
COMPUTER SCIENCE Capstone PBL

- Character Motion Synthesis and Character Control

4 - Motion Editing, Motion Synthesis

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Fall 2022

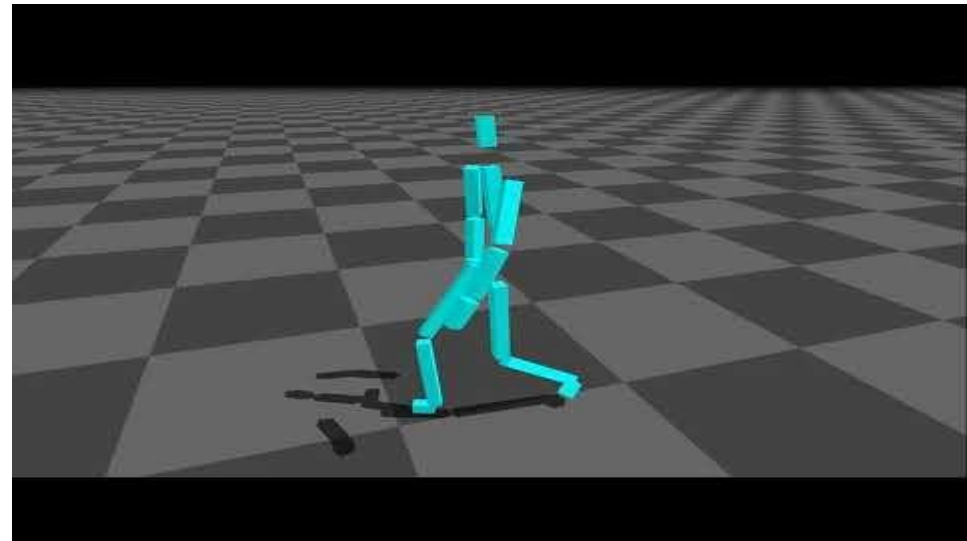
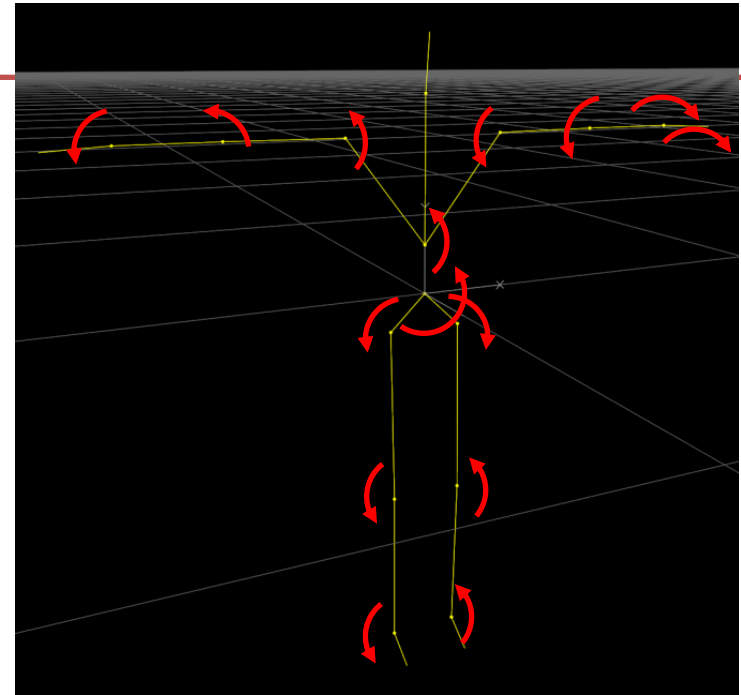
Today's Topics

- 3D Orientation & Rotation
- Inverse Kinematics
 - Limb IK
- Motion Editing Techniques
 - Posture / Motion Difference
 - Motion Warping
- Intro to More Motion Editing Techniques
 - Interpolation of Postures
 - Time Warping
 - Motion Stitching
 - Motion Blending
- Intro to Data-Driven Motion Synthesis
 - Motion Graph
 - Motion Matching
- Intro to Deep Learning-Based Motion Synthesis

3D Orientation & Rotation

Recall: Skeletal Motion

- "*Motion*": time-varying data
 - internal joint motion
 - w.r.t. default frame of each joint
 - the frame after applying joint offset to the parent frame
 - usually **rotation**
 - position and **orientation** of skeletal root
 - w.r.t. global frame
 - usually the pelvis part



Orientation vs. Rotation and Position vs. Translation

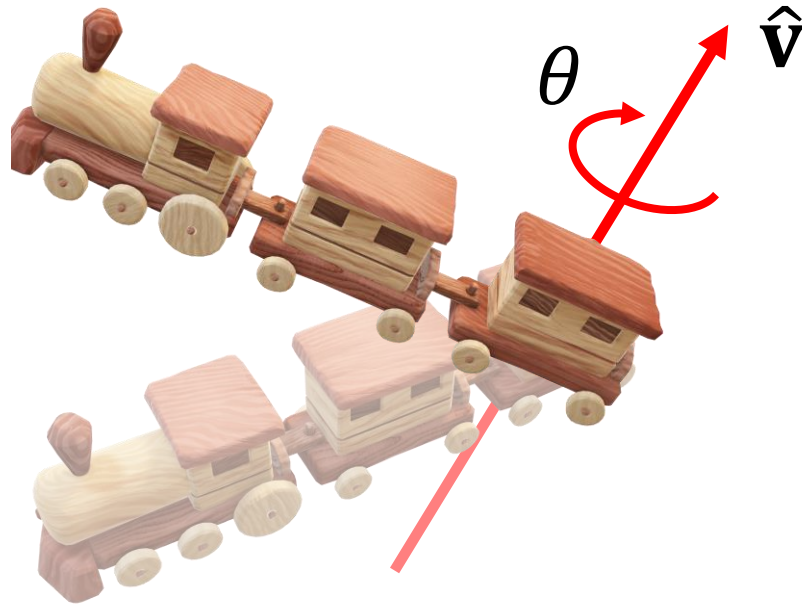
- **Orientation & Position** - *state*
 - Position: The state of being located.
 - **Orientation**: The state of being oriented. (angular position)
- **Rotation & Translation** - *movement*
 - Translation: Linear movement (difference btwn. positions)
 - **Rotation**: Angular movement (difference btwn. orientations)
- This relationship is analogous to *point* vs. *vector* in *coordinate-invariant geometric programming*.
 - point: position
 - vector: difference between two points

Describing 3D Rotation & Orientation

- Describing 3D rotation & orientation is not as intuitive as the 2D case.
- Several ways to describe 3D rotation and orientation
 - Euler angles
 - Rotation vector (Axis-angle)
 - Rotation matrices
 - Unit quaternions

3D Rotation

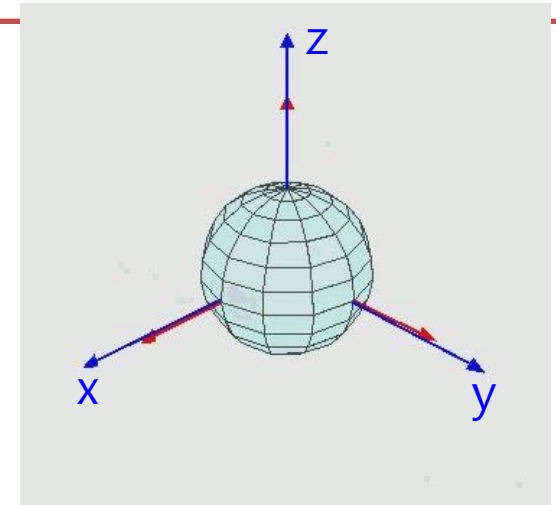
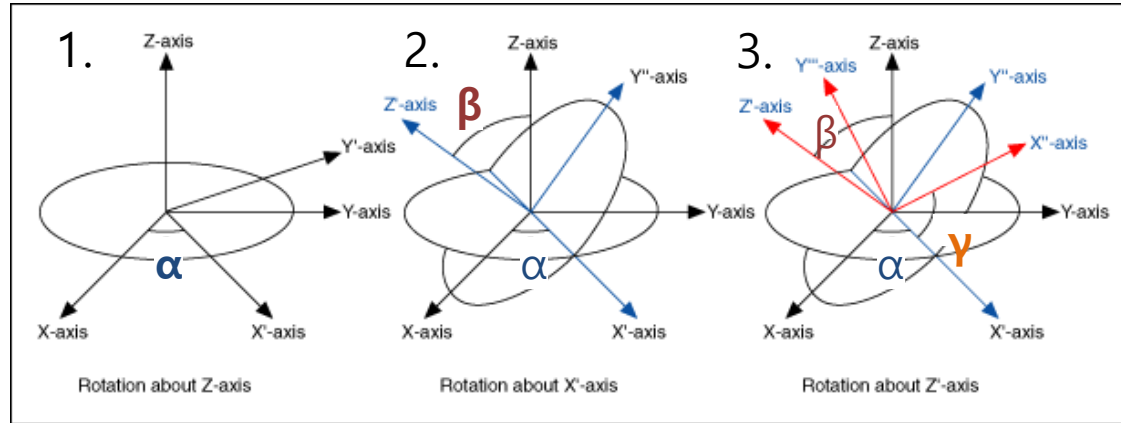
- For any 3D rotation, we can always find a fixed *axis* of rotation and an *angle* about the axis.



Euler Angles

- Express any arbitrary 3D rotation using **three rotation angles about three principle axes.**
- Possible 12 combinations
 - XYZ, XYX, XZY, XZX
 - YZX, YZY, YXZ, YXY
 - ZXY, ZXZ, ZYX, ZYZ
 - (Combination is possible as long as the same axis does not appear consecutively.)

Example: ZXZ Euler Angles



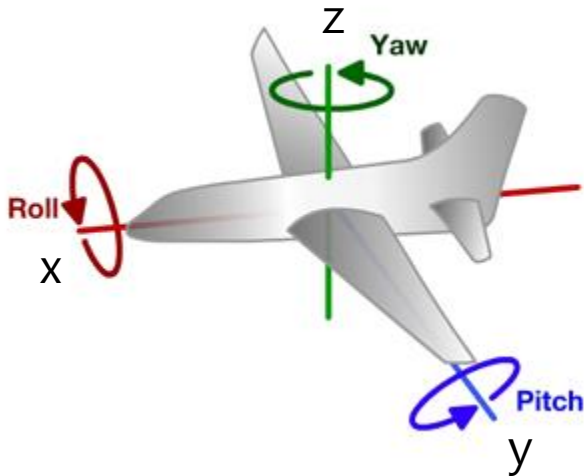
- 1. Rotate about Z-axis by α
- 2. Rotate about X-axis of the new frame by β
- 3. Rotate about Z-axis of the new frame by γ

<https://commons.wikimedia.org/wiki/File:Euler2a.gif>

$$R = \begin{bmatrix} \cos \alpha & -\sin \alpha & 0 \\ \sin \alpha & \cos \alpha & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos \beta & -\sin \beta \\ 0 & \sin \beta & \cos \beta \end{bmatrix} \begin{bmatrix} \cos \gamma & -\sin \gamma & 0 \\ \sin \gamma & \cos \gamma & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$R = R_z(\alpha) R_x(\beta) R_z(\gamma)$$

Example: Yaw-Pitch-Roll Convention (ZYX Euler Angles)

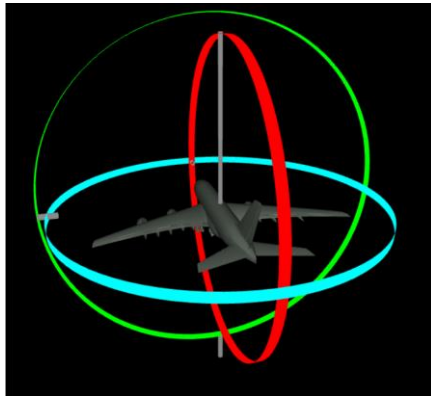


- Common for describing the orientation of aircrafts
- 1. Rotate about Z-axis by **yaw** angle
- 2. Rotate about Y-axis of the new frame by **pitch** angle
- 3. Rotate about X-axis of the new frame by **roll** angle

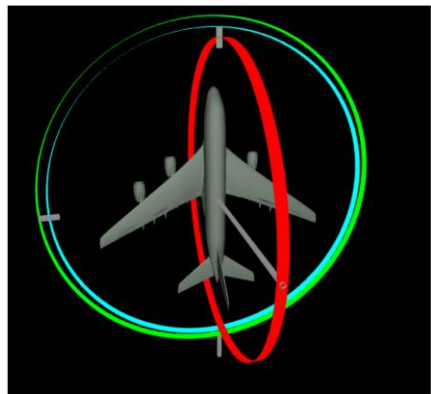
$$R = R_z(\text{yaw}) R_y(\text{pitch}) R_x(\text{roll})$$

Gimbal Lock

- Euler angles temporarily lose a DOF when the two axes are aligned.



Normal configuration:
The object can rotate freely.

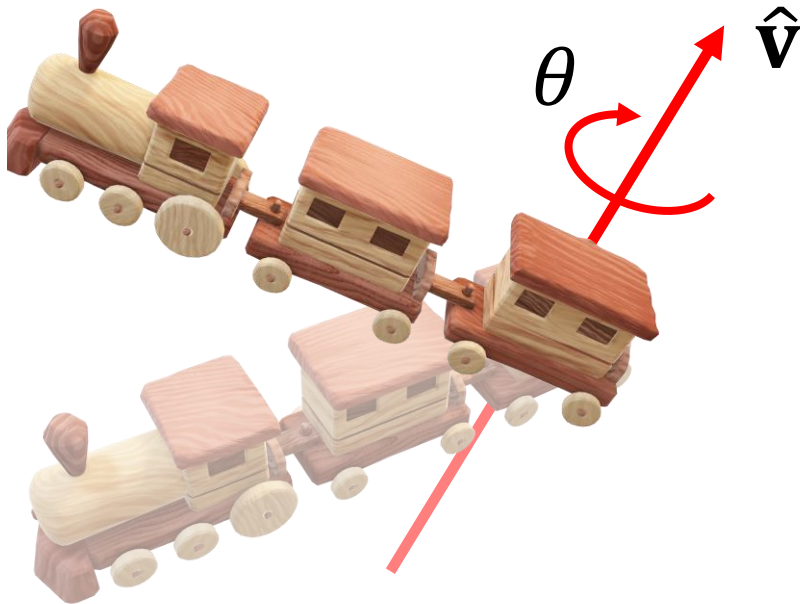


Gimbal lock configuration:
The object can not rotate in one direction.

Try:

- <https://compsci290-s2016.github.io/CoursePage/Materials/EulerAnglesViz/index.html>
- Set pitch to 90°

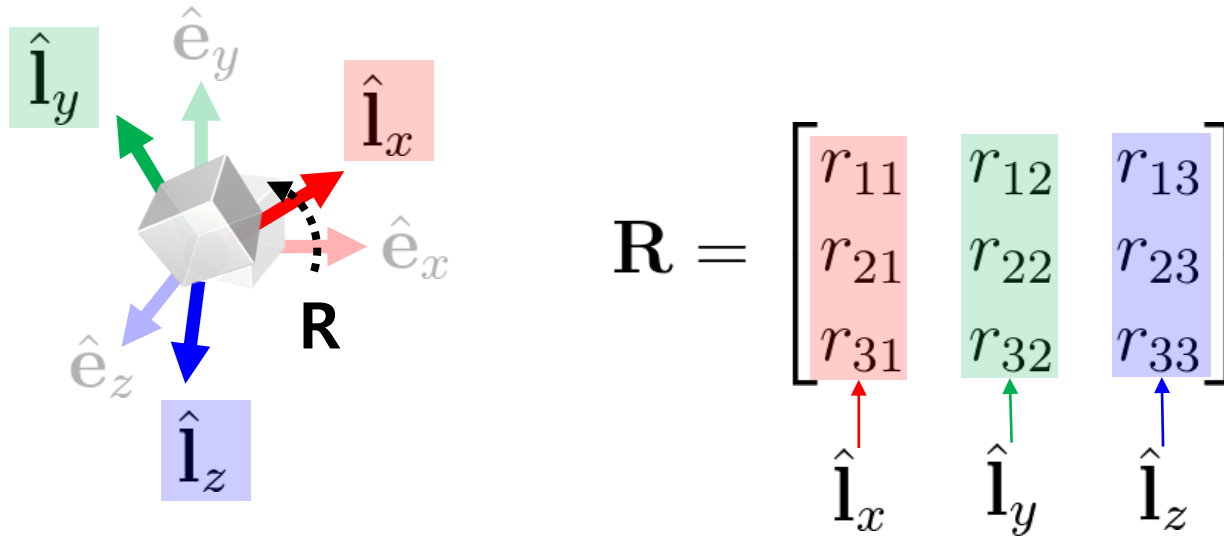
Rotation Vector (Axis-Angle)



$\hat{\mathbf{v}}$: rotation axis (unit vector)
 θ : scalar angle

- Rotation vector: $\mathbf{v} = \theta \hat{\mathbf{v}} = (x, y, z)$
- Axis-Angle: $(\theta, \hat{\mathbf{v}})$

Rotation Matrix



- A rotation matrix defines
 - **Orientation** of new rotated frame or,
 - **Rotation** from a global frame to be that rotated frame

Rotation Matrix

- A square matrix \mathbf{R} is a rotation matrix if and only if

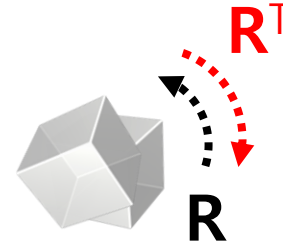
$$\boxed{1. \mathbf{R}\mathbf{R}^T = \mathbf{R}^T\mathbf{R} = \mathbf{I}} \&\& \boxed{2. \det(\mathbf{R}) = 1}$$

- A rotation matrix is an **orthogonal matrix with determinant 1**.
 - Sometimes it is called *special orthogonal matrix*
 - A set of **rotation matrices of size 3** forms a *special orthogonal group, $SO(3)$*

Geometric Properties of Rotation Matrix

- \mathbf{R}^T is an inverse rotation of \mathbf{R} .

– Because, $\mathbf{R}\mathbf{R}^T = \mathbf{I} \iff \mathbf{R}^{-1} = \mathbf{R}^T$



- $\mathbf{R}_1\mathbf{R}_2$ is a rotation matrix as well (composite rotation).

– proof) $(\mathbf{R}_1\mathbf{R}_2)^T(\mathbf{R}_1\mathbf{R}_2) = \mathbf{R}_2^T\mathbf{R}_1^T\mathbf{R}_1\mathbf{R}_2 = \mathbf{R}_2^T\mathbf{R}_2 = \mathbf{I}$

and $\det(\mathbf{R}_1\mathbf{R}_2) = \det(\mathbf{R}_1) \cdot \det(\mathbf{R}_2) = 1$

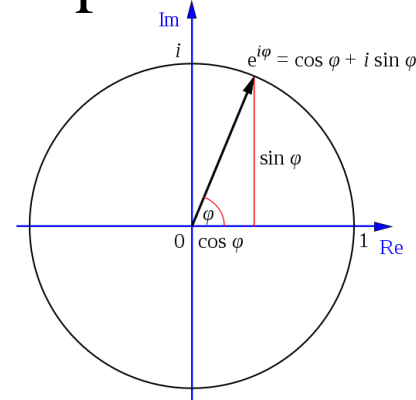
- The length of vector \mathbf{v} is not changed after applying a rotation matrix \mathbf{R} .

– proof) $\|\mathbf{R}\mathbf{v}\|^2 = (\mathbf{R}\mathbf{v})^T(\mathbf{R}\mathbf{v}) = \mathbf{v}^T\mathbf{R}^T\mathbf{R}\mathbf{v} = \mathbf{v}^T\mathbf{v} = \|\mathbf{v}\|^2$

$$\boxed{\mathbf{v}^T} \boxed{\mathbf{v}} = \mathbf{v} \cdot \mathbf{v}$$

Quaternions

- Complex numbers can be used to represent 2D rotations.
- $z = x + yi$, where $i^2 = -1$
 - Euler's formula: $e^{i\varphi} = \cos \varphi + i \sin \varphi$
- Basic idea: Quaternion is its extension to 3D space.
- $q = w + xi + yj + zk$
- , where $i^2 = j^2 = k^2 = ijk = -1$
 $ij = k, \quad jk = i, \quad ki = j$
 $ji = -k, \quad kj = -i, \quad ik = -j$



Quaternions

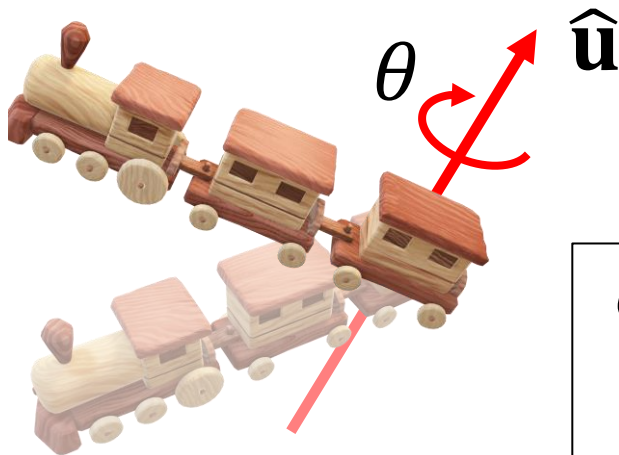
- $q = w + xi + yj + zk$
 - w is called a *real part* (or *scalar part*).
 - $xi + yj + zk$ is called an *imaginary part* (or *vector part*).

- Notation:

$$\begin{aligned} q &= w + xi + yj + zk \\ &= (w, x, y, z) \\ &= (w, \mathbf{v}) \end{aligned}$$

Unit Quaternions

- *Unit quaternions* represent 3D rotations.
- $q = w + ix + jy + kz$,
- , where $w^2 + x^2 + y^2 + z^2 = 1$
- Rotation about axis $\hat{\mathbf{u}}$ by angle θ :



$$\mathbf{q} = \left(\cos \frac{\theta}{2}, \hat{\mathbf{u}} \sin \frac{\theta}{2} \right)$$

$$\begin{aligned} q &= w + xi + yj + zk \\ &= (w, x, y, z) \\ &= (w, \mathbf{v}) \end{aligned}$$

Unit Quaternions

- A 3D position (x, y, z) is represented as a *pure imaginary quaternion* $(0, x, y, z)$.
- If $\mathbf{p} = (0, x, y, z)$ is rotated about axis $\hat{\mathbf{v}}$ by angle θ , then the rotated position $\mathbf{p}' = (0, x', y', z')$ is:



Unit Quaternions

Identity $\mathbf{q} = (1, 0, 0, 0)$

Multiplication	$\begin{aligned}\mathbf{q}_1 \mathbf{q}_2 &= (w_1, \mathbf{v}_1)(w_2, \mathbf{v}_2) \\ &= (w_1 w_2 - \mathbf{v}_1 \cdot \mathbf{v}_2, w_1 \mathbf{v}_2 + w_2 \mathbf{v}_1 + \mathbf{v}_1 \times \mathbf{v}_2)\end{aligned}$
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Inverse
$$\begin{aligned}\mathbf{q}^{-1} &= (w, -x, -y, -z)/(w^2 + x^2 + y^2 + z^2) \\ &= (-w, x, y, z)/(w^2 + x^2 + y^2 + z^2)\end{aligned}$$

- $\mathbf{q}_1 \mathbf{q}_2$: rotate by \mathbf{q}_1 then \mathbf{q}_2 w.r.t. local frame
or rotate by \mathbf{q}_2 then \mathbf{q}_1 w.r.t. global frame
- $\mathbf{p}' = \mathbf{q}_1 \mathbf{q}_2 \mathbf{p} (\mathbf{q}_1 \mathbf{q}_2)^{-1} = \mathbf{q}_1 (\mathbf{q}_2 \mathbf{p} \mathbf{q}_2^{-1}) \mathbf{q}_1^{-1}$

Which Representation to Use?

- General recommendation: Use **rotation matrices** or **unit quaternions**.
- Because they provide accurate "addition", "subtraction", and interpolation of rotations.
- In addition, Euler angles have the gimbal lock issue.

"Addition" of Rotations

- Rotation matrix, Unit quaternion:
 - $\mathbf{R}_1 \mathbf{R}_2$ (or $\mathbf{q}_1 \mathbf{q}_2$)
 - Rotate by \mathbf{R}_1 (or \mathbf{q}_1), then by \mathbf{R}_2 (or \mathbf{q}_2) w.r.t. local frame.
 - (Element-wise addition does NOT even produce a rotation matrix or unit quaternion.)
- Euler angles:
 - $(\alpha_1, \beta_1, \gamma_1) + (\alpha_2, \beta_2, \gamma_2) = (\alpha_1 + \alpha_2, \beta_1 + \beta_2, \gamma_1 + \gamma_2)$?
 - Does **NOT** mean rotate by $(\alpha_1, \beta_1, \gamma_1)$, then by $(\alpha_2, \beta_2, \gamma_2)$!
- Rotation vector:
 - $\mathbf{v}_1 + \mathbf{v}_2$?
 - Does **NOT** mean rotate by \mathbf{v}_1 , then by \mathbf{v}_2 !

"Subtraction" of Rotations

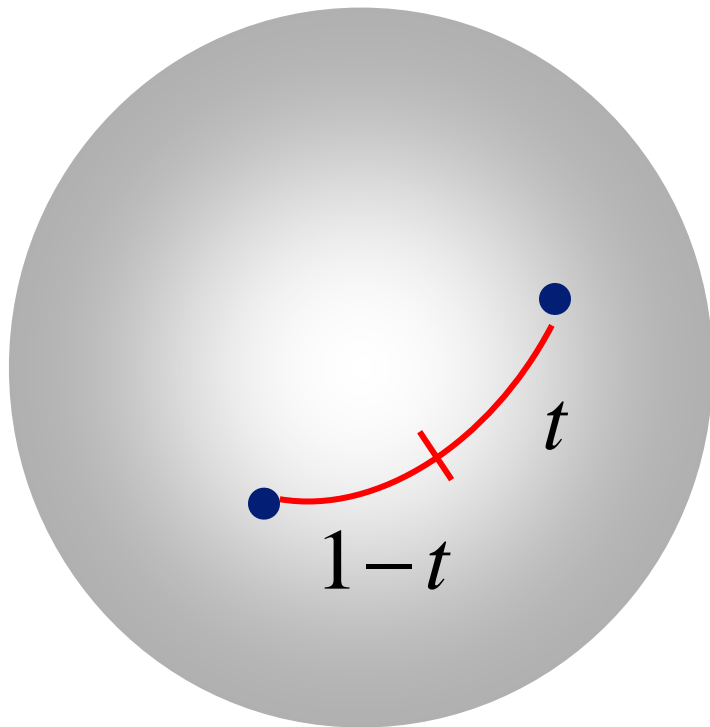
- Rotation matrix, Unit quaternion:
 - $\mathbf{R}_1^T \mathbf{R}_2$ (or $\mathbf{q}_1^{-1} \mathbf{q}_2$)
 - Rotational difference: A rotation matrix that rotate a frame \mathbf{R}_1 (or \mathbf{q}_1) to be coincident with the frame \mathbf{R}_2 (or \mathbf{q}_2) when applied w.r.t. the frame \mathbf{R}_1 (or \mathbf{q}_1)
 - Because $\mathbf{R}_1 (\mathbf{R}_1^T \mathbf{R}_2) = \mathbf{R}_2$
 - (Element-wise subtraction does NOT even produce a rotation matrix or unit quaternion.)
- Euler angles:
 - $(\alpha_2, \beta_2, \gamma_2) - (\alpha_1, \beta_1, \gamma_1) = (\alpha_2 - \alpha_1, \beta_2 - \beta_1, \gamma_2 - \gamma_1) ?$
 - Does **NOT** mean difference between rotation $(\alpha_1, \beta_1, \gamma_1)$ and $(\alpha_2, \beta_2, \gamma_2) !$
- Rotation vector:
 - $\mathbf{v}_2 - \mathbf{v}_1 ?$
 - Does **NOT** mean difference between rotation \mathbf{v}_1 and $\mathbf{v}_2 !$

Interpolation of Rotations

- Can we just linear interpolate each element of
 - Euler angles
 - Rotation vector
 - Rotation matrix
 - Unit quaternion
- ?
- → No!
- The right answer: **slerp**

Slerp

- **Slerp** [Shoemake 1985]
 - Spherical linear interpolation
 - Linear interpolation of two orientations



“t” refers power, not transpose

$$\begin{aligned}\text{slerp}(\mathbf{R}_1, \mathbf{R}_2, t) &= \mathbf{R}_1 (\mathbf{R}_1^T \mathbf{R}_2)^t \\ &= \mathbf{R}_1 \exp(t \cdot \log(\mathbf{R}_1^T \mathbf{R}_2))\end{aligned}$$

Slerp

- $\text{slerp}(\mathbf{R}_1, \mathbf{R}_2, t) = \mathbf{R}_1(\mathbf{R}_1^T \mathbf{R}_2)^t$
 $= \mathbf{R}_1 \exp(t \cdot \log(\mathbf{R}_1^T \mathbf{R}_2))$
 - $\exp()$: **rotation vector** to **rotation matrix**
 - $\log()$: **rotation matrix** to **rotation vector**
- Implication
 - $\mathbf{R}_1^T \mathbf{R}_2$: difference between orientation \mathbf{R}_1 and \mathbf{R}_2 ($\mathbf{R}_2(-)\mathbf{R}_1$)
 - \mathbf{R}^t : scaling rotation (scaling rotation angle)
 - $\mathbf{R}_a \mathbf{R}_b$: add rotation \mathbf{R}_b to orientation \mathbf{R}_a ($\mathbf{R}_a(+)\mathbf{R}_b$)

Exp & Log

- **Exp (exponential): rotation vector to rotation matrix**

- Given normalized rotation axis $\mathbf{u}=(u_x, u_y, u_z)$, rotation angle θ

$$R = \begin{bmatrix} \cos \theta + u_x^2 (1 - \cos \theta) & u_x u_y (1 - \cos \theta) - u_z \sin \theta & u_x u_z (1 - \cos \theta) + u_y \sin \theta \\ u_y u_x (1 - \cos \theta) + u_z \sin \theta & \cos \theta + u_y^2 (1 - \cos \theta) & u_y u_z (1 - \cos \theta) - u_x \sin \theta \\ u_z u_x (1 - \cos \theta) - u_y \sin \theta & u_z u_y (1 - \cos \theta) + u_x \sin \theta & \cos \theta + u_z^2 (1 - \cos \theta) \end{bmatrix}$$

(Rodrigues' rotation formula)

- **Log (logarithm): rotation matrix to rotation vector**

Given rotation matrix \mathbf{R} , compute axis \mathbf{v} and angle θ

$$\theta = \cos^{-1}((R_{11} + R_{22} + R_{33} - 1)/2)$$

$$v_1 = (R_{32} - R_{23})/(2 \sin \theta)$$

$$v_2 = (R_{13} - R_{31})/(2 \sin \theta)$$

$$v_3 = (R_{21} - R_{12})/(2 \sin \theta)$$

→ But this formula has a singularity at $\theta=k\pi$, where k is an integer.

Algorithm for Log

Algorithm for Computing the Logarithm of a Rotation Matrix

Objective: Given $R \in SO(3)$, find $\omega \in \mathbb{R}^3$, $\|\omega\| = 1$, and $\theta \in [0, \pi]$ such that

$$R = e^{[\omega]\theta} = I + \sin \theta [\omega] + (1 - \cos \theta)[\omega]^2. \quad (3.62)$$

(i) If $\text{tr } R = 3$, then set $\omega = 0$, $\theta = 0$.

(ii) If $\text{tr } R = -1$, then set $\theta = \pi$, and ω to any of the three following vectors that is nonzero:

$$\omega = \frac{1}{\sqrt{2(1 + r_{33})}} \begin{bmatrix} r_{13} \\ r_{23} \\ 1 + r_{33} \end{bmatrix} \quad (3.63)$$

or

$$\omega = \frac{1}{\sqrt{2(1 + r_{22})}} \begin{bmatrix} r_{12} \\ 1 + r_{22} \\ r_{32} \end{bmatrix} \quad (3.64)$$

or

$$\omega = \frac{1}{\sqrt{2(1 + r_{11})}} \begin{bmatrix} 1 + r_{11} \\ r_{21} \\ r_{31} \end{bmatrix}. \quad (3.65)$$

(iii) Otherwise set $\theta = \cos^{-1} \left(\frac{\text{tr } R - 1}{2} \right) \in [0, \pi)$ and $[\omega] = \frac{1}{2 \sin \theta} (R - R^T)$.

See section 3.2.3.3 of "Modern Robotics" for detail:
<http://hades.mech.northwestern.edu/images/2/25/MR-v2.pdf>

Exp & Log

- However, you can just use `scipy.spatial.transform.Rotation` for `exp()` and `log()`.
 - <https://docs.scipy.org/doc/scipy/reference/generated/scipy.spatial.transform.Rotation.html>

Slerp

- Quaternion slerp:

- $\text{slerp}(\mathbf{q}_1, \mathbf{q}_2, t) = \mathbf{q}_1(\mathbf{q}_1^{-1}\mathbf{q}_2)^t$

- Geometric slerp (equivalent):

- $\text{slerp}(\mathbf{q}_1, \mathbf{q}_2, t) = \frac{\sin((1-t)\varphi)}{\sin \varphi} \mathbf{q}_1 + \frac{\sin(t\varphi)}{\sin \varphi} \mathbf{q}_2$

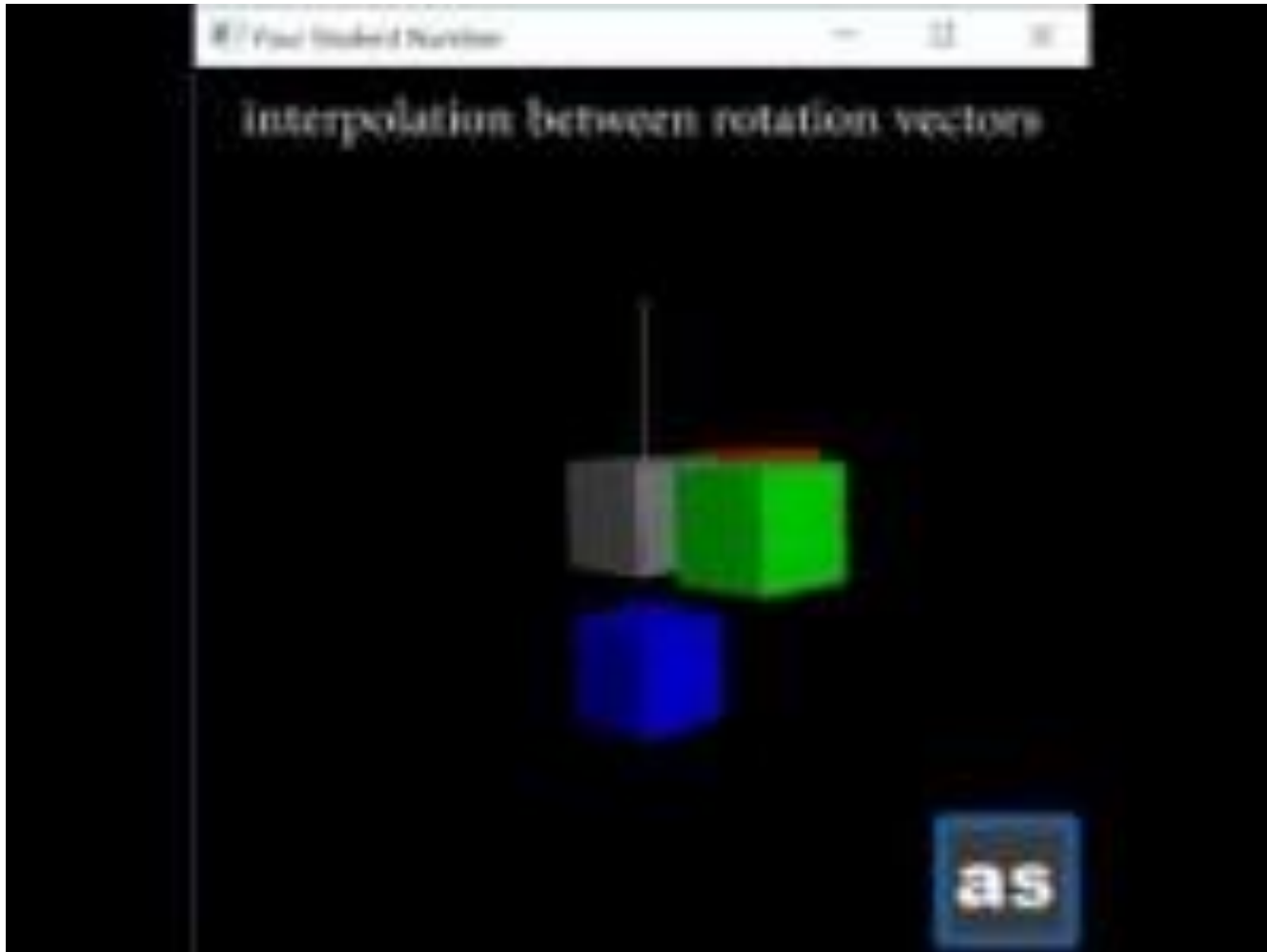
- φ : the angle subtended by the arc ($\cos \varphi = \mathbf{q}_1 \cdot \mathbf{q}_2$)

- No slerp for Euler angles or rotation vector representation!

- They need to be converted to rotation matrix or unit quaternions to slerp.

Interpolation of Rotations

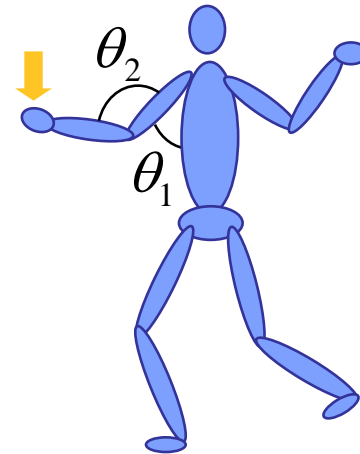
- Start orientation (ZYX Euler angles): $R_z(-90) R_y(90) R_x(0)$
- End orientation (ZYX Euler angles): $R_z(0) R_y(0) R_x(90)$



Limb IK

Inverse Kinematics

- Given the desired position and possibly orientation of the end effector,
- What is the set of joint orientations for other joints to generate the desired end effector configuration?

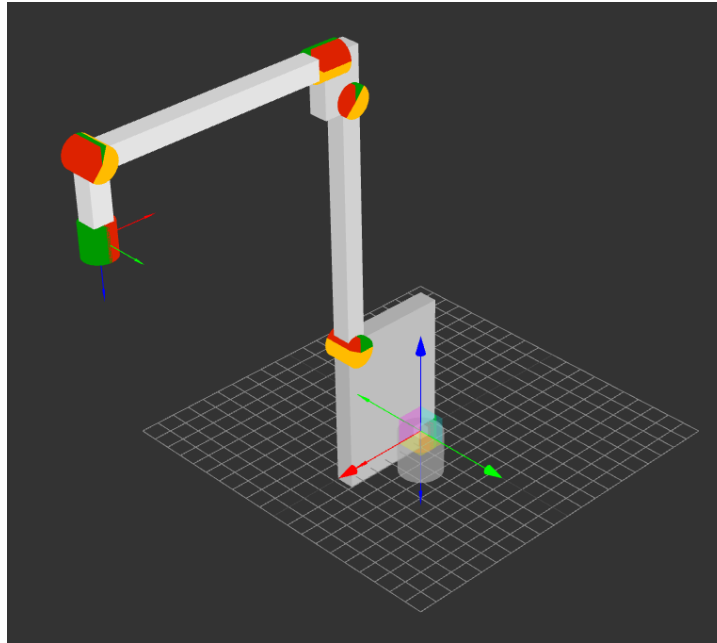


$$\theta_i = F^{-1}(\mathbf{p}, \mathbf{q})$$

Inverse Kinematics

: Given the position & orientation of end-effector, compute joint angles

[Practice] FK / IK Online Demo



<http://robot.glumb.de/>

- Forward kinematics : Open “angles” menu and change values
- Inverse kinematics : Move the end-effector position by mouse dragging

Inverse Kinematics

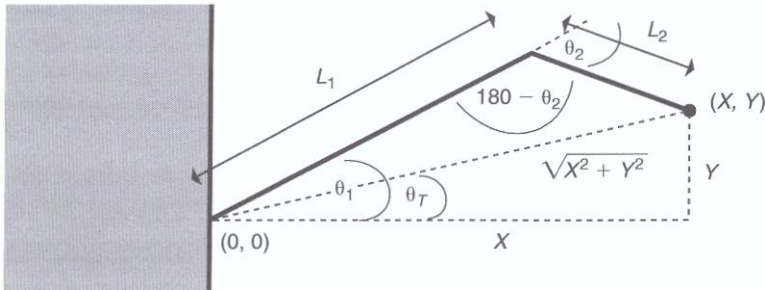
- More challenging than forward kinematics
- An IK problem might have:
 - Multiple solutions
 - Infinitely many solutions
 - No solutions
- IK problems generally do not have analytic solutions.

Inverse Kinematics

- Analytic solver
 - Limb IK (Foot IK)
- Numerical solver
 - Jacobian Transpose IK
 - Pseudo-inverse Jacobian
 - Cyclic coordinate descent (CCD)
 - ...

Simple Analytic Solution

- For sufficiently simple mechanisms (two links in 2D plane), the joint angles can be analytically calculated.
- For example, given L_1 , L_2 , and (X, Y) ,
- What is θ_1 and θ_2 ?



$$\cos(\theta_T) = \frac{X}{\sqrt{X^2 + Y^2}}$$

$$\theta_T = \arccos\left(\frac{X}{\sqrt{X^2 + Y^2}}\right)$$

$$\cos(\theta_1 - \theta_T) = \frac{L_1^2 + X^2 + Y^2 - L_2^2}{2L_1\sqrt{X^2 + Y^2}}$$

(cosine rule)

$$\theta_1 = \arccos\left(\frac{L_1^2 + X^2 + Y^2 - L_2^2}{2L_1\sqrt{X^2 + Y^2}}\right) + \theta_T$$

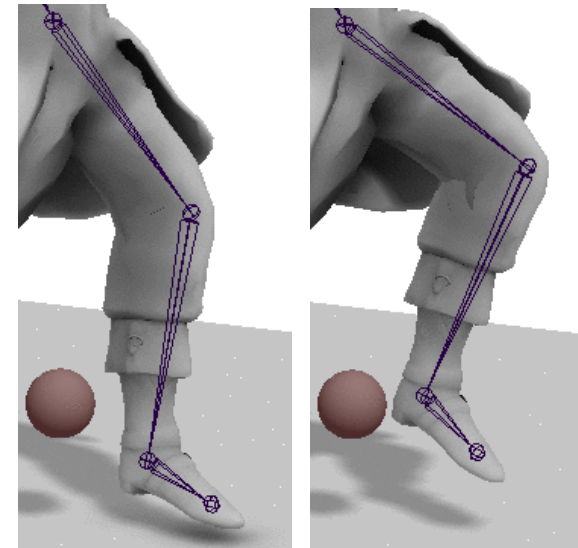
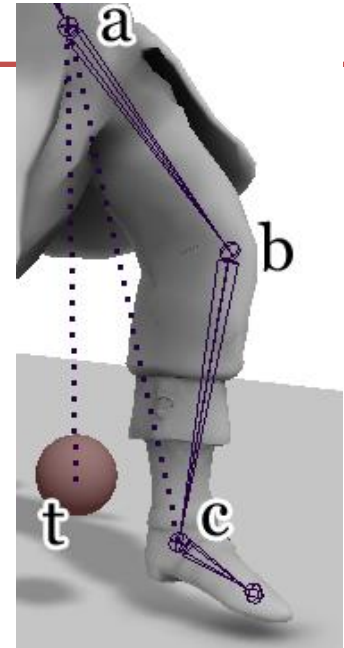
$$\cos(180 - \theta_2) = -\cos(\theta_2) = \frac{L_1^2 + L_2^2 - (X^2 + Y^2)}{2L_1L_2}$$

(cosine rule)

$$\theta_2 = \arccos\left(\frac{L_1^2 + L_2^2 - X^2 - Y^2}{2L_1L_2}\right)$$

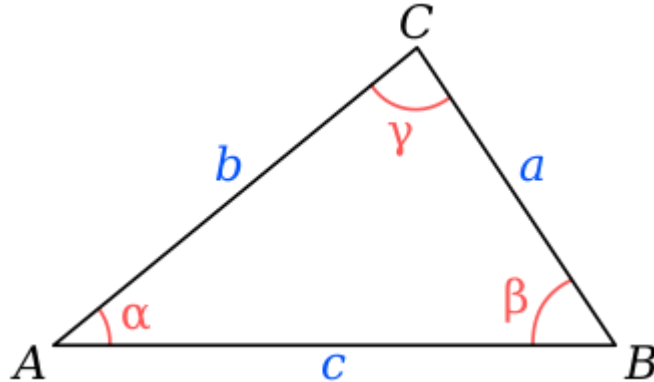
Limb IK

- Actually, this simple analytic solution alone is not very useful.
- Instead, limb IK (or foot IK) is often used to correct leg joints so that the feet can touch the ground.
- 두 단계로 나누어서 생각 (t: target position)
- 1단계: line ac와 line at의 길이가 같아지도록 joint a와 b의 각도를 조절
- 2단계: line ac와 line at가 일치하도록 joint a의 각도를 조절



1단계: line ac와 line at의 길이가 같아지도록 joint a와 b의 각도를 조절

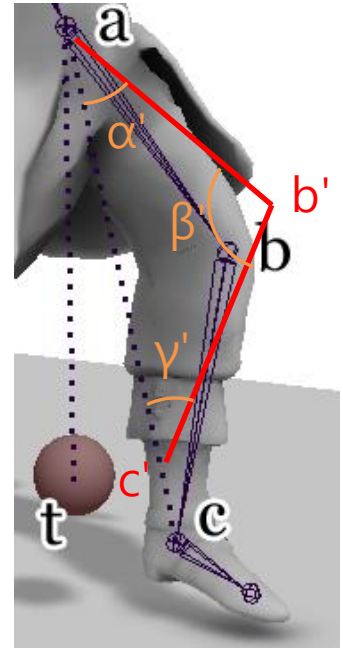
- cosine 법칙 이용



$$c^2 = a^2 + b^2 - 2ab \cos C$$

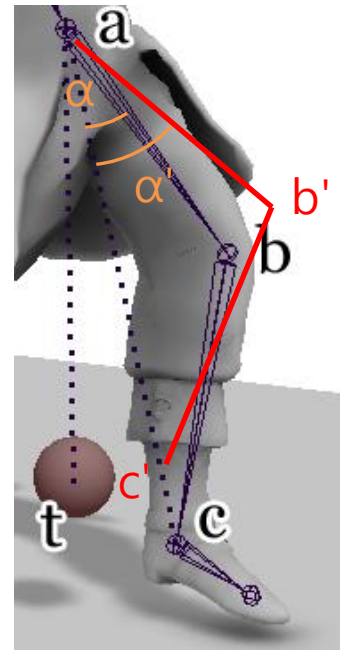
$$\cos C = \frac{a^2 + b^2 - c^2}{2ab}$$

- c'을 ac'길이==at길이 가 되도록 하는 ac 상의 위치, b'을 그때의 관절 b 위치라 하면,
- (모든 변의 길이를 구할 수 있기 때문에)
cosine 법칙을 이용해서 α' , β' , γ' 을 계산 가능



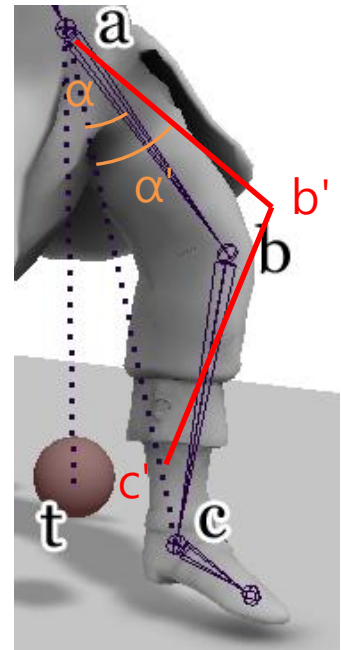
1단계: line ac와 line at의 길이가 같아지도록 joint a와 b의 각도를 조절

- Joint a: ($ab \times ac$ 의 unit vector)인 $axis_{\{g\}}$ 에 대해 $(\alpha - \alpha')$ 만큼 회전하는 rotation $R_{diff}^{\{g\}}$ 을 현재 joint a의 orientation에 더하면 (rotation matrix의 곱) 됨.
 - $R_{diff}^{\{g\}}$: a, b, c의 global position으로 구한 $axis_{\{g\}}$ 에 대한 회전을 뜻하는 rotation matrix는 global frame $\{g\}$ 에서 적용되는 회전을 의미하게 된다.
 - 하지만 human motion에서 joint a의 orientation은 일반적으로 joint a's default frame $\{ad\}$ (parent joint frame에서 link transform만 적용된 후의 frame) 대해서 표현되므로,
 - 모션으로부터 joint a의 orientation을 바로 구하면 $\{ad\}$ 에 대해 표현된 $R_a^{\{ad\}}$ 을 얻게 되며, 새로운 orientation을 설정할 때도 $\{ad\}$ 에 대해 표현된 new $R_a^{\{ad\}}$ 를 설정해야 한다.
 - $R_{diff}^{\{g\}}$ 와 $R_a^{\{ad\}}$ 를 어떻게 더해서 new $R_a^{\{ad\}}$ 를 구할 수 있을까?



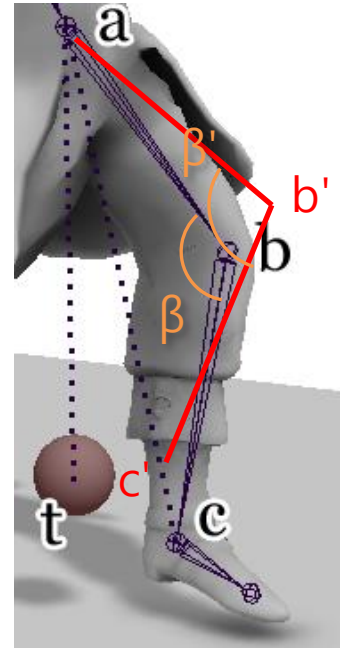
1단계: line ac와 line at의 길이가 같아지도록 joint a와 b의 각도를 조절

- 방법 1: local frame에 대한 rotation을 오른쪽에 곱함
 - forward kinematics로 joint a의 global orientation $R_a^{\{g\}}$ 를 구해서
 - $R_a^{\{g\}-1}$ 를 $axis^{\{g\}}$ 에 곱해서 joint a's frame $\{a\}$ 에 대해 표현된 $axis^{\{a\}}$ 로 만들고 (joint a's default frame $\{ad\}$ 가 아님에 주의),
 - 이것에 대해 $(\alpha - \alpha')$ 만큼 회전하는 $R_{diff}^{\{a\}}$ 을 구해서
 - new $R_a^{\{ad\}} = R_a^{\{ad\}} R_{diff}^{\{a\}}$
 - new $R_a^{\{ad\}}$ 를 joint a의 orientation으로 설정하면 됨.
- 방법 2: global frame에 대한 rotation을 왼쪽에 곱함
 - new $R_a^{\{g\}} = R_{diff}^{\{g\}} R_a^{\{g\}}$
 - new $R_a^{\{g\}}$ 로부터 new $R_a^{\{ad\}}$ 를 계산하여 이를 joint a의 orientation으로 설정하면 됨.



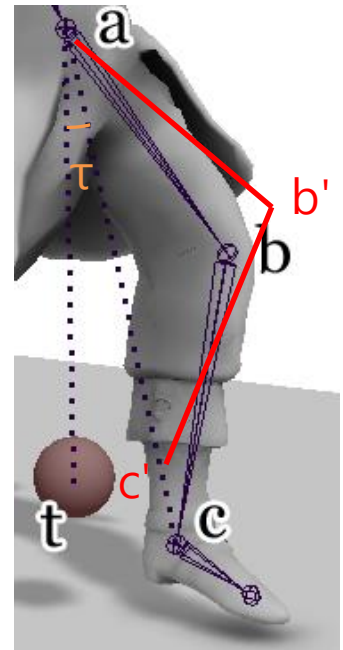
1단계: line ac와 line at의 길이가 같아지도록 joint a와 b의 각도를 조절

- Joint b: 마찬가지로 방법으로 β 가 β' 로 바뀌도록 하는 R_{diff} 를 구해서 $R_b^{\{bd\}}$ 를 업데이트 할 수 있음.



2단계: line ac와 line at가 일치하도록 joint a의 각도를 조절

- Joint a: ($ac' \times at$ 의 unit vector)인 $axis_{\{g\}}$ 에 대해 τ 만큼 회전하는 rotation $R_{diff}^{\{g\}}$ 을 현재 joint a의 orientation 에 앞과 마찬가지로 방식으로 더하면 됨.
 - τ 는 cosine 법칙을 이용해서 구할 수도 있지만, 벡터 내적 (inner product)을 이용해 구할 수도 있다.
- 참고: <http://theorangeduck.com/page/simple-two-joint>
 - (unit quaternion을 이용해 기술되어 있음)



Example - Human figure



<https://youtu.be/bt3hTDwiTi0>

Posture / Motion Difference

Posture Difference

- "Motion": time-varying data
 - internal joint orientation: $\mathbf{R}_1(t), \dots, \mathbf{R}_n(t)$
 - position and orientation of skeletal root: $\mathbf{p}_0(t), \mathbf{R}_0(t)$
- Posture (pose): "motion" at a single frame

Posture Difference

- Let's think about "posture difference" **d**.
- Valid operations (**o**: posture, **d**: difference):

$\mathbf{o}_1 + \mathbf{d} = \mathbf{o}_2$	
$\mathbf{o}_2 - \mathbf{o}_1 = \mathbf{d}$	
$c * \mathbf{d}_1 = \mathbf{d}_2$ (scalar c)	
$\mathbf{d}_1 + \mathbf{d}_2 = \mathbf{d}_3$	

Posture Difference

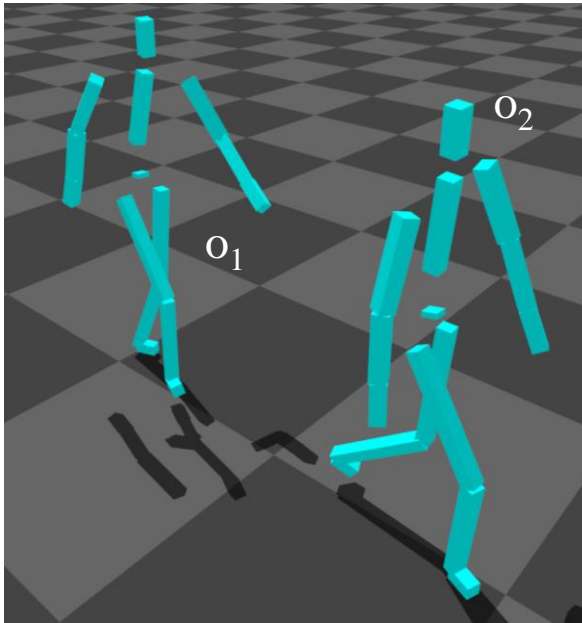
- Let's think about "posture difference" \mathbf{d} .
- Valid operations (\mathbf{o} : posture, \mathbf{d} : difference):

$\mathbf{o}_1 + \mathbf{d} = \mathbf{o}_2$	$\mathbf{p}_0^{\mathbf{o}_1} + \mathbf{p}_0^{\mathbf{d}} = \mathbf{p}_0^{\mathbf{o}_2}$ $\mathbf{R}_i^{\mathbf{o}_1} \mathbf{R}_i^{\mathbf{d}} = \mathbf{R}_i^{\mathbf{o}_2}$
$\mathbf{o}_2 - \mathbf{o}_1 = \mathbf{d}$	$\mathbf{p}_0^{\mathbf{o}_2} - \mathbf{p}_0^{\mathbf{o}_1} = \mathbf{p}_0^{\mathbf{d}}$ $(\mathbf{R}_i^{\mathbf{o}_1})^T \mathbf{R}_i^{\mathbf{o}_2} = \mathbf{R}_i^{\mathbf{d}}$
$c * \mathbf{d}_1 = \mathbf{d}_2$ (scalar c)	$c * \mathbf{p}_0^{\mathbf{d}_1} = \mathbf{p}_0^{\mathbf{d}_2}$ $\exp(c * \log(\mathbf{R}_i^{\mathbf{d}_1})) = \mathbf{R}_i^{\mathbf{d}_2}$
$\mathbf{d}_1 + \mathbf{d}_2 = \mathbf{d}_3$	$\mathbf{p}_0^{\mathbf{d}_1} + \mathbf{p}_0^{\mathbf{d}_2} = \mathbf{p}_0^{\mathbf{d}_3}$ $\mathbf{R}_i^{\mathbf{d}_1} \mathbf{R}_i^{\mathbf{d}_2} = \mathbf{R}_i^{\mathbf{d}_3}$

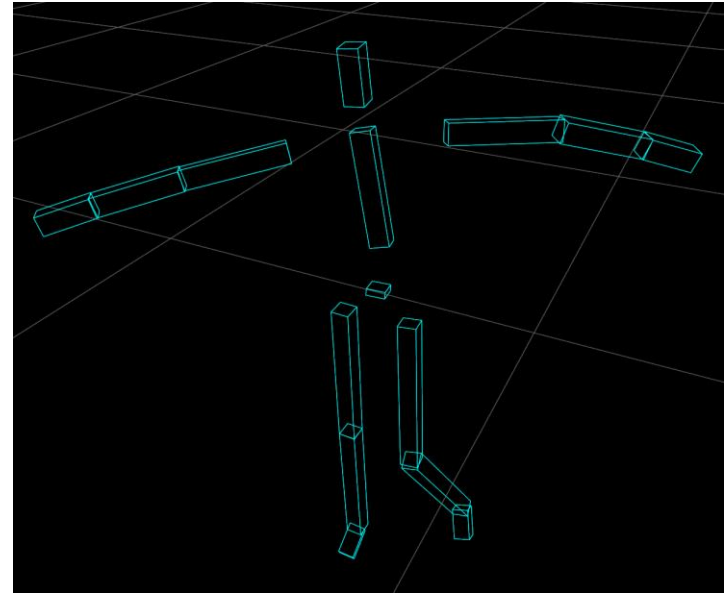
- Invalid operations:
 - $\mathbf{o}_1 + \mathbf{o}_2 = ?$
 - $c * \mathbf{o}_1 = ?$
 - $\mathbf{d} + \mathbf{o}_1 = ?$ (not commutative)

Motion Difference

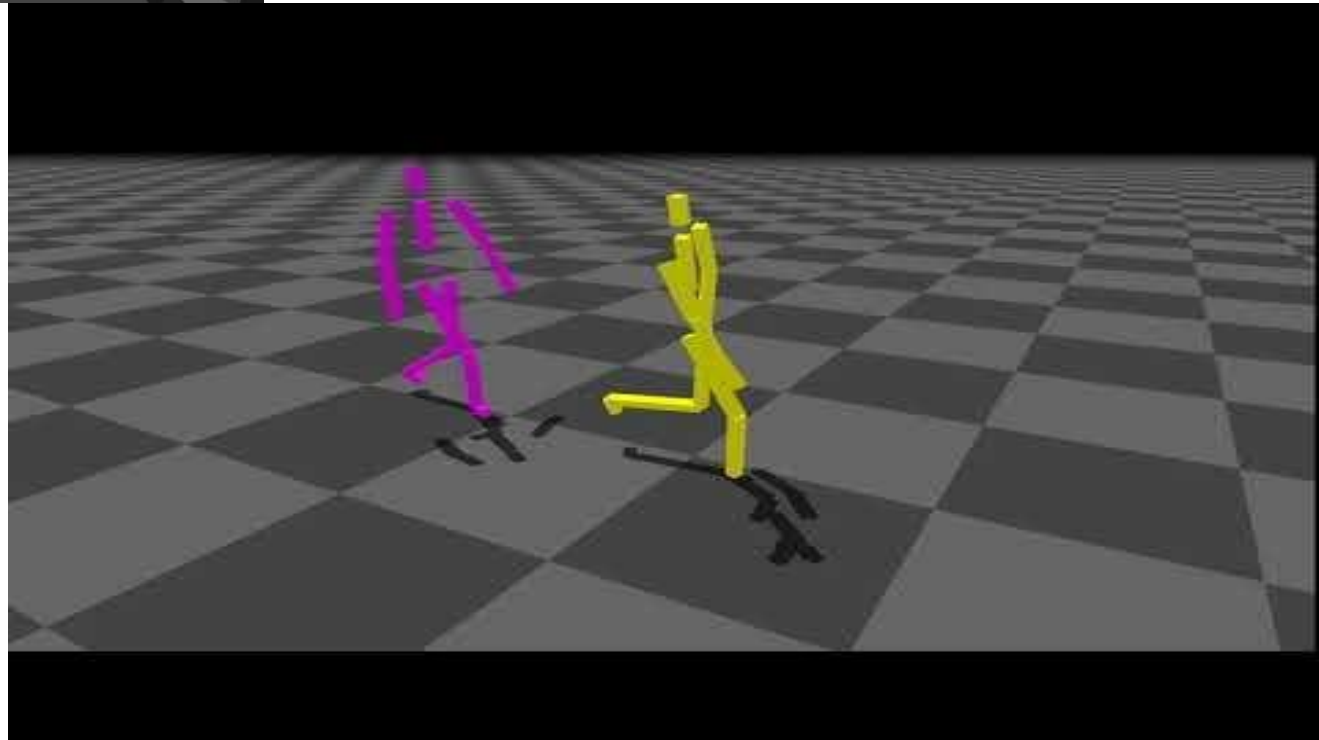
- Just compute posture differences between postures at the same frame for every frame of both motions.
 - for every t ,
 - $\mathbf{m}_1(t) + \mathbf{d}(t) = \mathbf{m}_2(t)$
 - $\mathbf{m}_2(t) - \mathbf{m}_1(t) = \mathbf{d}(t)$
 - $c * \mathbf{d}_1(t) = \mathbf{d}_2(t)$ (scalar c)
 - $\mathbf{d}_1(t) + \mathbf{d}_2(t) = \mathbf{d}_3(t)$
- Motion difference is sometimes called *motion displacement mapping*.



$$\mathbf{d} = \mathbf{o}_2 - \mathbf{o}_1$$



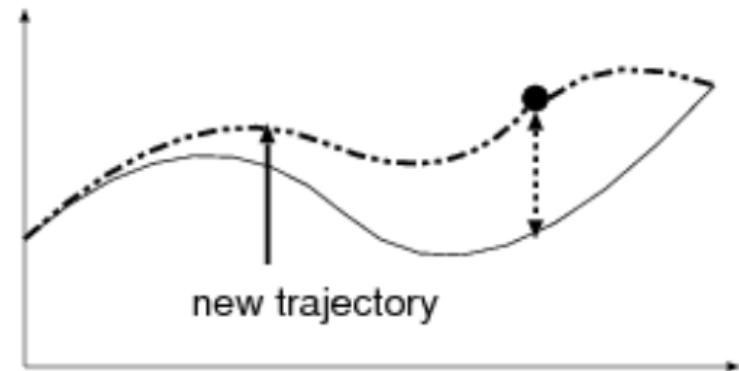
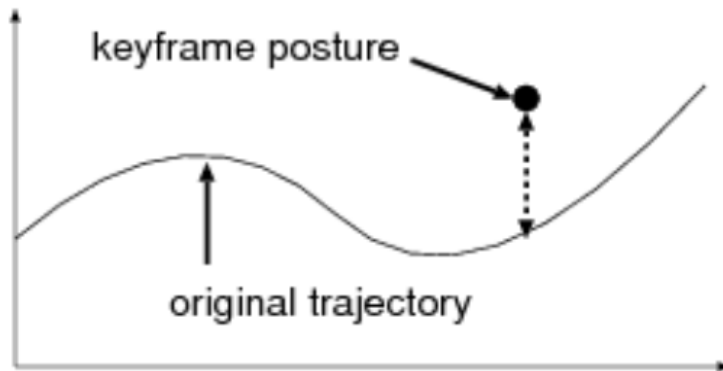
$$\mathbf{m}_1(t) + \mathbf{d} = \mathbf{m}_2(t)$$



Motion Warping

Motion Warping

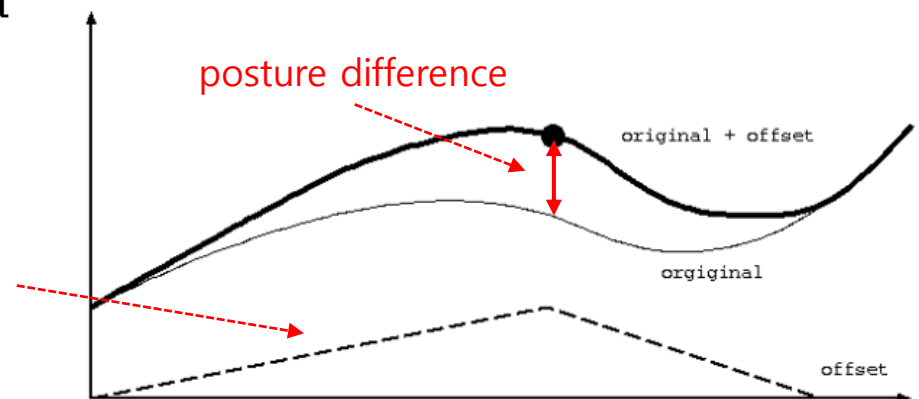
- Adding offset to the data so the constraint is satisfied



Warped Motion = original motion + offset

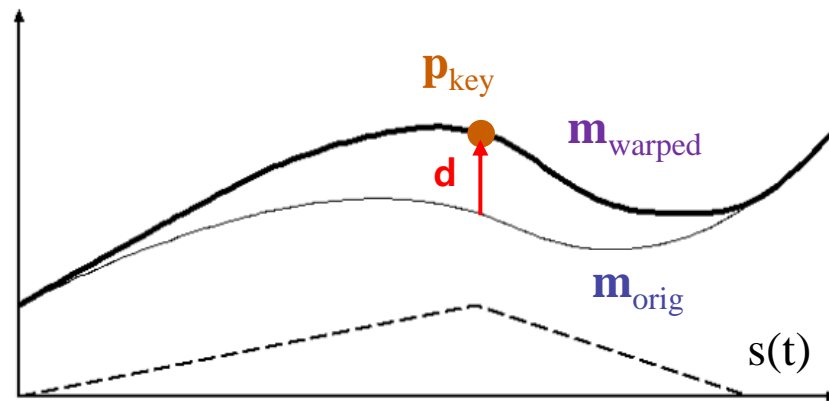
Offset can be a simple 1D motion

scale of posture difference
added to the original motion



Motion Warping

- Set the keyframe posture so that the constraint is satisfied.



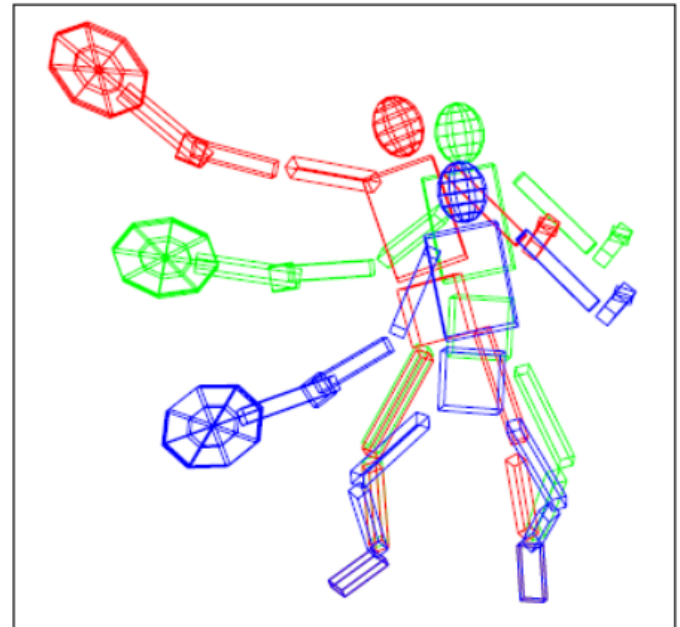
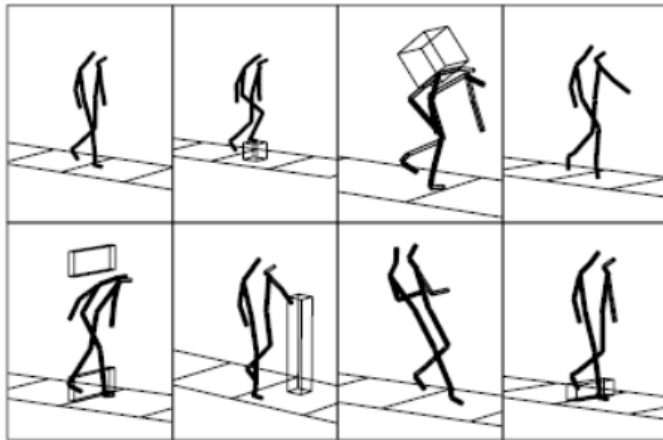
$$\mathbf{d} = \mathbf{p}_{key}(T_{key}) - \mathbf{m}_{orig}(T_{key})$$

(\mathbf{p}_{key} : key frame posture at time T_{key})

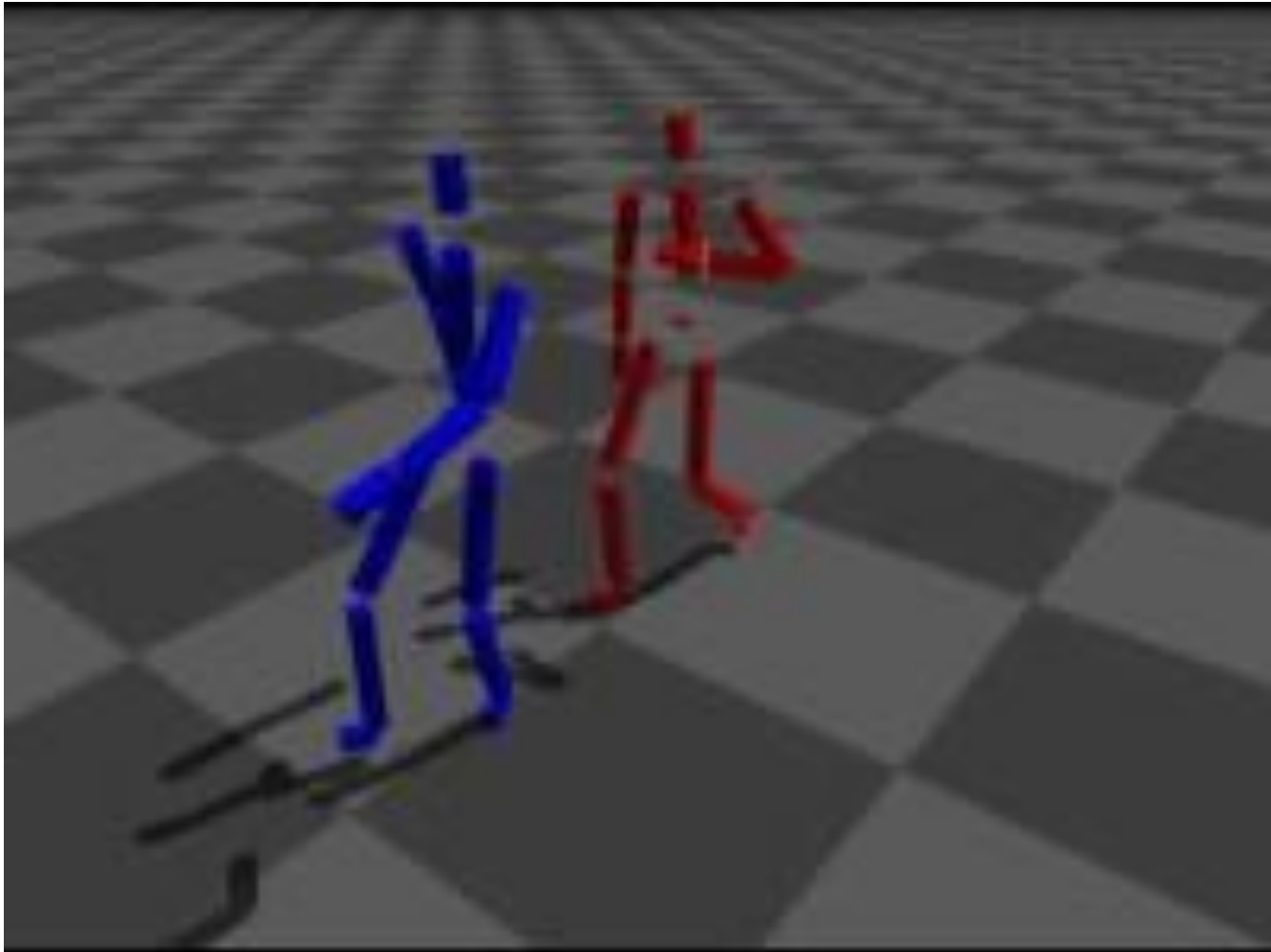
$$\mathbf{m}_{warped}(t) = \mathbf{m}_{orig}(t) + s(t) \cdot \mathbf{d}$$

Motion Warping

- Edit the captured motion a little bit so that it satisfies the requirements
 - Effective for changing the location the hand or the foot passes



Example - Motion Warping with Limb IK



<https://youtu.be/VH4QuV2mFcq>

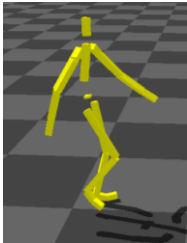
Discussion #1

- Go to <https://www.slido.com/>
- Join #pbl-ys
- Click "Polls"
- 아래의 형식으로 적어서 제출할 것.
 - 이름: 자신의 의견 blah blah ...

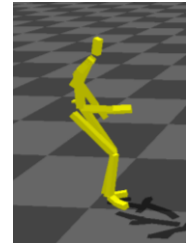
Intro to More Motion Editing Techniques

Interpolation of Postures

pose A



pose B



poses for $0 < t < 1$?



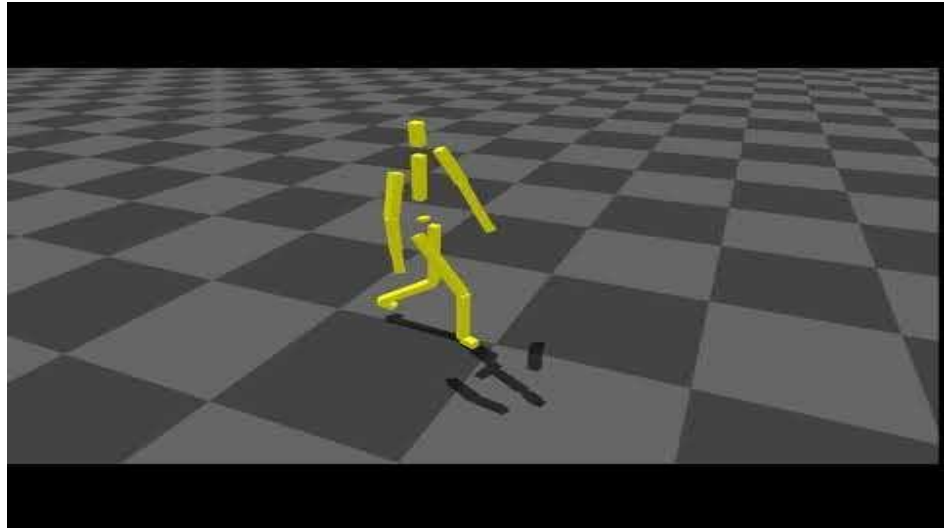
- Simple solution: linear interpolation
- Linear interpolation of two poses by:
 - Linear interpolation of root positions
 - Slerp of root & joint orientations

Time Warping

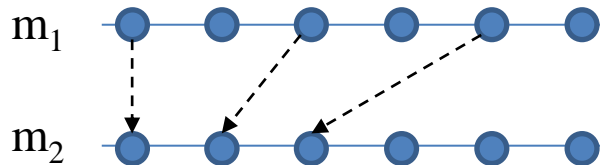
$$\mathbf{m}_2(t) = \mathbf{m}_1(s(t))$$

\mathbf{m}_1 (original motion)

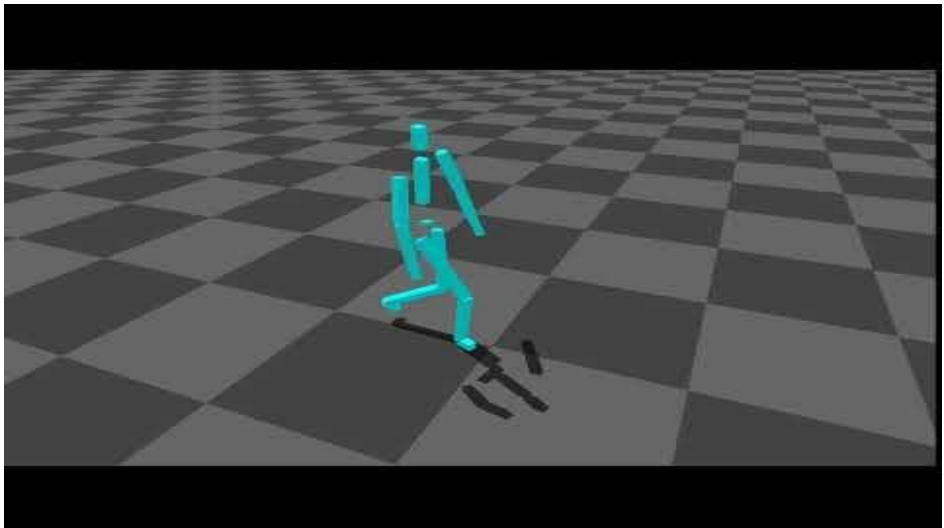
<https://youtu.be/ViYTecPLrho>



$$\mathbf{m}_2(t) = \mathbf{m}_1(2*t)$$



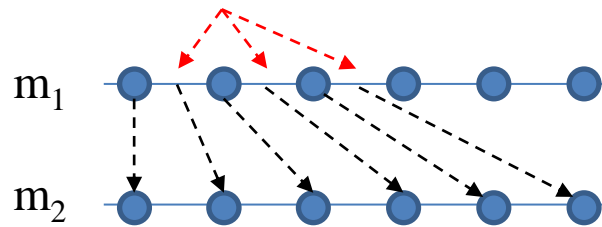
<https://youtu.be/dStMgTRR5iw>



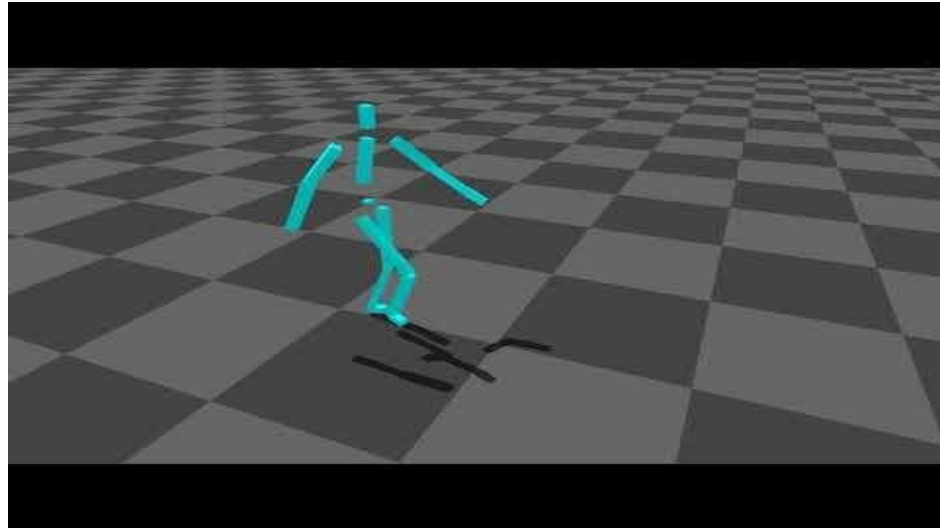
Time Warping

$$\mathbf{m}_2(t) = \mathbf{m}_1(0.5 * t)$$

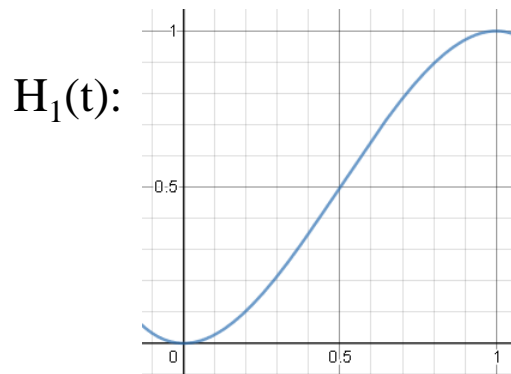
compute interpolated poses



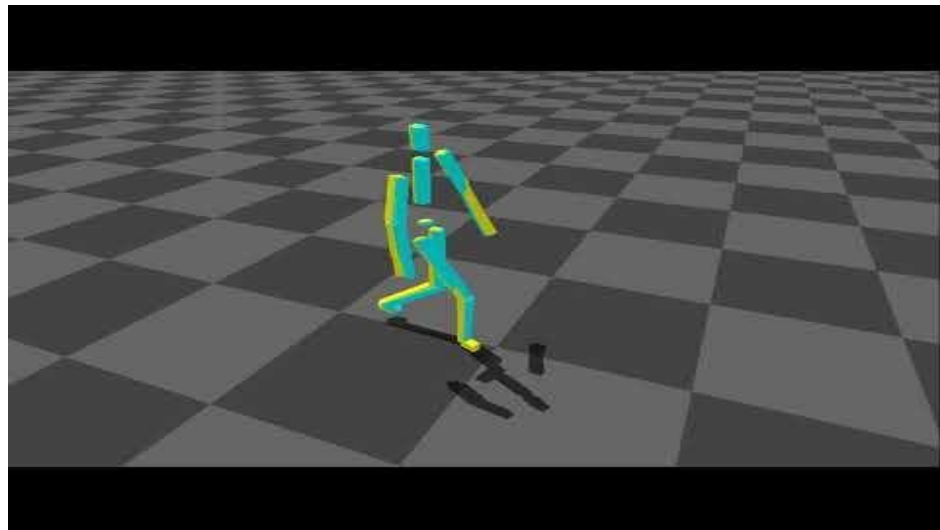
<https://youtu.be/c3ZI7vMqQMM>



$$\mathbf{m}_2(t) = \mathbf{m}_1(H_1(t))$$

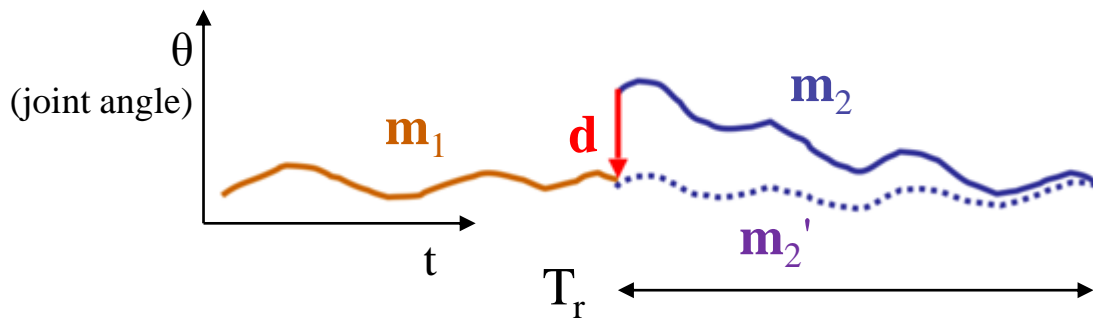


<https://youtu.be/nAJ-rGNBXSc>



Motion Stitching

- Editing the second motion (\mathbf{m}_2) so that it connects seamlessly after the first motion (\mathbf{m}_1).

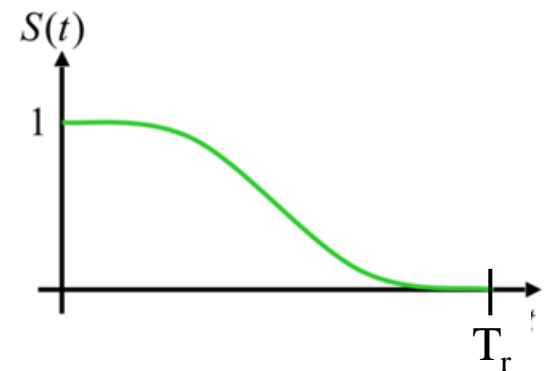


$$\mathbf{d} = \mathbf{m}_1(T_1) - \mathbf{m}_2(0)$$

(T_1 : last frame time of \mathbf{m}_1 , 0: first frame time)

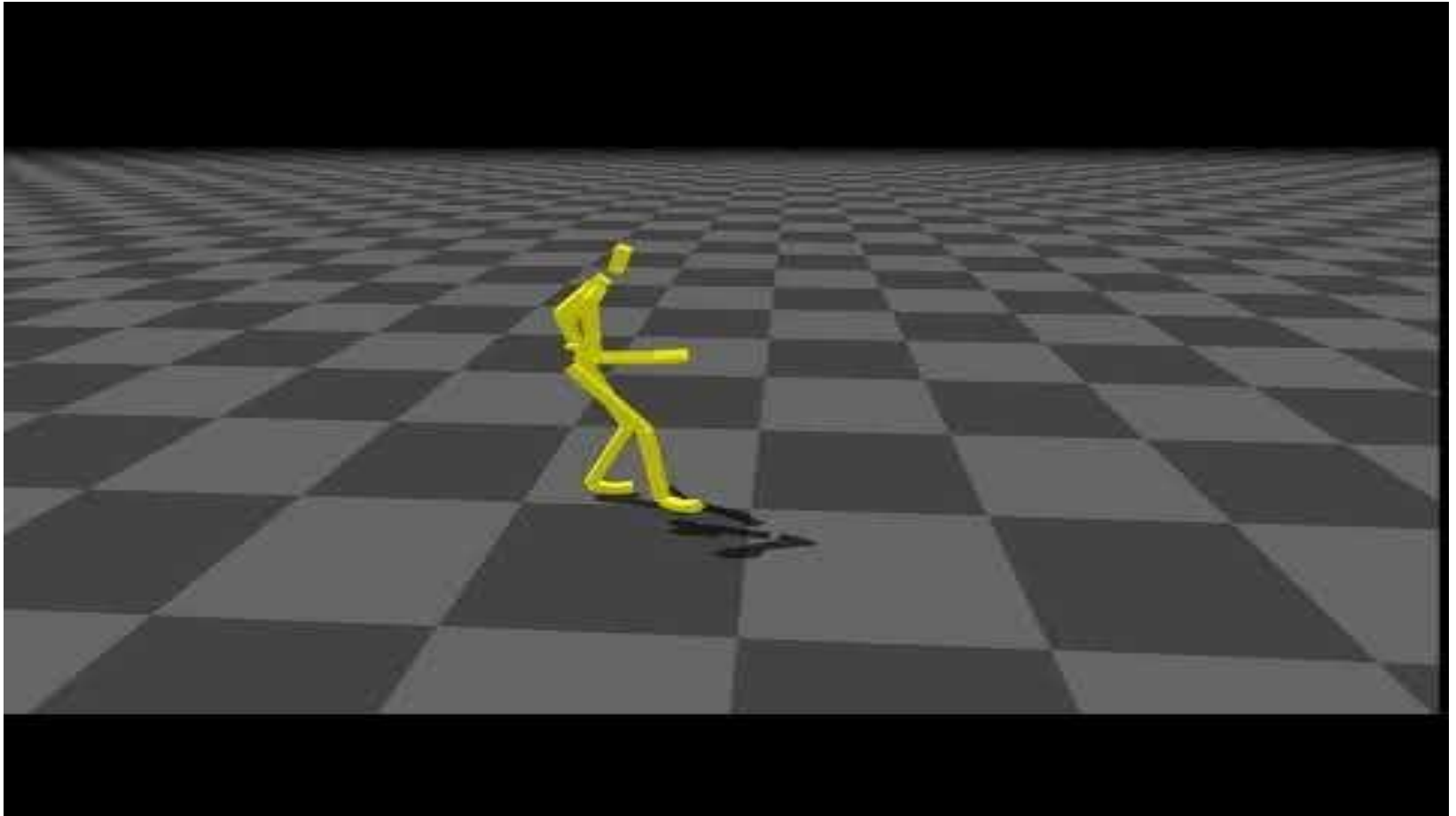
$$\mathbf{m}_2'(t) = \mathbf{m}_2(t) + s(t) \cdot \mathbf{d}$$

$s(t)$: transition function



T_r : transition duration
(transition length)

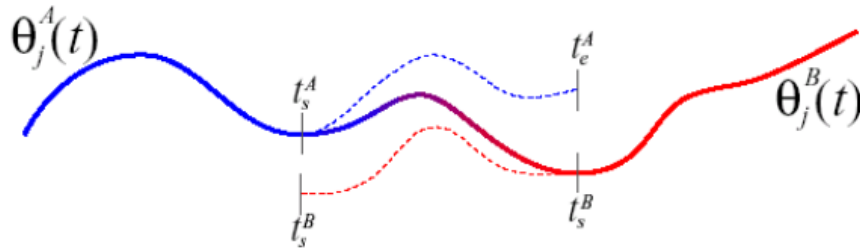
Motion Stitching



https://youtu.be/3UBO_CDFjOI

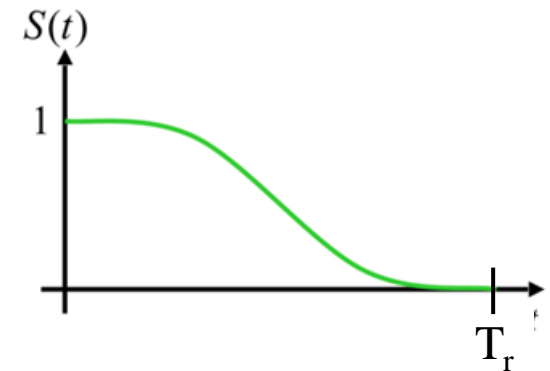
Motion Blending

- Interpolation between motions

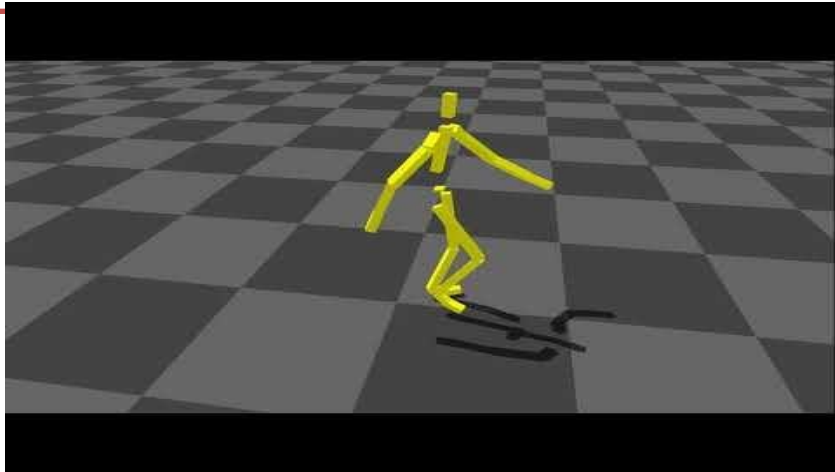


$$\mathbf{m}(t) = (1-s(t)) \cdot \mathbf{m}_1(t) + s(t) \cdot \mathbf{m}_2(t)$$

$s(t)$: transition function

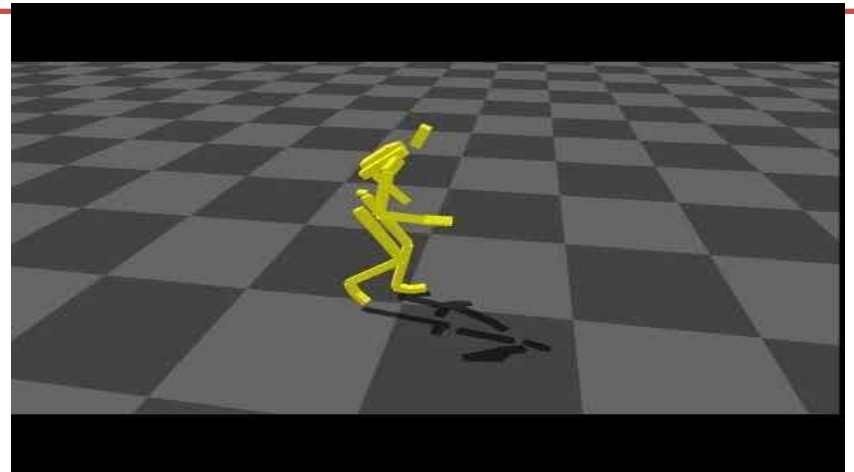


Motion Blending



cycle0

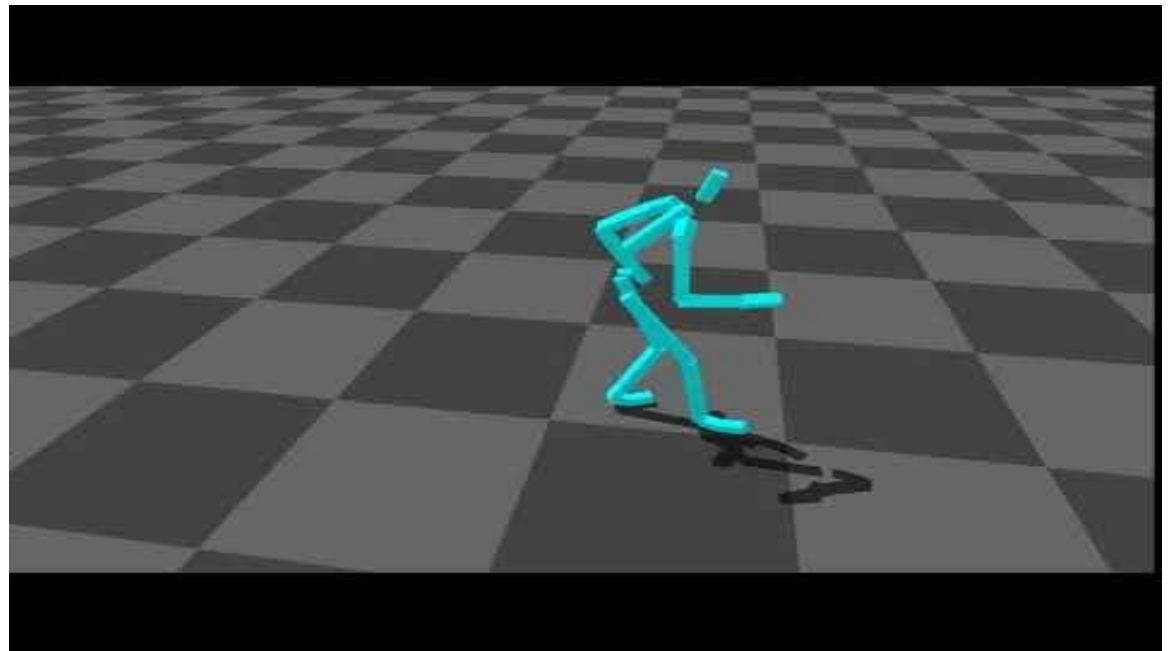
<https://youtu.be/T0xRiNzQ2To>



cycle1

<https://youtu.be/fKCgIgOKI-E>

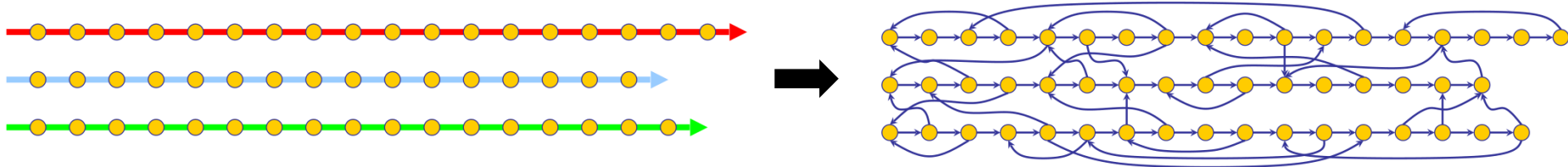
blended motion
(motion0 + blended cycle + motion1)



https://youtu.be/_y7BtHh6Yvw

Intro to Data-Driven Motion Synthesis

Motion Graph [Lee et al. 2002] [Kovar et al. 2002] [Arikan&Forsyth 2002]



- Consideration for creating transitions:
 - Contact states, pose similarity, avoiding dead-ends
- Once a motion graph is constructed, you can find a series of transitions passing through...
 - Specified poses
 - Specified locations / continuous path
 - Specified poses and times
 - ...
- by using graph search algorithms (such as Dijkstra, A*, ...) or dynamic programming.
- Motion editing techniques facilitates smooth transitions.

Motion Graph

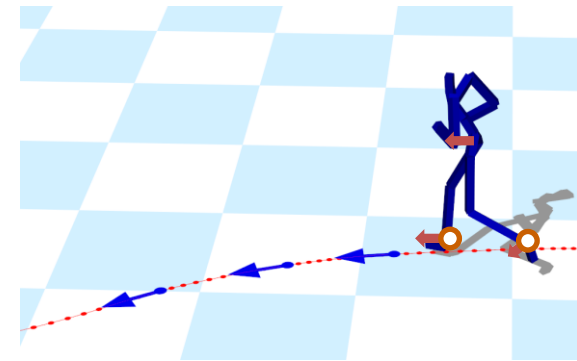


[Lee et al. 2002]

<http://graphics.cs.cmu.edu/projects/Avatar/>

Motion Matching [Büttner and Clavet 2015]

- *Motion DB* stores the pose for each frame of motion data.
- *Feature DB* stores extracted "features" for each frame of motion data.
 - Feature: (current state, future information)



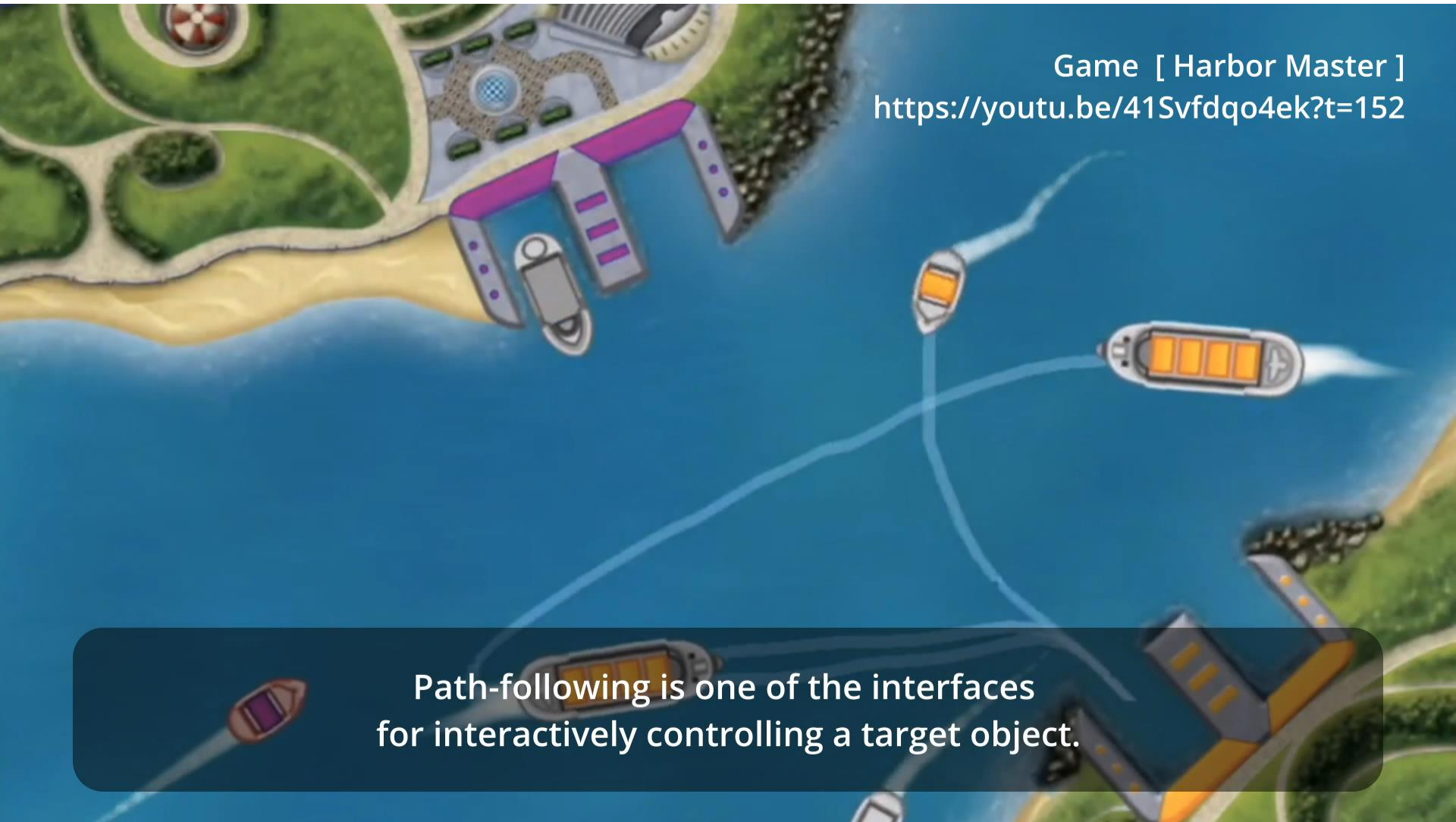
- Matching (performed periodically):
 - Query q : (current character state, future information created by user input)
 - Search for the frame j^* that corresponds to the feature closest to the query q .
- , then motions are played sequentially from the j^* frame in the *motion DB*.
- Motion editing techniques facilitates smooth transitions.

Motion Matching



<https://youtu.be/qBbCjuJpE9o>

(Submitted Paper) Interactive Character Path-Following using Motion Matching

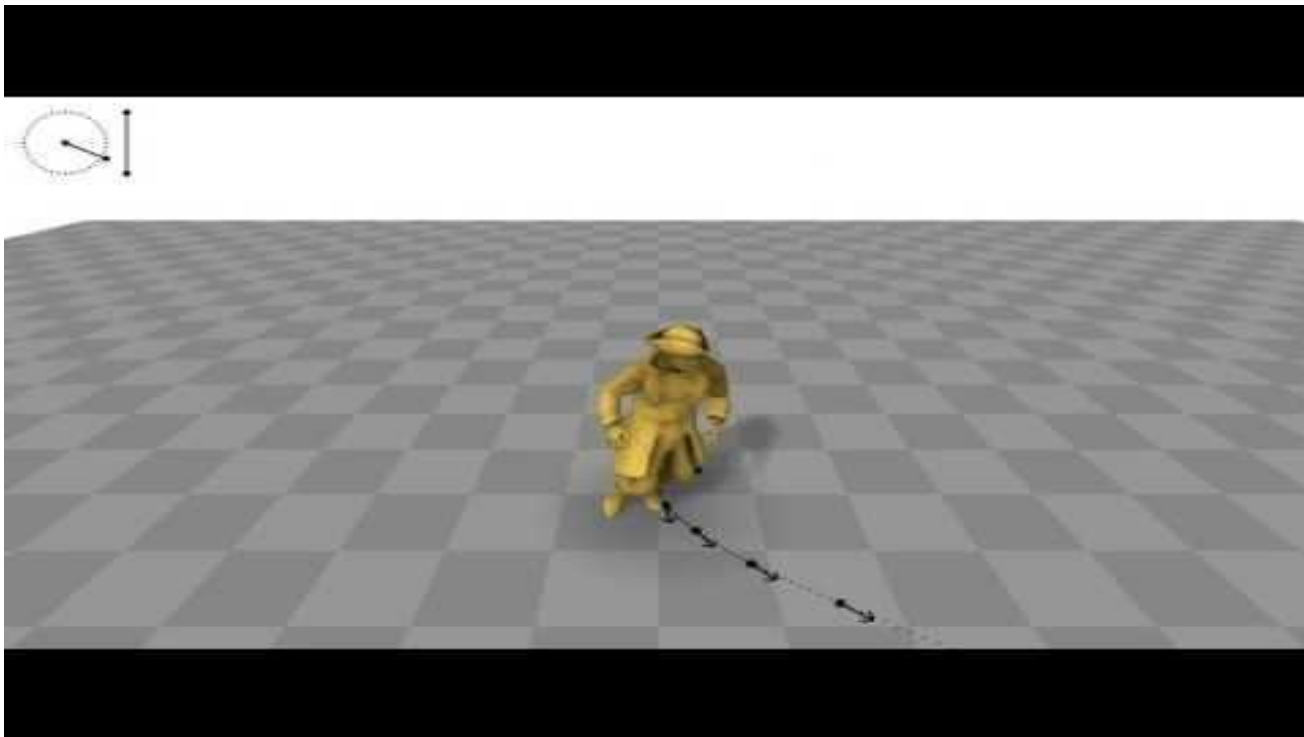


Intro to Data-Driven Motion Synthesis

Deep Motion Synthesis - Problems

- Character control
 - [Holden 2017], [Lee 2018], [Zhang 2018], [Starke 2019], [Henter 2020], [Ling 2020], [Starke 2020], [Starke 2021], [Lee 2021], [Cho 2021]

<https://youtu.be/Ul0Gilv5wvY>



[Holden 2017]

Problems

- Motion prediction
 - [Holden 2015], [Holden 2016], [Harvey 2020]

<https://youtu.be/fTV7sXqO6ig>



[Harvey 2020]

Problems

- Style transfer
 - [Holden 2016], [Aberman 2020b]

<https://youtu.be/m04zuBSdGrc>



[Aberman 2020b]

Problems

- Sound to motion
 - [Yoon 2020], [Valle-Pérez 2021]

<https://youtu.be/uBnCePehA-Y>



[Valle-Pérez 2021]

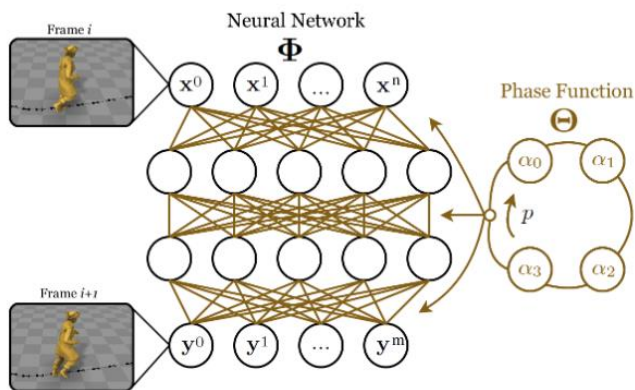
Learning Approaches

- Learns a model generating new motions
 - Discriminative models with FFNN, CNN, or RNN structures
 - Generative models with GAN, VAE, or Flow
 - Manifold learning with Autoencoders
 - Reinforcement learning to learn a policy
 - ...

Supervised Learning in Deep Motion Synthesis

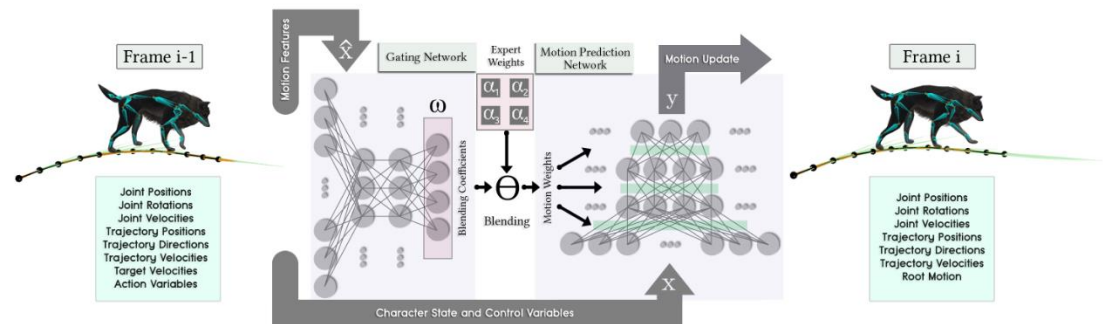
- *Feed-forward neural networks* are widely used to output the next pose.
 - Training data: motion capture data set
 - Input: pose at frame i , control input, ...
 - Output: pose at frame $i+1$, ...

<https://youtu.be/UI0Gily5wvY>



[Holden 2017]

<https://youtu.be/uFJvRYtjQ4c>

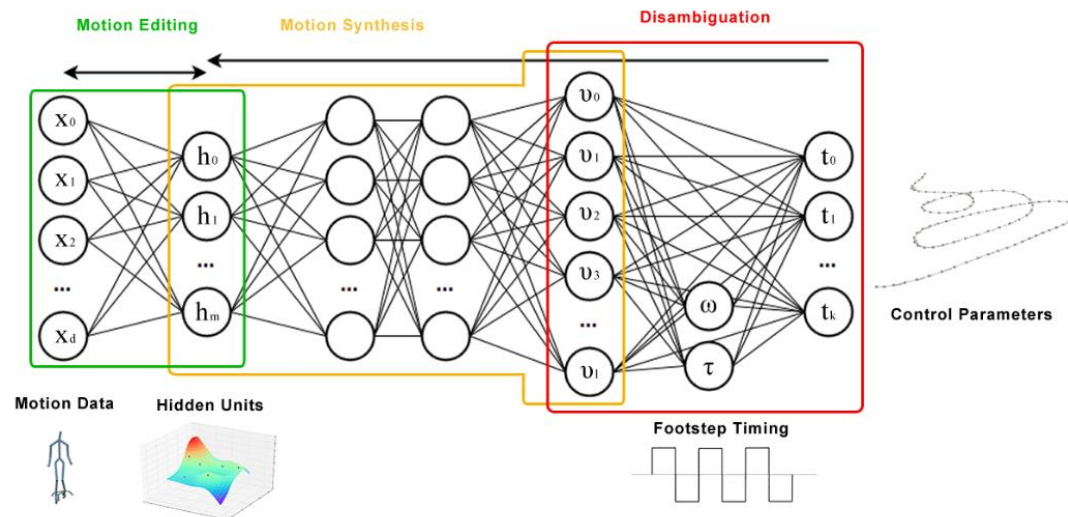


[Zhang 2018]

Supervised Learning in Deep Motion Synthesis

- *Convolutional neural networks* are used to output the new motion.
 - Input: some input sequence
 - Output: output motion (pose sequence)

<https://youtu.be/urf-AAIwNYk>

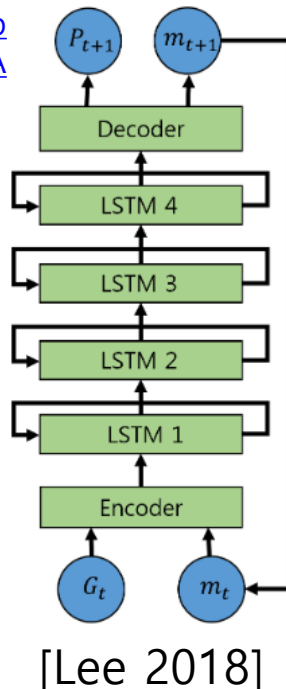


[Holden 2016]

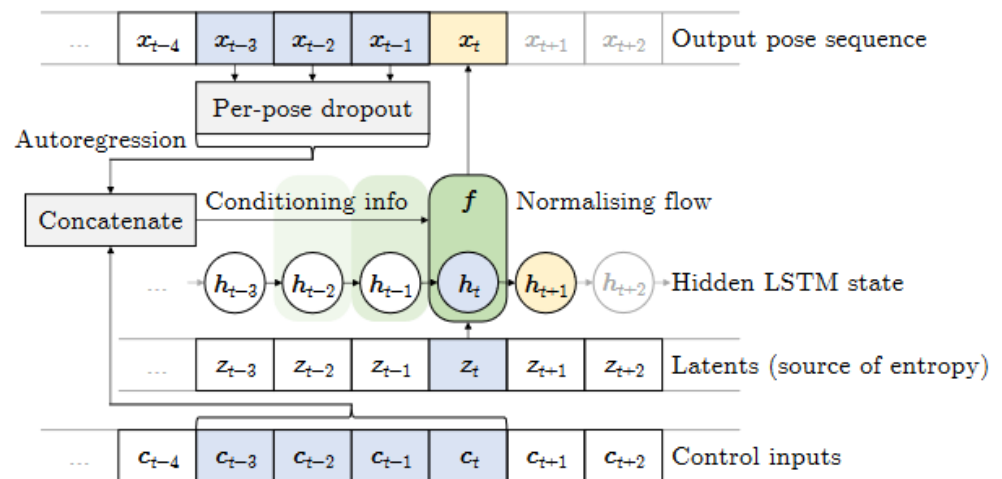
Supervised Learning in Deep Motion Synthesis

- *Recurrent neural networks* are used to output the next pose.
 - Input: pose at frame i , *hidden state* (representing past information), ...
 - Output: pose at frame $i+1$, ...

<https://youtu.be/pe-YTvavbtA>



<https://youtu.be/pe-YTvavbtA>



[Henter 2020]

Unsupervised Learning in Deep Motion Synthesis

- AutoEncoder

<https://youtu.be/dLopOB6D9co>

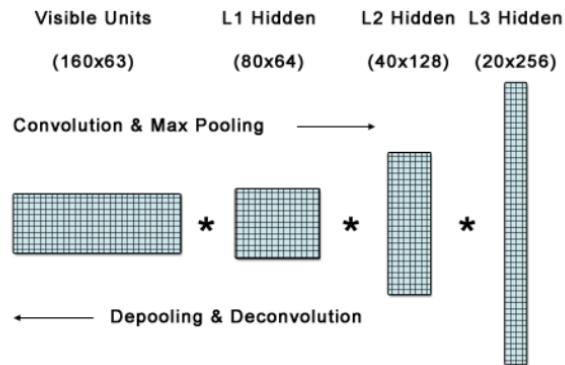
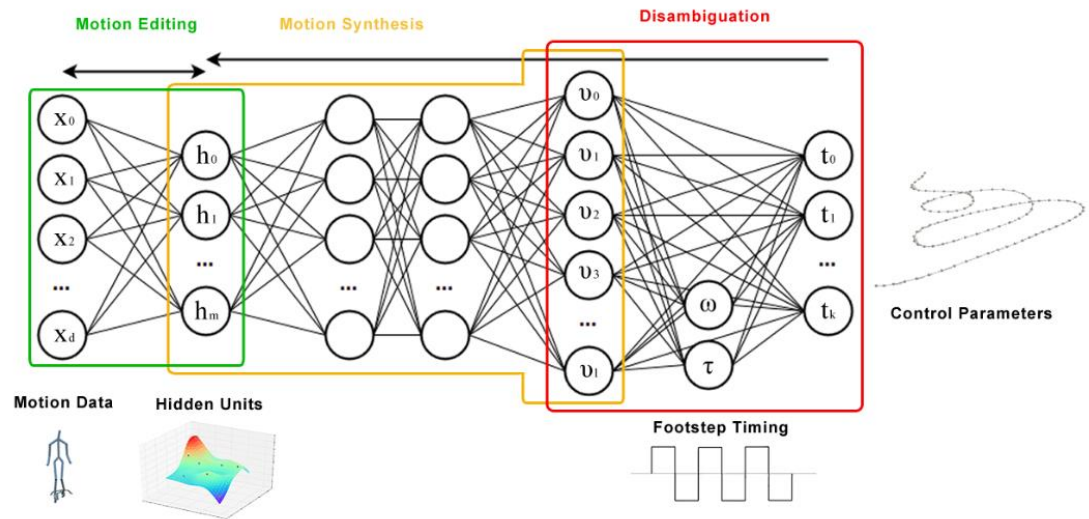


Figure 3: Units of the Convolutional Autoencoder. The input to layer 1 is a window of 160 frames of 63 degrees of freedom. After the first convolution and max pooling this becomes a window of 80 with 64 degrees of freedom. After layer 2 it becomes 40 by 128, and after layer 3 it becomes 20 by 256.

[Holden 2015]

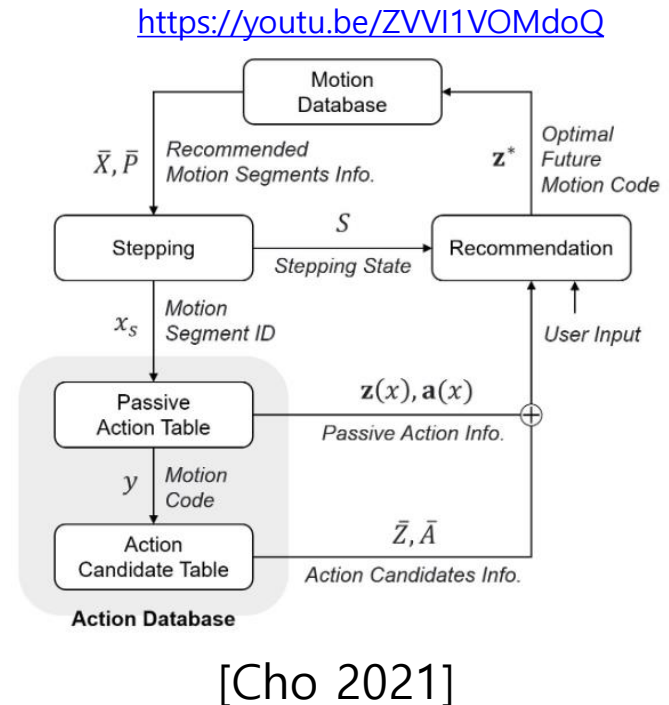
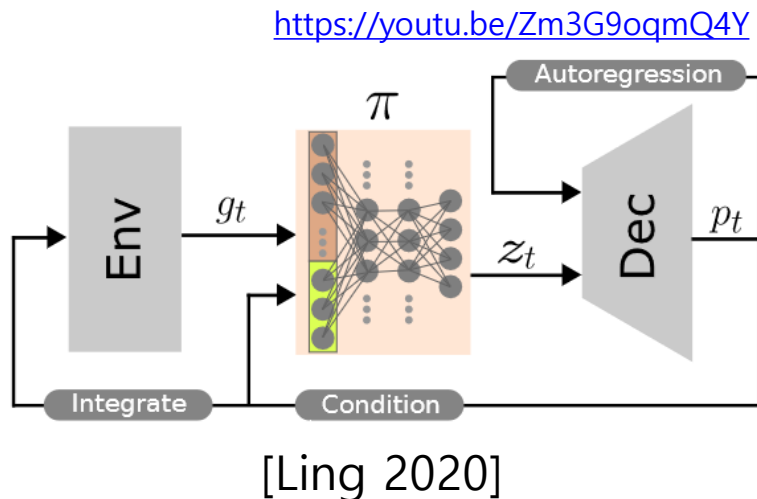
<https://youtu.be/urf-AAIwNYk>



[Holden 2016]

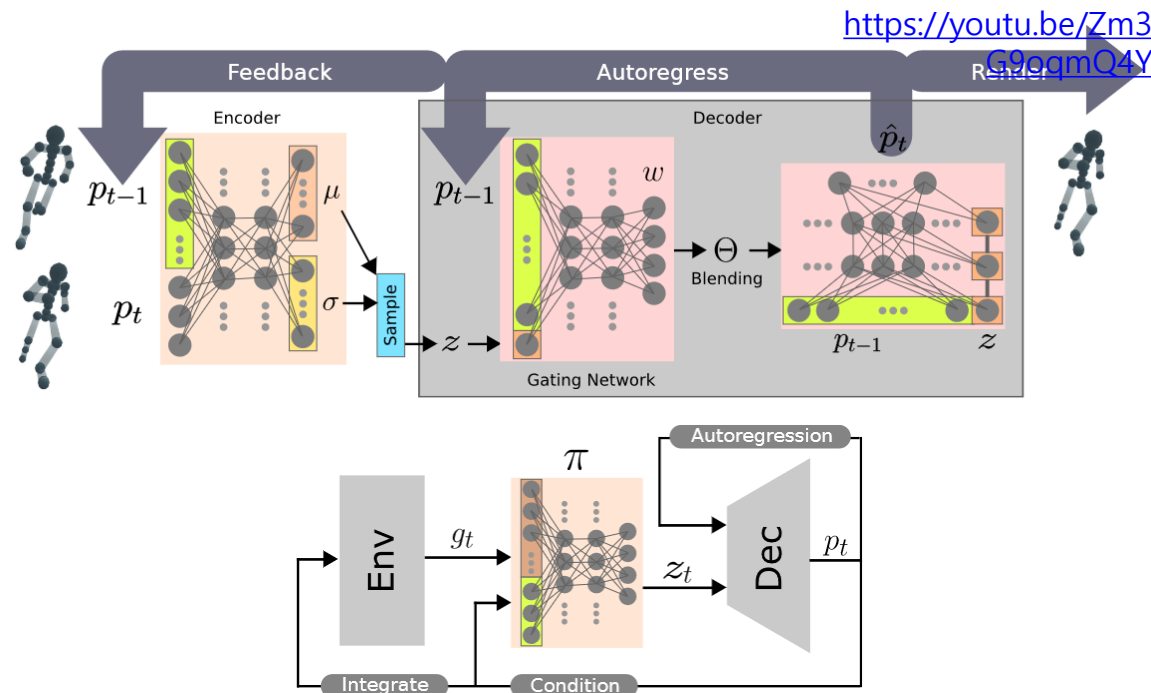
Reinforcement Learning in Deep Motion Synthesis

- Deep reinforcement learning (DRL) has been mainly used for physically-simulated character control.
- But recently it has been gradually applied to motion synthesis tasks.



Generative Model in Deep Motion Synthesis

- Generative models are used to generate different motions for the same control signal
 - because human motion is not completely deterministic



[Ling 2020]

Next Time

- Next week:
 - Preparing Project 3 (No class)
- The week after next:
 - Presentation: Project 3