a chair
  - human body parts
  - trigger of a gun
  - opening door
  - actually almost anything you can think of in a game...



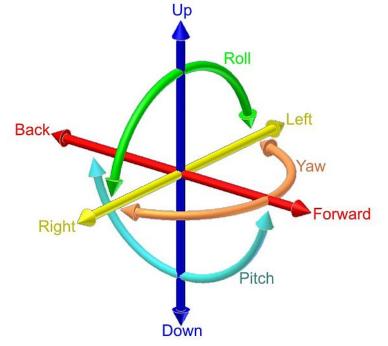
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## Degrees of Freedom

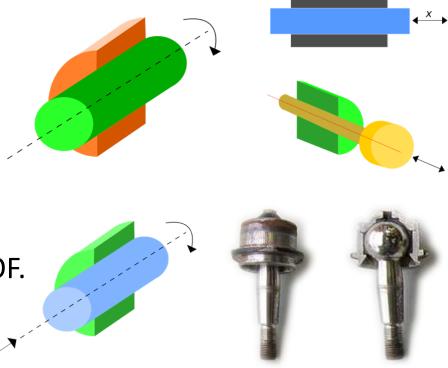
 Description of allowed movement by specifying degrees of freedom (DOF)

- Translational (3 DOF max)
- Rotational (3 DOF max)



### Kinematic Pair

- Kinematic pair: connection between two bodies that imposes constraints on their relative movement.
  - Lower pair: constraint on a point, line or plane
    - Revolute pair, or hinged joint: 1 rotational DOF.
    - Prismatic joint, or slider: 1 translational DOF.
    - Screw pair: 1 coordinated rotation/translation DOF.
    - Cylindrical pair: 1 translational + 1 rotational DOF.
    - Spherical pair, or ball-and-socket joint: 3 rotational DOF.
    - Planar pair: 3 translational DOF.
  - Higher pair: constraint on a curve or surface.

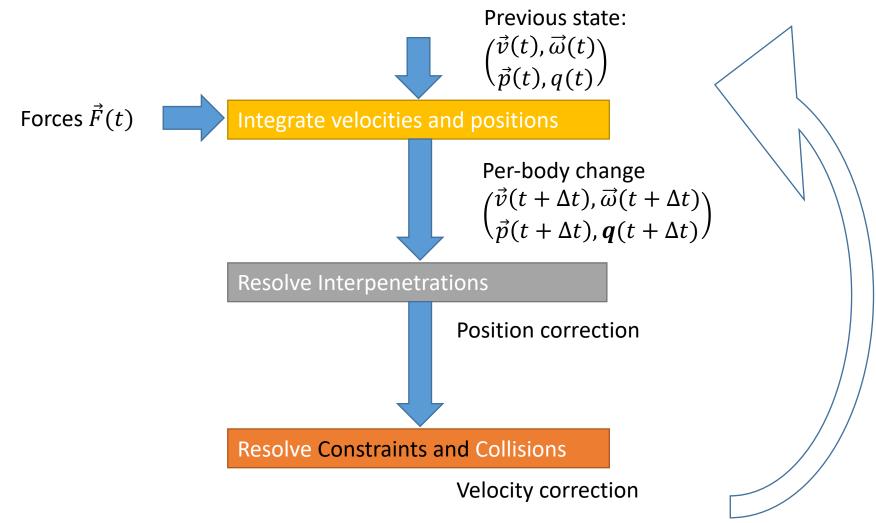


### Constraints Resolution

- Energy of components on constrained d.o.f. can be lost.
  - Converted into heat and sound  $(E_o)$
  - As opposed to putting pressure on the joint, causing degradation.
- Project net force on the unconstrained degrees of freedom.
- Not true for soft bodies!
  - Deforming accordingly...
- (Lagrangian) Approximation: no loss of energy.

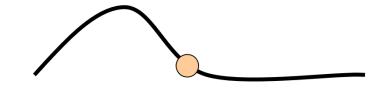
https://www.youtube.com/watch?v=t ZB0JViW68

# Constrained Game-Engine Loop



### Constraint Representation

- Often with an implicit function C(p, p', p'') = 0.
- Example: position constraint on a curve: C(x, y) = 0
  - Constraints dependent only on position are called *holonomic*.



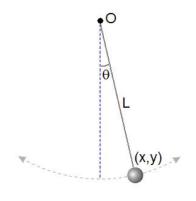
- Equality constraints: C(p, p', p'') = 0
- Inequality constraints:  $C(p, p', p'') \ge 0$ 
  - Examples?

# Differential Equality Constraints

• Our purpose: Move from  $\vec{p}(t)$  to  $\vec{p}(t+\Delta t)$  and conserve:

$$C(\vec{p}(t)) = C(\vec{p}(t + \Delta t)) = 0$$

- Algebraic interpretation: the following are conserved:
  - $C(\vec{p}(t)) = 0$  (position)
  - $dC/_{dt} = 0$  (constraint velocity)
  - $d^2C/_{dt^2} = 0$  (constraint acceleration)
- Because function should be constant 0!



$$C(p) = |p| - L$$

## Velocity Constraint

• Chain rule (single constraint, single bodies):

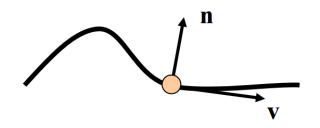
$$\frac{dc}{dt} = \frac{\partial c}{\partial p} \cdot \frac{dp}{dt} = \frac{\partial c}{\partial p} \cdot \vec{v} = 0$$

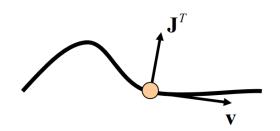


$$C = \{c_1, \dots c_n\}, P = \{p_1, \dots p_m\}$$

- We get the Jacobian  $J_{n\times m}=\left(\frac{\partial C_i}{\partial p_i}\right)$
- Kinematic velocity constraint:

$$J\vec{v}=0$$



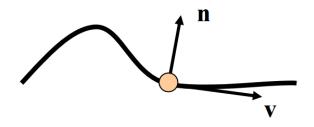


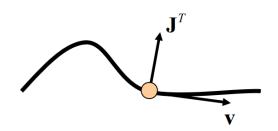
$$C(p,p',p'')=0$$

### Velocity Constraint

• J $\vec{v}=0$  means " $\vec{v}$  is in the right null space of J"

- Dimension of null space ⇔ degrees of valid movement.
- Full rank of J →
   null space empty →
   no allowed movement!
  - Physical example: move on two curves at the same time.
- How do we adhere to allowed velocity all the time?



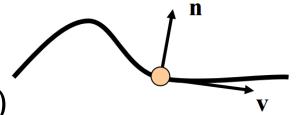


$$C(p,p',p'')=0$$

### **Acceleration Constraint**

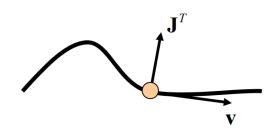
- Acceleration constraint:  $\frac{d^2C}{dt^2} = \frac{dJ}{dt}\vec{v} + J\vec{a} = 0$
- Net external force:  $\overrightarrow{F_E}$ .
- Mass matrix:

$$M_{3mx3m} = diag(M_1, M_1, M_1, \cdots, M_m, M_m, M_m)$$



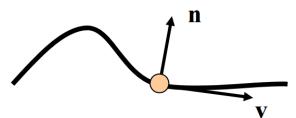
- To obey the constraint, we need to assume a virtual force  $\overrightarrow{F_c}$ .
- Dynamic force equation:

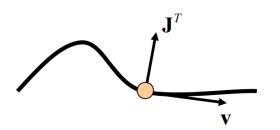
$$\frac{dJ}{dt}\vec{v} + JM^{-1}(\vec{F_E} + \vec{F_C}) = 0$$



### The Constraint Force

- The constraint force diverts free (unconstrained) movement to valid movement.
- Geometric intuition: constraint force "kills" invalid movement and nothing else.
- Principle of virtual work:  $\overrightarrow{F_c}$  does no work on  $\overrightarrow{v}$ .
- Motivation: (energy-wise) lossless constraint.
- Also: since it never causes illegal movement!
  - Where have we seen this?



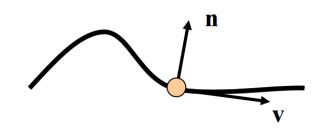


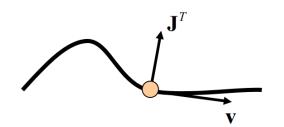
### The Constraint Force

- Principle of virtual work:  $\overrightarrow{F_c} \cdot \overrightarrow{v} = 0$ .
- We have that  $J\vec{v}=0$ .
  - $\rightarrow \overrightarrow{F_c}$  is spanned by the rows of J:

$$\overrightarrow{F_c} = \sum_i \lambda_i J(I,.) = J^T \lambda$$

- $\lambda$  : Lagrange Multiplier.
- How do we get  $\lambda$  in order to compute  $\overrightarrow{F_c}$ ?





$$\overrightarrow{F_c} = J^T \lambda$$

### Direct Solution

• To get : 
$$\frac{d^2C}{dt^2} = 0$$
, we need to solve for the Lagrange Multiplier: 
$$\frac{dJ}{dt}\vec{v} + JM^{-1}(\vec{F_E} + \vec{F_C}) = \frac{dJ}{dt}\vec{v} + JM^{-1}(\vec{F_E} + J^T\lambda)$$

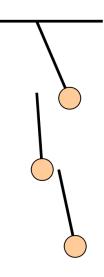
• We get:

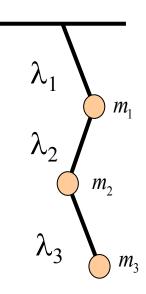
$$\int M^{-1}J^{T}\lambda = \frac{dJ}{dt}\vec{v} + JM^{-1}\overrightarrow{F_{E}}$$
Square  $|C| \times |C|$  matrix

- Disadvantages:
  - Expensive linear solve
  - $J, \frac{dJ}{dt}$  change at each point p!
  - Might need complicated derivatives:  $\frac{dJ}{dt}$ .
  - Discrete integration combined with this will make an inexact solution.

## Alternative: Sequential impulses

- Previously: working with forces to get valid acceleration.
- Now: working with impulses to get valid velocity.
- Sequential impulses (SI)
  - Applying impulses at each constraint iteratively to correct velocity.
  - Quite stable and converges to a global solution.
- Pro:
  - Easier friction and collision handling.
  - Velocity rather than acceleration.
  - In a time step, impulse  $\Leftrightarrow$  force.
- Con: velocity constraints are not precise → might produce position drift → breaking the constraint.





# Sequential impulses

#### Step 1

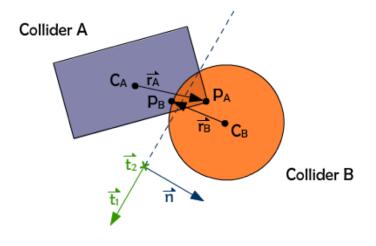
Integrate applied forces, yielding to tentative velocities

#### Step 2

Apply impulses sequentially for all constraints to correct the velocity errors

#### Step 3

Use the new velocities to update the positions



http://allenchou.net/wp-content/uploads/2013/12/contacts-figure.png

# Computing the Impulses

- A single constraint i currently has  $J_i v \neq 0$ .
  - $J_i$  row vector, normal to constraint at closest point.
- Change velocity by  $\Delta v$  to get  $J_i(v + \Delta v) = 0$ .
- $\Delta P_i = \int_t^{t+\Delta t} \overrightarrow{F_c} dt$ . For infinitesimal time step:

$$\Delta P_i \parallel \overrightarrow{F_c} \rightarrow \Delta P_i = J_i^T \lambda_i = M \Delta v$$

- (a different Lagrange multiplier that we need to compute)
- We compute  $\lambda_i$ :

$$J_i(v + M^{-1}J_i^T \lambda_i) = 0$$
$$\lambda_i = \frac{-J_i v}{J_i M^{-1}J_i^T}$$

### Sequential Integration

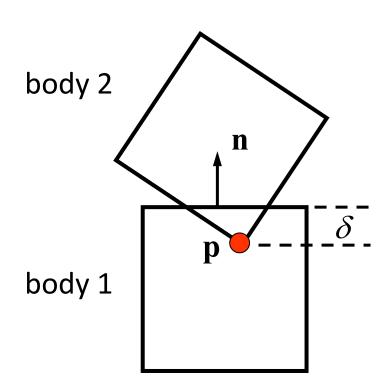
 As solving a constraint for a body may influence the solving for another one, we need an iterative process

```
• while (!done) {
      for all constraints c do solve c;
}
```

 Convergence is usually ensured by reaching either a maximal amount of iterations or a minimal change in every constraints

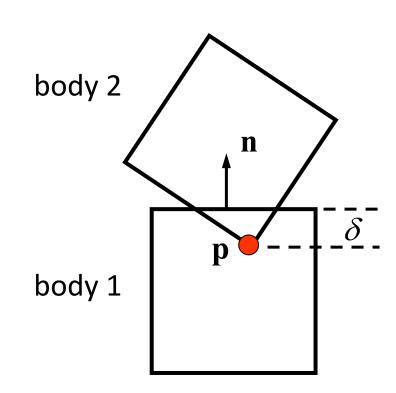
## Inequality Constraints

- Constraints are inactive when C(p) > 0.
- When they becomes active, solve for them.
  - Otherwise ignore.
- Example: collision and interpenetration.



### Collision Constraint

- In previous iteration:  $x_i$  is out of the object.
- After position integration:  $x_{i+1}$  is penetrating.
- Penetration normal:  $\hat{n}$  (pointing outwards)
- Must make sure: the projected point  $x_{i+1}$  is non-penetrating.
- Constraint:  $C(x_{i+1}) = (x_{i+1} x_i)\hat{n} \ge 0$ .
- Velocity constraint:  $\vec{v} \cdot \hat{n} \geq 0$
- Note: Can invalidate other constraints.



## Working with rigid bodies

- Constraints depend on chosen center x and orientation q: c(x,q) = 0.
- As such, velocity constraints are of the form:  $\frac{dC}{dt}(v,\omega)=0$
- Mass matrix consequently contains tensors of inertia:

$$M = diag\{M_1, M_1, M_1, I_1, \cdots, M_m, M_m, M_m, I_m\}$$

## Collision Resolution as Constraint Solving

$$\frac{dC}{dt} = (\vec{v}_{A+} - \vec{v}_{B+}) \cdot \hat{n} = (\vec{v}_{A+} + \vec{\omega}_{A+} \times \vec{r}_{A} - \vec{v}_{B+} - \vec{\omega}_{B+} \times \vec{r}_{B}) \cdot \hat{n} = \begin{pmatrix} \hat{n} \\ \vec{r}_{A} \times \hat{n} \\ -\hat{n} \\ -\vec{r}_{A} \times \hat{n} \end{pmatrix} \begin{pmatrix} \vec{v}_{A+} \\ \vec{\omega}_{A+} \\ \vec{v}_{B+} \\ \vec{\omega}_{B+} \end{pmatrix} = Jv$$

$$M = \begin{pmatrix} M_A & & & & \\ & I_A & & & \\ & & M_A & & \end{pmatrix}$$

$$\lambda_i = \frac{-Jv}{JM^{-1}J^T}$$

#### Handy Identities

$$\mathbf{A} \cdot (\mathbf{B} \times \mathbf{C}) =$$

$$\mathbf{C} \cdot (\mathbf{A} \times \mathbf{B}) =$$

$$\mathbf{B} \cdot (\mathbf{C} \times \mathbf{A})$$

$$n$$
 $v_{A+}$ 
 $v_{A-}$ 
 $v_{B-}$ 
 $v_{B-}$ 
 $v_{B+}$ 
 $v_{B-}$ 
 $v_{B-}$ 

# Collision Resolution as Constraint Solving

• We get the "well-known":

$$\lambda = \frac{-(1 + C_R)[(\bar{v}_{A-} - \bar{v}_{B-}) \cdot \hat{n}]}{\left(\frac{1}{m_A} + \frac{1}{m_B}\right) + [(\vec{r}_A \times \hat{n})^T I_A^{-1} (\vec{r}_A \times \hat{n}) + (\vec{r}_B \times \hat{n})^T I_B^{-1} (\vec{r}_B \times \hat{n})]}$$

- ...For  $C_R = 0$ .
  - Explanation: we solved for "minimum velocity to resolve".
- Elastic restitution is an extra velocity bias.

#### Handy Identities

$$\mathbf{A} \cdot (\mathbf{B} \times \mathbf{C}) =$$

$$\mathbf{C} \cdot (\mathbf{A} \times \mathbf{B}) =$$

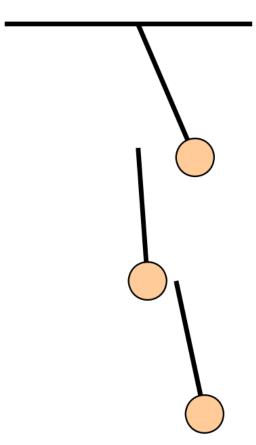
$$\mathbf{B} \cdot (\mathbf{C} \times \mathbf{A})$$

$$n - v_{A+}$$
 $v_{A-}$ 
 $v_{B-}$ 
 $v_{B-}$ 
 $v_{B-}$ 
 $v_{B-}$ 
 $v_{B-}$ 
 $v_{B-}$ 

$$\lambda_i = \frac{-Jv}{JM^{-1}J^T}$$

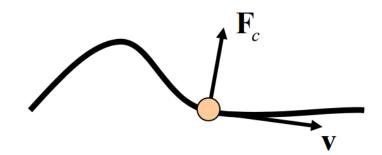
#### Position Drift

- Even if velocities are fixed, the position integration might drift
  - Since discrete time step.
- Solution: nothing simple
  - Like interpenetration resolution
- Possible easy fix: project positions as well!
  - More on that in position-based dynamics...



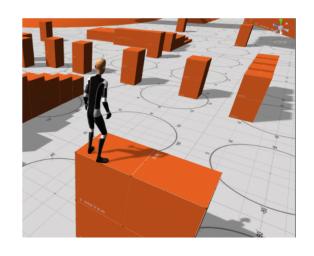
### Alternative: Reduced Coordinate Simulation

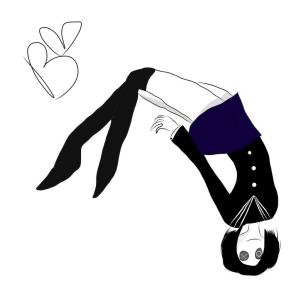
- Instead of breaking & correcting => only produce force / torque along a specific DOF.
  - Reducing possible movement.
- impulse-based is a full coordinate simulation.
- Pro: faster simulation.
  - less calculations.
  - less tuning.
- Con: more vulnerable to numerical instability.
  - might be difficult to parameterize the DOF system.
- For instance:  $Jv = 0 \Rightarrow$  use only null space of J!



#### Motion control

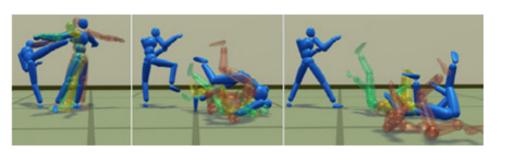
- Imagine you have rigid bodies moving and constrained correctly according to the forces you apply.
- Letting them 'live' on their own is fine for passive objects.
  - projectile, furniture, environmental objects etc.
- Living beings produce motion!
  - Otherwise they will just fall onto the ground at the beginning of the game





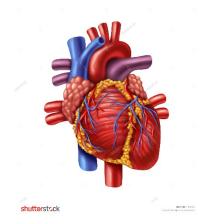
### **Motion Control**

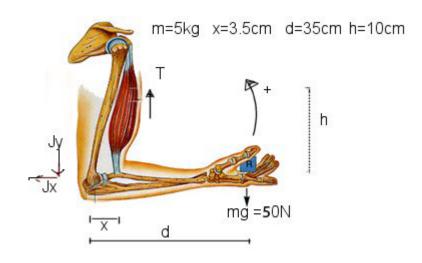
- Animated bodies are controlled using a mix of kinematics and dynamics.
- Kinematics: replay and slightly adapt pre-recorded motions.
- Dynamics: passively animate objects reacting to external forces (e.g. human ragdoll).
- A real-time controller switches from one to the other according to events, forces, poses *etc*.



### **Motion Control**

- Imagine that you want to actively actuate rigid bodies using forces and torques.
  - not yet in games but will probably in a near future...
- Actuators generate motion from within the bodies
  - Joint torques
  - External forces
  - Virtual forces
  - Muscle forces
  - contractile elements

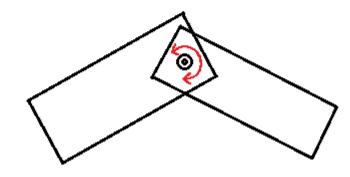




continuummechanics.org/cm/index.html

## Joint Torques

- Most straightforward actuation model.
- Joint torques directly generate torques for each actuated DOF.
- Assuming that there is a 'fake' motor at the location of the joint that can produce torque.
- e.g. increase a joint angle ⇔ positive torque.
- Very useful for tracking pre-recorded motions or any other error-based poses (e.g. balance pose)
  - Amount of applied torque depends directly on the error



http://files.maartenbaert.be/extremephysics/illustration-hinge-joint.png

### **External Forces**

- Applying external forces on the right object, local position, and for the right amount of time.
  - Difficult to produce realistic motions.
  - Does not really fit natural motions.
    - Motion originates from inside!
- Similar to a puppetry technique.
  - System control must be mastered.
- But very useful for the control of the global orientation and position of a complex system (root node).



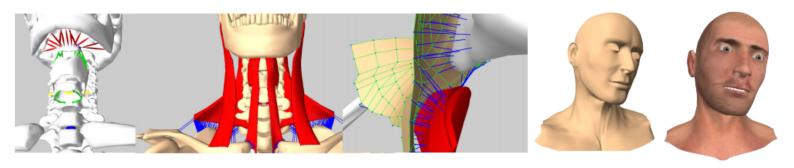
### Virtual Forces

- Not really an actuation method.
- Emulate the effect of applying an external force by computing the equivalent joint torque.
- Use the relation between joint rotation and position where a force is applied (by the Jacobian of the system).

### Muscle Forces

- Motion actually comes from the contraction of muscles.
  - they also produce torques at joints, but not in a 'fake motor' way.
- Emulating physical behavior to the model => "realistic" actuation of rigid bodies.
  - commonly used in biomechanics / motion analysis
  - the muscle model is usually a combination of non-linear springs and dampers.





[Lou et al. 2013] "Physics-based Human Neck Simulation"

### Controller design

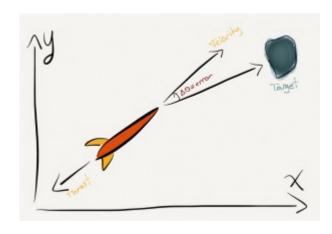
- Joint-space motion control.
  - defines and tracks kinematics target.

- Stimulus-response network control
  - genetically evolves controllers according to objectives.

- Constrained dynamics optimization control.
  - finds optimal torques through online optimization.

### PD Controller

- Proportional Derivative (PD) controller
  - Used to compute joint torques linearly proportional to the difference between the current state and the target state.
  - Based on joint orientation and angular velocity.



 $http://www.gamedev.net/uploads/monthly_10_2014/ccs-224713-0-48143500-1414694390.jpg$ 

### PD Controller

$$\tau = k_p(\theta_d - \theta) + k_v(\dot{\theta}_d - \dot{\theta})$$

#### Where:

- $\tau$  is the generated joint torque.
- $\theta_d$  and  $\theta$  the desired and current joint angles.
- $\dot{\theta}_d$  and  $\dot{\theta}$  the desired and current joint angular velocity.
- $k_p$  and  $k_v$  are the controller gain.
  - Regulating how responsive the controller is to deviation.

### PD Controller

- Con: The motion-controller designer needs to carefully define the gain values.
  - too high  $k_p$ : producing stiff unresponsive motions.
  - too low  $k_p$ : not track target correctly.
  - too high  $k_{\nu}$ : converging too slowly to the target.
  - too low  $k_{\nu}$ : producing oscillations.
- ...Expect to spend time fine tuning gains!

$$\tau = k_p(\theta_d - \theta) + k_v(\dot{\theta}_d - \dot{\theta})$$