

딥러닝으로 드래곤 길들이기

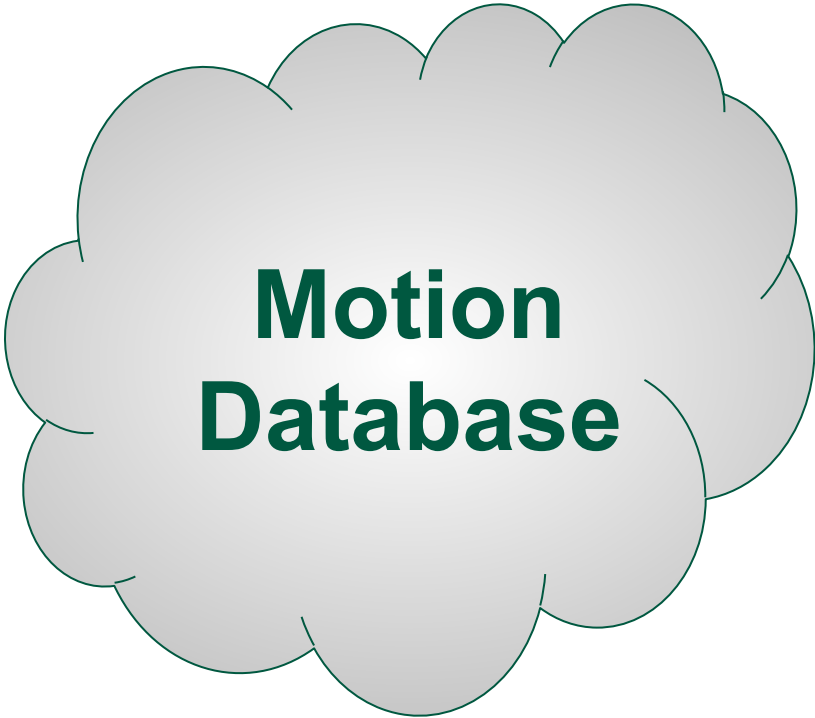
이제희 서울대학교 컴퓨터공학부



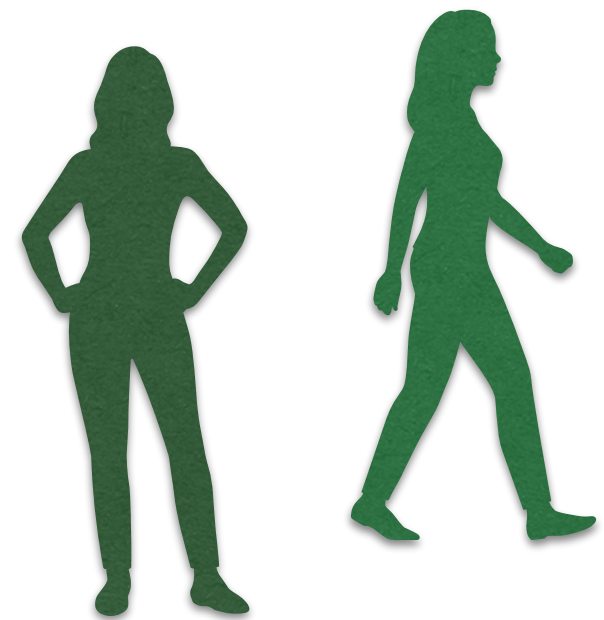
Motion Capture



Data-Driven Animation



**Motion
Database**



Data-Driven Animation

Degree of User Intervention

Unsupervised

Minimally-supervised

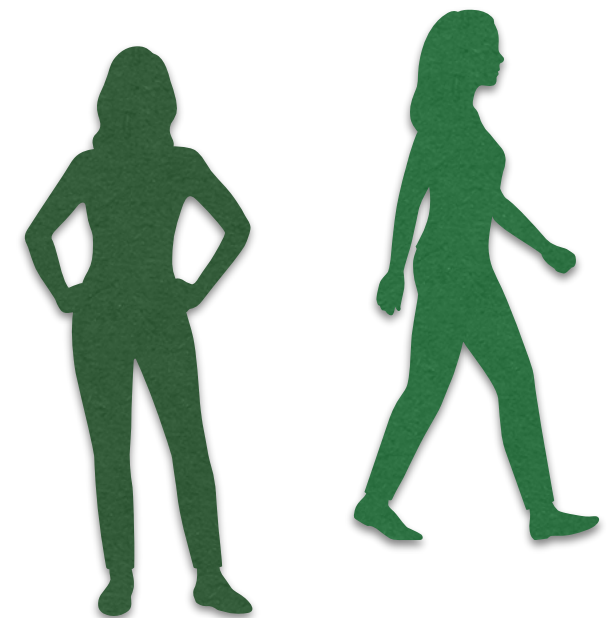
Supervised



**Motion
Database**

Data Size

small to very large



Model Complexity

Sticks and joints

3D animated character

Musculoskeleton

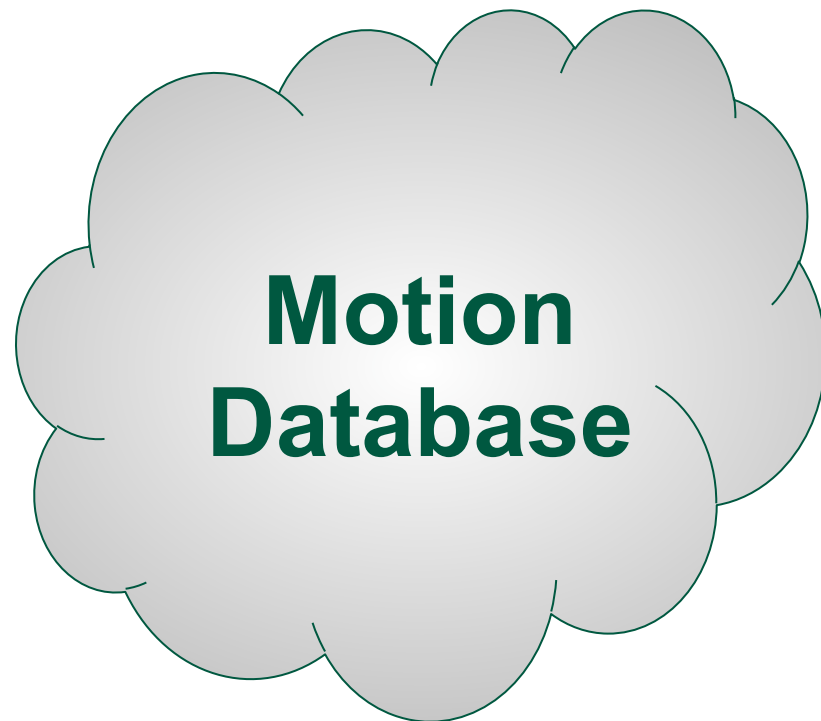
Data-Driven Animation

Degree of User Intervention

Unsupervised

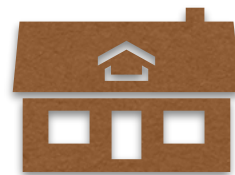
Minimally-supervised

Supervised

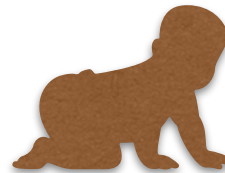


Data Size

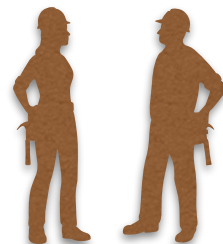
small to very large



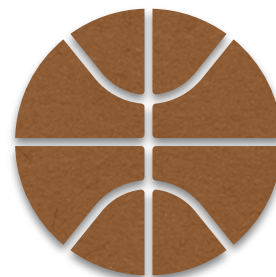
Environment



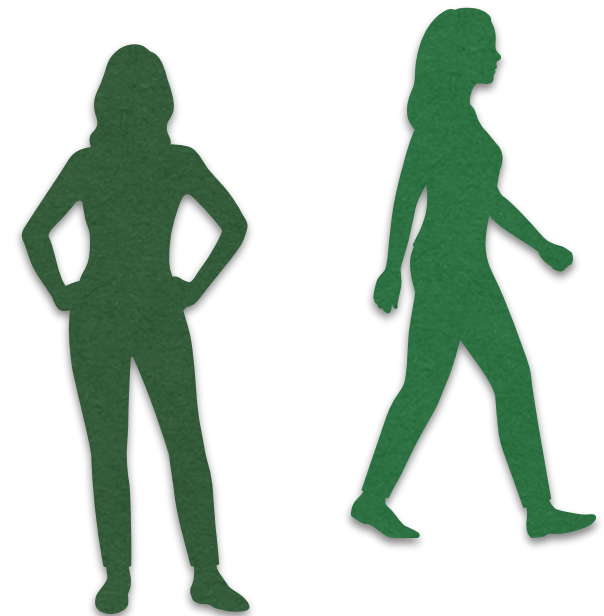
Body Types



Interaction



Syntax and Semantics



Model Complexity

Sticks and joints

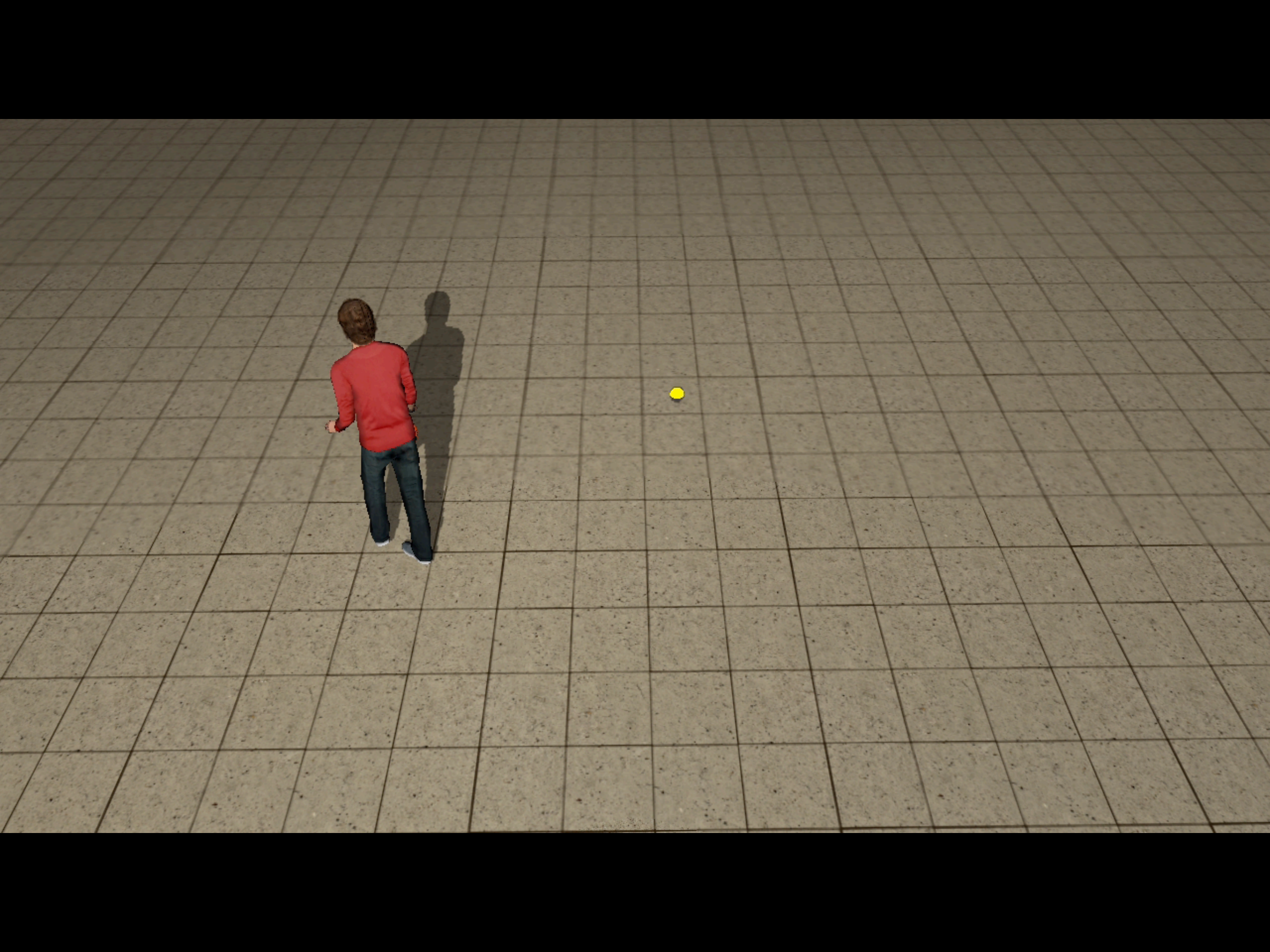
3D animated character

Musculoskeleton

Interactive Character Animation by Learning Multi-Objective Control



Kyungho Lee, Seyoung Lee and Jehee Lee
(ACM Transactions on Graphics 2018, SIGGRAPH Asia 2018)



Why Deep Learning?

Fast at runtime

Specialized chips will be available to all platforms soon

Compact in memory

We don't have to keep motion data

Everything goes into layers of networks

Easy to build

Learning process is largely automated

Interactive control

We still want to have direct control over the character's motion

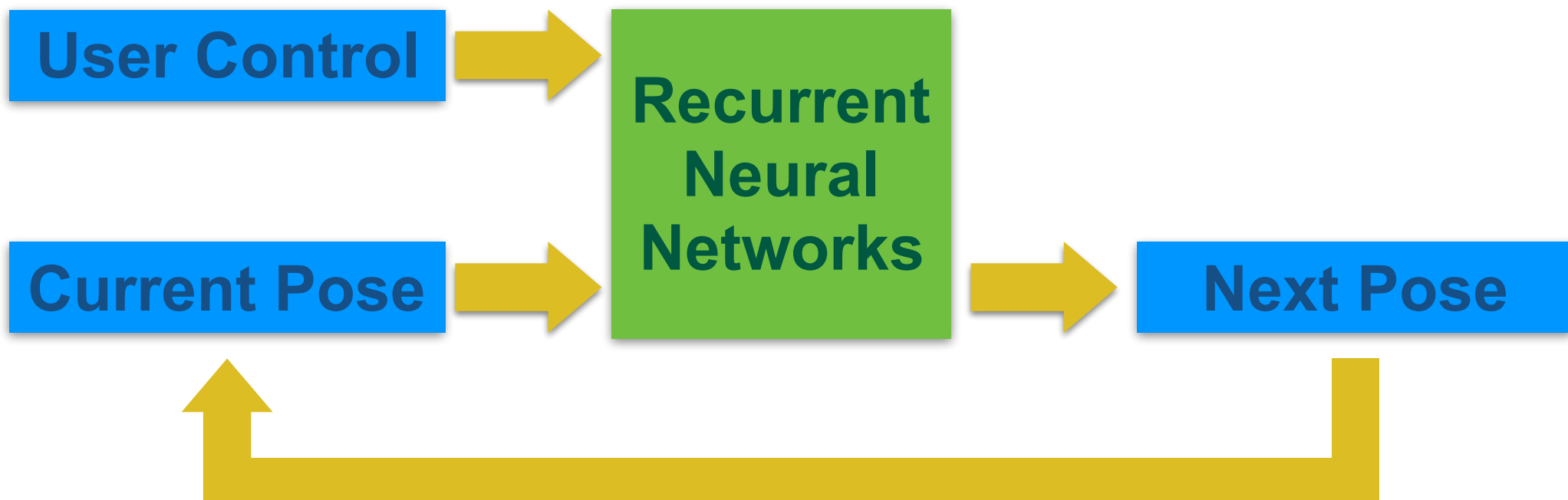
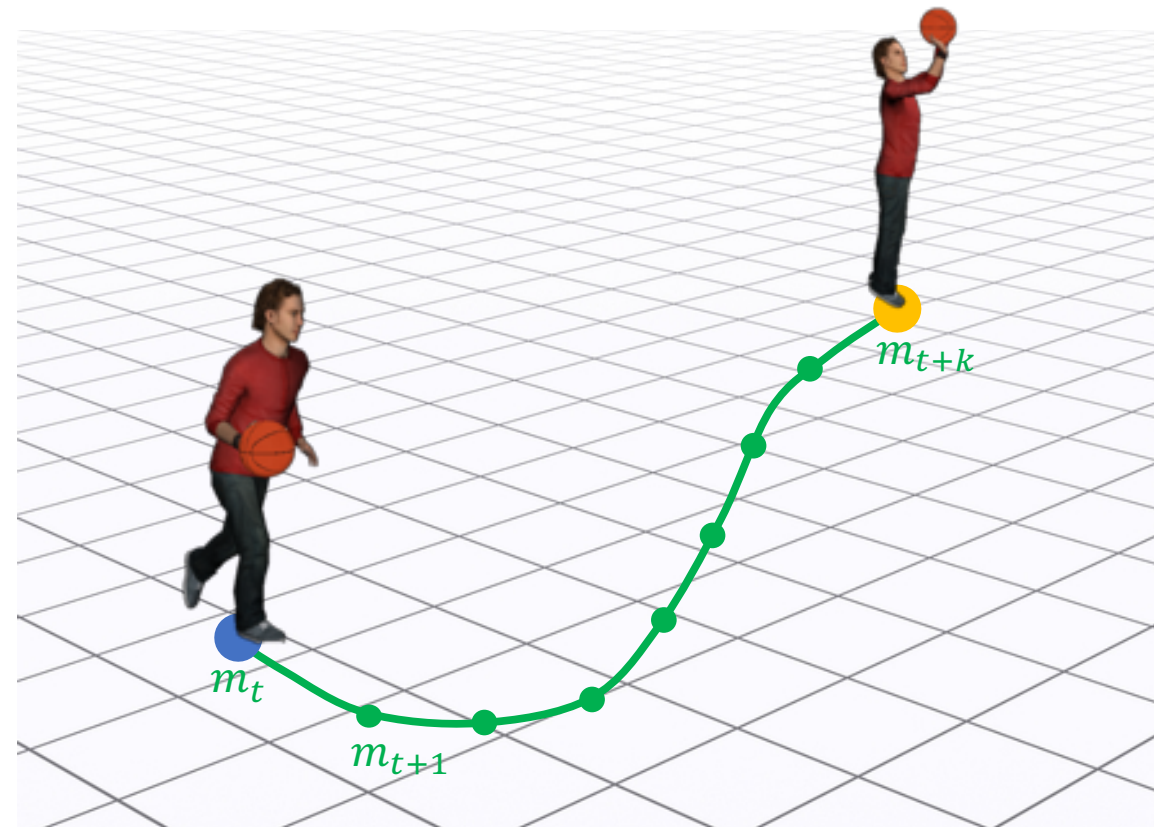
Motion Capture Data





RNN for Interactive Character Control

Target position,
direction,
velocity, ...



Issues in Motion Learning

Multiple actions in the data set

Dribble, shoot, pass, catch, pivot, run, ...

Size of motion data

We can never have enough data for deep learning

Segmentation and labeling

labor-intensive and time consuming

Multi-Objective Control

Learn Multiple actions simultaneously

Each action may have a different set of control parameters

Automatic data augmentation

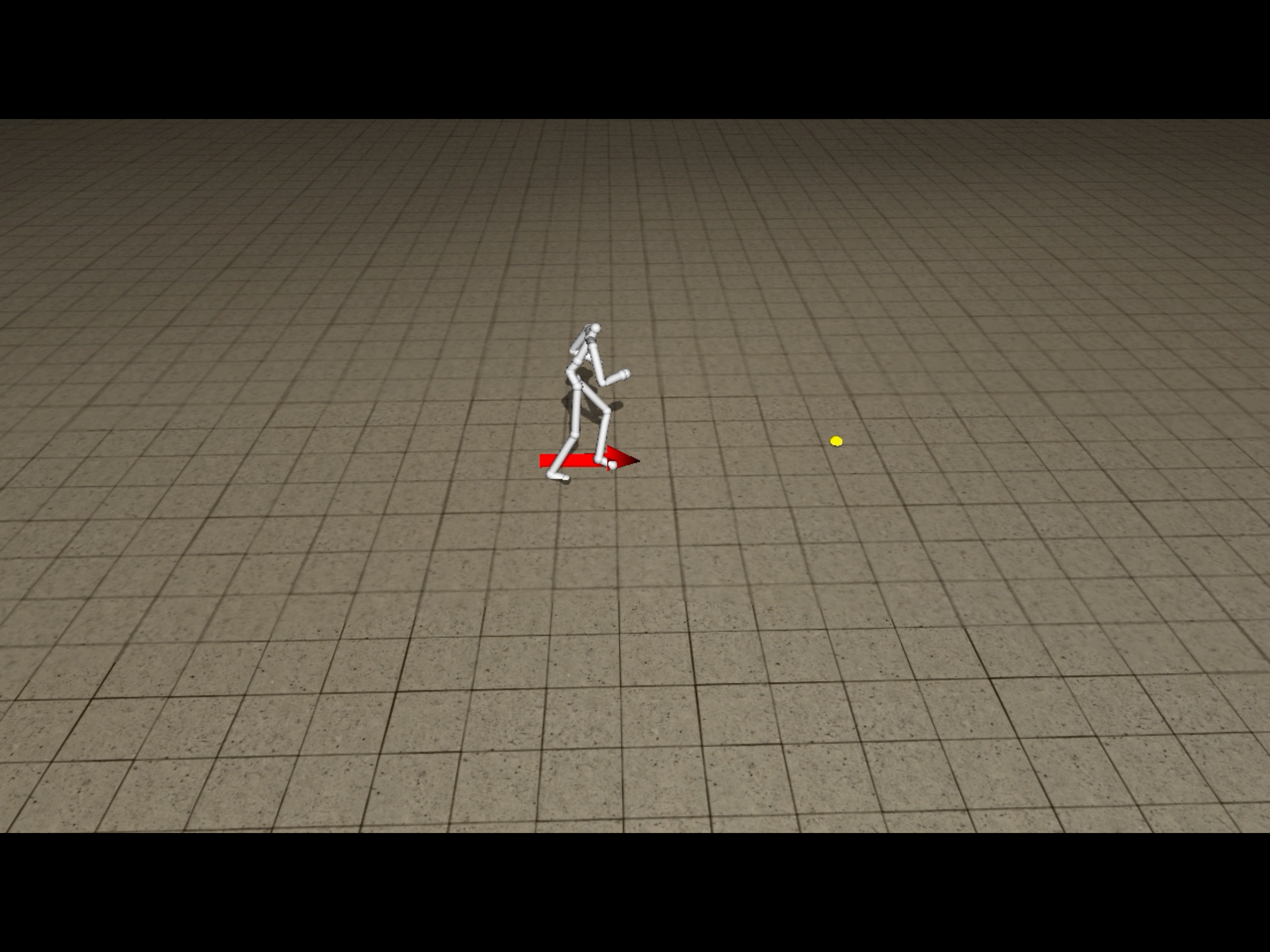
Enhance variabilities in combinatorial and spatiotemporal domains

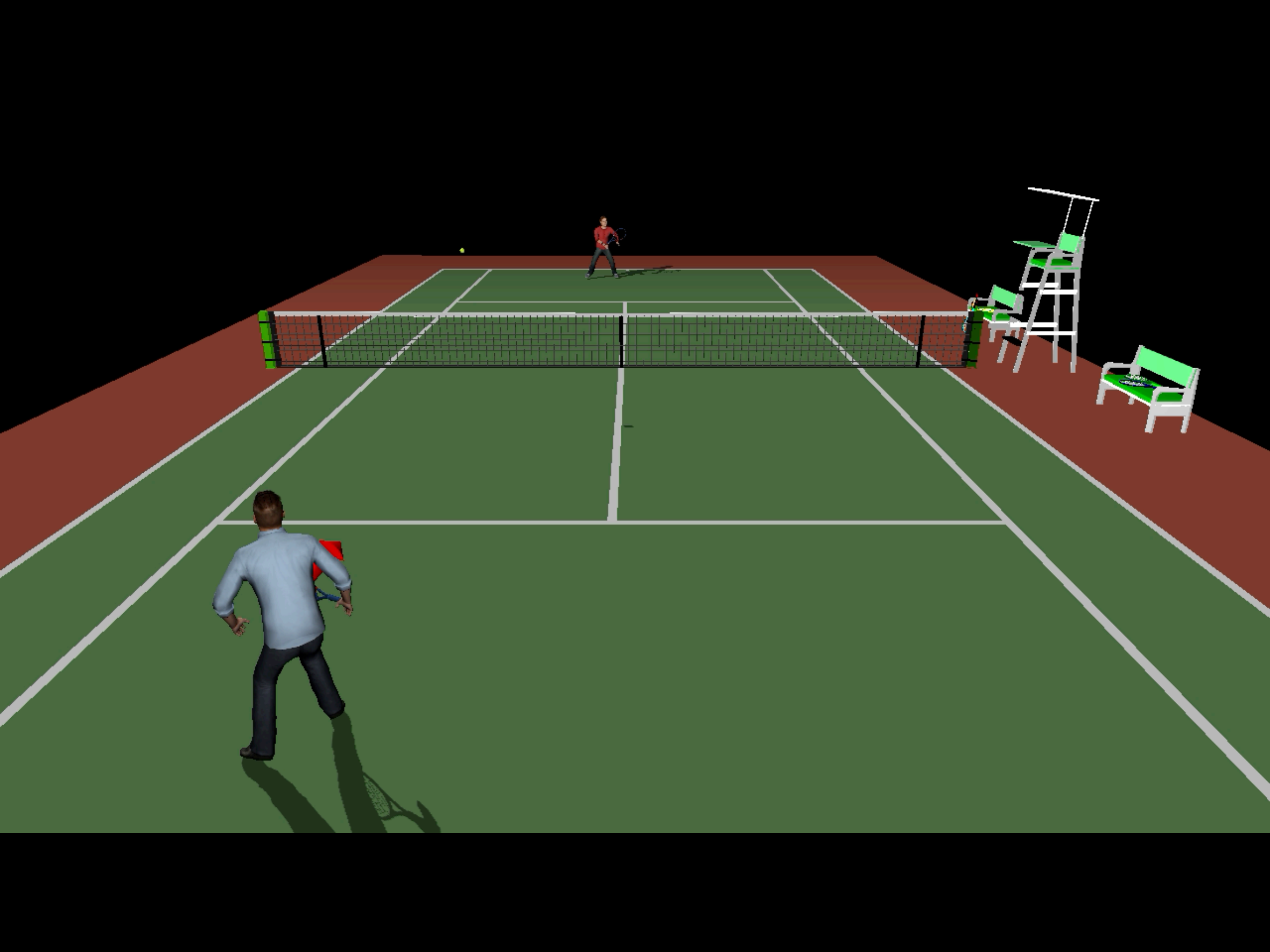
From very small (10 to 20 sec) to medium (20 min)

Minimal user intervention

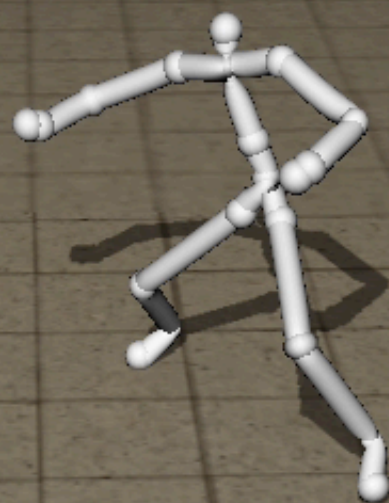
No segmentation, minimal labeling

Need more work to design syntactic and semantic rules





Gorilla



Physically Simulated



[SIGGRAPH 2010] Lee et al, Data-driven biped control

with Tracking Control Only



[SIGGRAPH 2010] Lee et al, Data-driven biped control

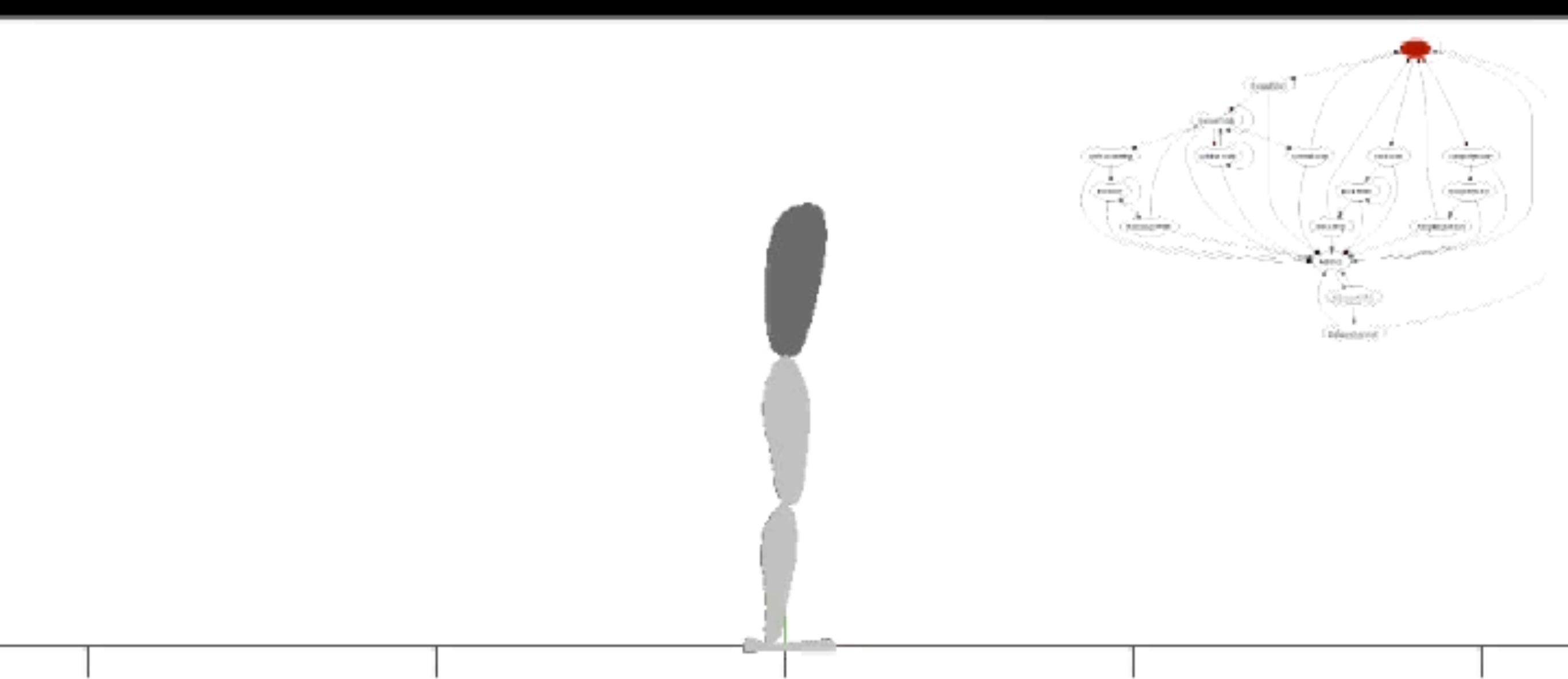
Physically Simulated



[SIGGRAPH 2010] Lee et al, Data-driven biped control



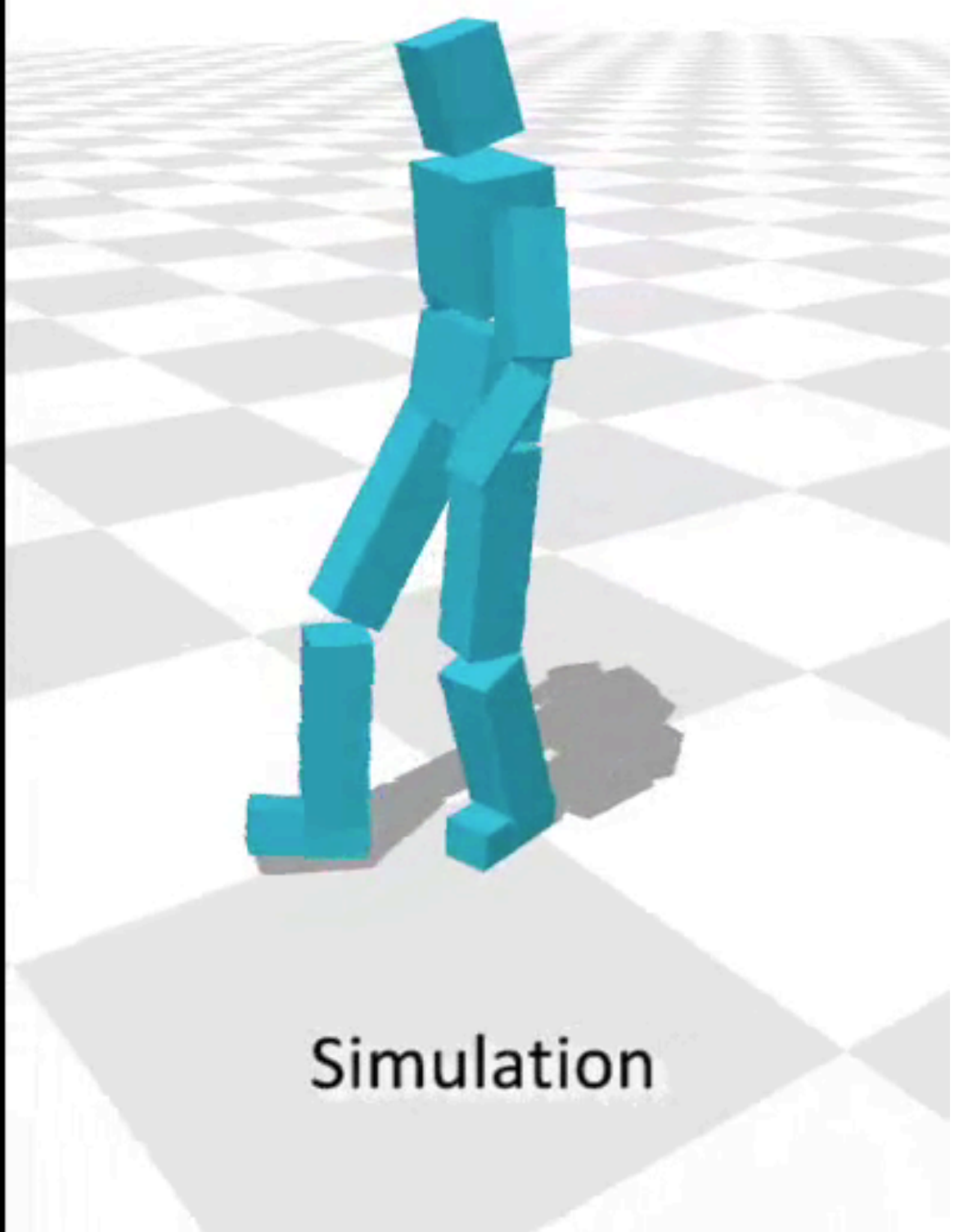
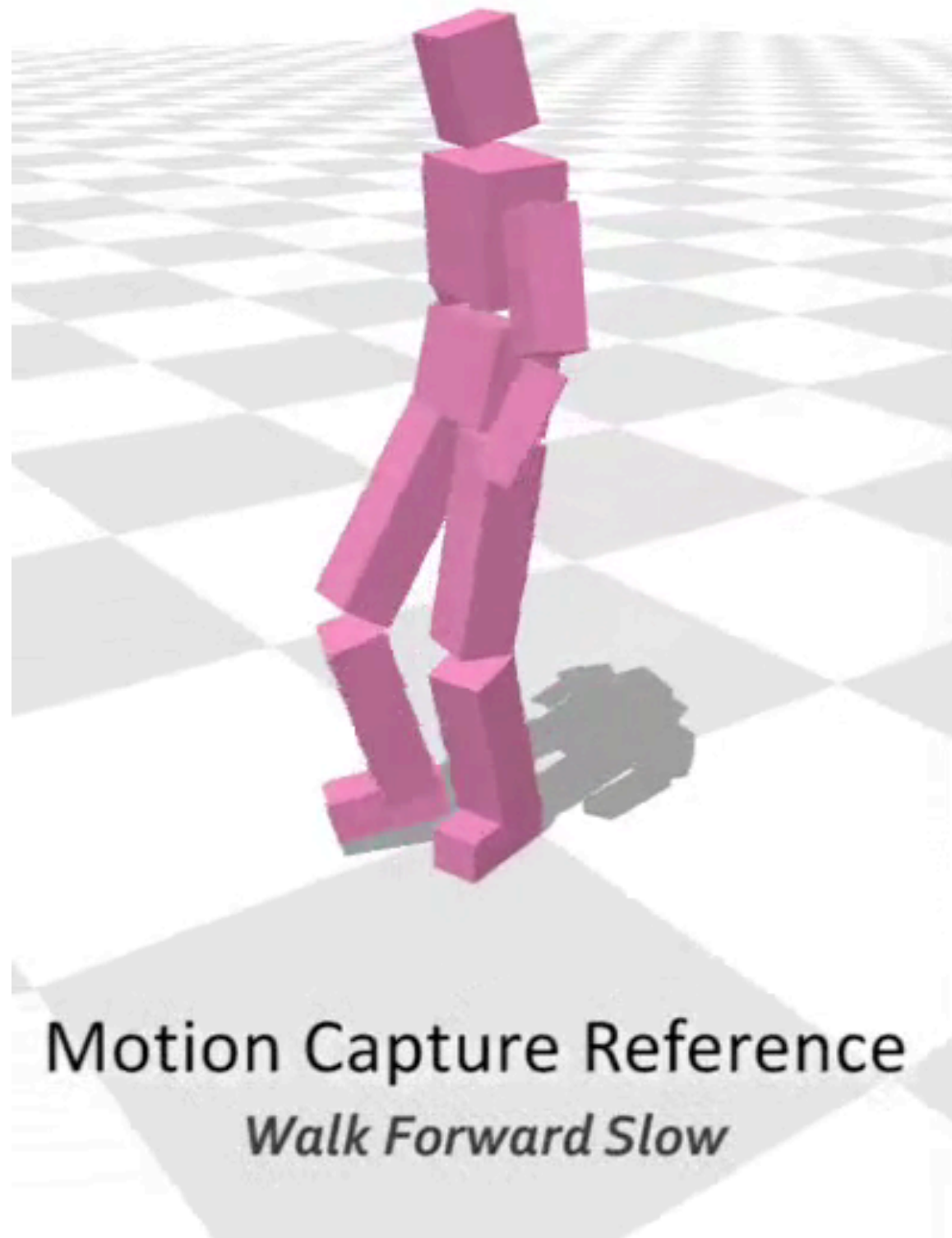
Hubo 2013



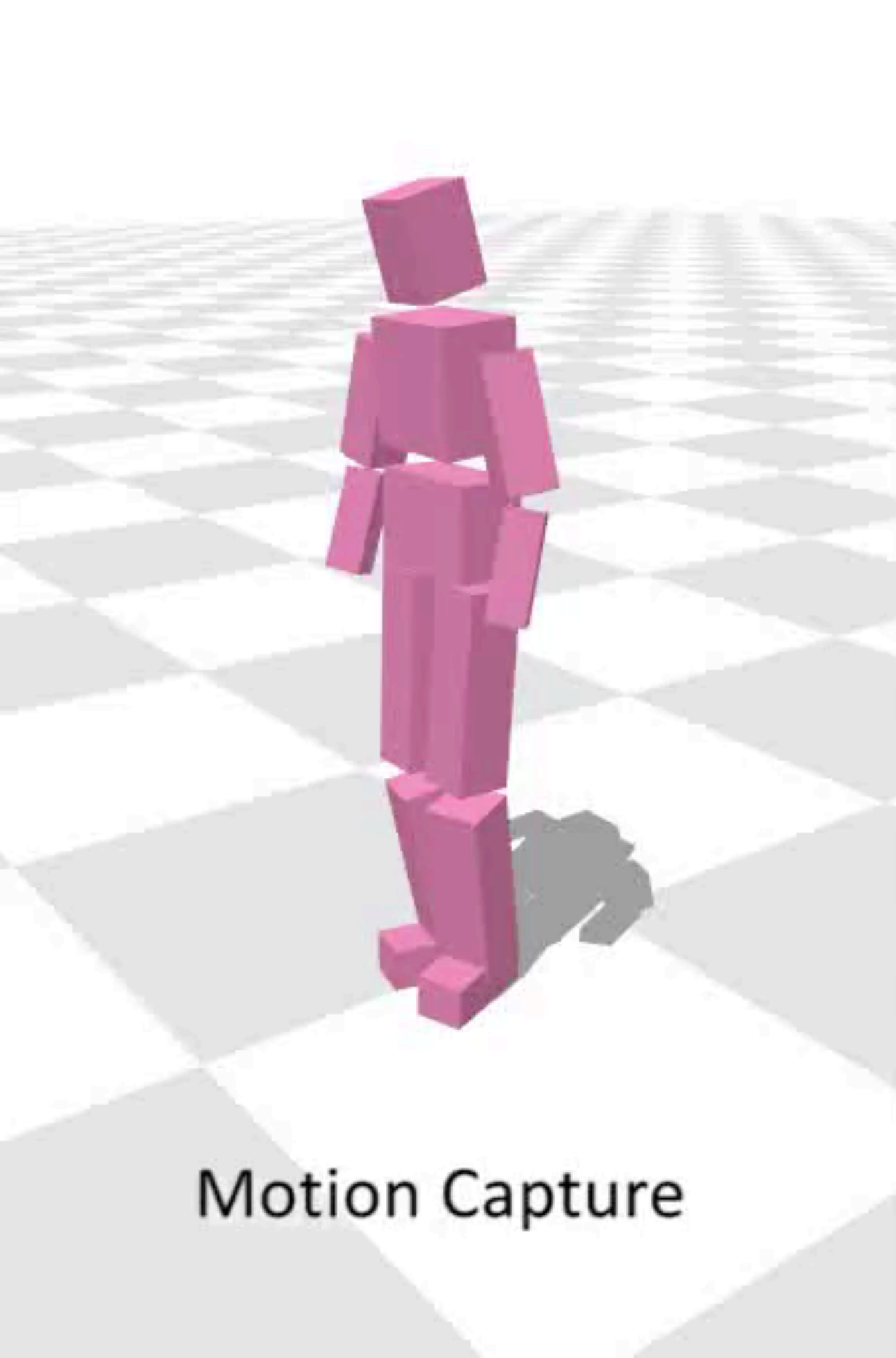
Stand

[SIGGRAPH 2007]

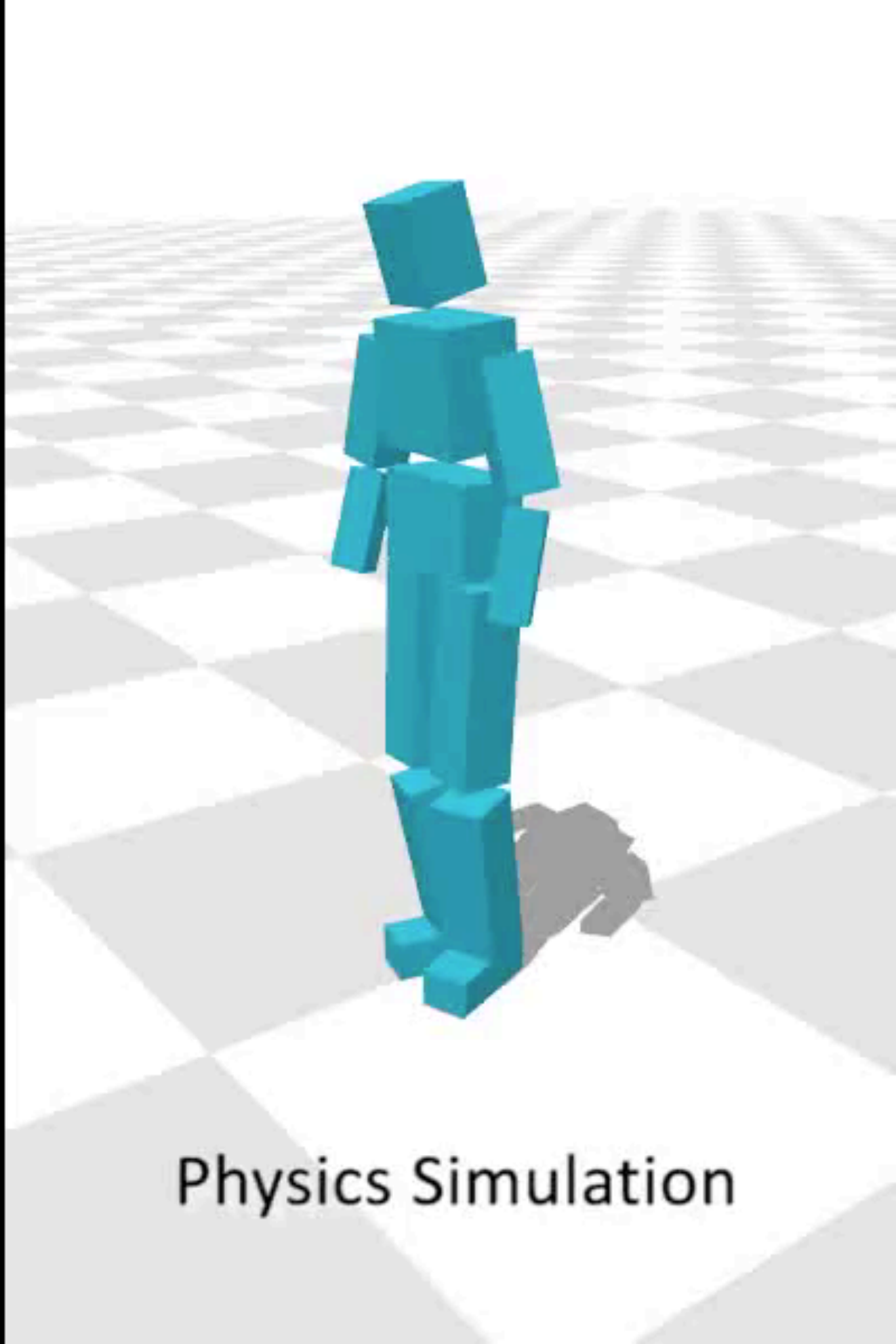
Sok et al, Simulating Biped Behaviors from Human Motion Data



[SIGGRAPH 2010] Lee et al, Data-driven biped control



Motion Capture



Physics Simulation

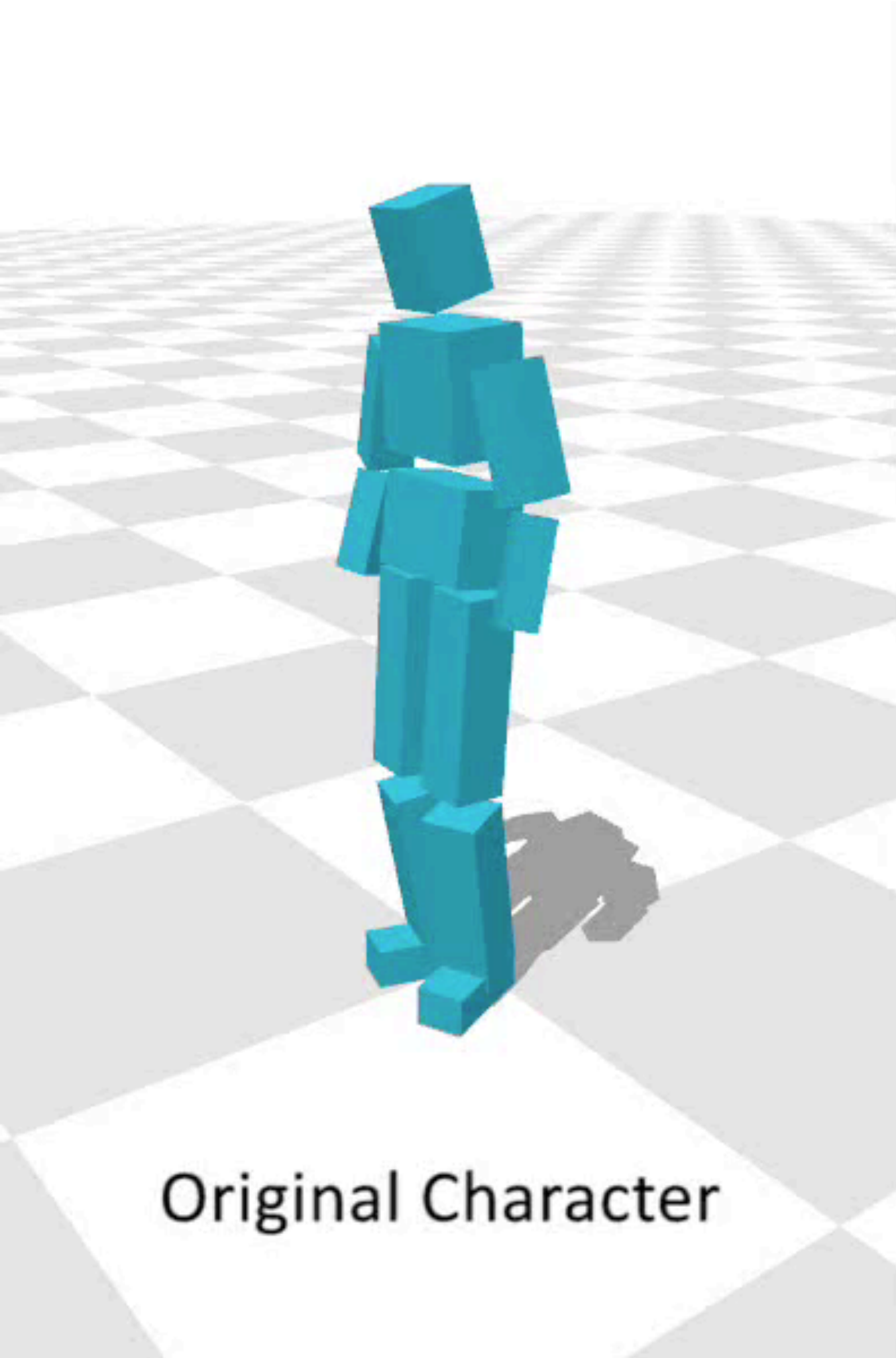
x 1



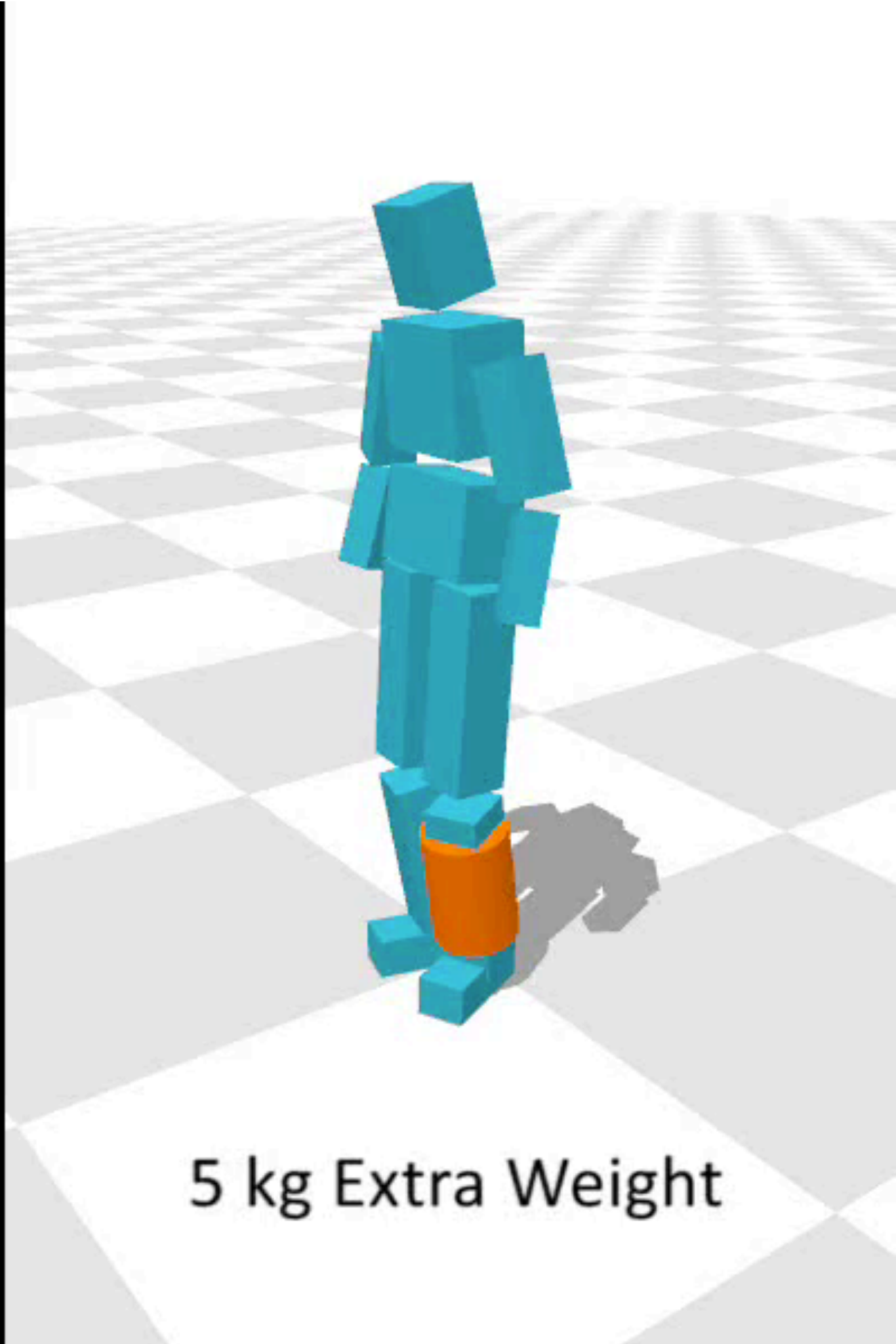
Motion Capture



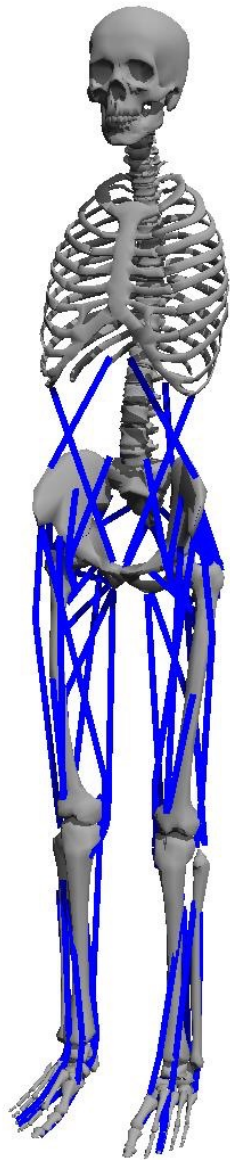
Physics Simulation



Original Character



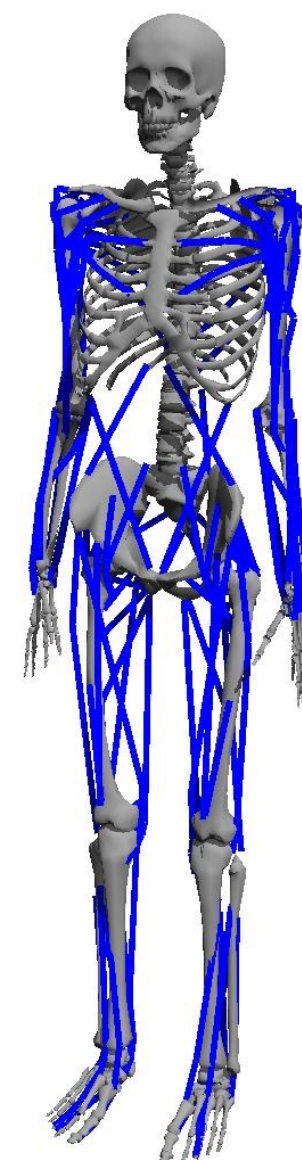
5 kg Extra Weight



Gait2562
(25 DOFs, 62 muscles)



Gait2592
(25 DOFs, 92 muscles)



Fullbody
(39 DOFs, 120 muscles)

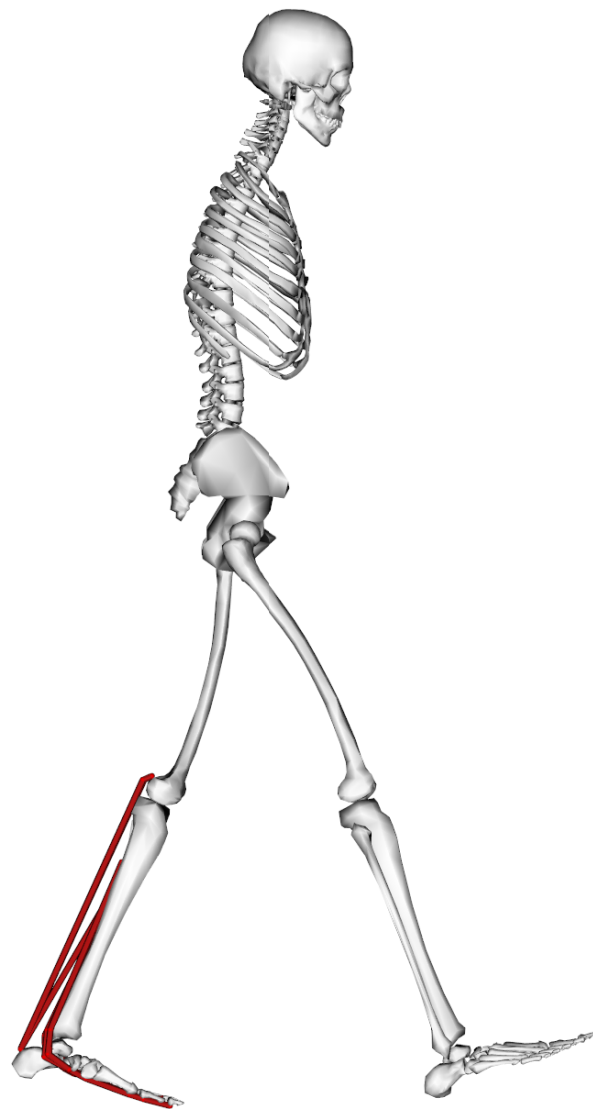
[SIGGRAPH Asia 2014] Lee et al, Many-Muscle Humanoids

Motion Capture Reference

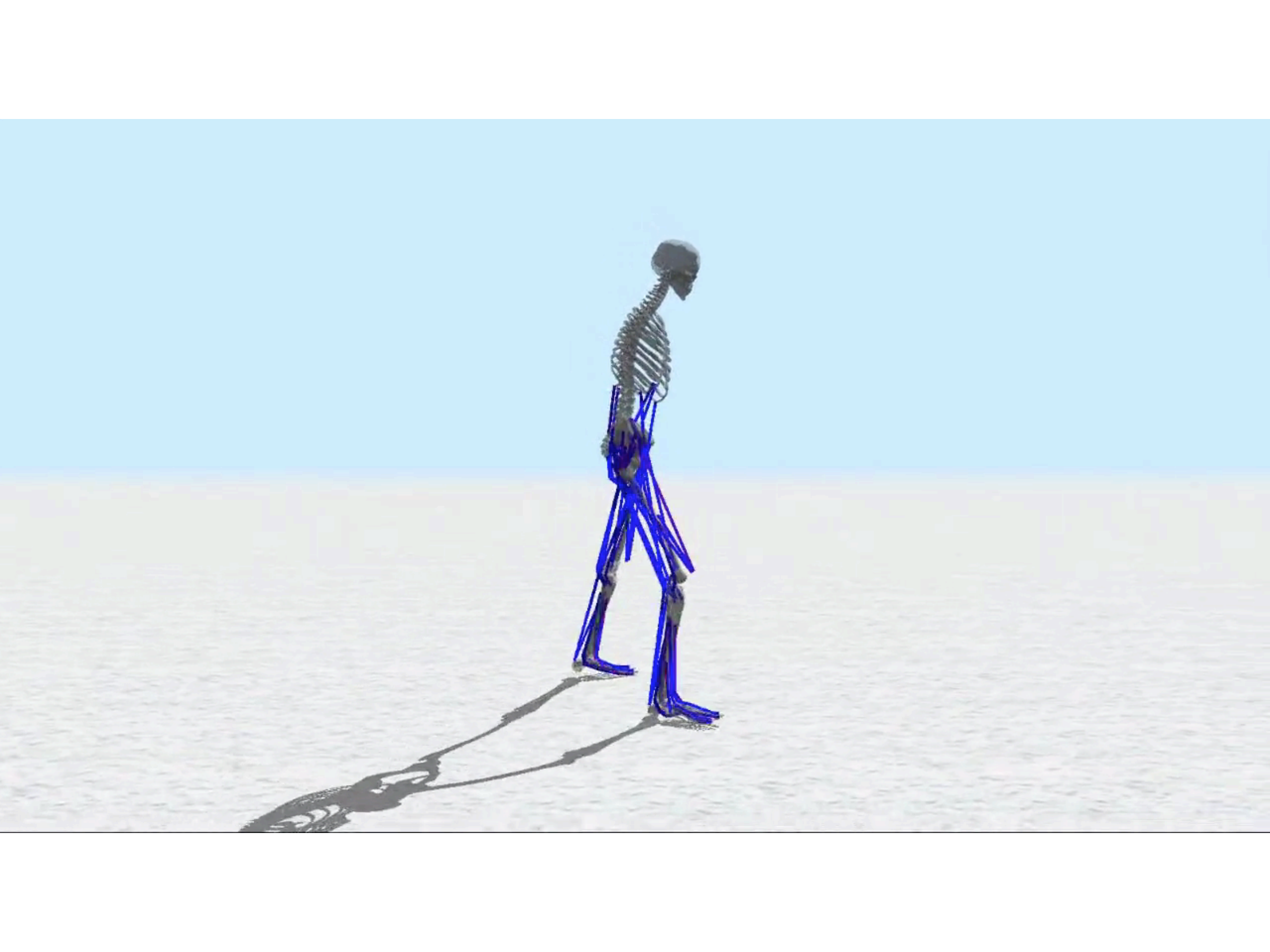
slow run

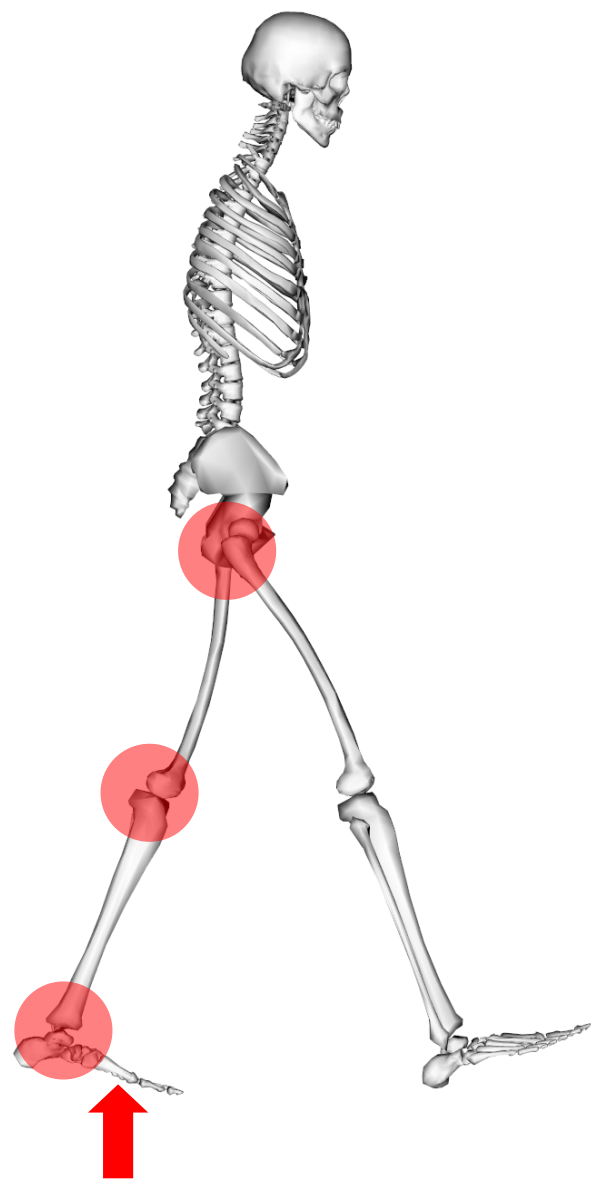


[SIGGRAPH Asia 2014] Lee et al, Many-Muscle Humanoids

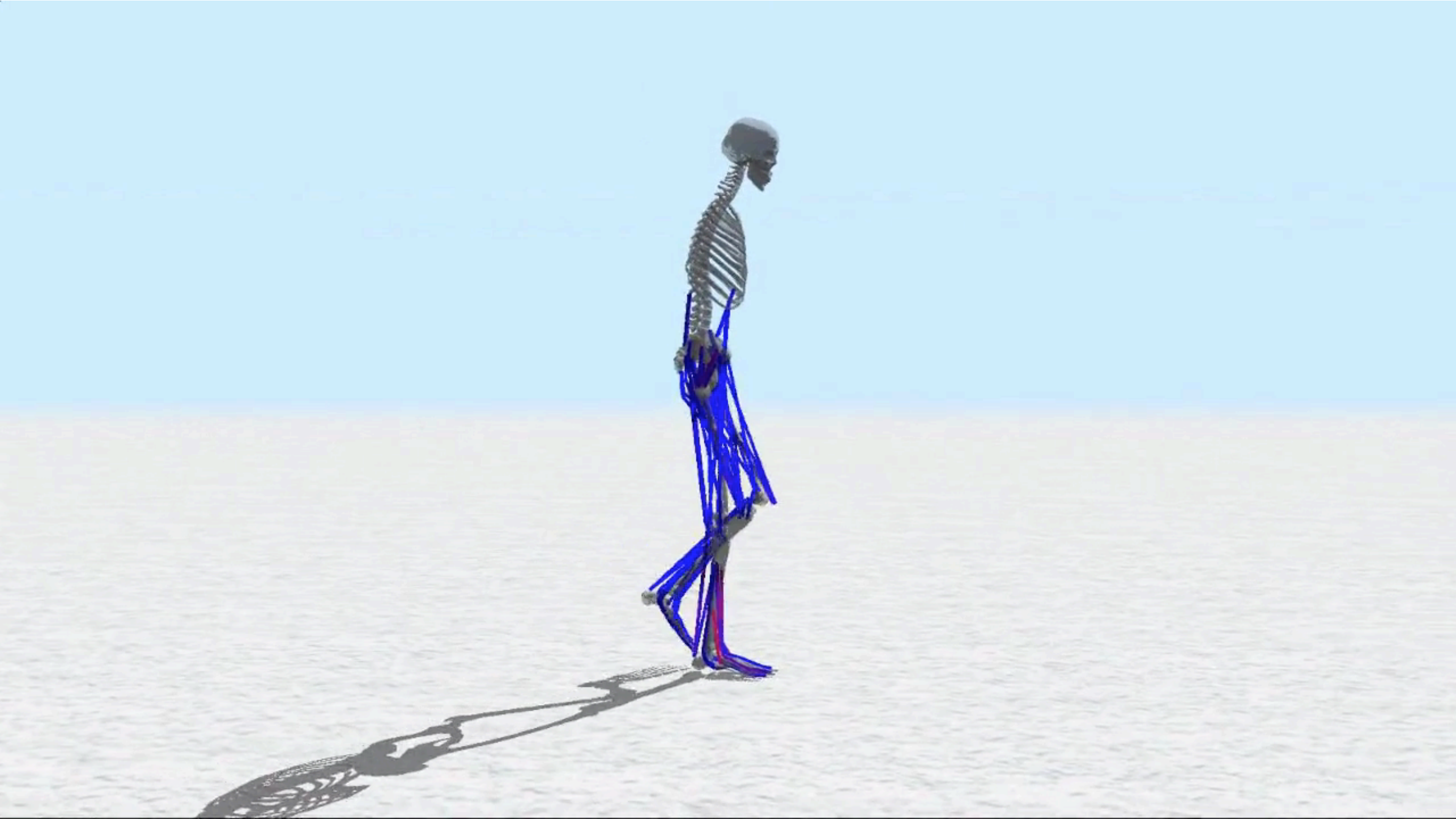


종아리 근육이 약화되면, ...

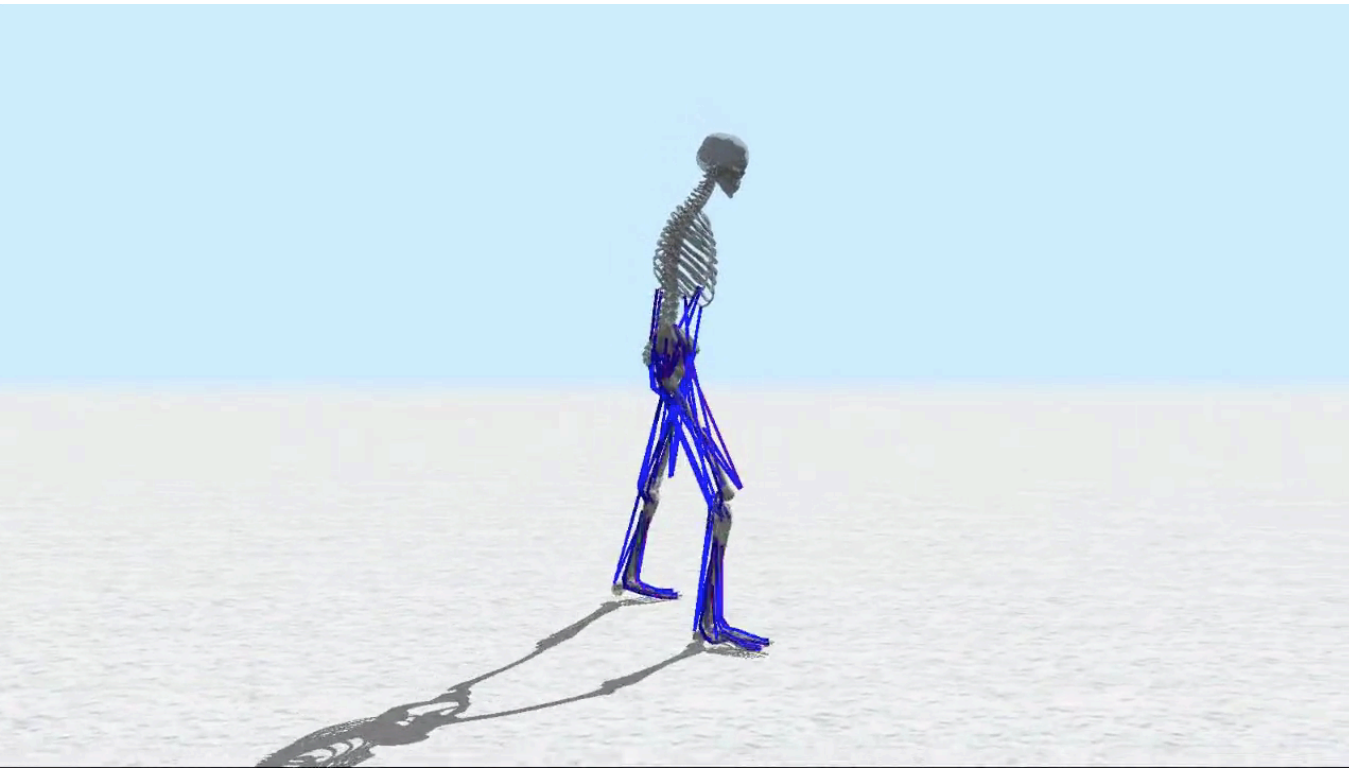




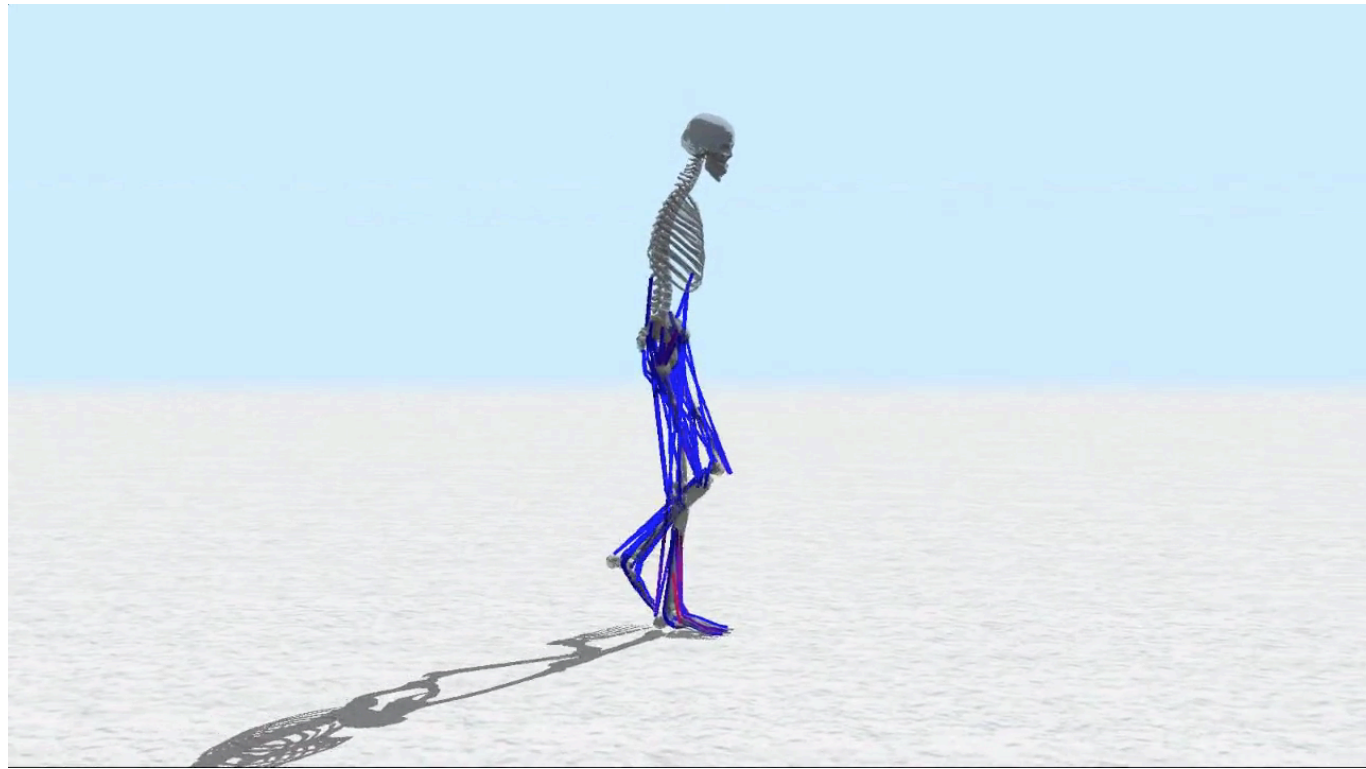
관절에 통증이 있으면, ...



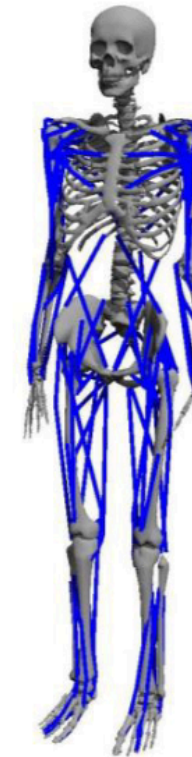
Painful Left Ankle Plantar Flexor



Painful Joints on Left Leg



보행 시뮬레이션을 위한 딥러닝



1980년대 – 2000년대
단순화된 모델
Inverted pendulum
로봇의 걸음걸이
DOF < 5

2007년
Fullbody dynamics
Motor/torque control
사람과 같은 걸음걸이
DOF < 10

2010년
Fullbody dynamics
Motor/torque control
다양한 동작
DOF < 100

2014년
근골격구조
선형 Hill type
생체역학적 동작
DOF < 1000

2017년 이후
근골격+피부
Volumetric FEM
해부학적 정밀도
DOF > 10000

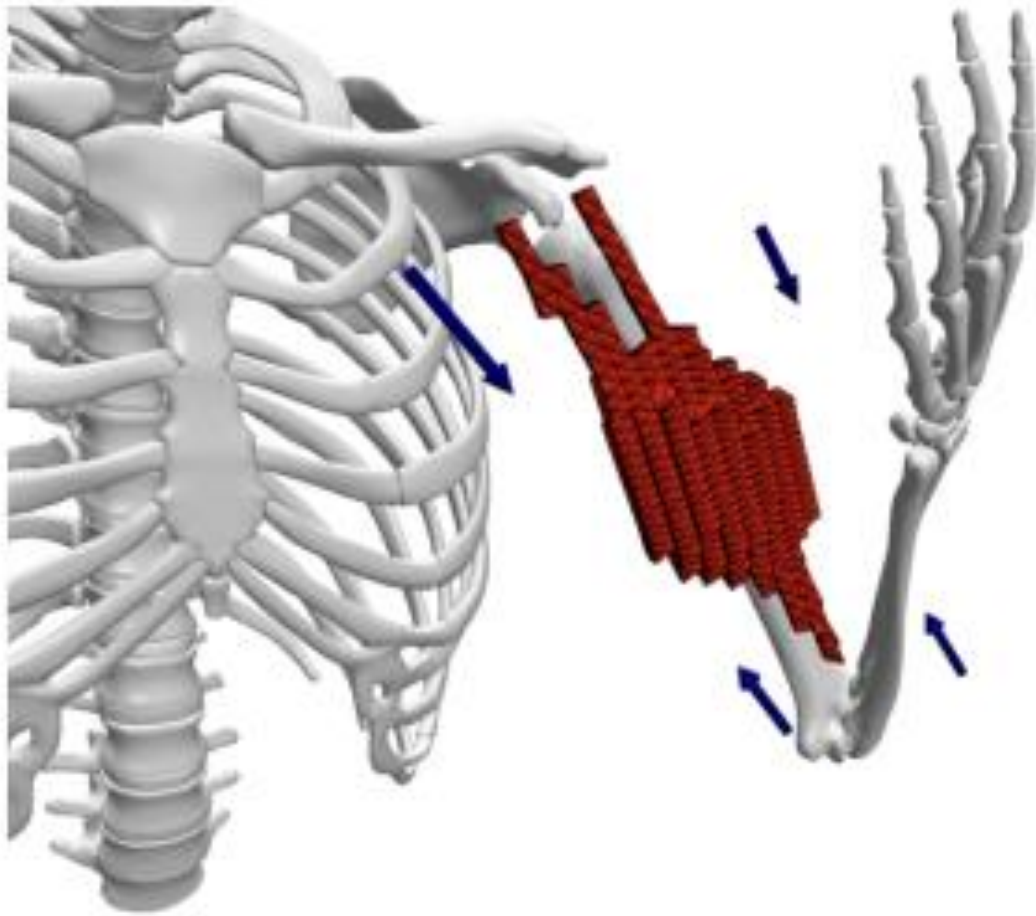
Closed-form solution

Feedback control, Optimization, Data-driven

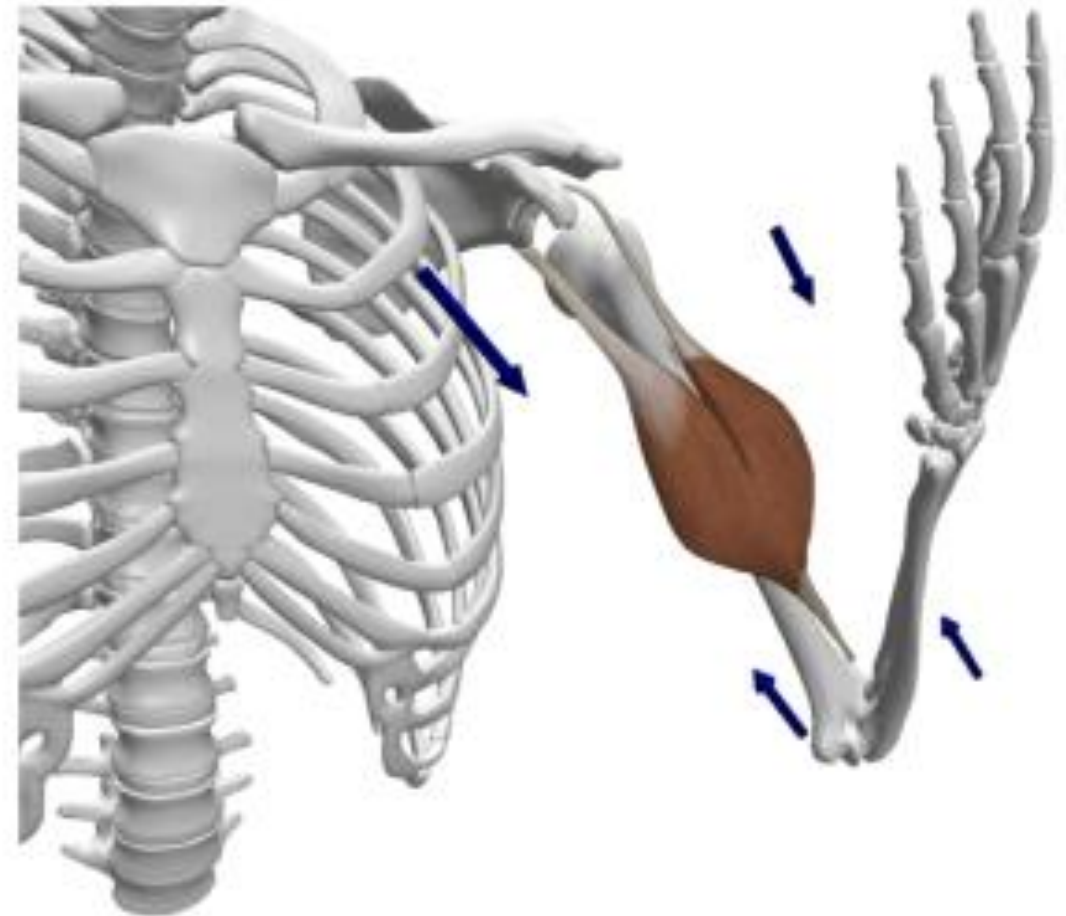
Direct policy learning
Local regression

Reinforcement learning
Deep neural network
End-to-end learning

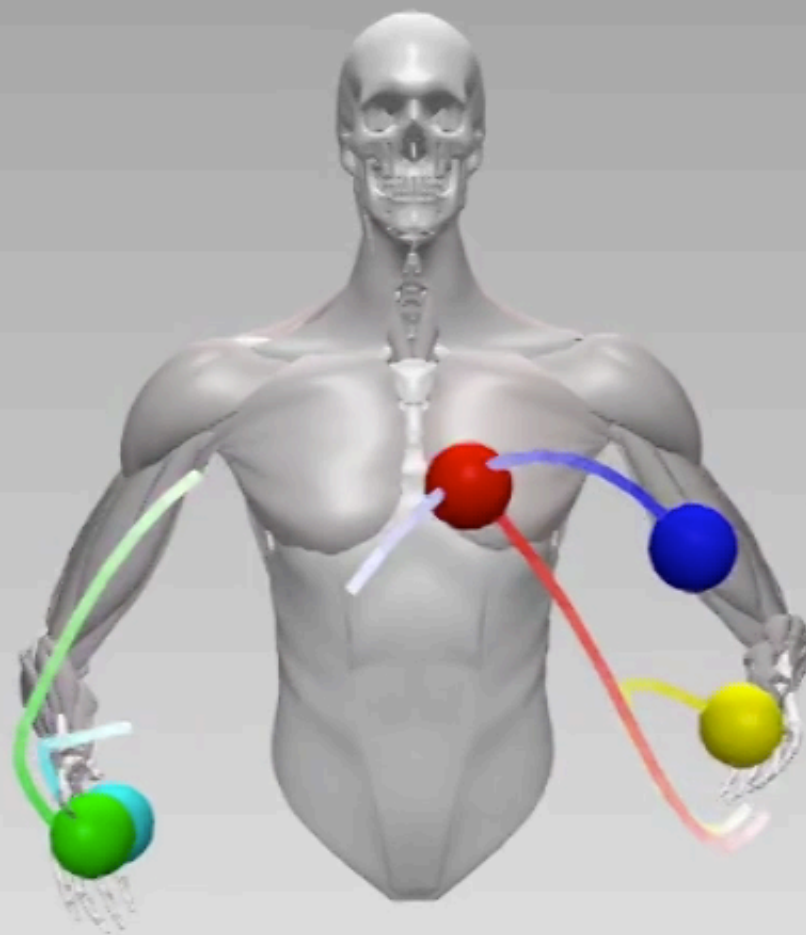
근골격신경 모델링

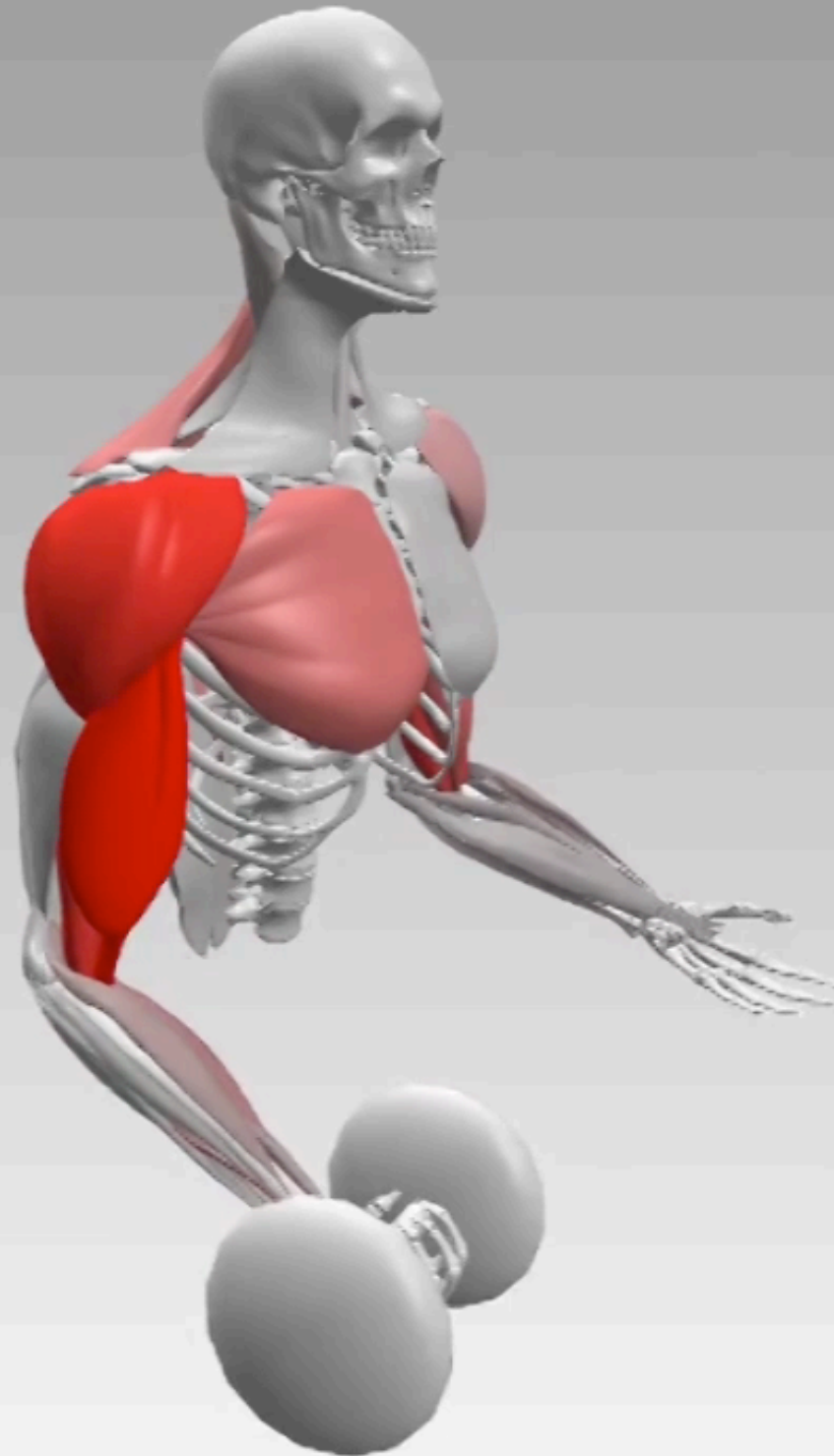
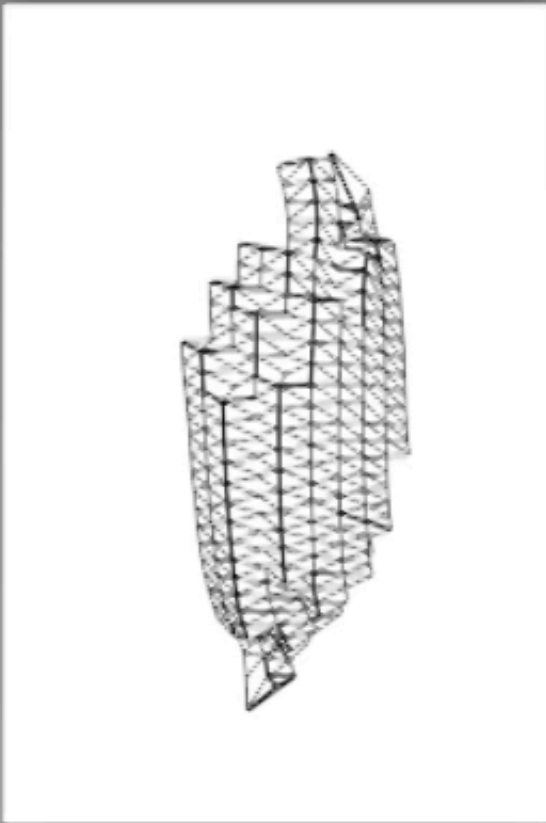


FEM simulation



Muscle contraction



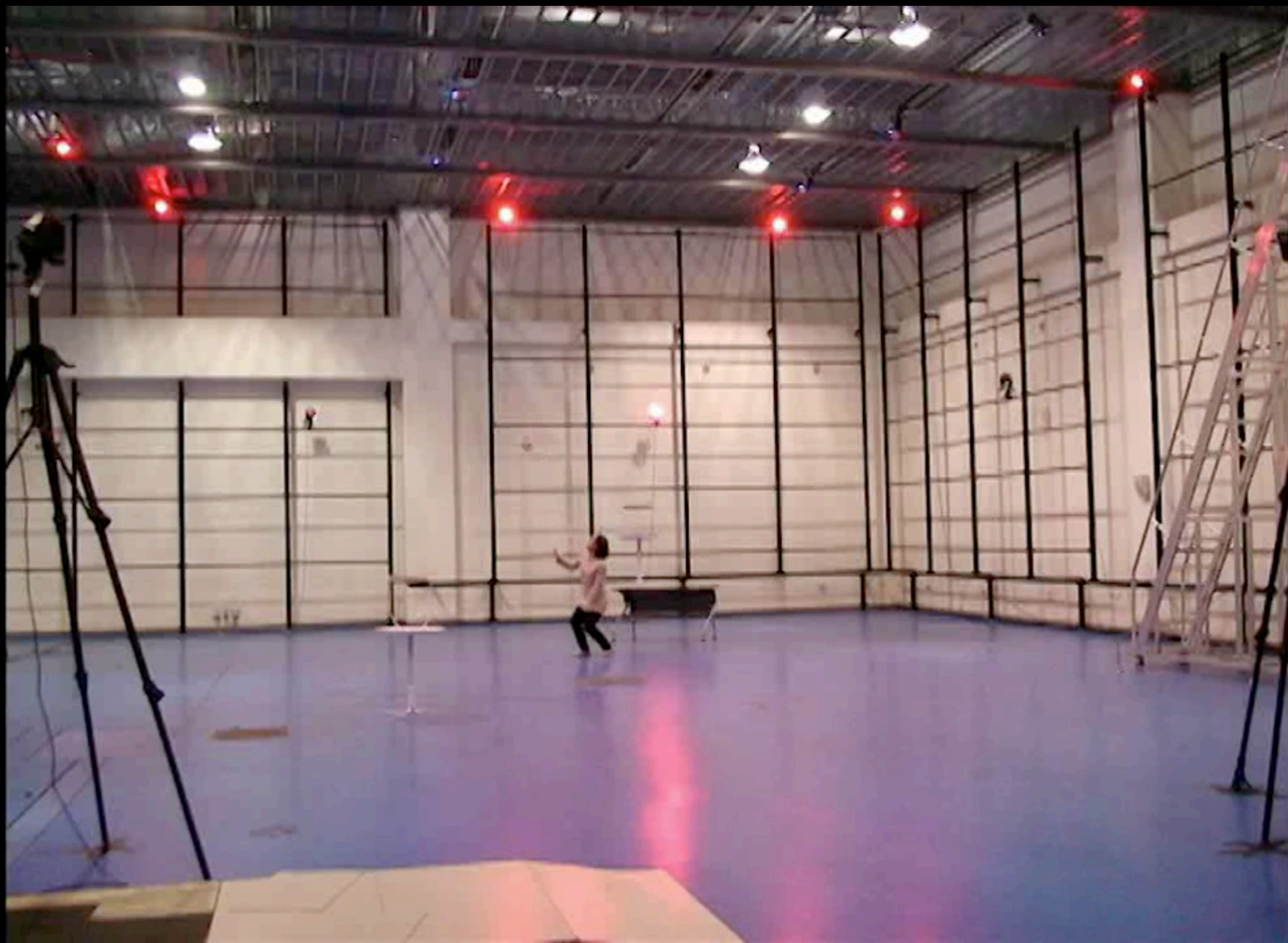


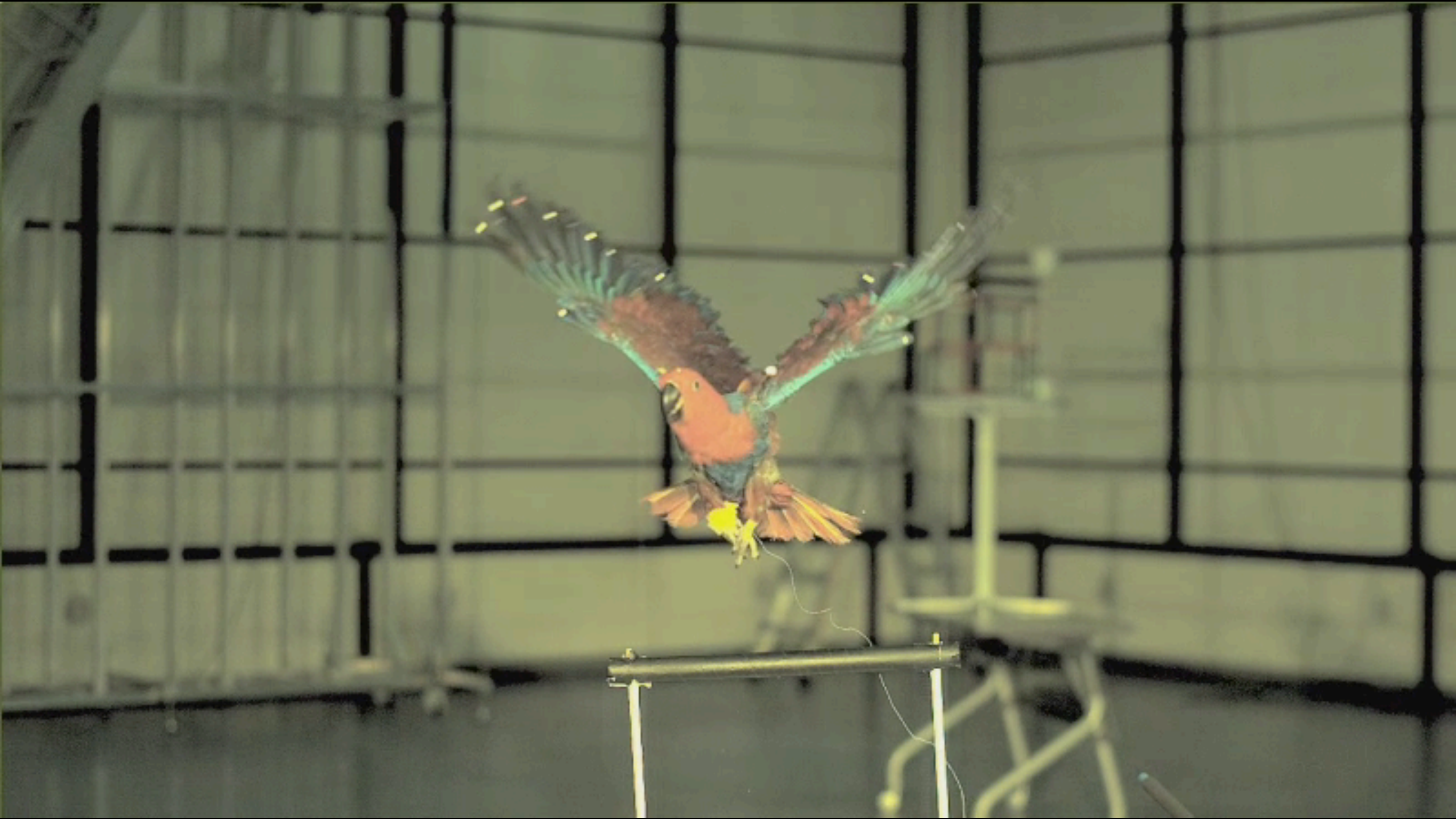
Data-driven Control of Flapping Flight



Collaboration with Eunjung Ju, Jungdam Won,
Byungkuk Choi, Junyong Noh, Min Gyu Choi
(ACM Transactions on Graphics 2013, presented at SIGGRAPH 2014)



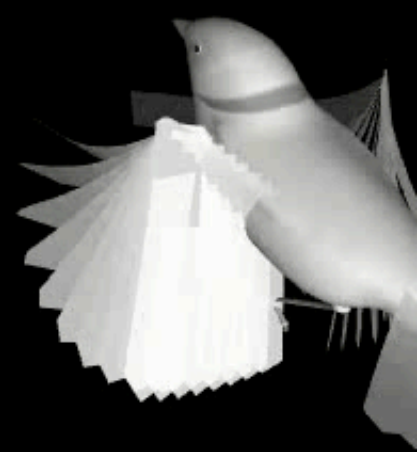










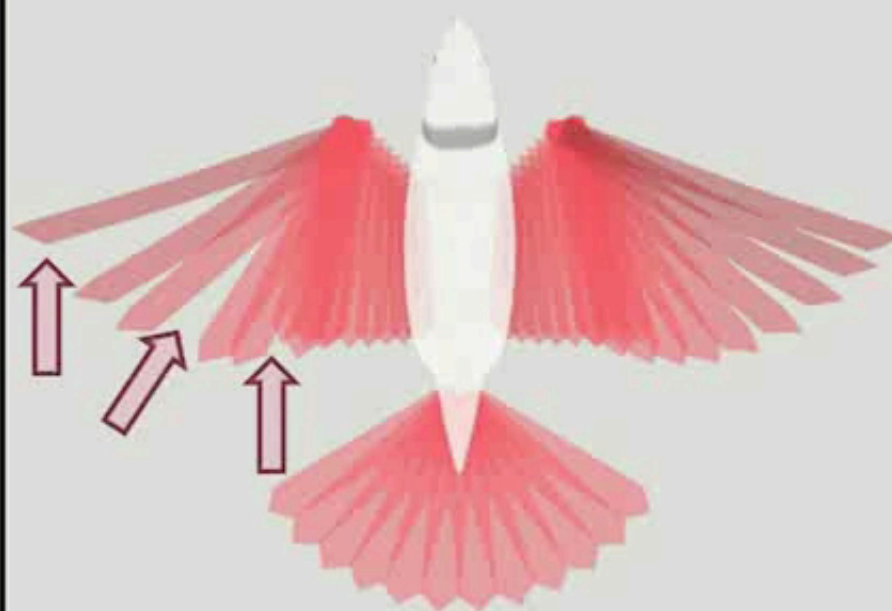


x 1/8

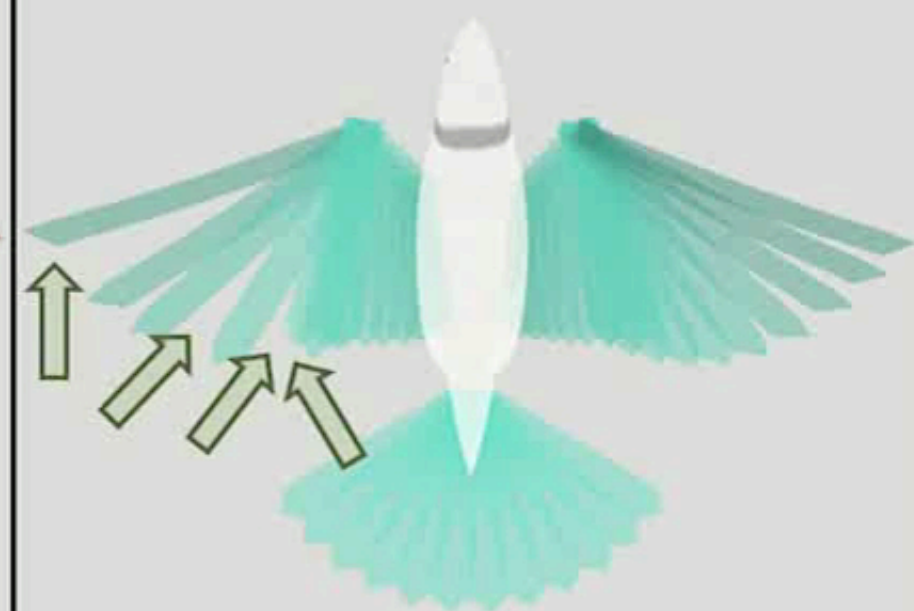




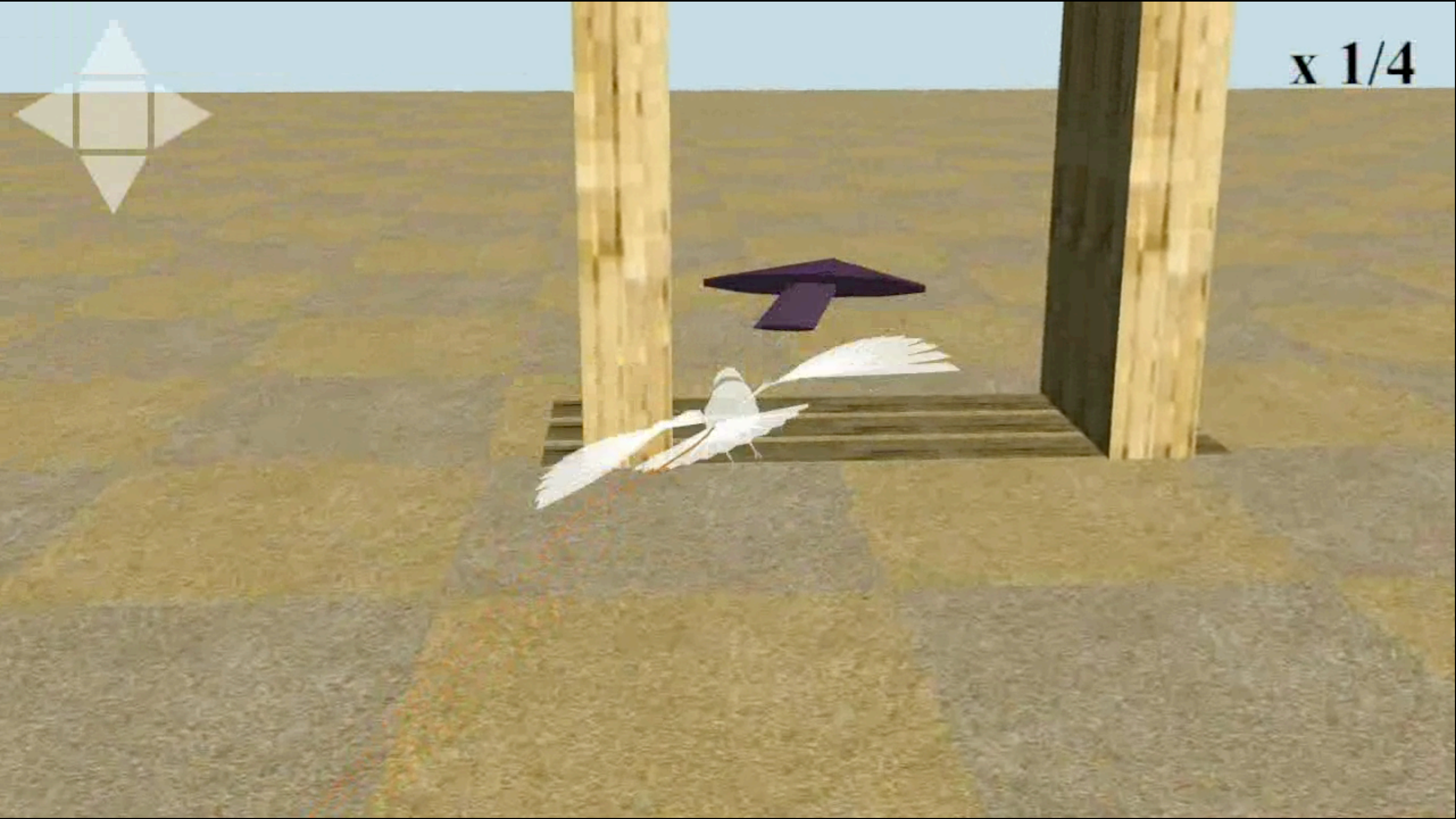
no missing feathers



3 feathers plucked

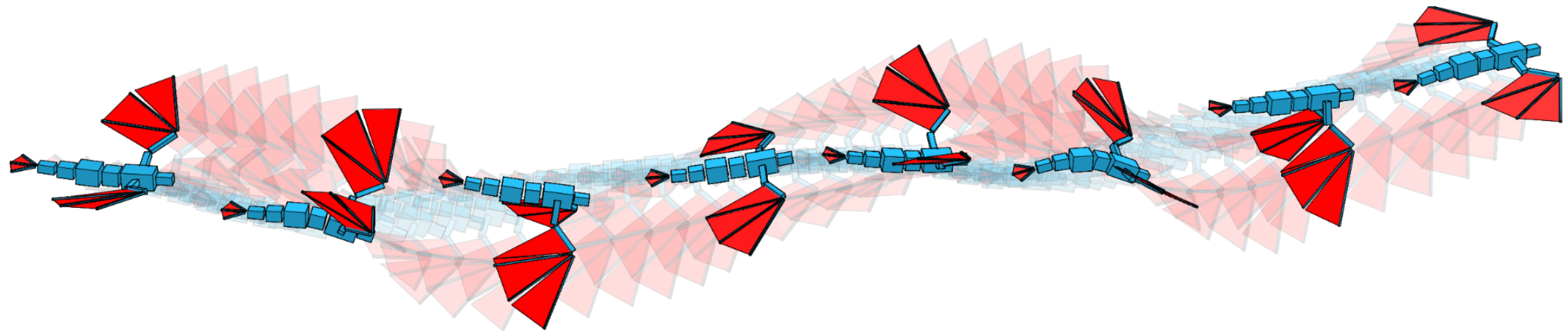


4 feathers plucked



x 1/4

How To Train Your Dragon: Example-Guided Control of Flapping Flight

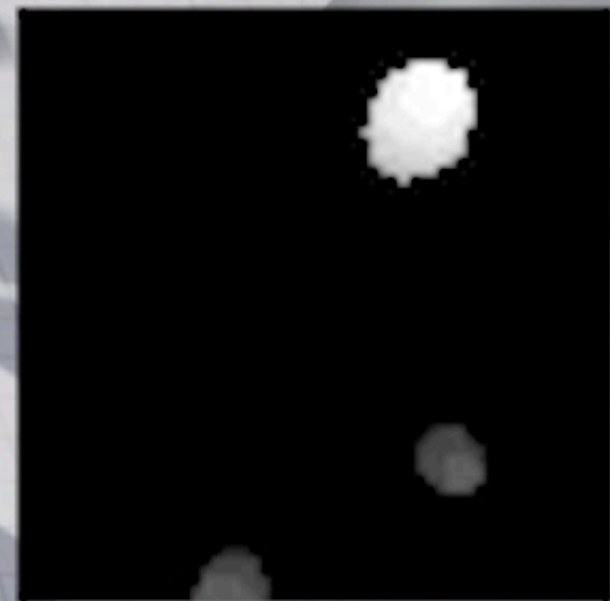
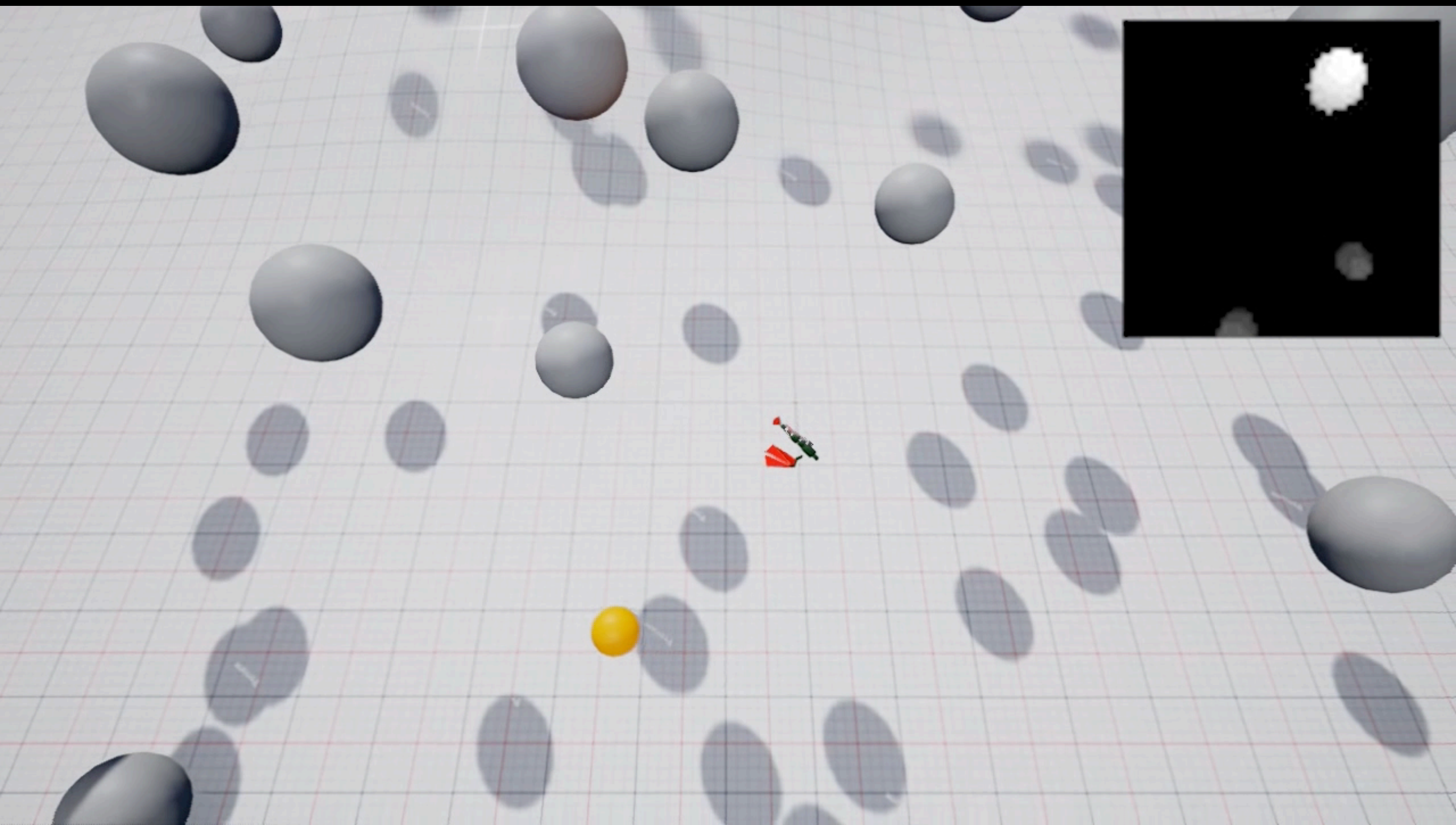


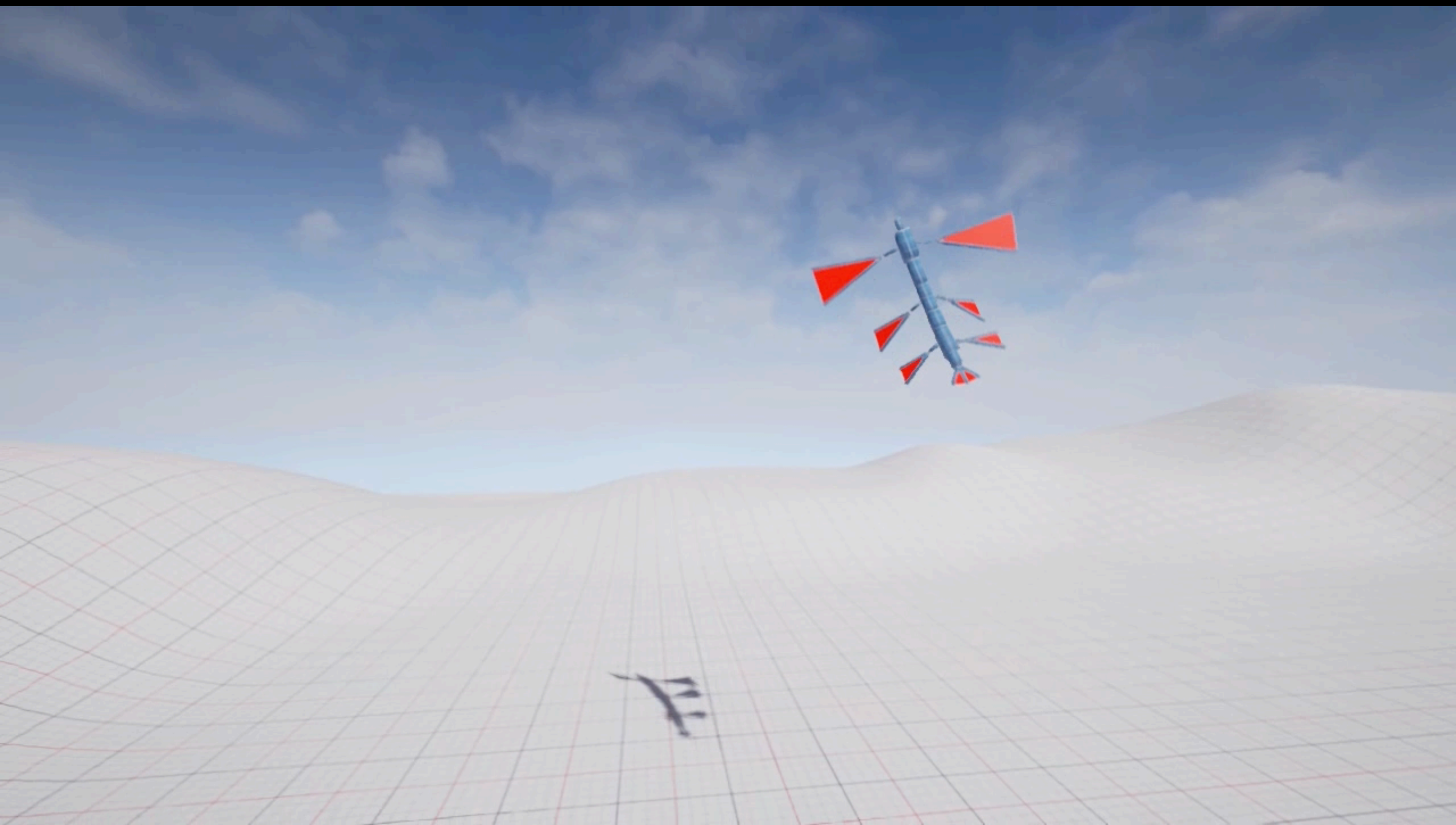
Jungdam Won Jongho Park Kwanyu Kim Jehee Lee

Seoul National University

[SIGGRAPH Asia 2017]





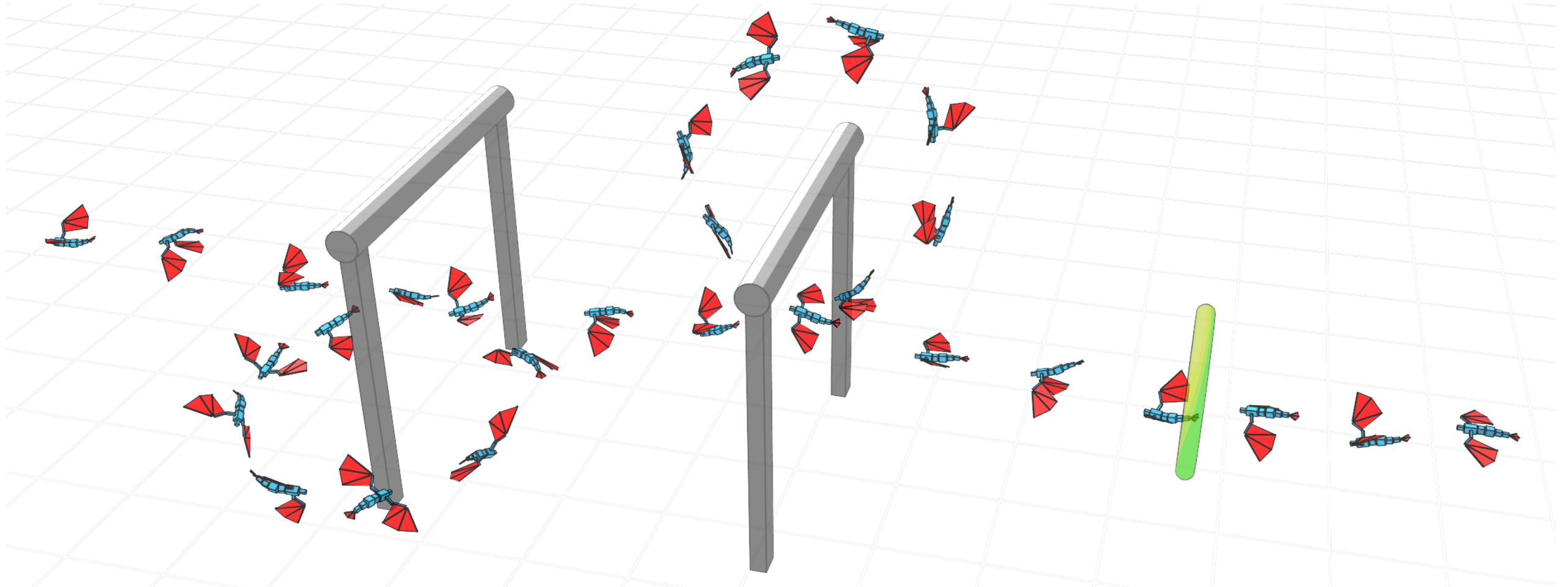








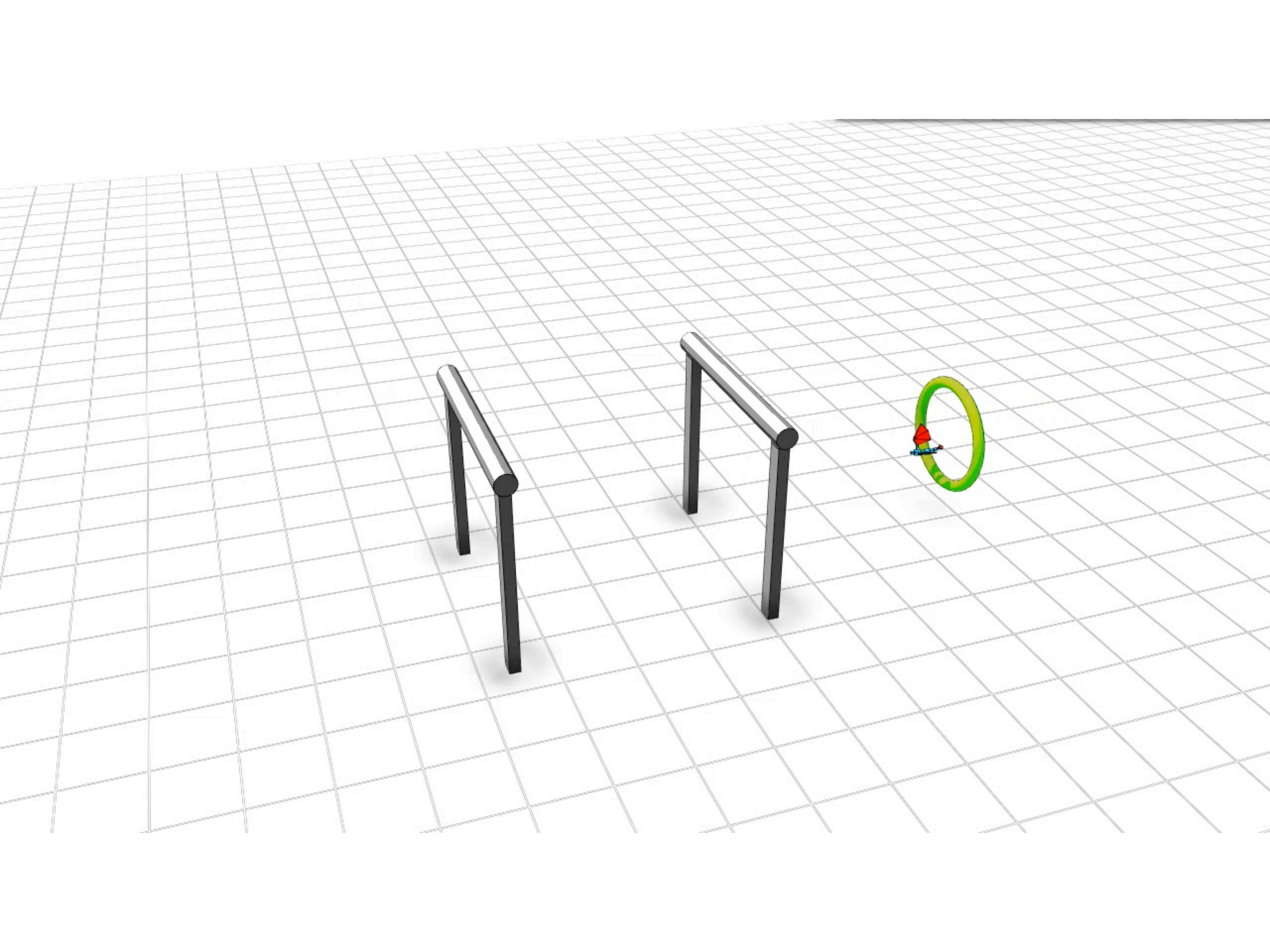
Aerobatic Control of Flying Creatures via Self-Regulated Learning

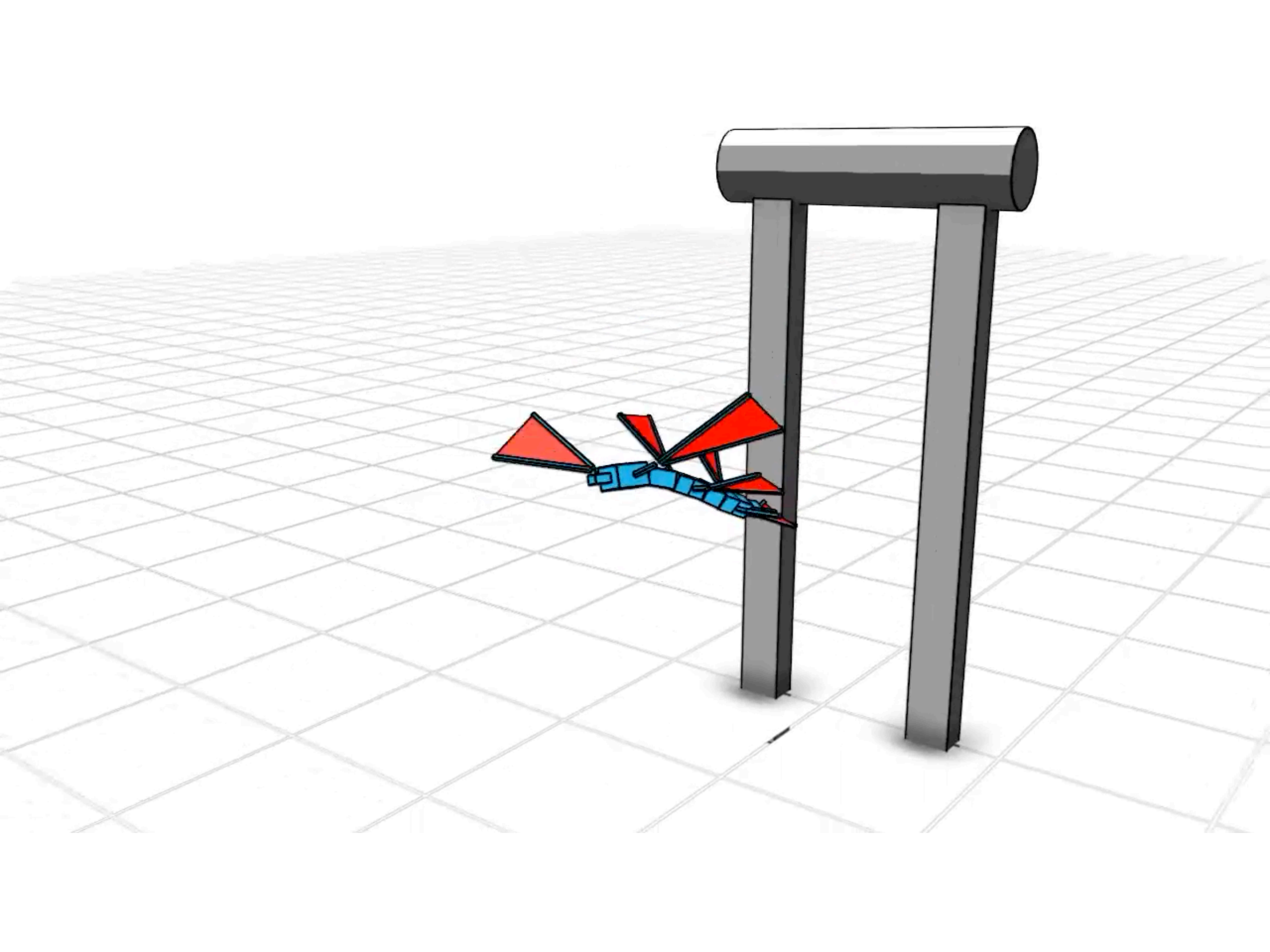


Jungdam Won Jungnam Park Jehee Lee

Seoul National University

[SIGGRAPH Asia 2018]





The papers and videos are available at
SNU Movement Research Lab at <http://mrl.snu.ac.kr>



Collaborators

Jungdam Won, Kyungho Lee, Seunghwan Lee, Ri Yu, Jungnam Park, Seyoung Lee,
Yoonsang Lee, Munseok Park, Jongho Park, Kwanyu Kim, Eunjung Ju, Manmyung Kim,
Kwangwon Sok, and many others