

Kinematic and Dynamic Adaptation of Human Motion for Imitation

Katsu Yamane

Disney Research, Pittsburgh

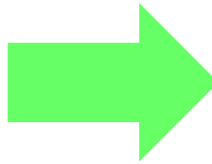


Carnegie Mellon University



Programming Robots through Imitation

- needs to adapt observed human motions to
 - ◆ robot kinematics / dynamics
 - different limb length, mass, actuation, ...
 - ◆ new environment / constraints
 - collision avoidance, task, ...



- different kinematics / dynamics
 - ◆ may be physically infeasible
 - ◆ may not be able to accomplish the task
- what can be modified / what has to be preserved
 - ◆ joint trajectory, endeffector trajectory, contact force, ...
 - ◆ usually task-dependent
- human adaptation
 - ◆ acquired by practice
 - ◆ once acquired, easily adapted to various scenes
 - ◆ sophisticated motor control? memory?

- common problem in robotics and graphics
 - ◆ usually formulated as an optimization problem
- synthesis
 - ◆ mathematical optimization [Gleicher et al. 1997] [Ude et al. 2004] [Liu et al. 2005]
 - ◆ learning [Kuniyoshi et al. 1994] [Atkeson, Schaal 1997] [Bentivegna et al. 2004]
 - ◆ task sequence [Nakaoka et al. 2007]
- analysis
 - ◆ motion graphs [Kovar et al. 2002] [Lee et al. 2002]
 - ◆ motor control [Mataric 2002]

1. adaptation techniques

	kinematics	dynamics
1) dynamics filter [Yamane, Nakamura 2003]	slightly different	
2) synthesizing manipulation tasks [Yamane, Kuffner, Hodgins 2004]	very different environments and characters	(quasi-static)
3) motorized marionette [Yamane, Hodings, Brown 2003]	limited mobility	different actuation mechanism

2. understanding human adaptation

detailed neuro-musculoskeletal human model

[Yamane et al. 2005] [Murai et al. 2008]

Adaptation Techniques

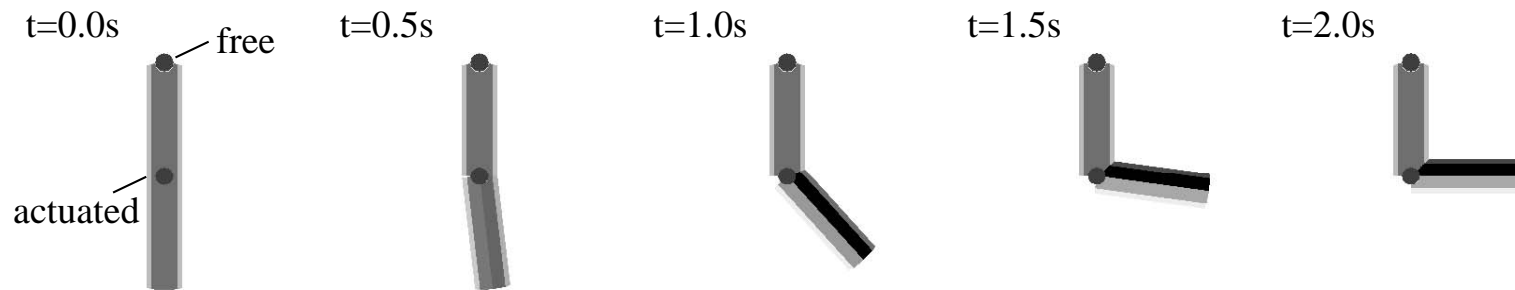
1. Dynamics Filter

[Yamane and Nakamura 2003]

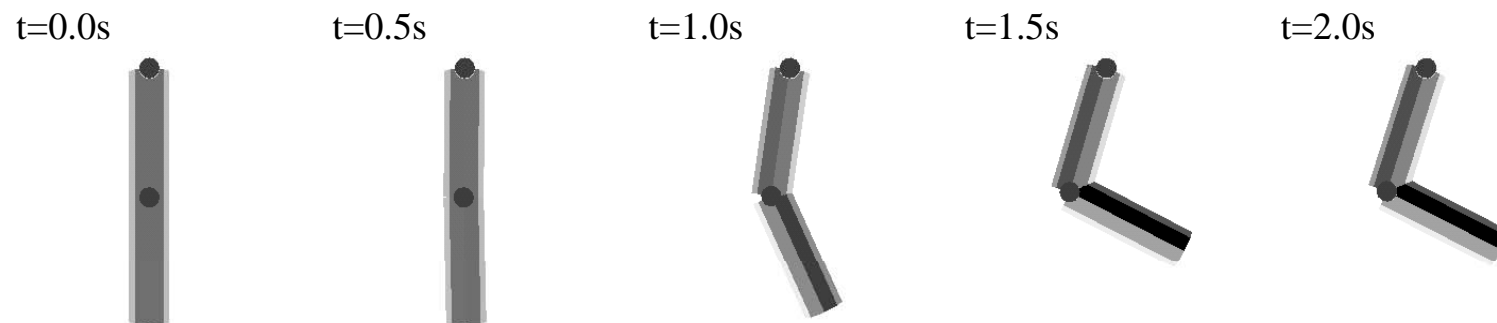
Dynamics Filter

- convert a physically infeasible motion to a feasible one
- simple example

original



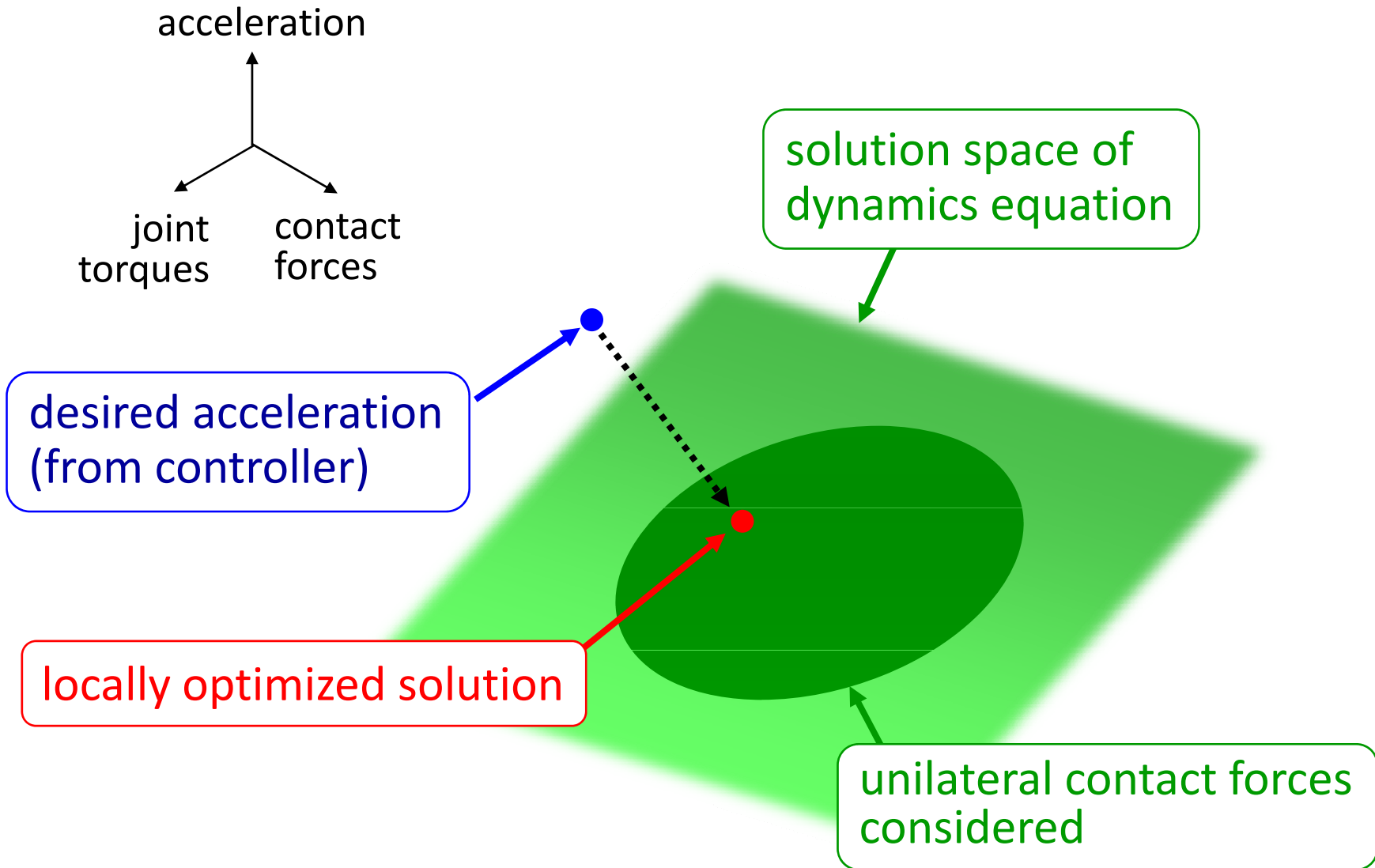
converted



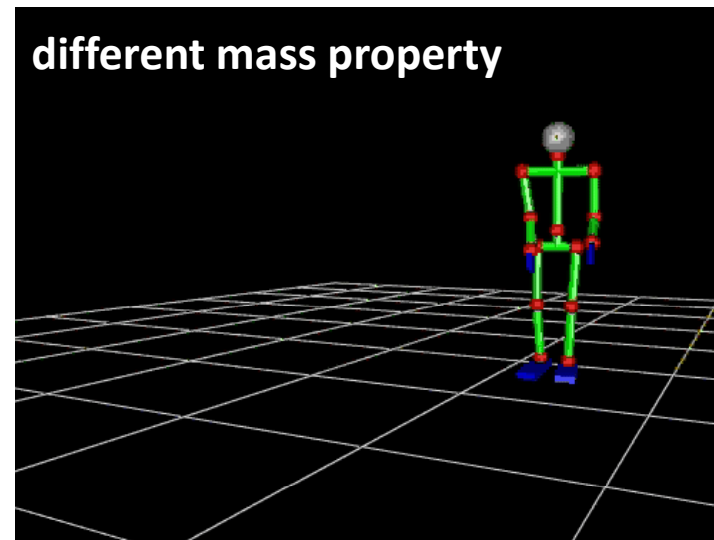
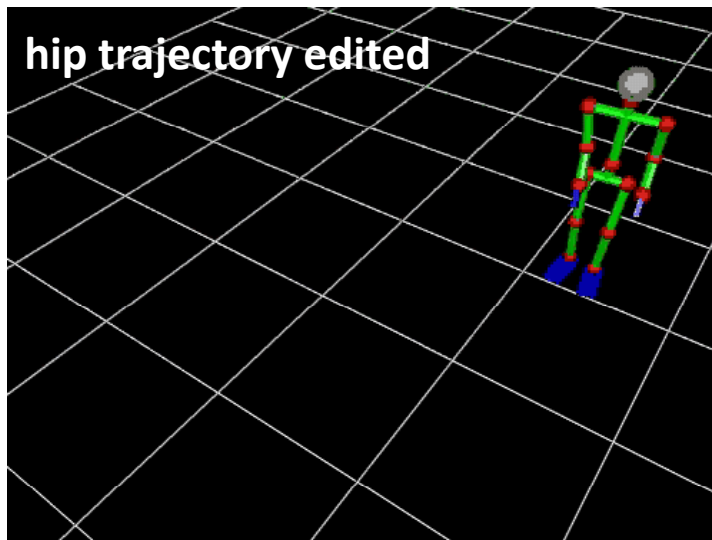
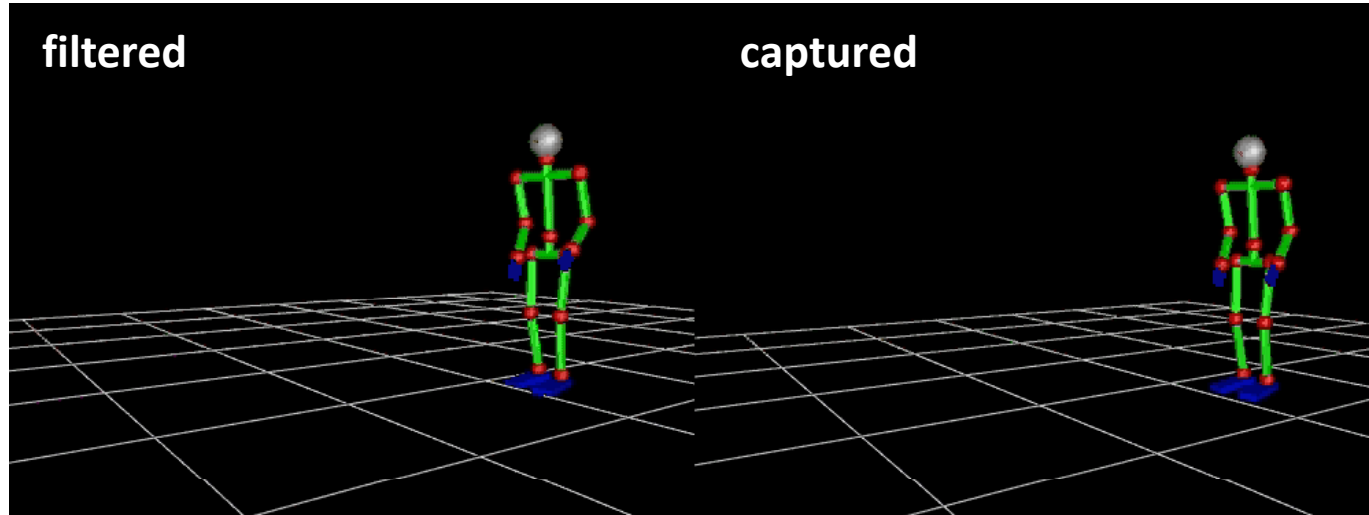
Dynamics Filter

- convert a physically infeasible motion to a feasible one
 - ◆ reasons for infeasibility
 - measurement error
 - different kinematic/dynamic parameters
 - manual editing
- technique
 - ◆ obtain desired accelerations by a feedback controller in joint/Cartesian spaces
 - ◆ project the accelerations onto feasible space by local (online) optimization
 - ◆ limitation: can only cope with small differences

Dynamics Filter: Concept



Dynamics Filter: Examples



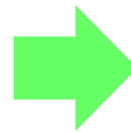
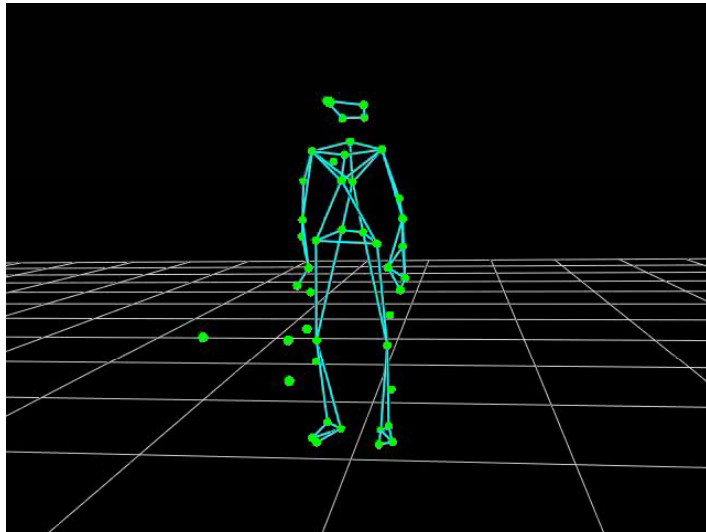
Adaptation Techniques

2. Synthesizing Manipulation Tasks

[Yamane, Kuffner, Hodgins 2004]

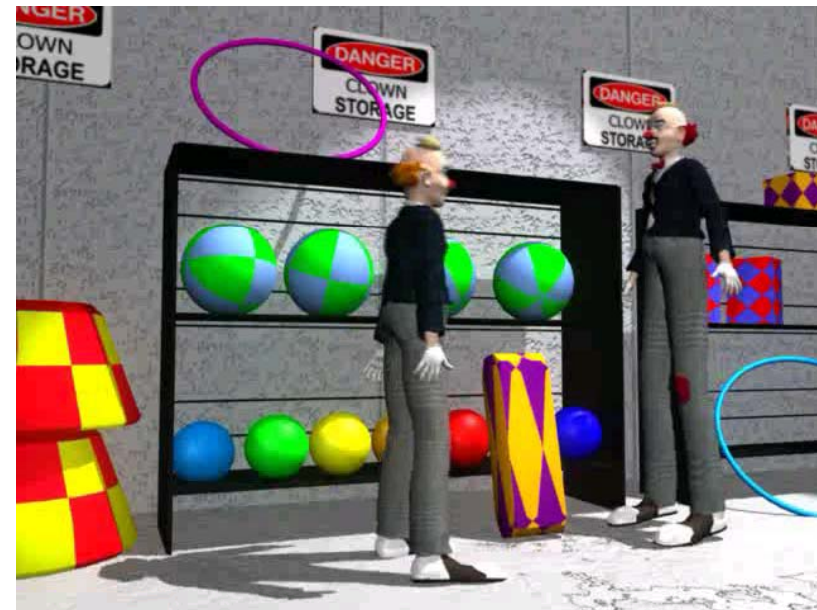
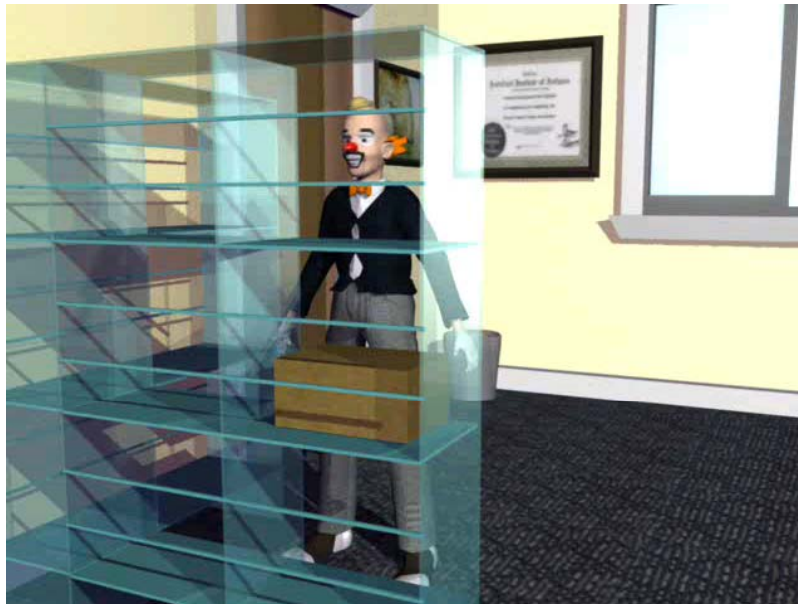
Synthesizing Manipulation Tasks

- adapt example motions to
 - ◆ new objects
 - ◆ new environment
 - ◆ new task (start/goal positions)



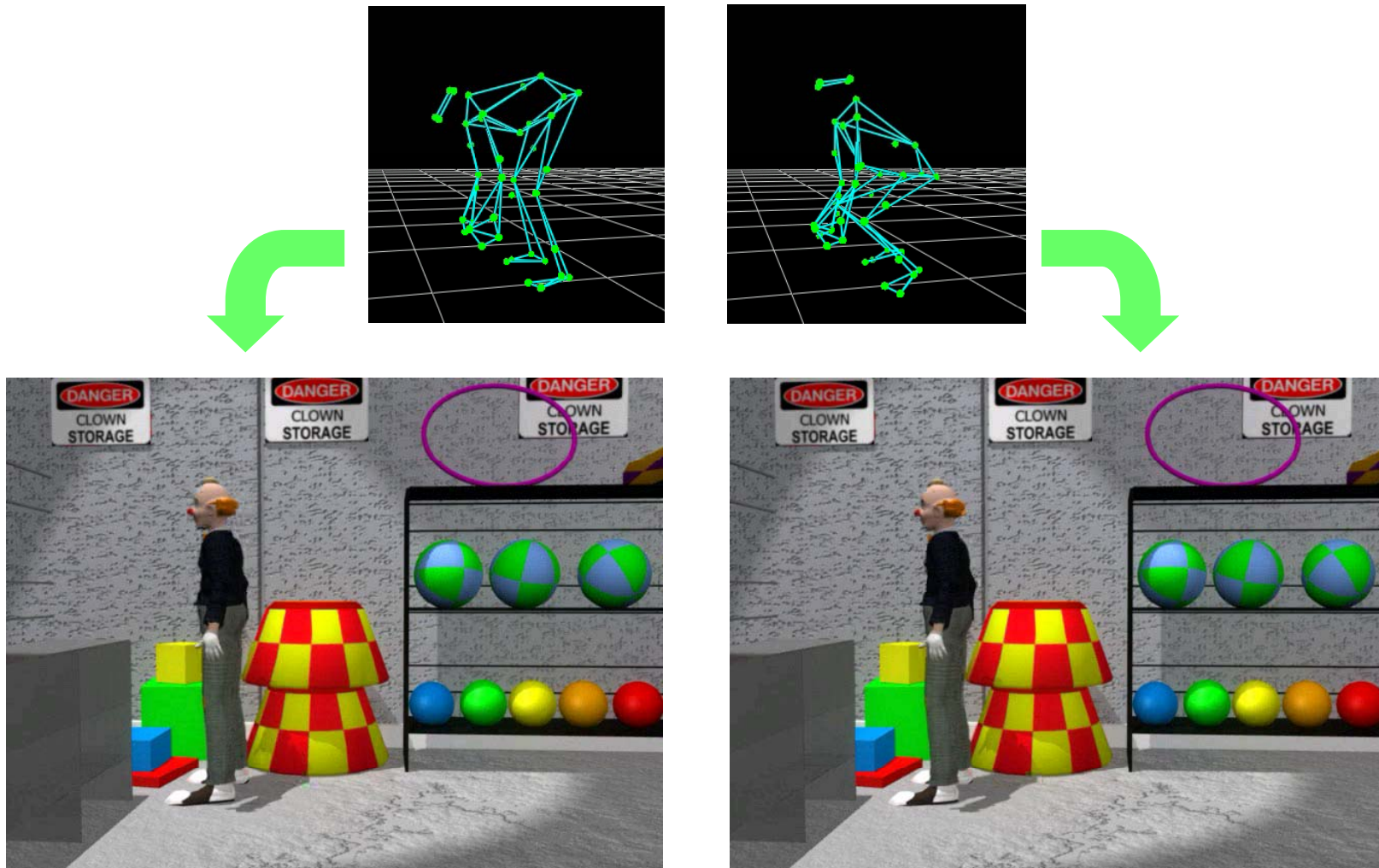
Synthesizing Manipulation Tasks

- combination of model and data
 - ◆ motion planning (RRT) [LaValle and Kuffner 2000] and inverse kinematics (UTPoser) [Yamane and Nakamura 2003]
 - ◆ relatively small data set (four pick-and-place examples)
- results



Synthesizing Manipulation Tasks

- using motion capture to bias IK



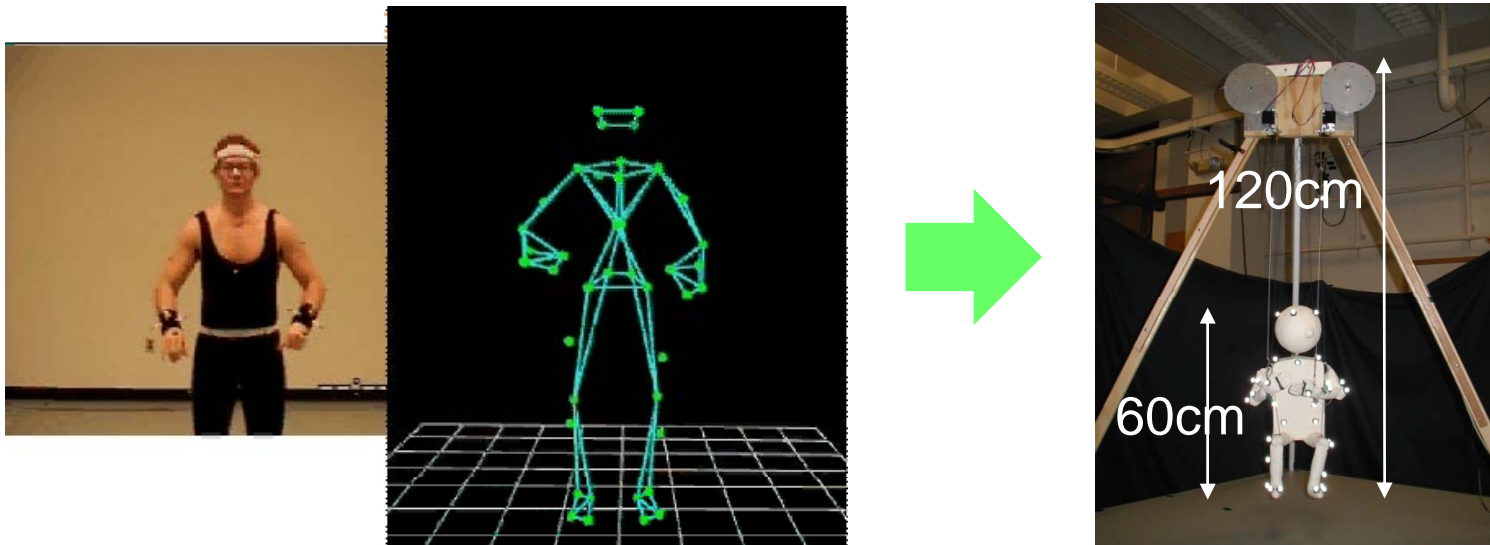
Adaptation Techniques

3. Motorized Marionette

[Yamane, Hodgins, Brown 2003]

Motorized Marionette

- programming marionettes by imitation
 - ◆ inexpensive device for entertainment



- issues
 - ◆ different actuation mechanism
 - ◆ very limited mobility

Motorized Marionette

■ issues and solutions

- ◆ mapping marker data to marionette's motion range
 - inverse kinematics with string length/direction constraints
- ◆ undesired swing
 - linear string model and feedforward swing suppression controller



Understanding Human Adaptation

Neuro-Musculoskeletal Human Model

Human Adaptation

- when human learns from observation
 - ◆ understands the demonstrator's intention
 - ◆ acquires the motion through learning (practice)
 - ◆ once acquired, instantly adapted to new environments, constraints and disturbances

- questions
 - ◆ how do we understand the demonstrator's intention?
 - ◆ what is the mechanism to realize instant adaptation?

Understanding Demonstrator's Intention

■ what has to be preserved?

- ◆ joint trajectory
- ◆ endeffector trajectory
- ◆ external force
- ◆ compliance
- ◆ something else?



■ automatic extraction is an open issue

- ◆ observing joint trajectories is not enough (obviously)
- ◆ analyzing multimodal data is essential

Instant Adaptation

■ hierarchy of controllers

◆ high-level motor control in the brain

- large (100ms+) delay: too slow to cope with disturbance?

◆ low-level reflex in the spinal cord

- sensors

- somatosensory information (muscle length, tension)
- touch, temperature, pain

- smaller (~30ms) delay

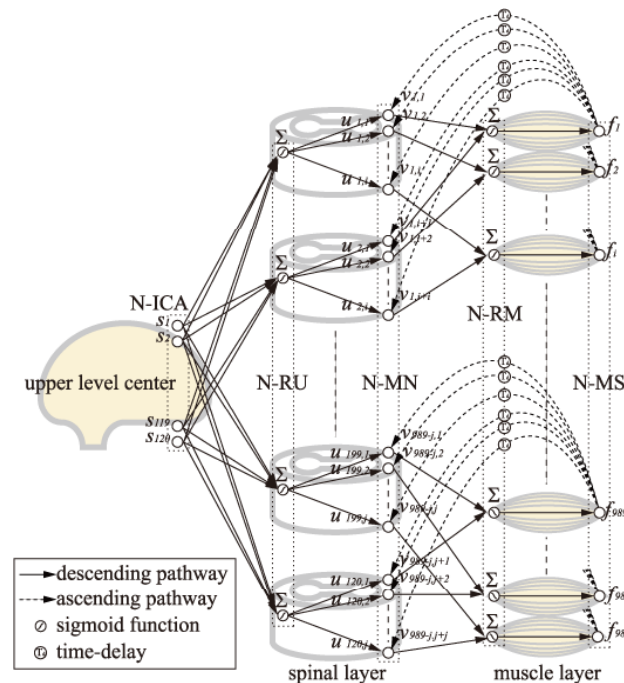
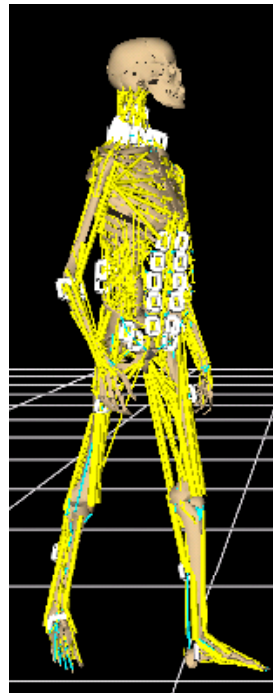
- most humanoid controllers run at or faster than 1KHz

→ sophisticated mechanism should exist

Towards Human-Level Adaptation

■ models for analysis

- ◆ detailed musculoskeletal model for estimating the somatosensory information [Yamane et al. 2005]
- ◆ neuromuscular network model with somatosensory reflex [Murai et al. 2008]



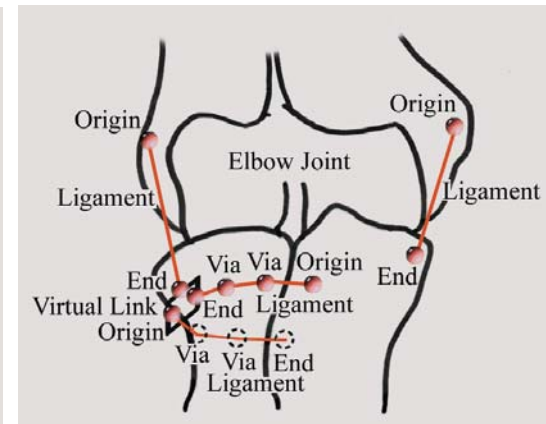
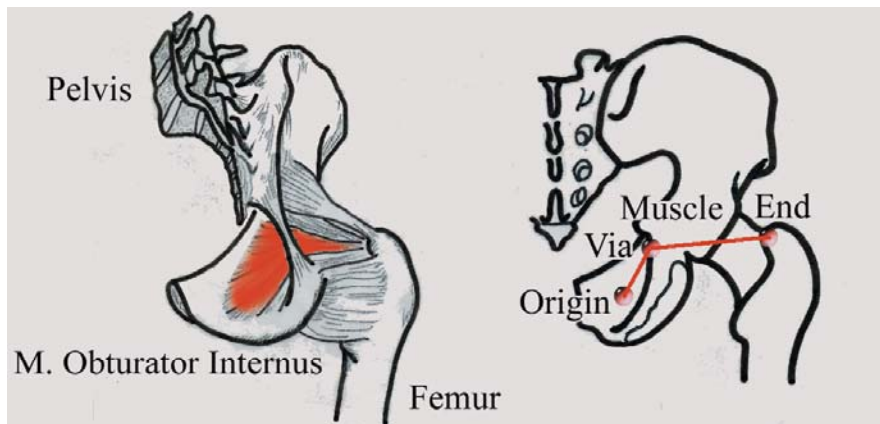
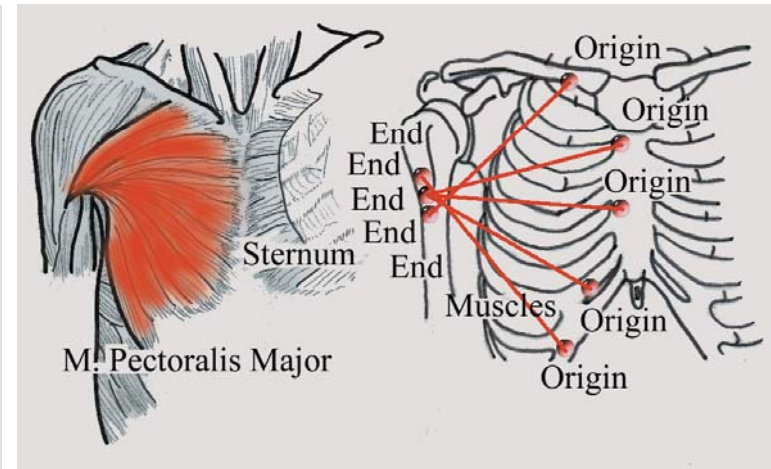
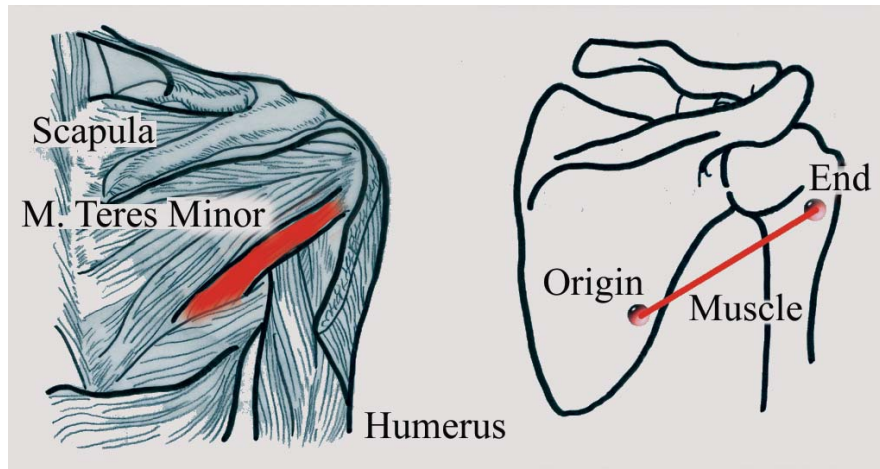
Musculoskeletal Model

- skeleton: 155 DOF
 - ◆ 200 bones → 53 groups
 - ◆ composed of mechanical joints
 - ◆ hand/foot fingers not included
- actuator: more than 1,000 wires
 - ◆ 997 muscles: linear actuators
 - ◆ 50 tendons: connect muscles and bones
 - ◆ 117 ligaments: constrain the bones
- algorithms
 - ◆ inverse kinematics → muscle length / velocity
 - ◆ inverse dynamics → muscle tension estimation



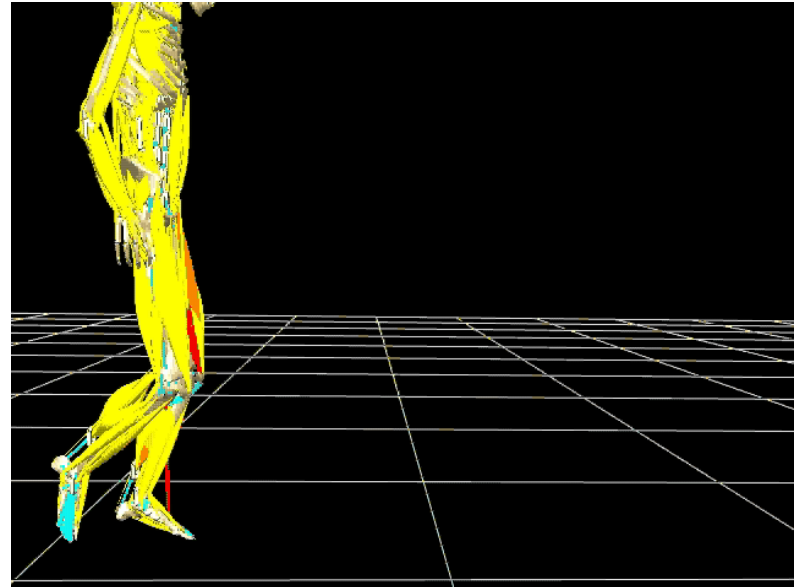
Musculo-Tendon Network

- mass-less, zero-radius wires with via-points

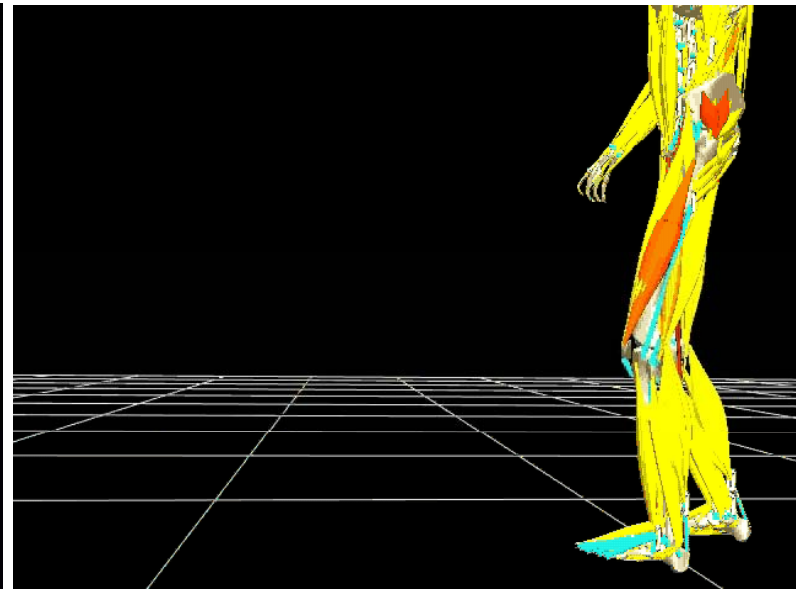


Example: Muscle Tension Estimation

toe
walk

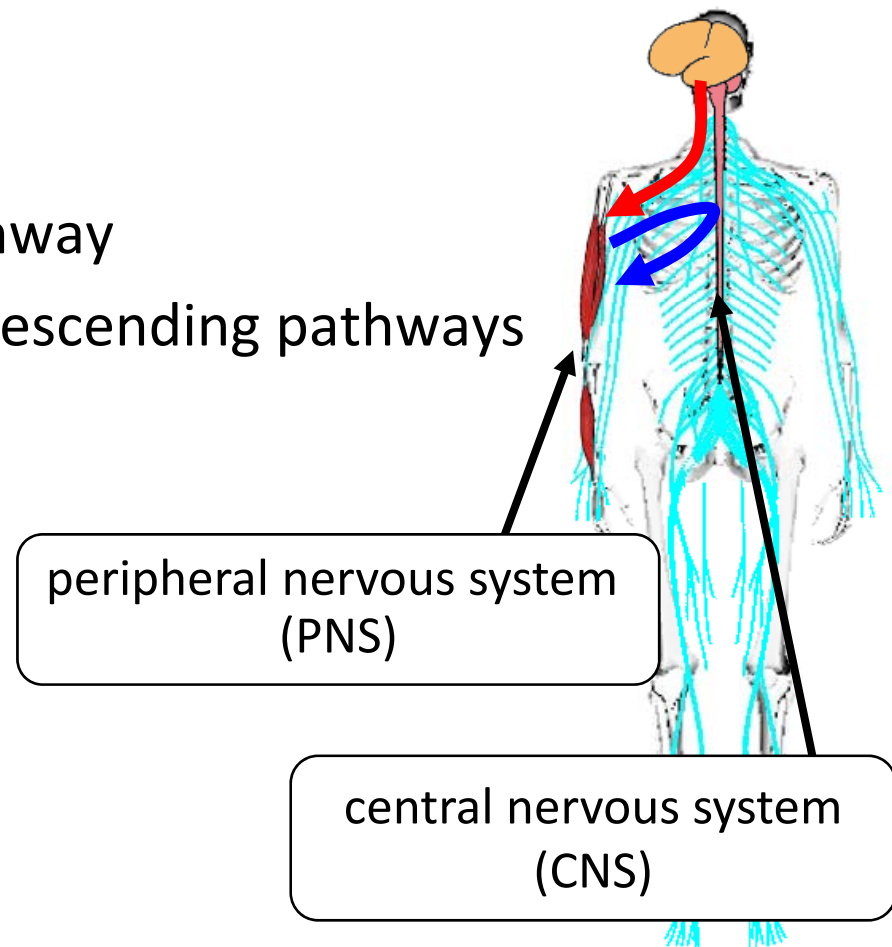


heel
walk



Neuromuscular Network Model

- input and output of the model
 - ◆ input: motor command signals at spinal nerve rami
 - ◆ output: muscle tension
- two paths:
 - ◆ **CNS→PNS**: descending pathway
 - ◆ **PNS→PNS**: ascending and descending pathways (somatic reflex network)



Identifying the Motor Command Signals

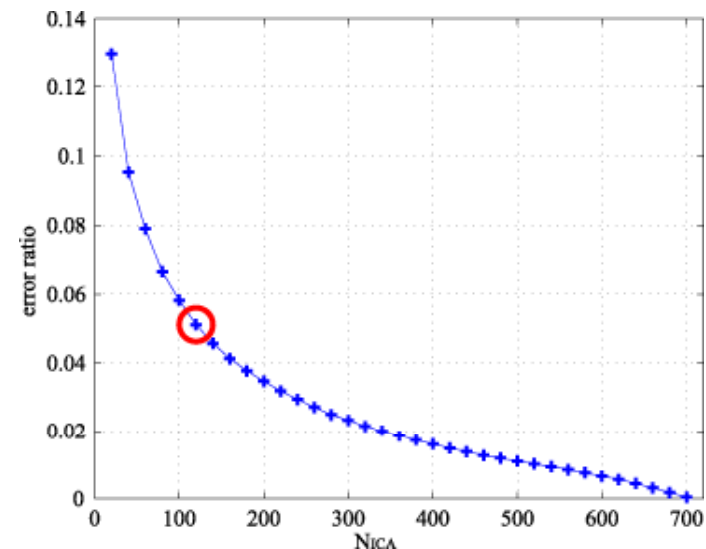
- independent component analysis (ICA)
 - ◆ estimate mutually independent signal sources
 - ◆ order of the independent signals is undefined

$$\mathbf{f} = \mathbf{W}_{ICA} \mathbf{s}$$

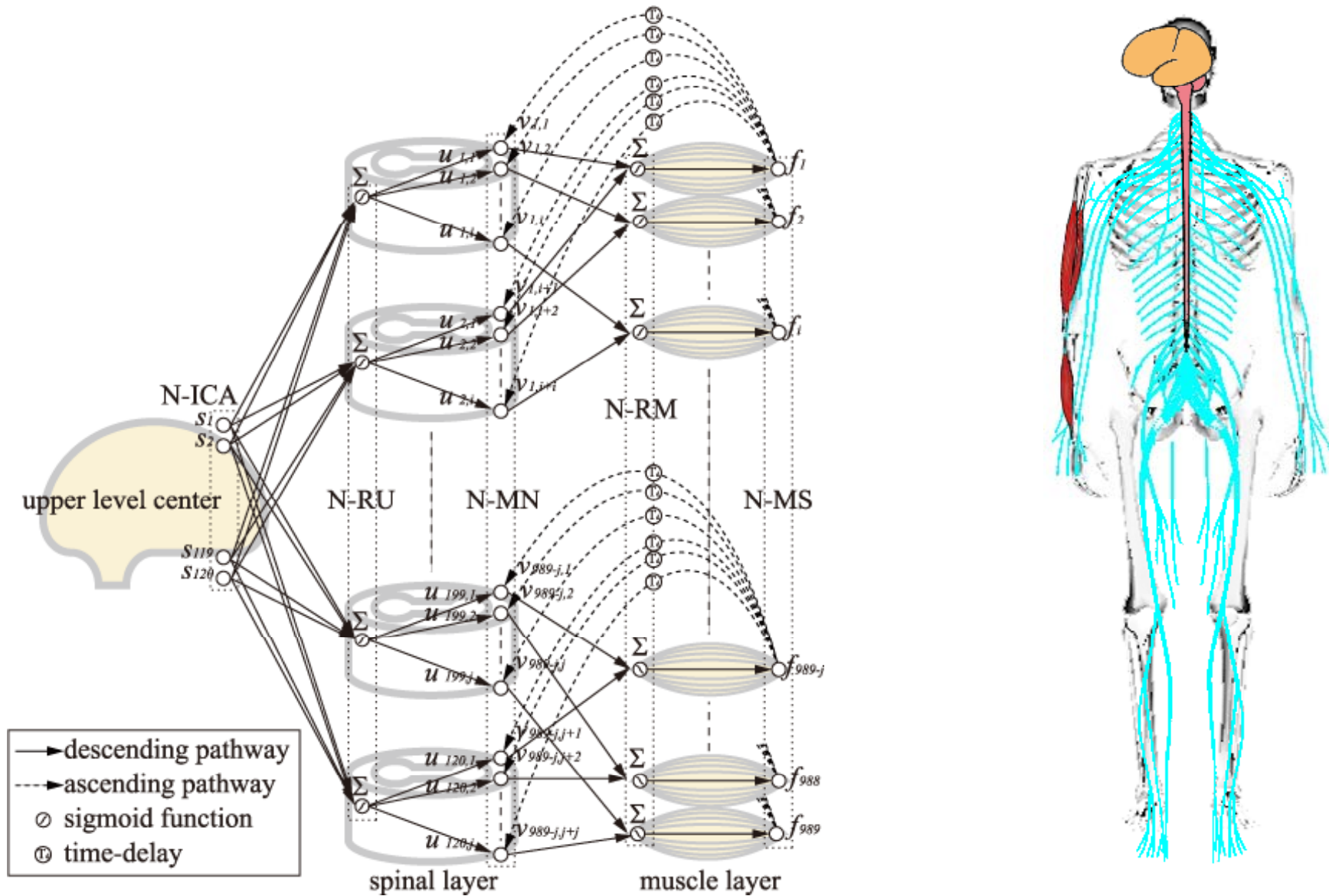
muscle tensions

independent signals
=motor command signals

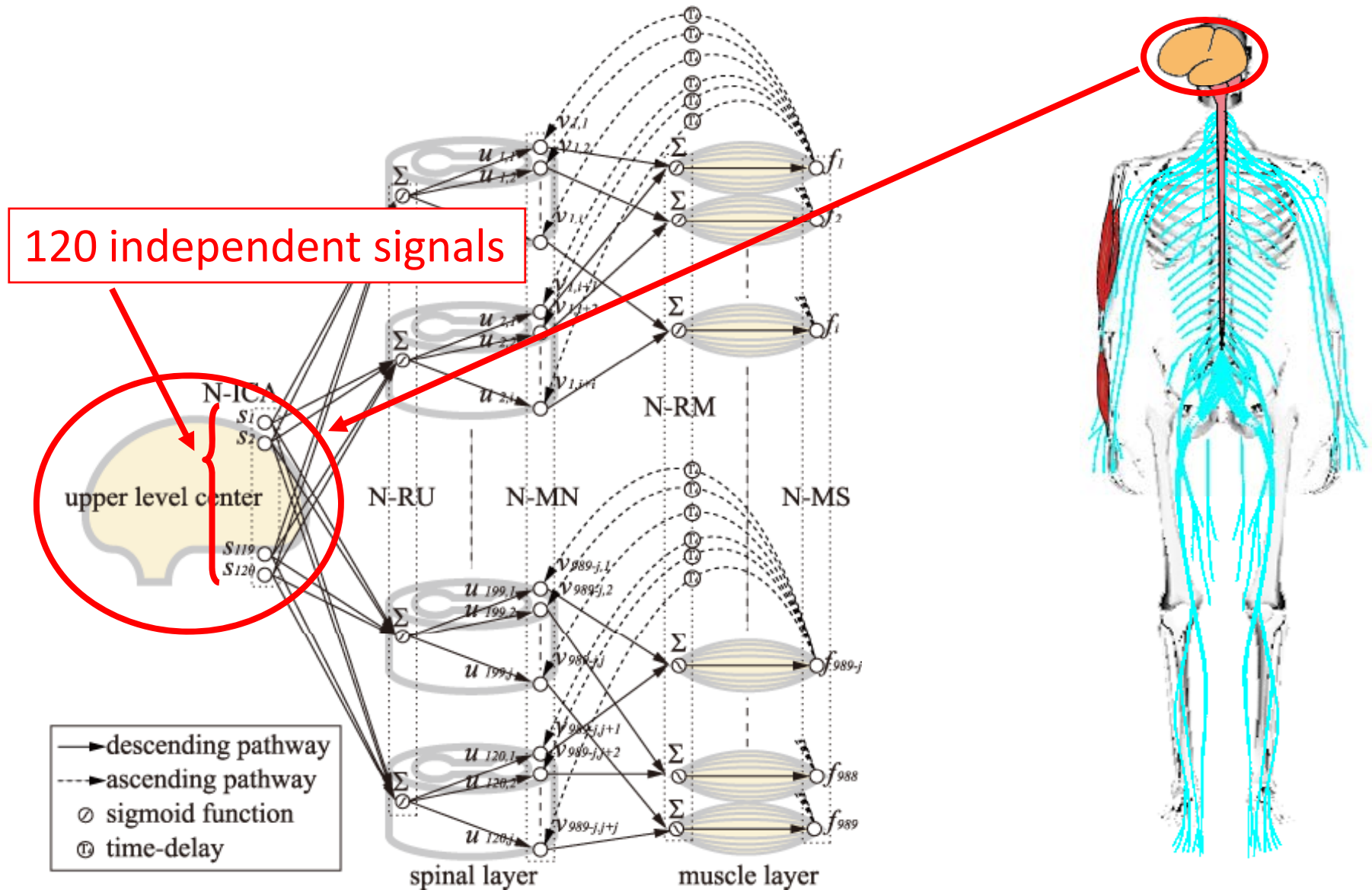
- dimension of \mathbf{s} ?
 - ◆ 120, the number of relevant spinal nerve rami, is enough!



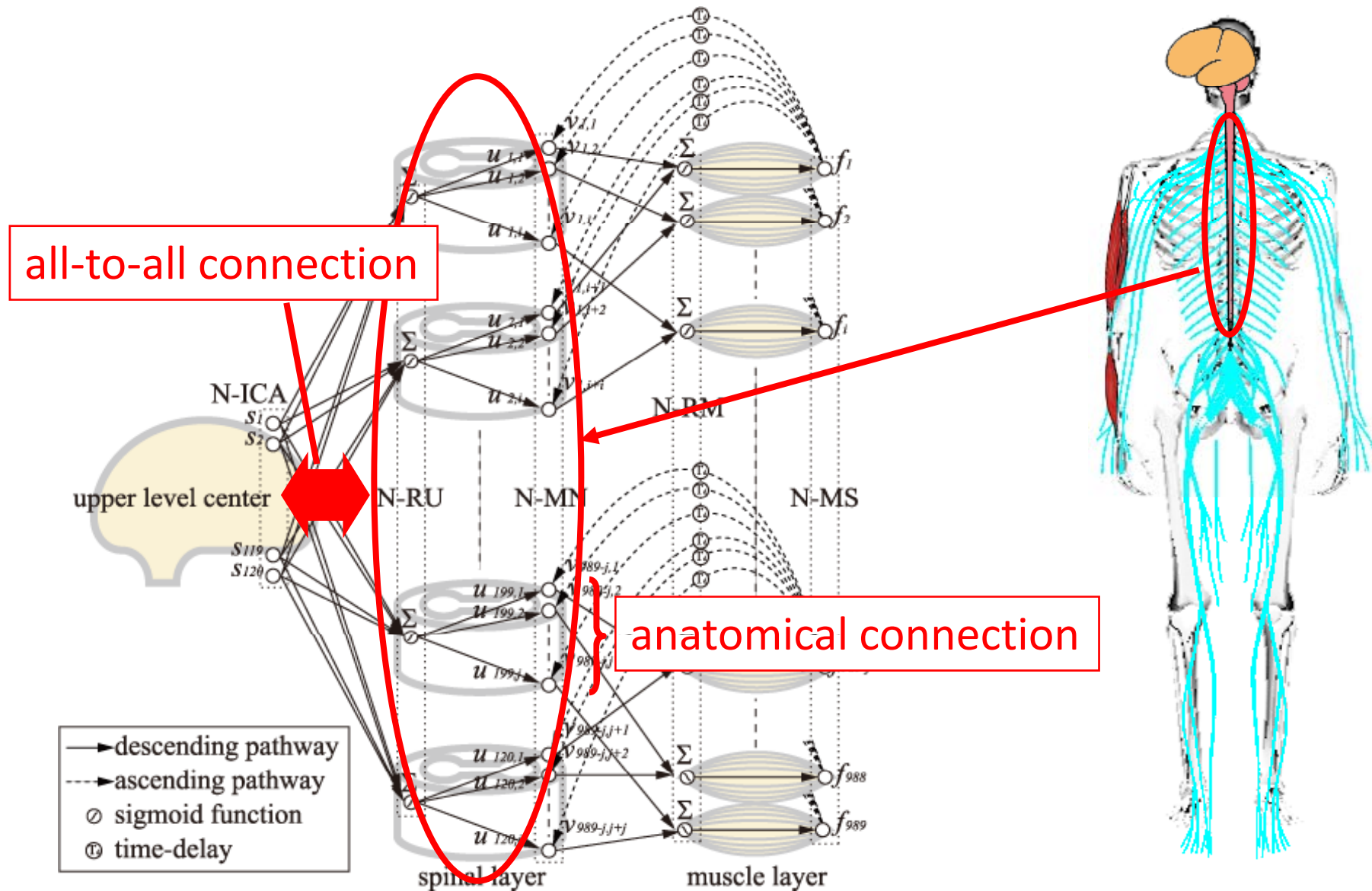
Neuromuscular Network Model



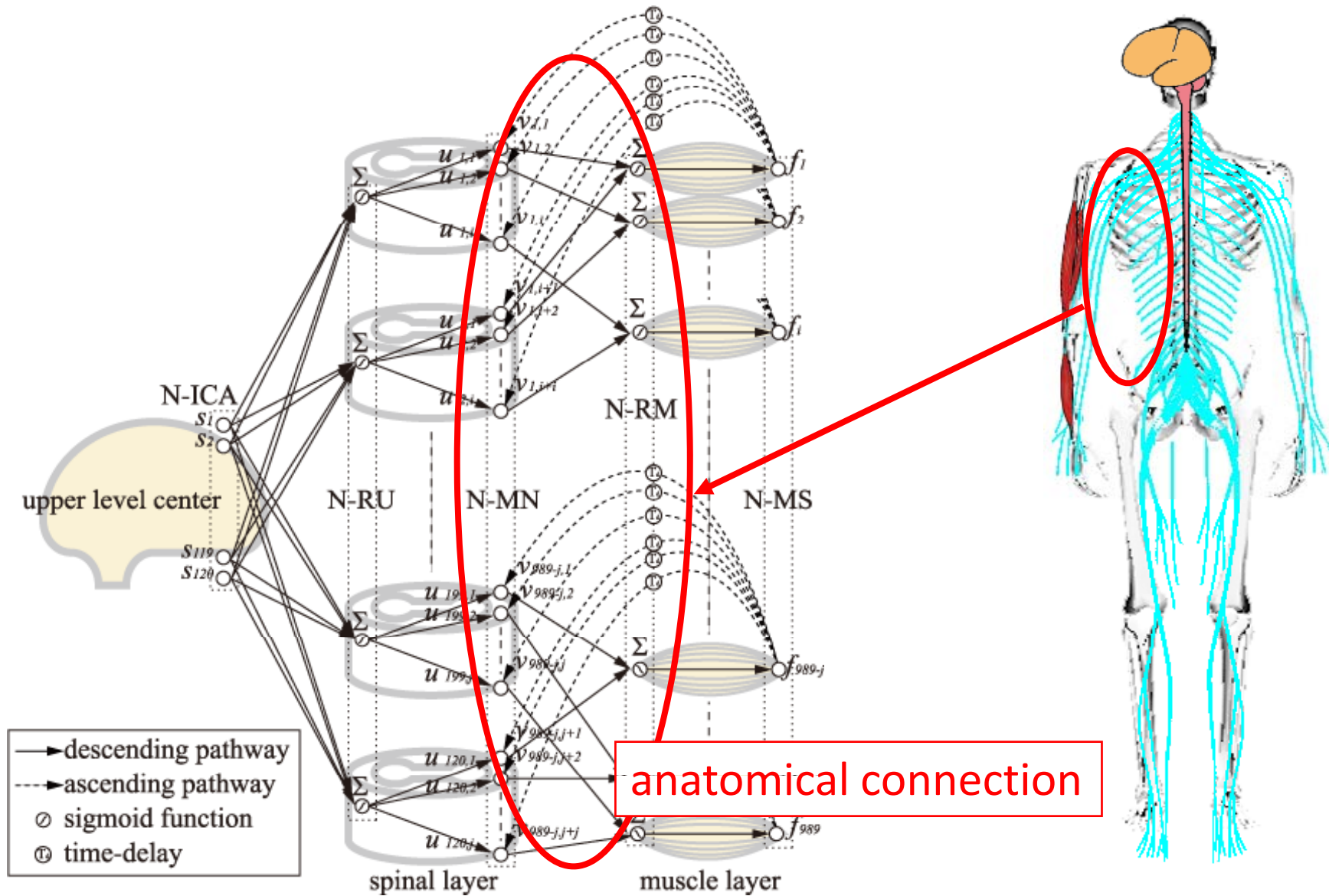
Neuromuscular Network Model



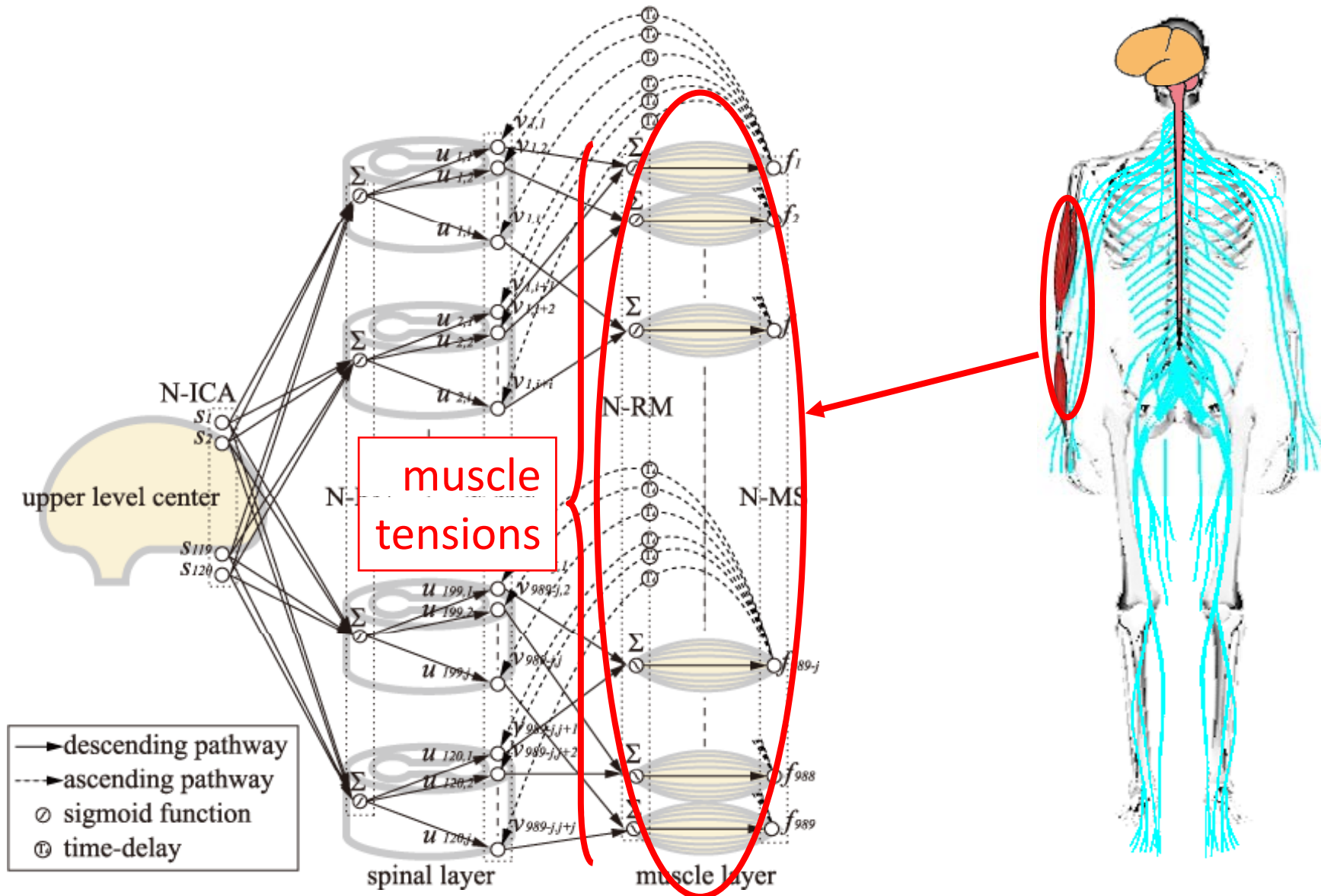
Neuromuscular Network Model



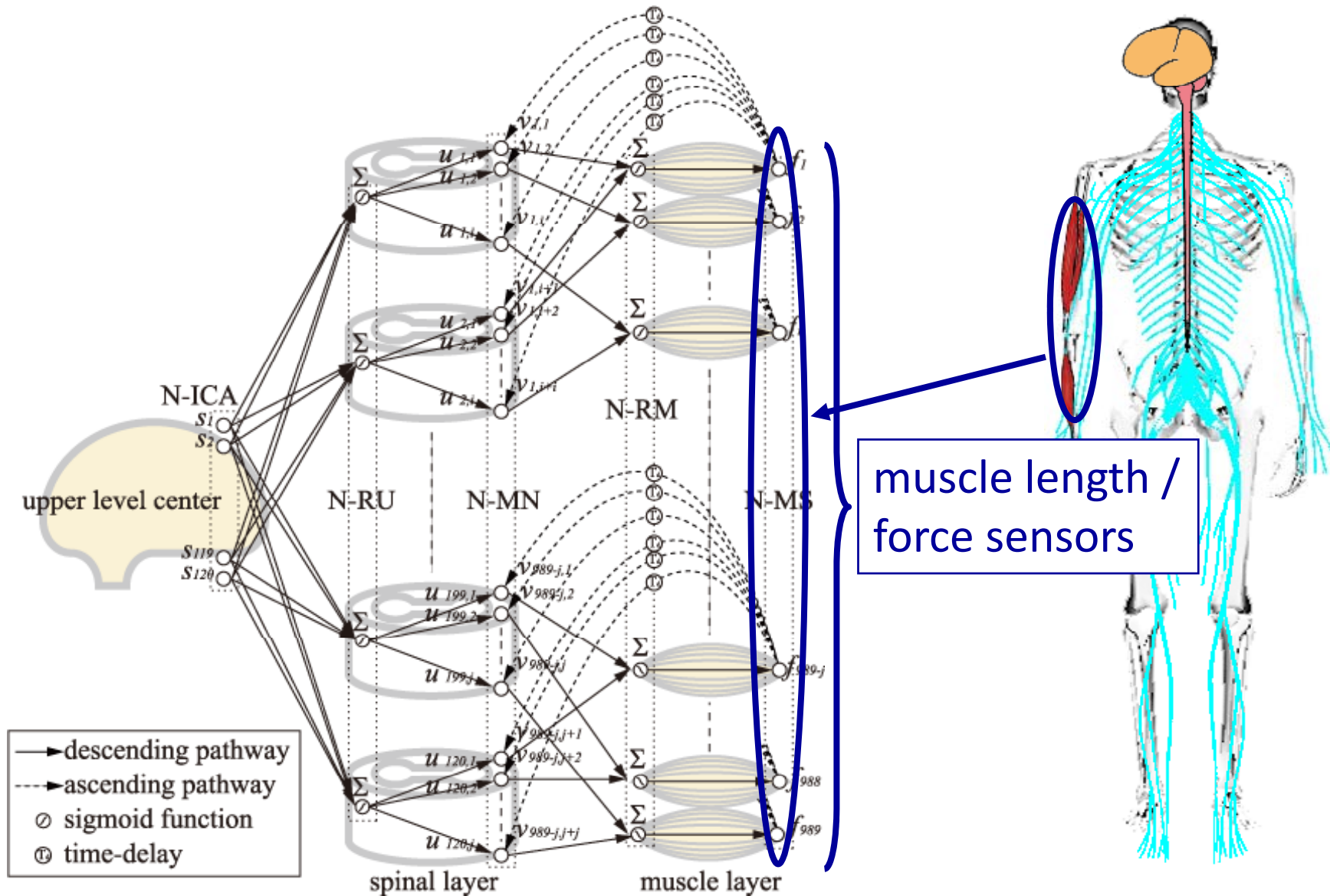
Neuromuscular Network Model



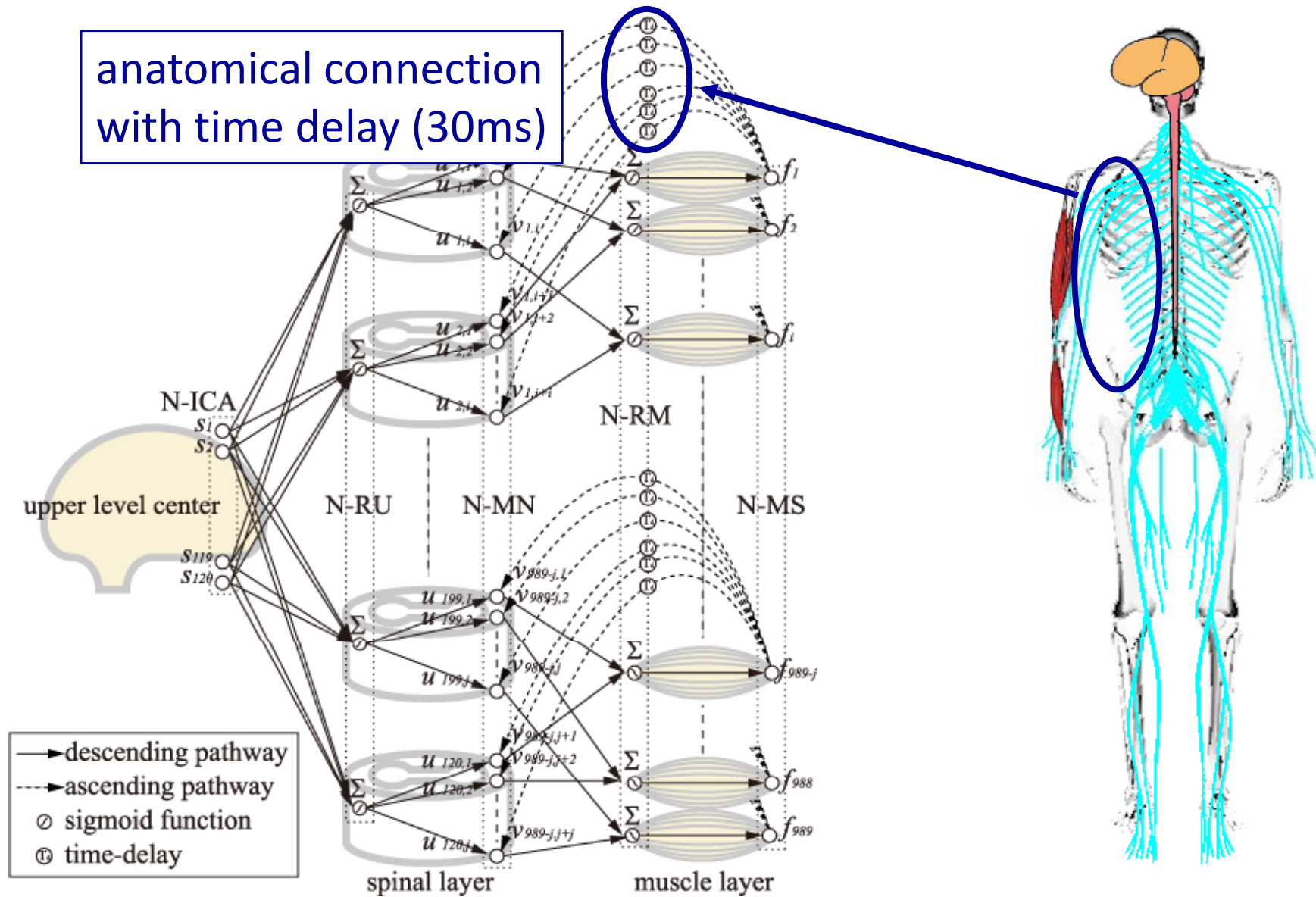
Neuromuscular Network Model



Neuromuscular Network Model



Neuromuscular Network Model



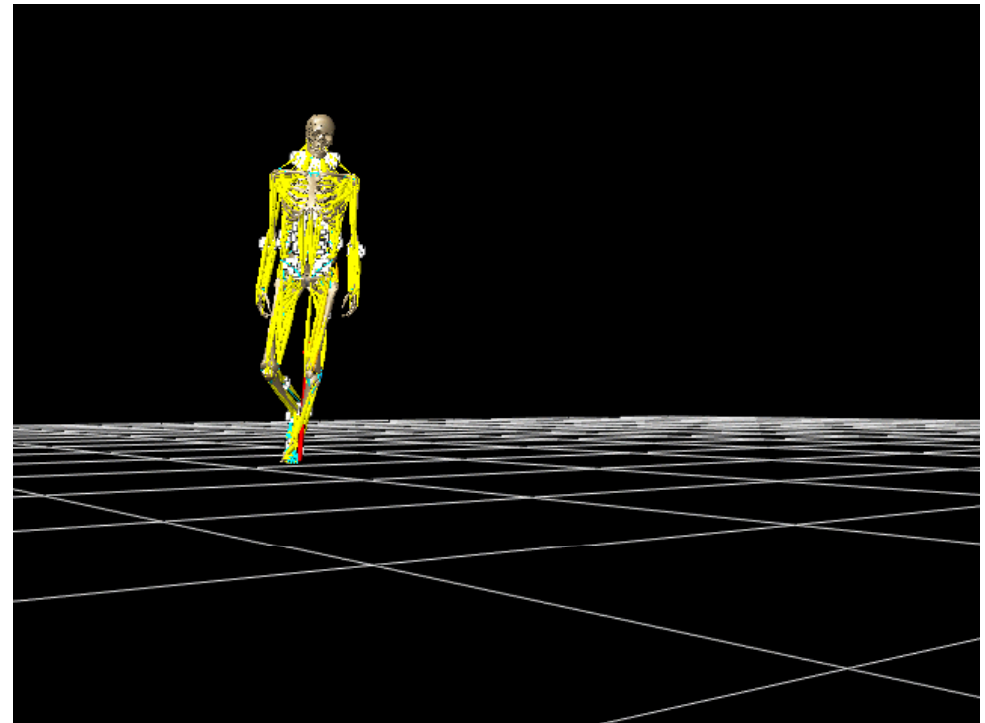
Identification

■ training data

- ◆ walk (2000 frames, 10 seconds)
- ◆ muscle tensions from inverse dynamics
- ◆ independent signals from ICA

■ training

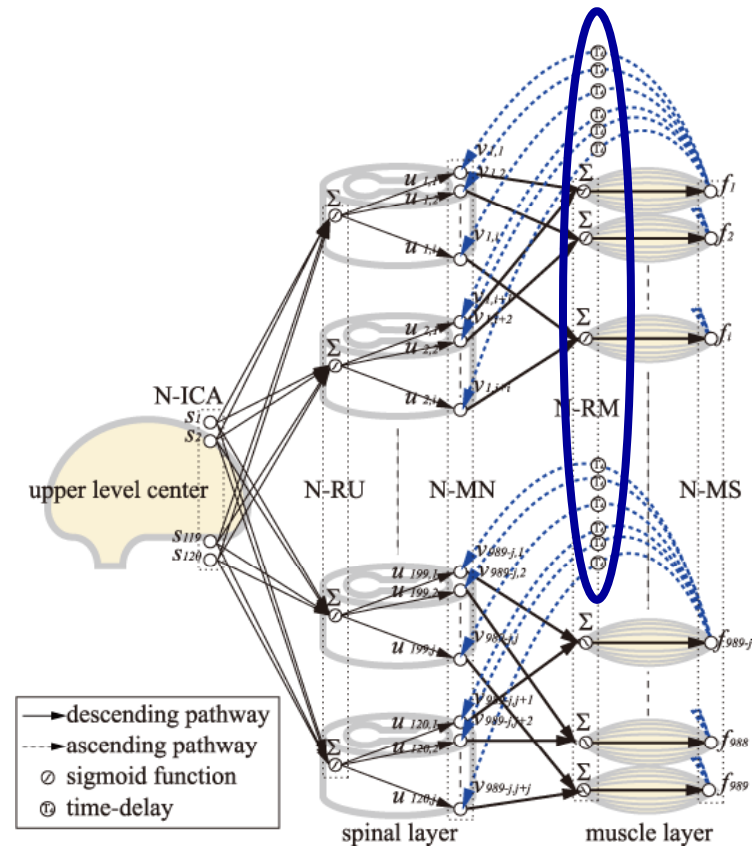
- ◆ 5000 cycles
- ◆ error
 - average: 2.59%
 - variance: 0.34%



Results: Weight Parameter of Reflex Loop

- from Iliacus: agonist for hip flexion

Iliacus	
muscle	weight
Iliacus	4.20E+00
Sartorius	2.60E-01
Rectus Femoris	3.94E+00
Pectineus	4.60E-01
Gracilis	-4.06E-01
Adductor Longus	-1.69E-01
Adductor Brevis	-3.59E-01
Adductor Magnus	-4.10E-02

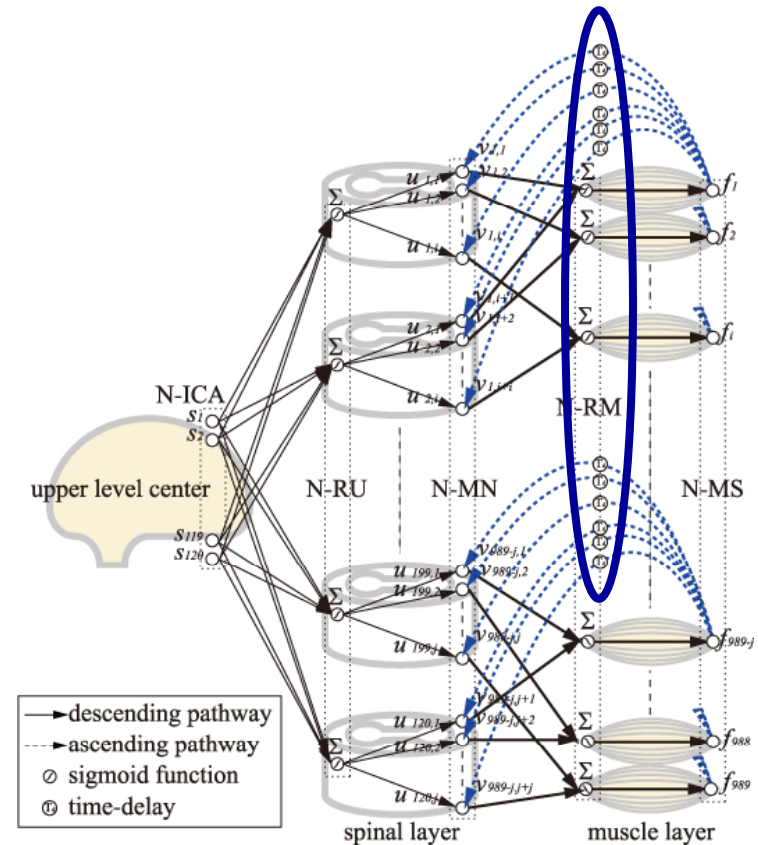


classified as agonist muscles
 for hip flexion in sports science

Results: Weight Parameter of Reflex Loop

- from Tensor Fasciae Latae: agonist for hip flexion

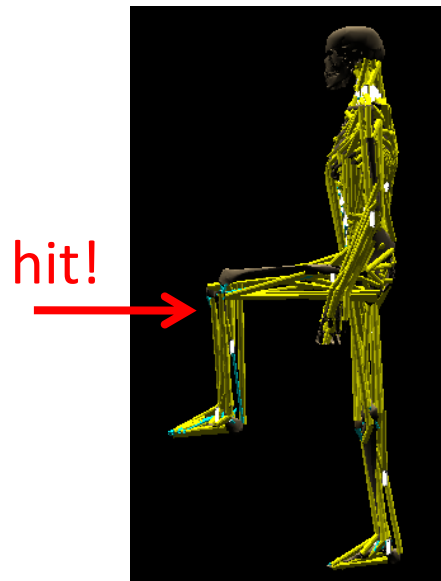
Tensor Fasciae Latae	
muscle	weight
Tensor Fasciae Latae	2.93E-01
Gluteus Maximus	-1.28E-01
Biceps Femoris	-4.53E-01
Semitendinosus	-1.02E+00
Semimembranosus	-1.13E+00
Gluteus Medius	-4.94E-02
Gluteus Minimus	-9.30E-01



classified as antagonist muscles for hip flexion in sports science

Results: Patellar Tendon Reflex

- patellar tendon reflex: stretch reflex of quadriceps



- motion adaptation techniques
 - ◆ **dynamics filter**: adaptation to small differences in kinematics and dynamics
 - ◆ **synthesizing manipulation tasks**: adaptation to different kinematics and environment
 - ◆ **motorized marionette**: completely different kinematics and actuation mechanism

- understanding human adaptation
 - ◆ **musculoskeletal and neuromuscular network models**

Acknowledgements

- Yoshihiko Nakamura (Univ. of Tokyo)
- Akihiko Murai (Univ. of Tokyo)
- Masaya Hirashima (Univ. of Tokyo)
 - ◆ dynamics filter
 - ◆ neuro-musculoskeletal human model

- Jessica Hodgins (CMU)
- CMU Graphics Lab
 - ◆ synthesizing manipulation tasks
 - ◆ motorized marionette