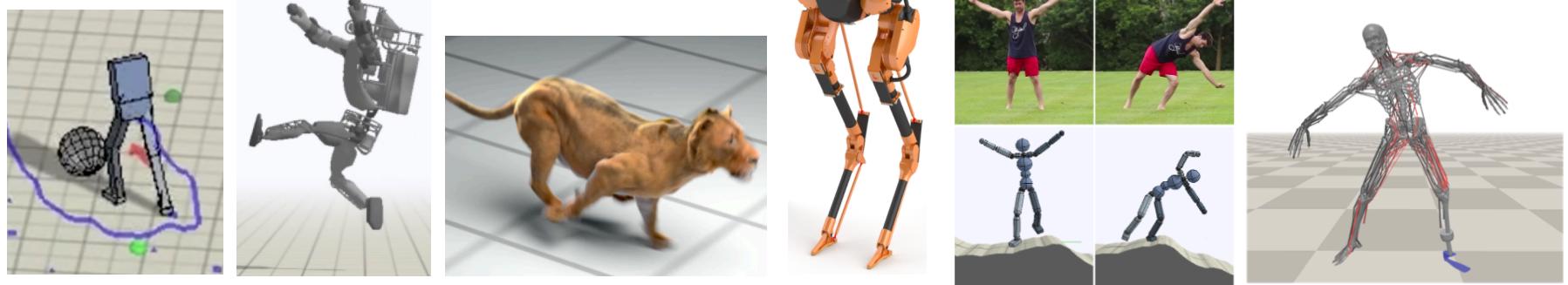


(LEARNED) PHYSICS-BASED HUMAN MOVEMENT: SHARED MODELS FOR ANIMATION, ROBOTICS, VISION, AND BIOMECHANICS

Michiel van de Panne

Dept of Computer Science
University of British Columbia

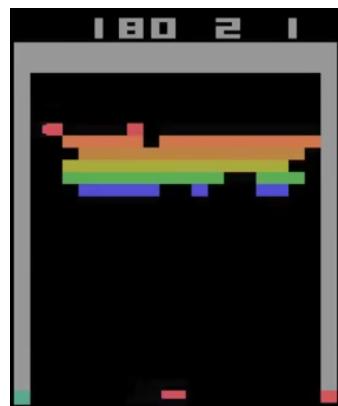
“The best way to predict the future is to create it.”



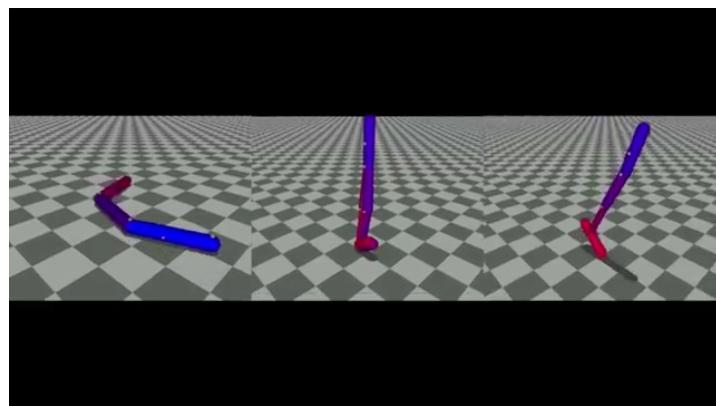
Dynamic Motion Control



DEEP RL RESULTS



[Mnih et al. 2015]

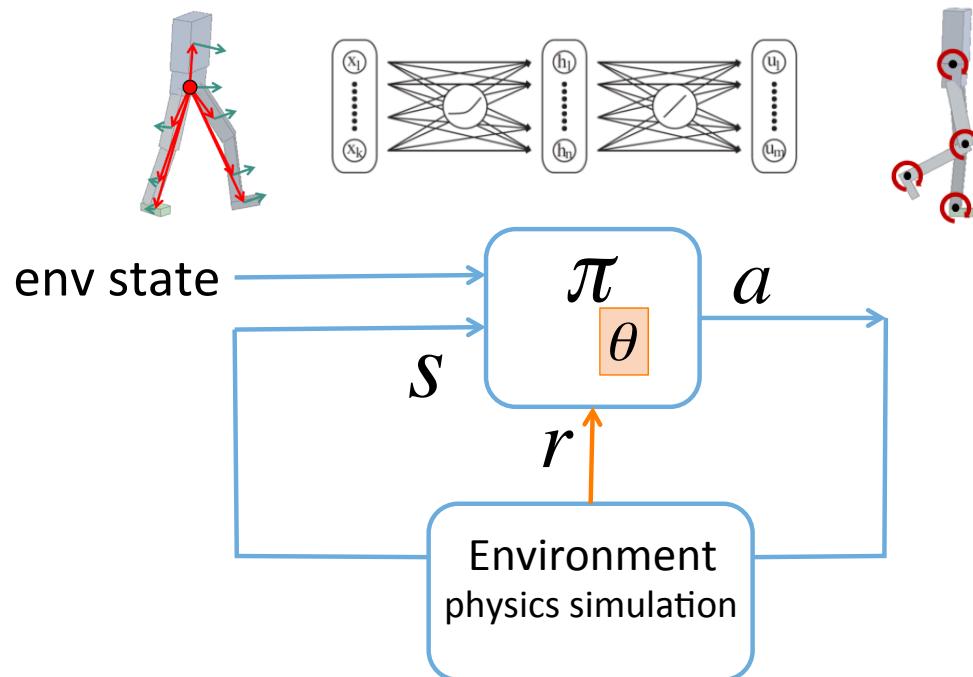


[Schulman et al. 2016]



[Chebotar et al. 2017]

REINFORCEMENT LEARNING FOR LOCOMOTION CONTROL

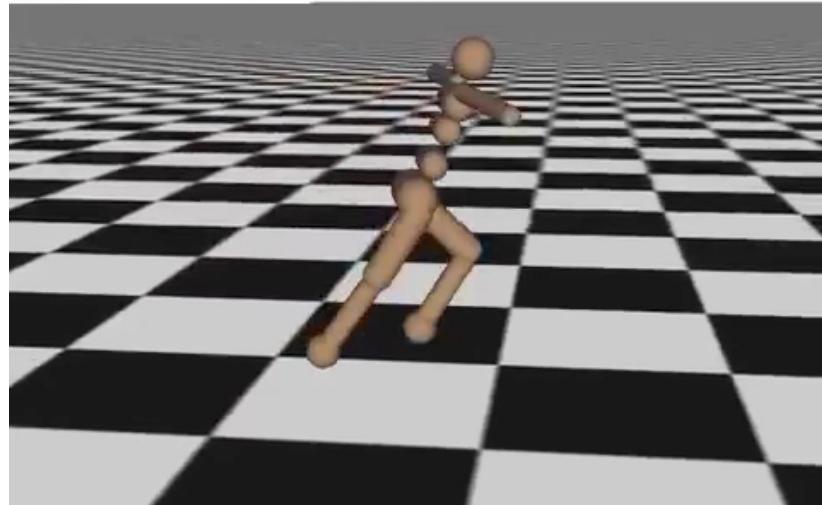


In principle:

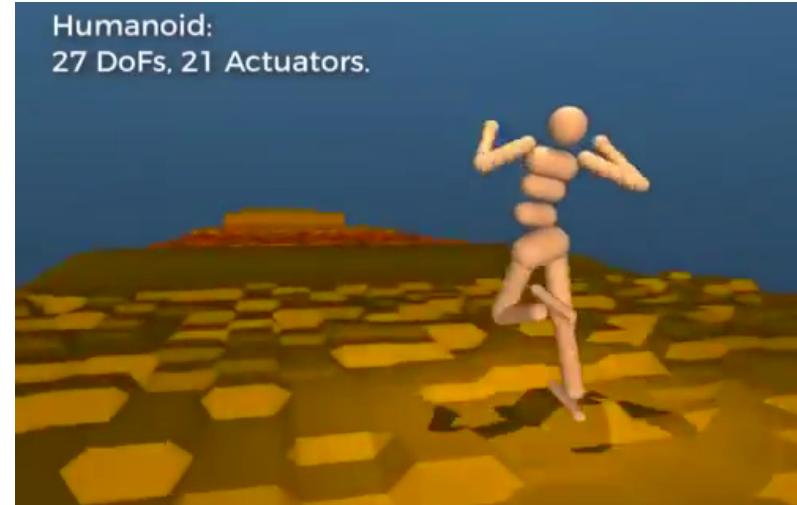
- specify rewards
- “train” using RL algorithm

$$\max_{\theta} \mathbb{E}\left[\sum_{t=0}^H R(s_t) | \pi_\theta\right]$$

MOTION QUALITY

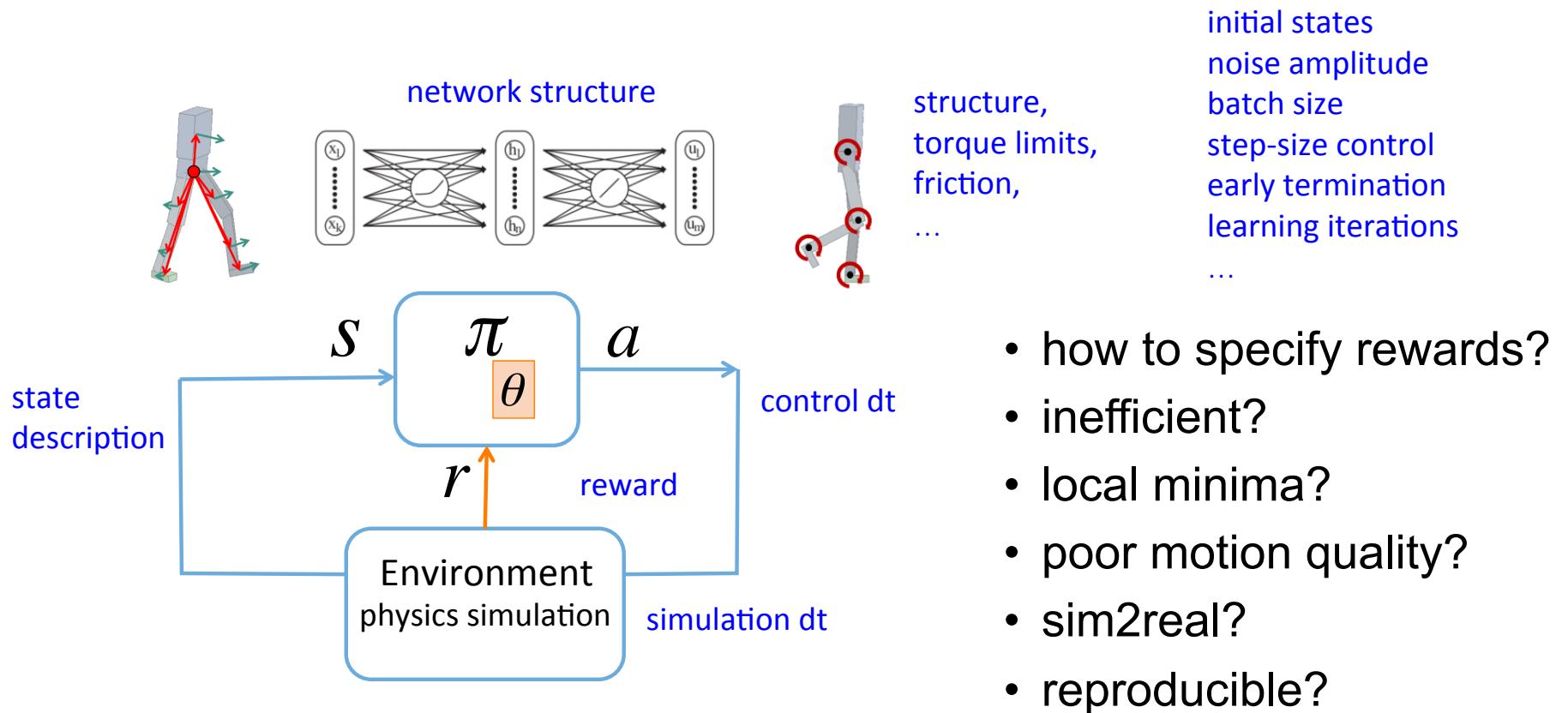


[Schulman et al. 2016]

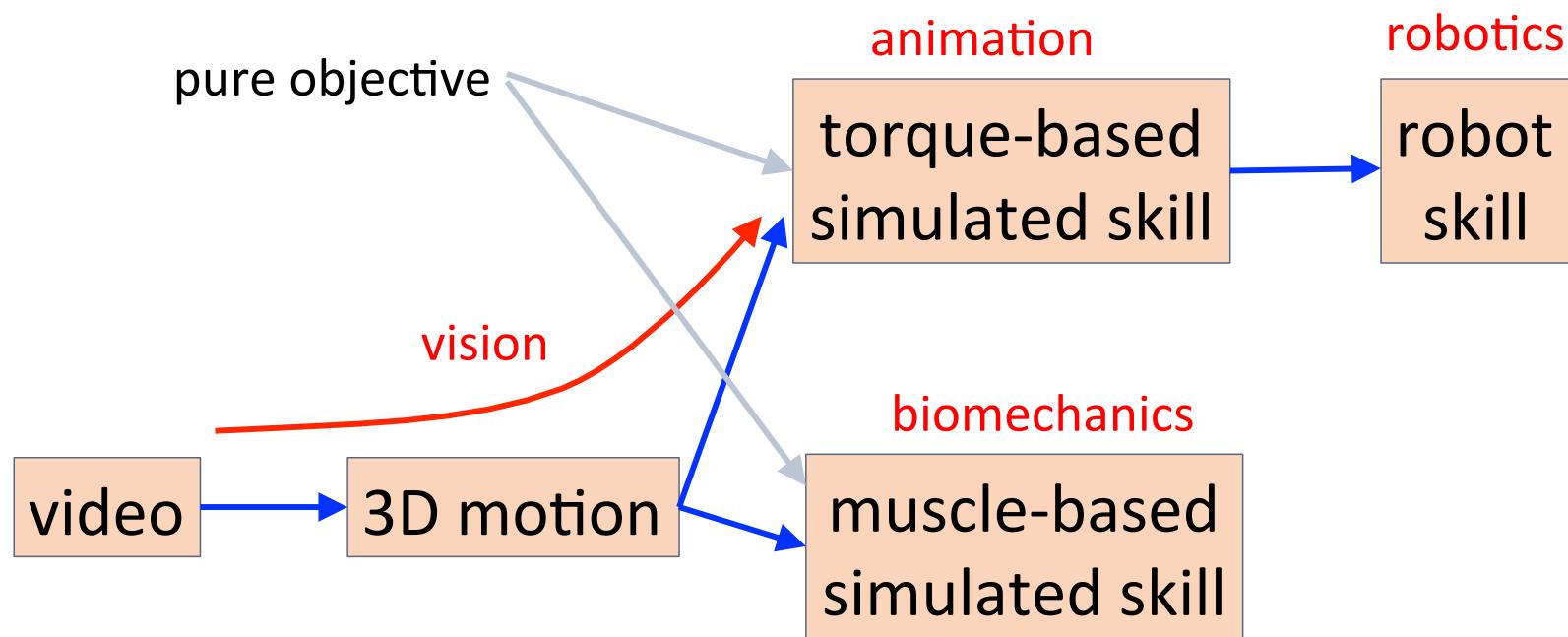


[Heess et al. 2017]

IN PRACTICE



OVERVIEW

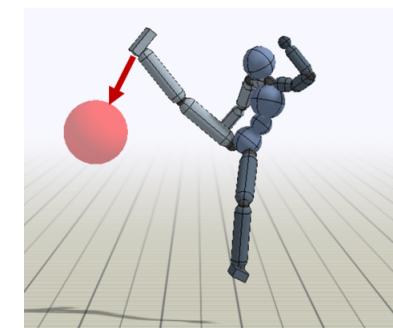
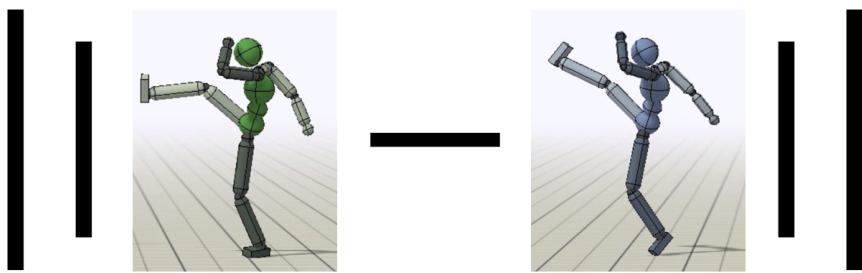


MOTION IMITATION

$$r_t = \omega^I r_t^I + \omega^G r_t^G$$

Imitation Objective

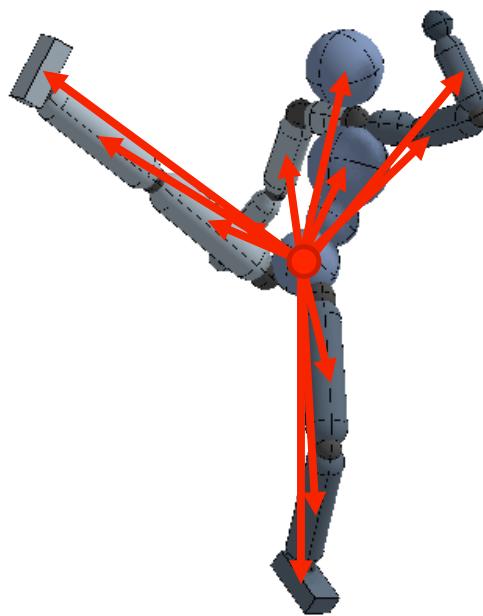
Task Objective



STATE + ACTION

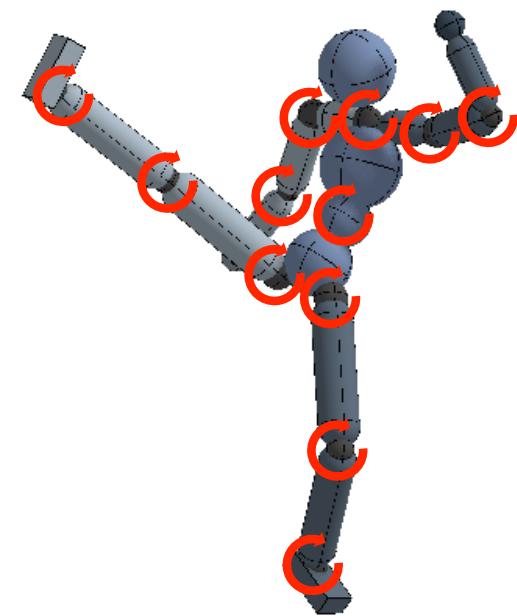
State: 197 D

- link positions
- link velocities



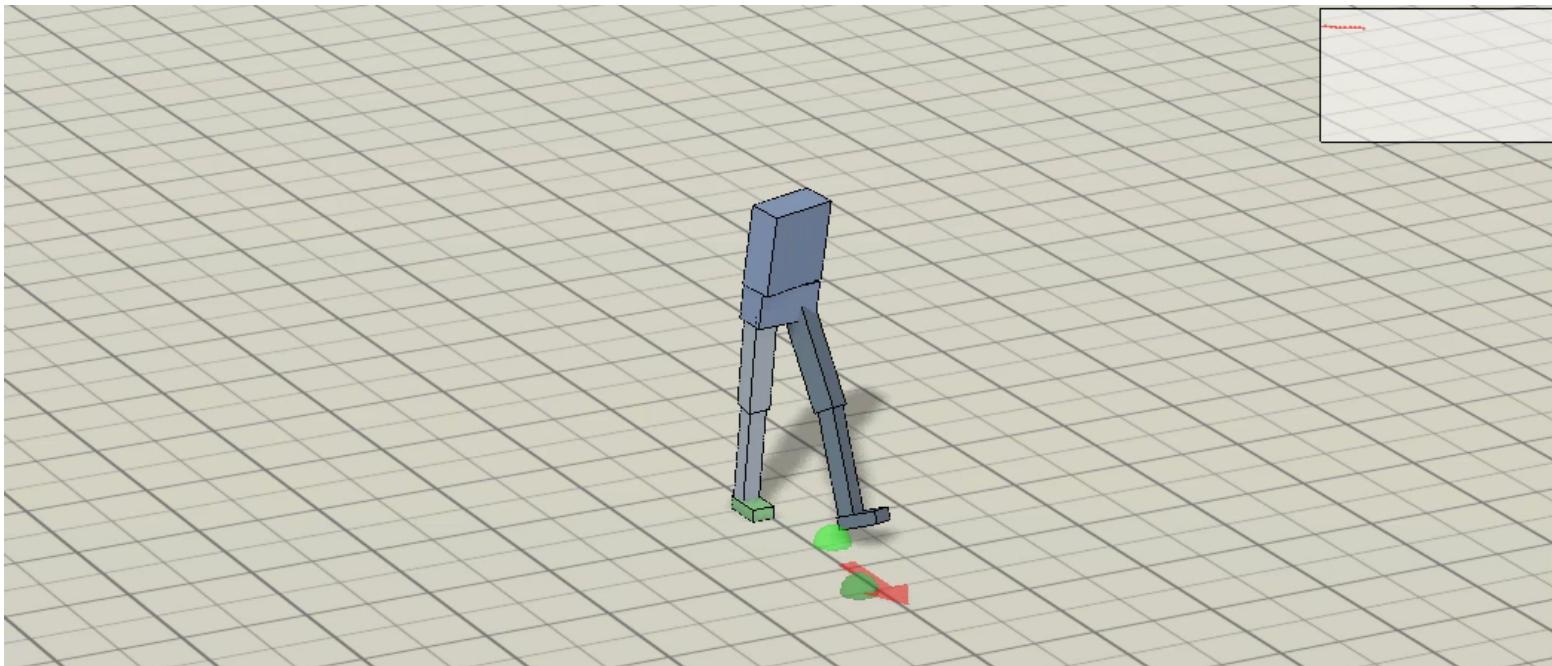
Action: 36 D

- PD targets



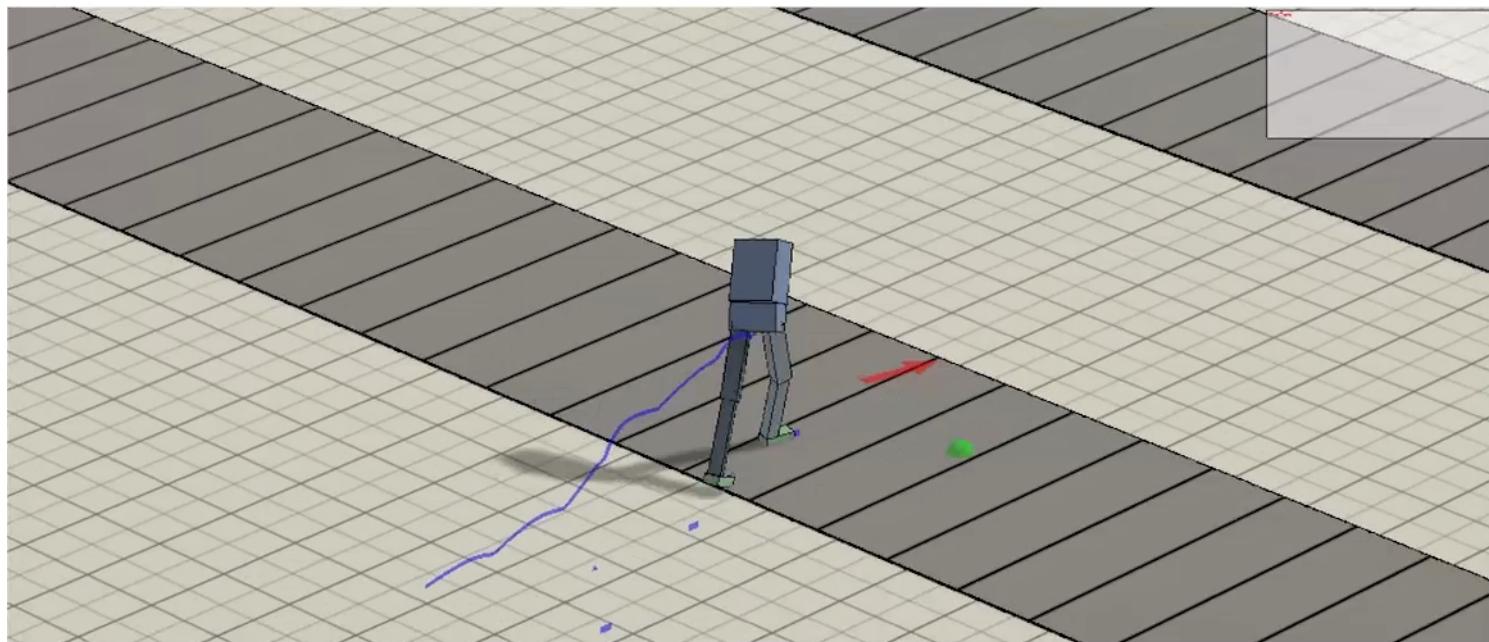
WALKING

[DeepLoco: SIGGRAPH 2017]

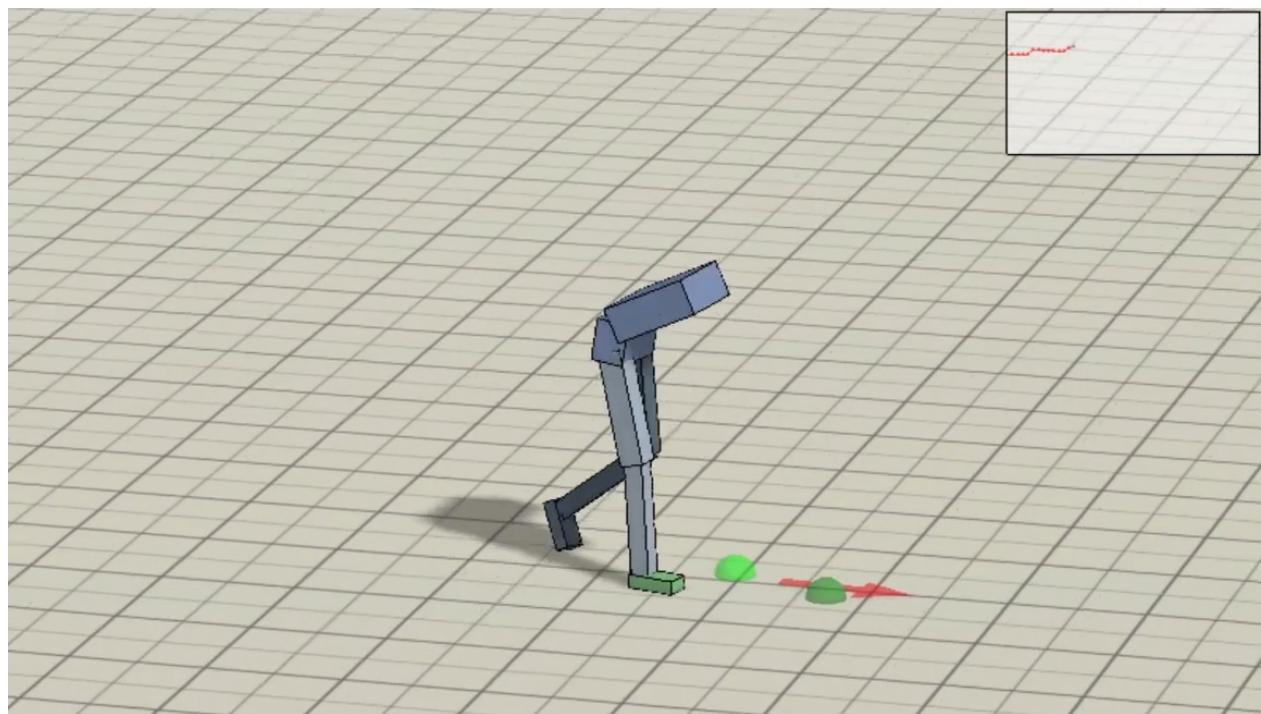


[DeepLoco: SIGGRAPH 2017]

Walking on Conveyor Belts



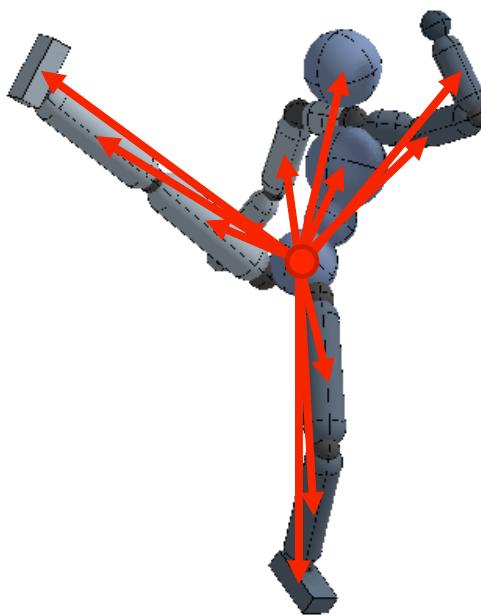
Forward Lean



STATE + ACTION

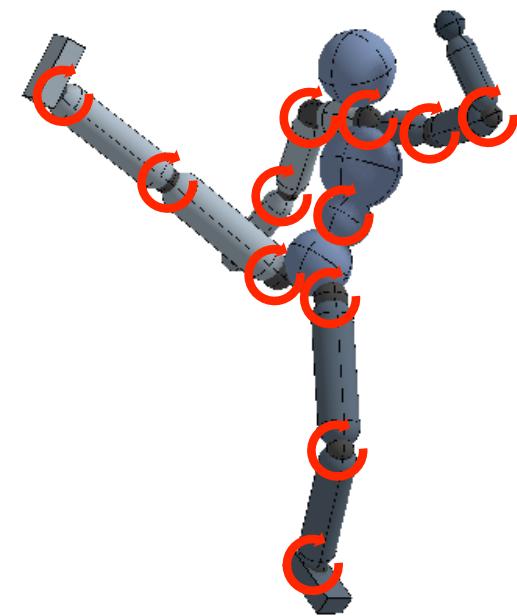
State: 197 D

- link positions
- link velocities

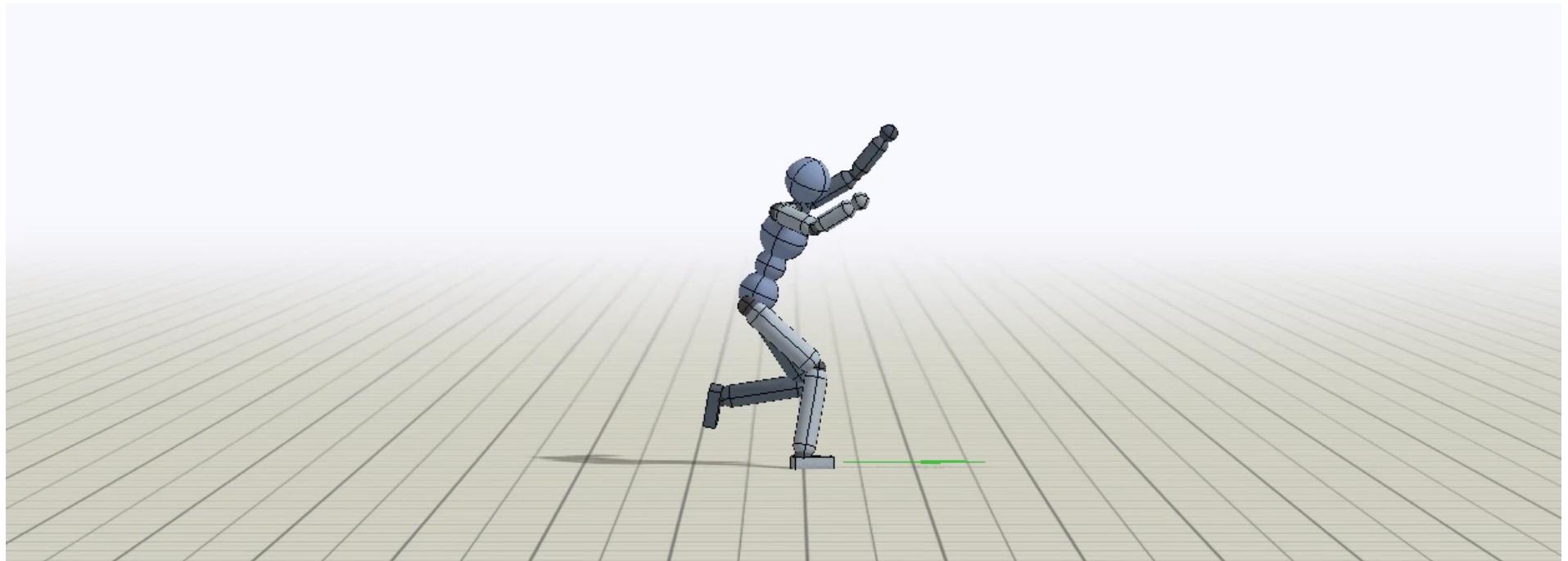


Action: 36 D

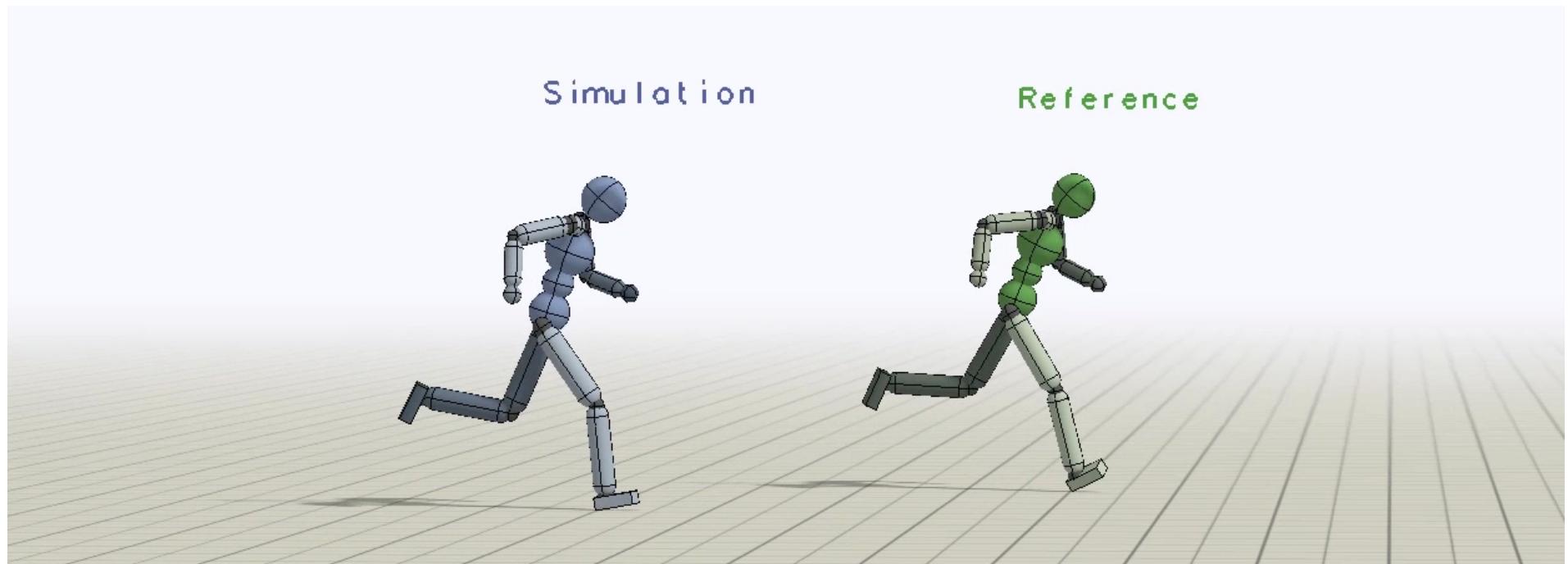
- PD targets



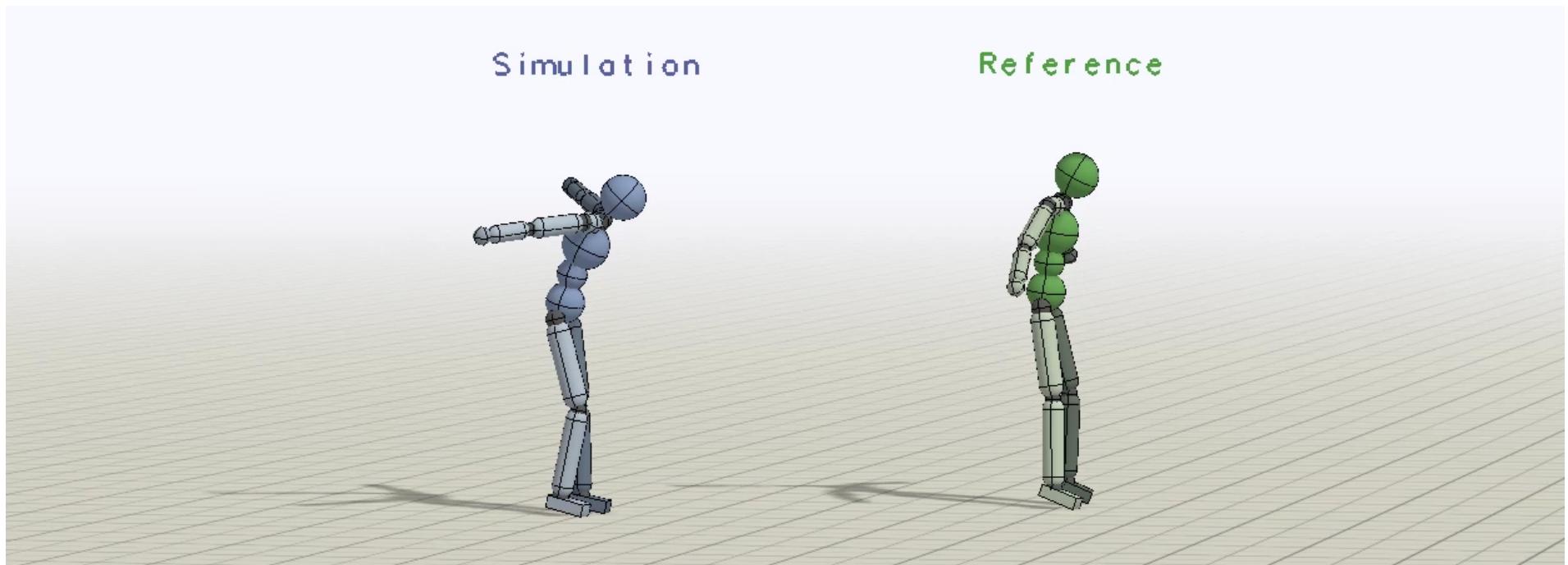
NO REFERENCE MOTION



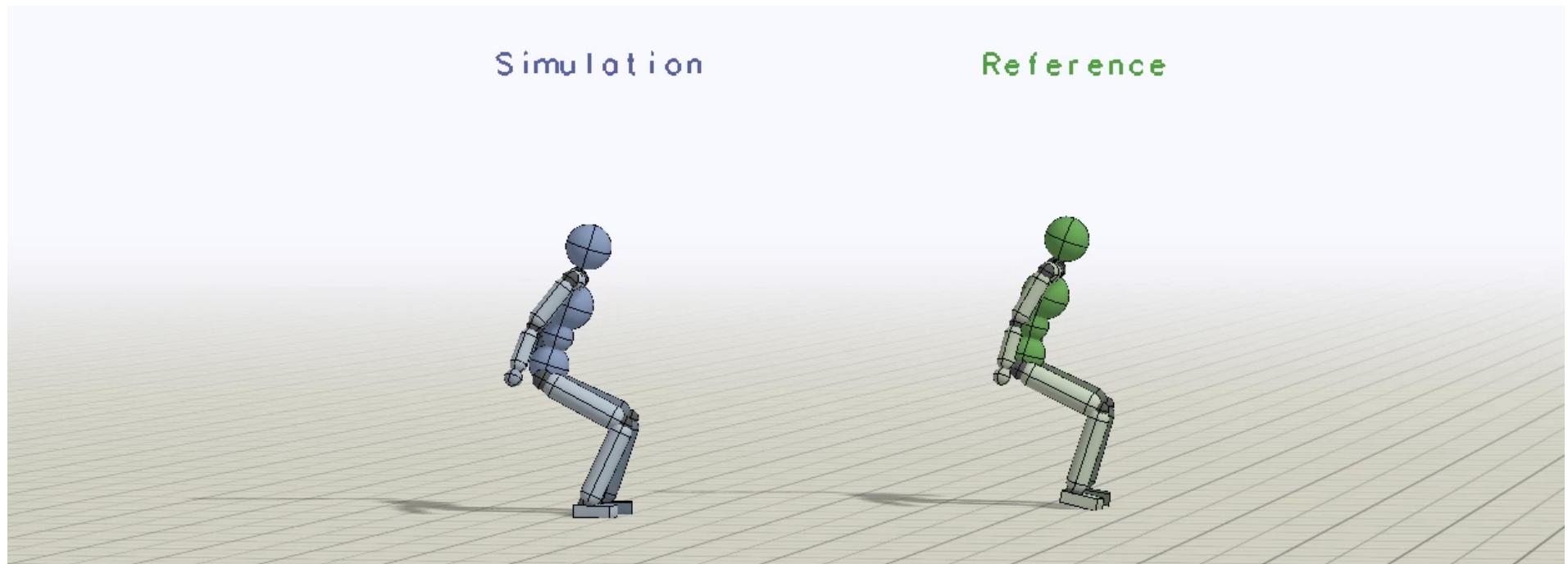
HUMANOID: RUN



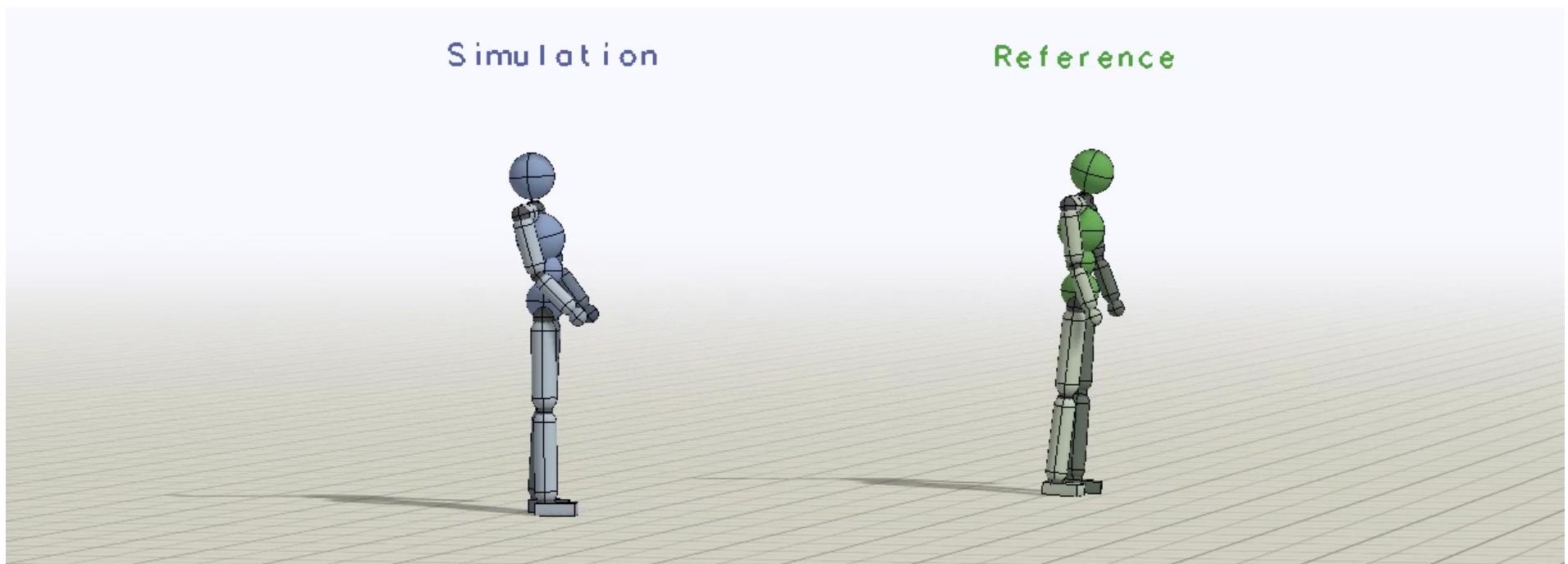
HUMANOID: BACKFLIP



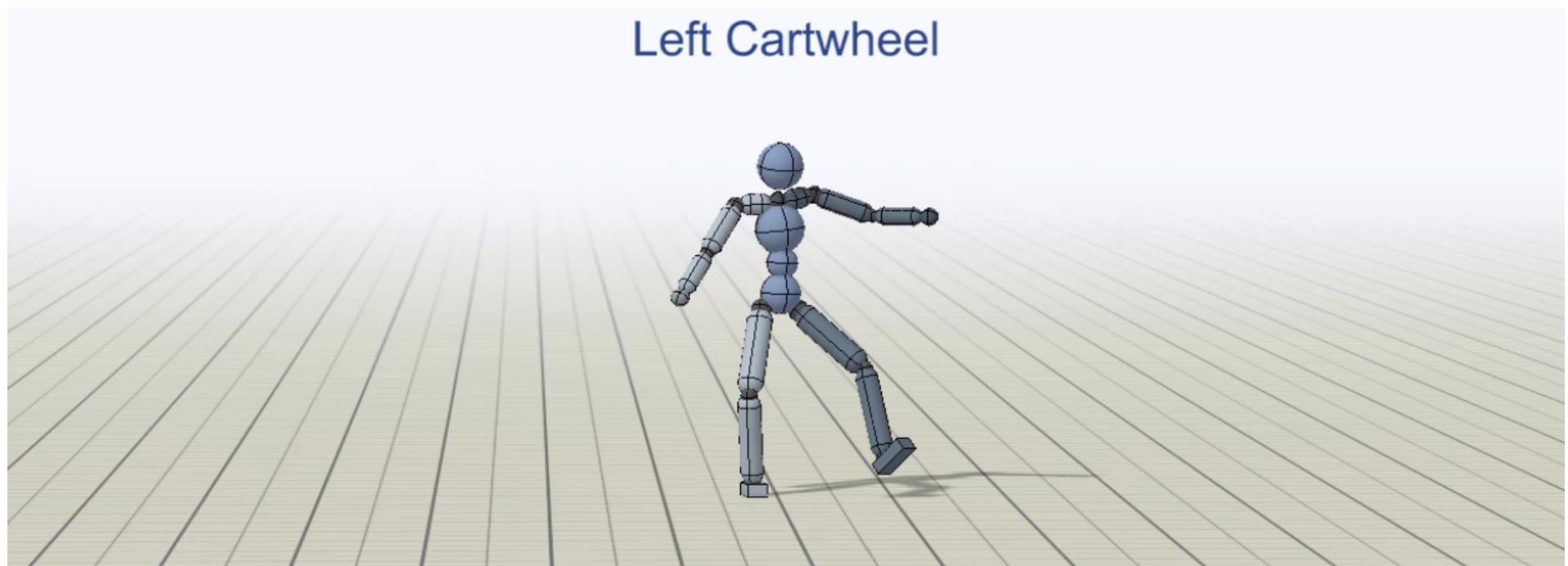
LOCAL MINIMA



WITH REFERENCE STATE INITIALIZATION



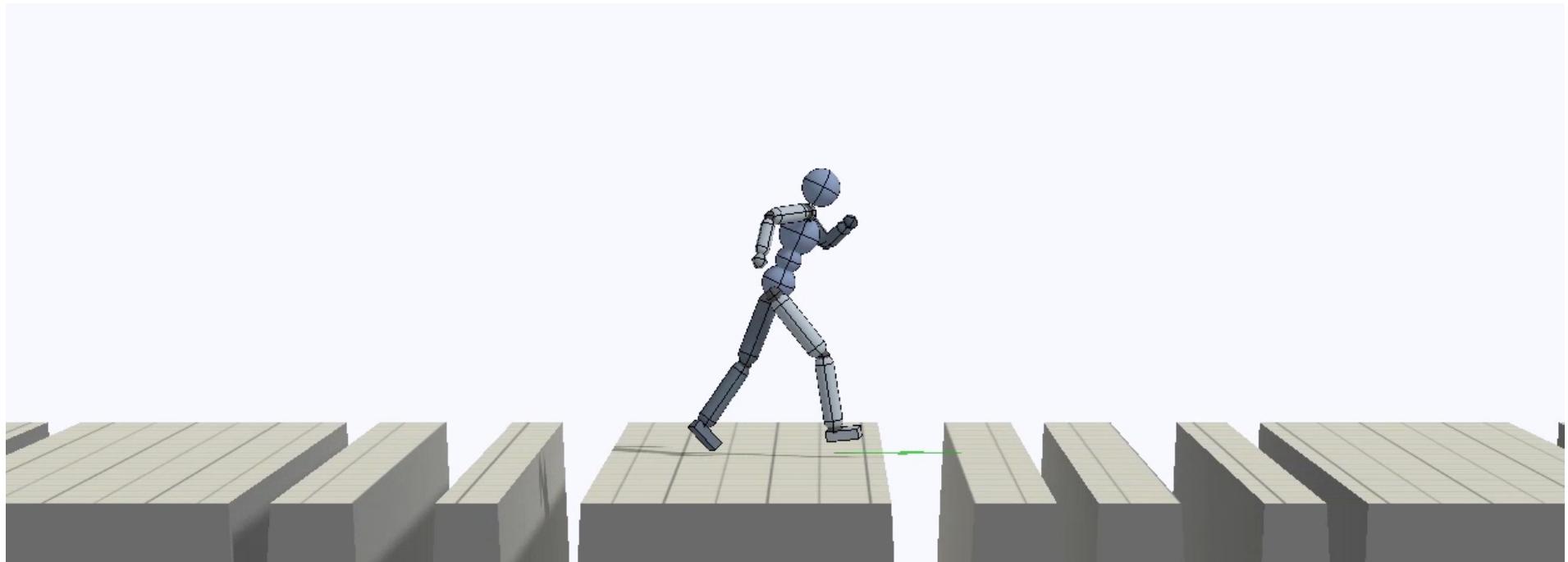
MULI-CLIP INTEGRATION



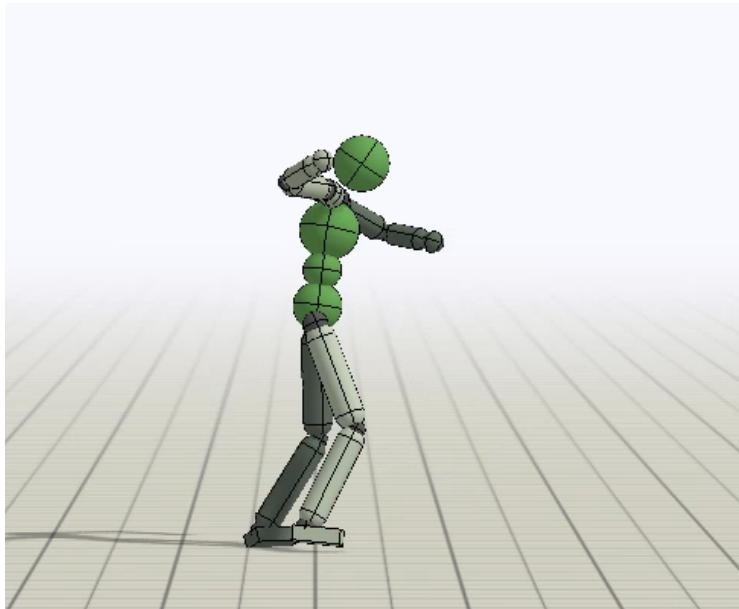
HUMANOID: BALANCE BEAM



HUMANOID: RUN – DENSE GAPS

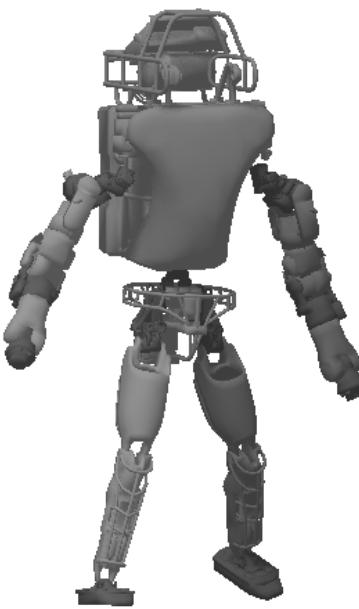


CHARACTER RETARGETING



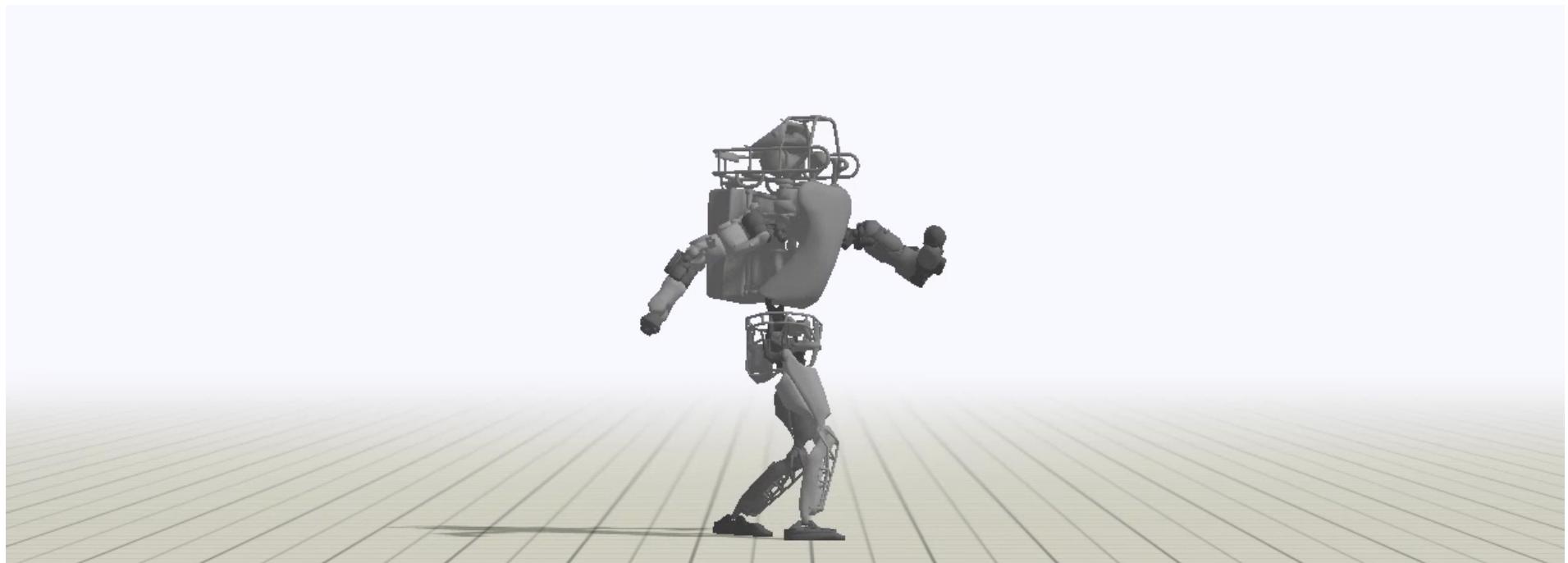
Reference Motion

+

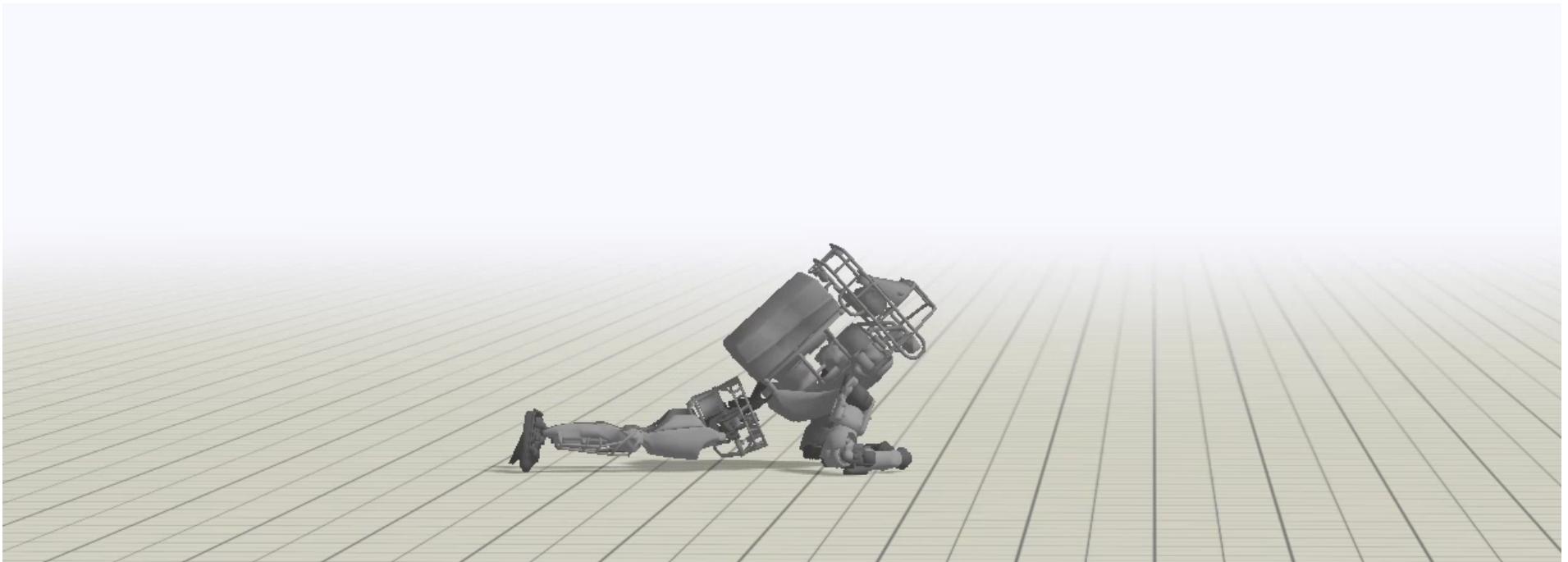


Atlas

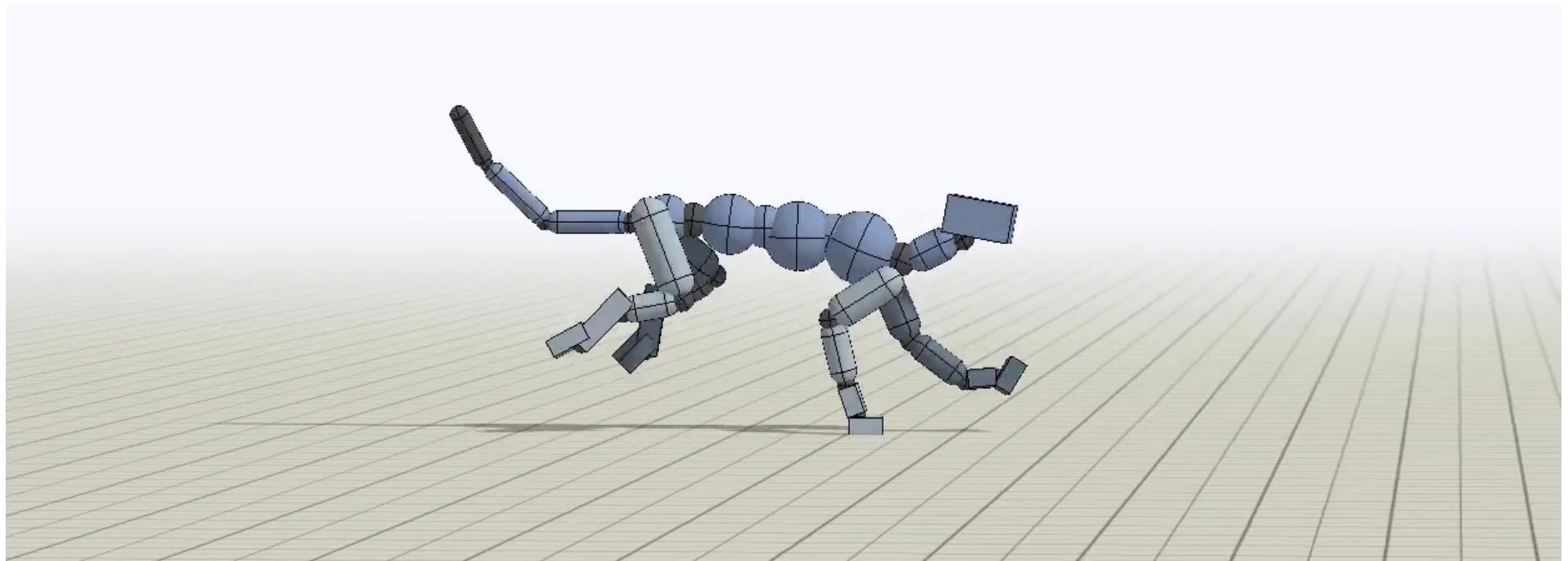
ATLAS: SPINKICK



ATLAS: GETUP- FACEDOWN



SIMULATED LION





Ziva Dynamics

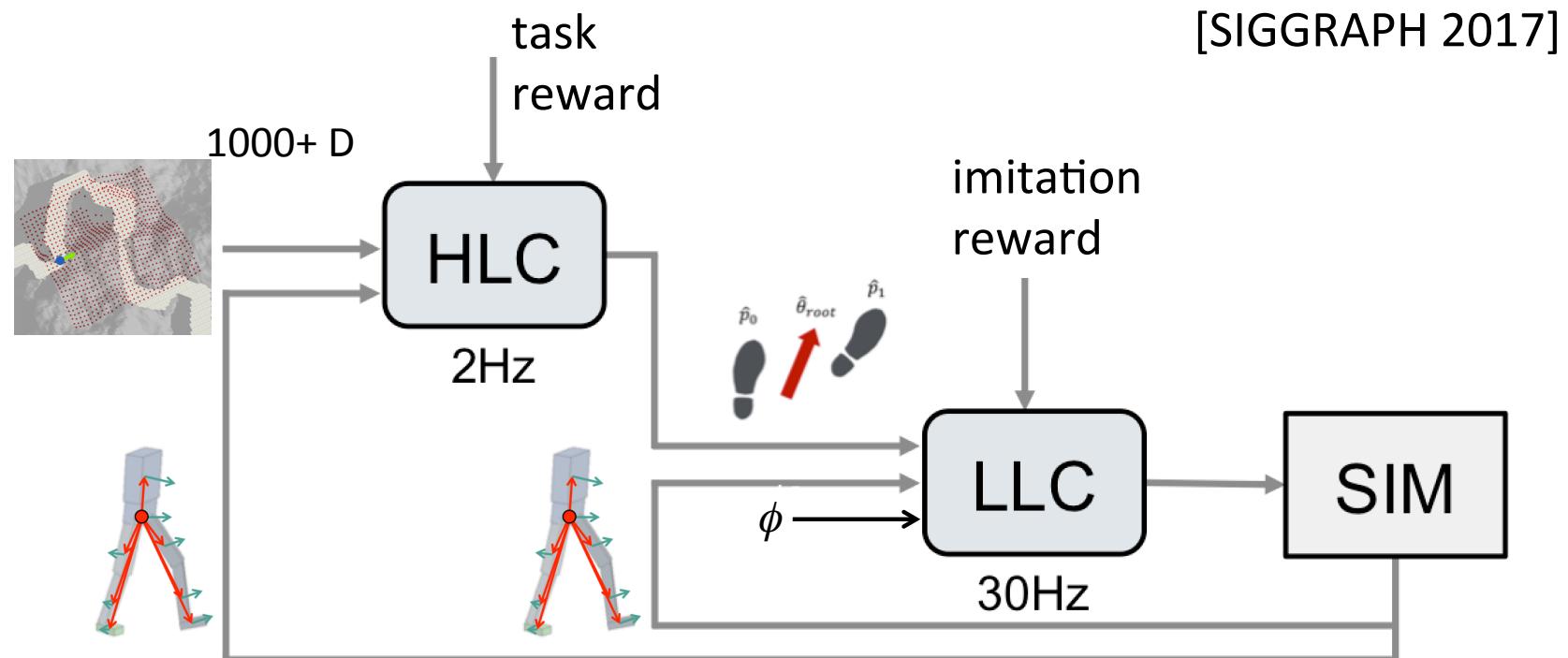


ZIVA

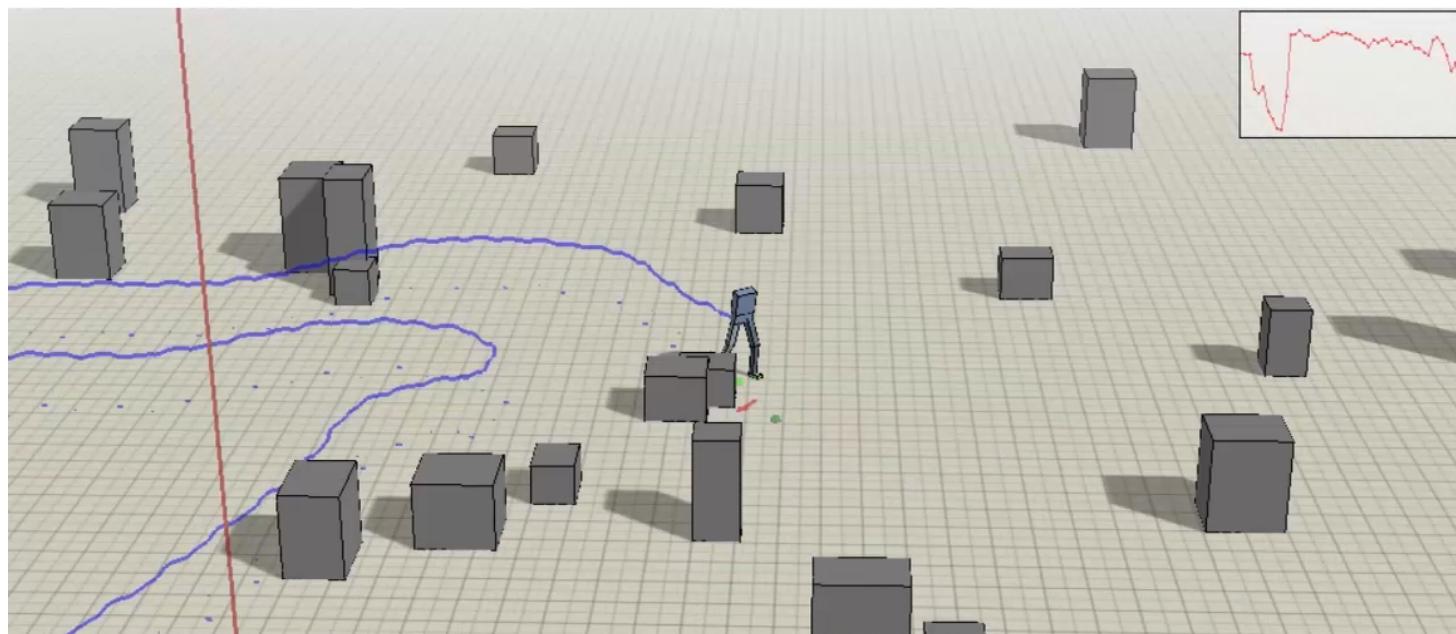


ZIVA

DEEPLOCO: HIERARCHICAL RL

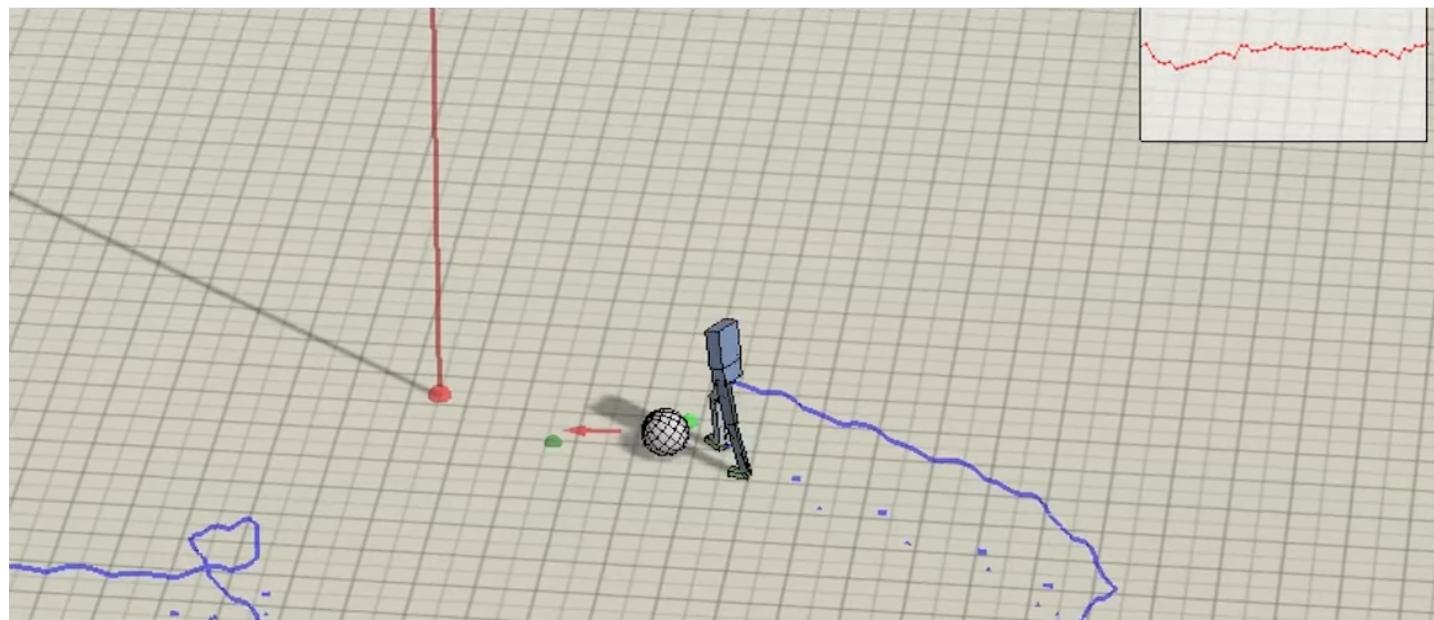


Dynamic Obstacles



HLC

Soccer Dribbling



SKILLS FROM VIDEO: REINFORCEMENT LEARNING OF PHYSICAL SKILLS FROM VIDEO

Transactions on Graphics (Proc. ACM SIGGRAPH Asia 2018)

[SIGGRAPH ASIA 2018]

Xue Bin Peng

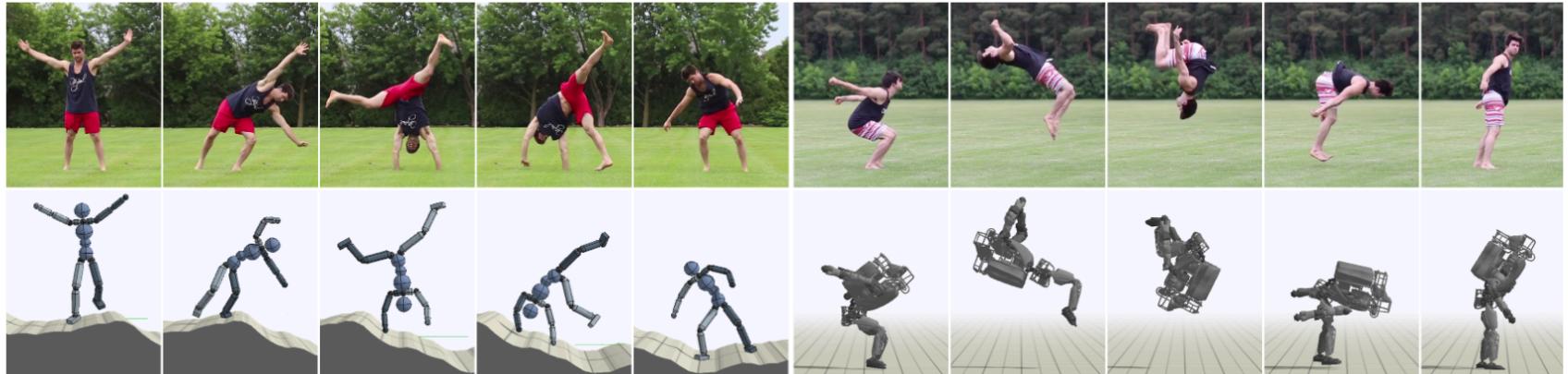
Angjoo Kanazawa

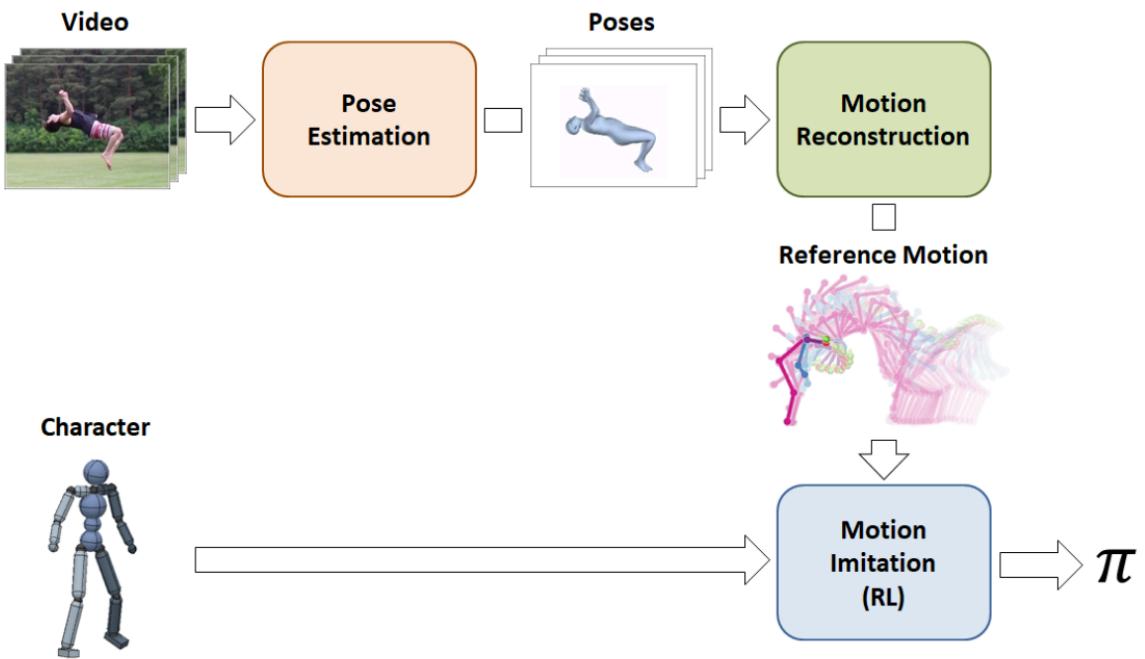
Jitendra Malik

Pieter Abbeel

Sergey Levine

University of California, Berkeley

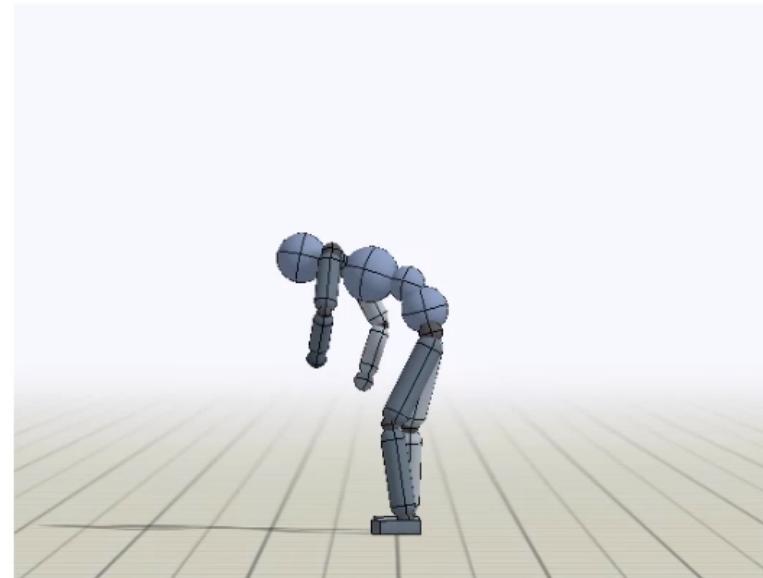




SKILLS FROM VIDEOS



Video



Simulation

SKILLS FROM VIDEOS

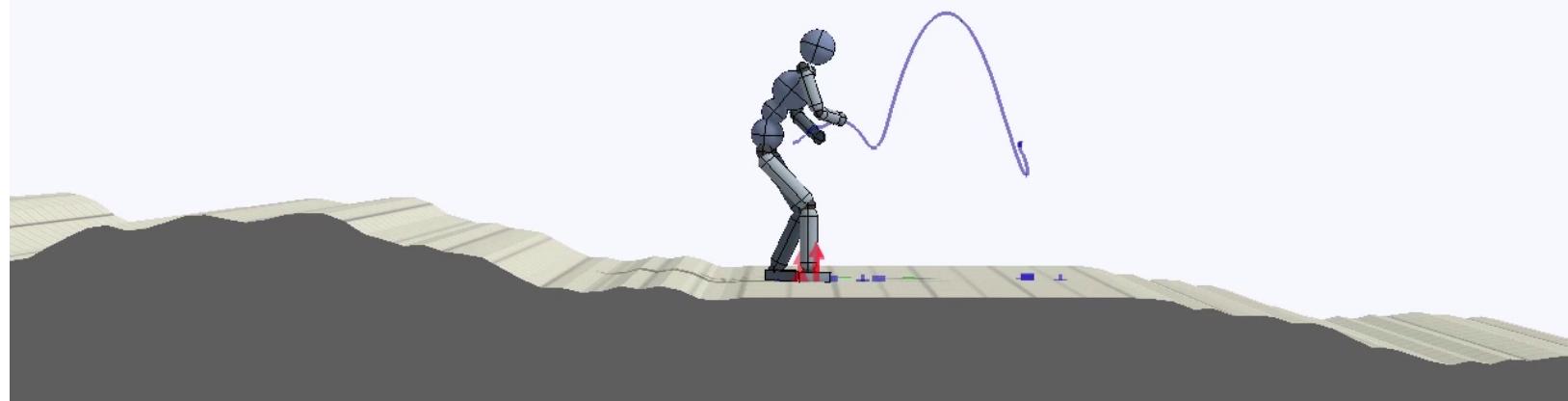


Video



Simulation

Retargeting

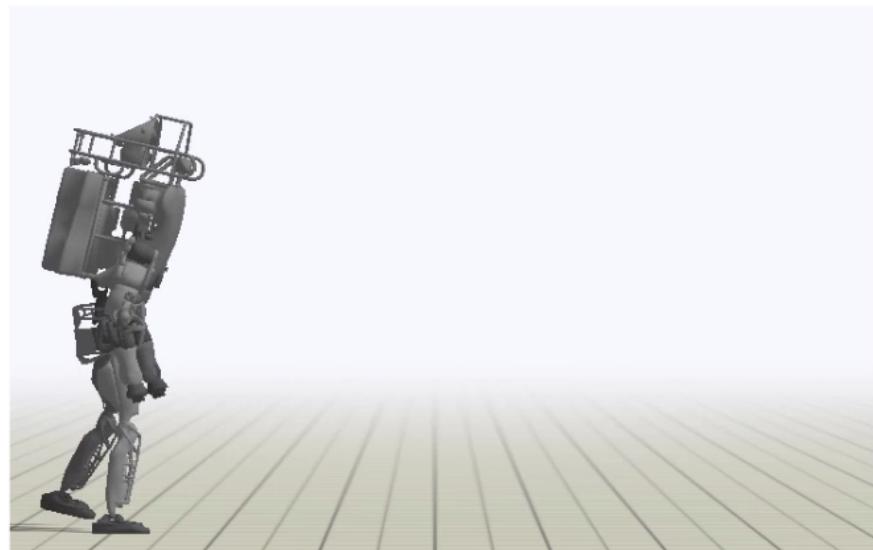


and retarget to different environments.

Character Retargeting



Video: Jump



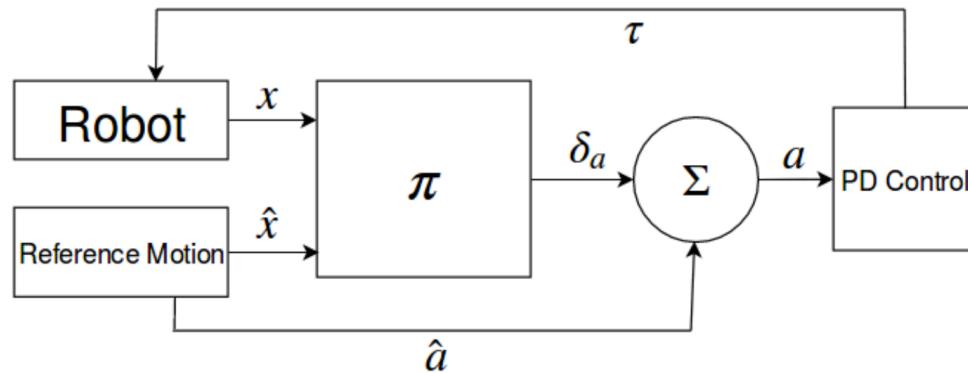
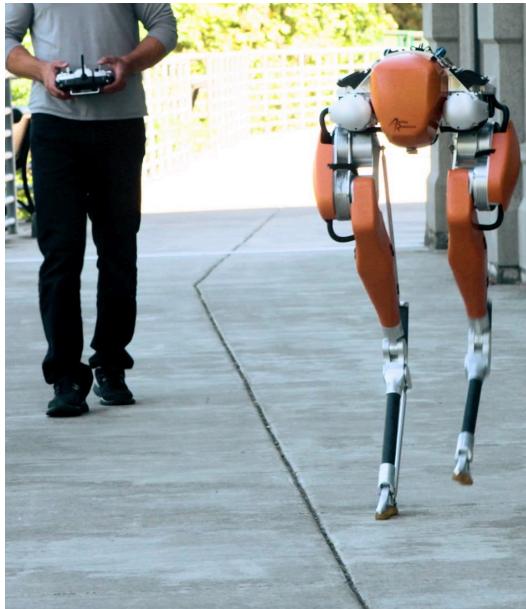
Policy

MOTION COMPLETION

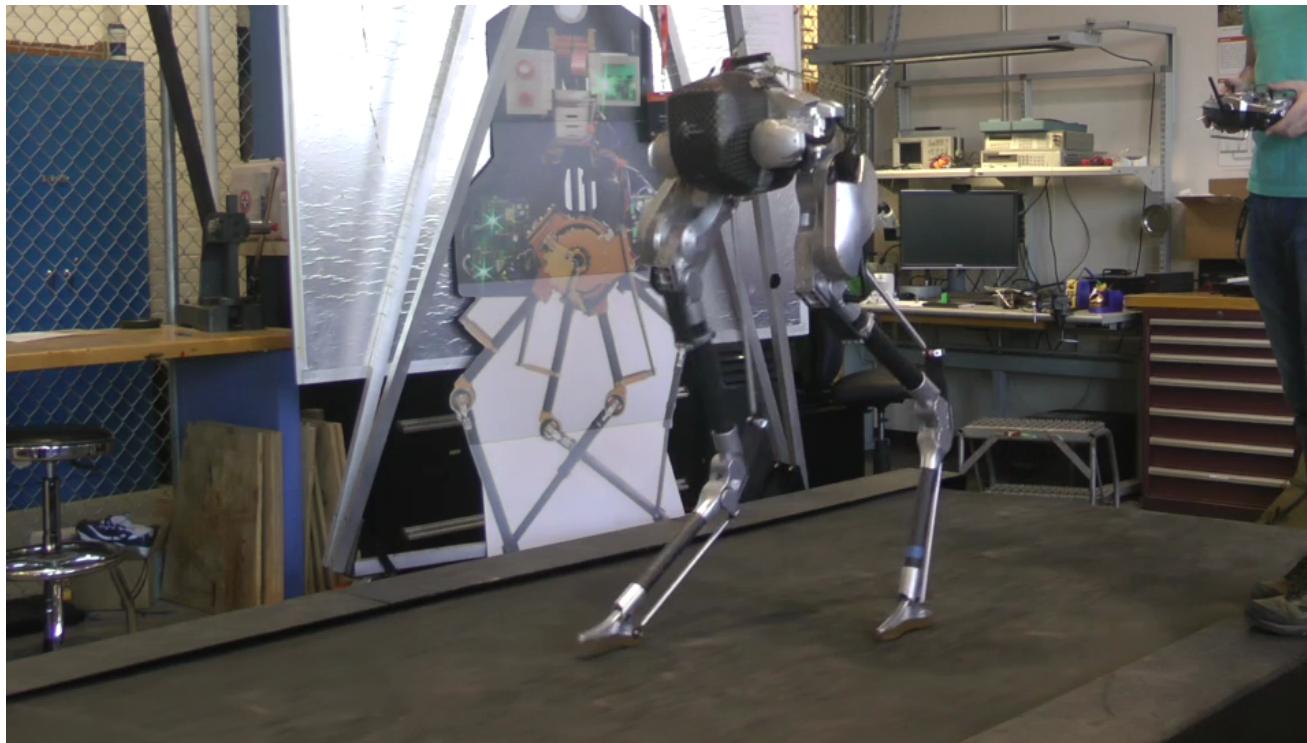
Environment Retargeting



FEEDBACK CONTROL FOR CASSIE WITH DEEP REINFORCEMENT LEARNING



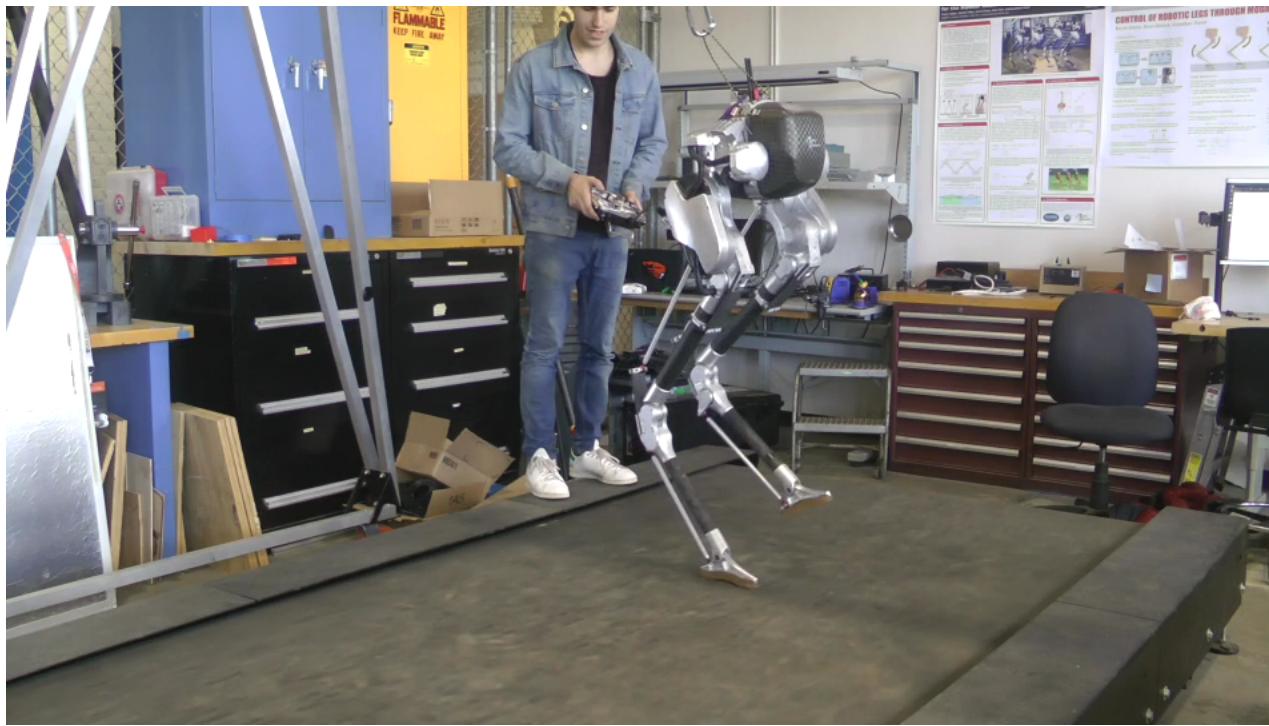
WALK AND STUMBLE



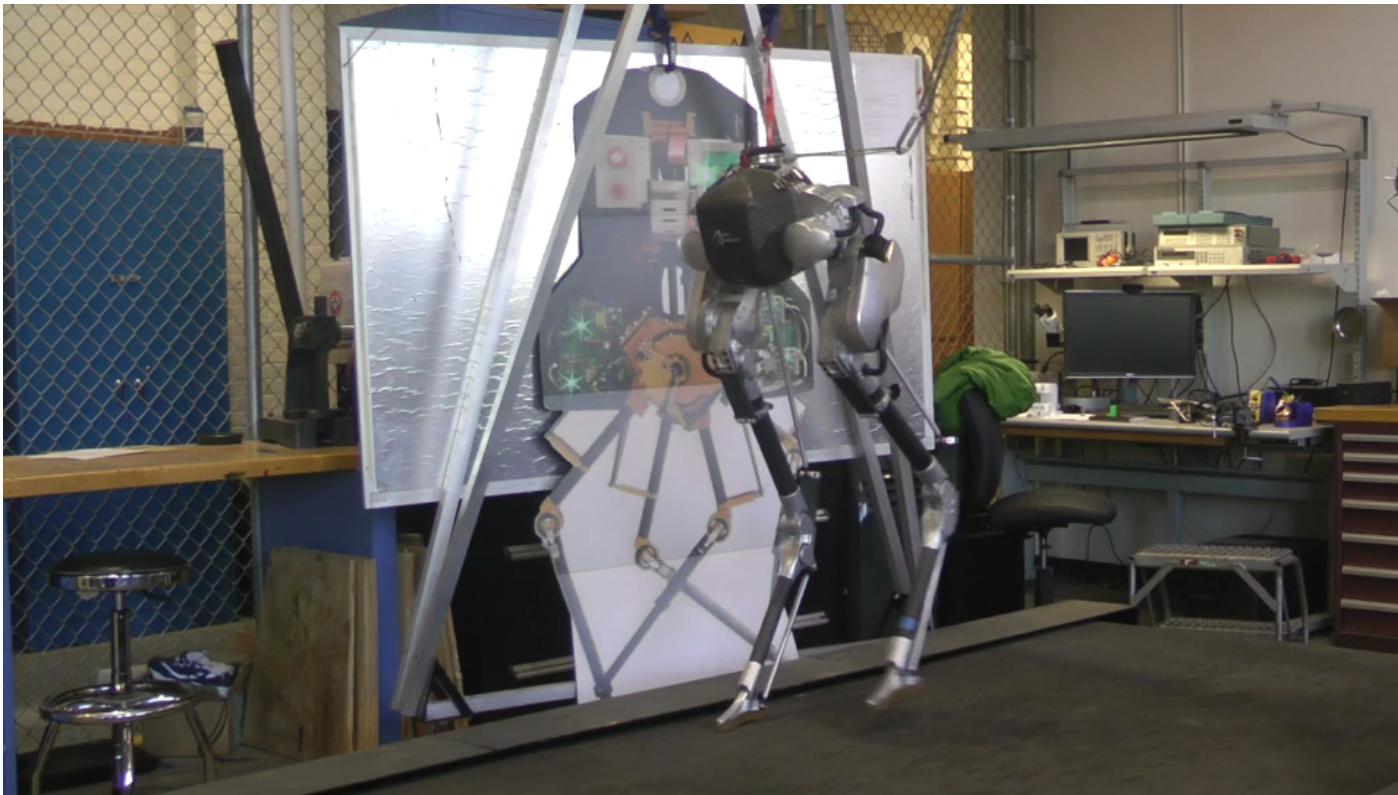
BACKWARDS WALK



SIDE-STEPS



SMOOTH STYLE



HIGH-STEPPING



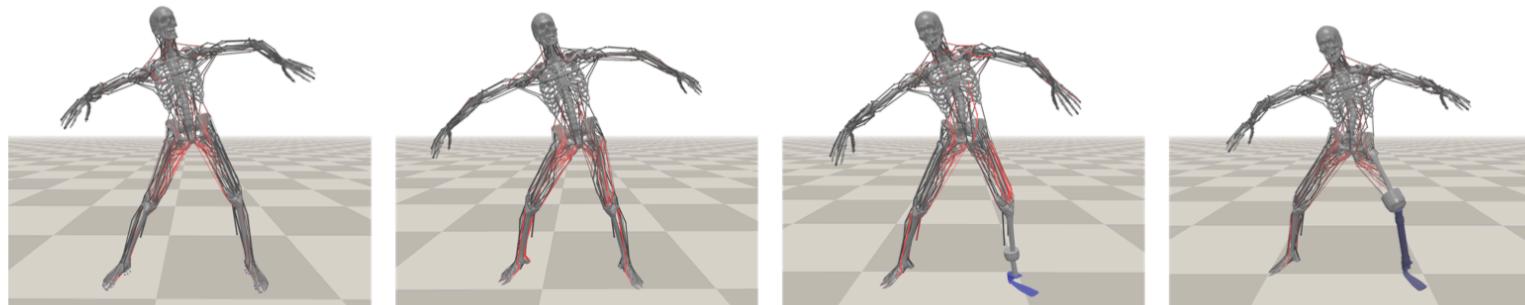
PERTURBATION TEST

Perturbation Test

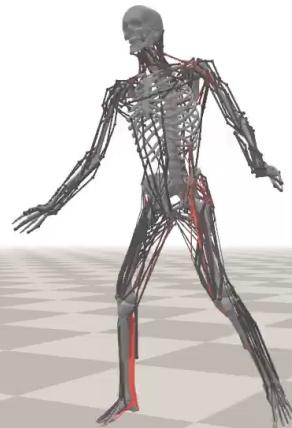
DEEP-MIMIC FOR BIOMECHANICAL MODELS

Scalable Muscle-Actuated Human Simulation and Control

SIGGRAPH 2019 Conditional Accept, Seoul National University

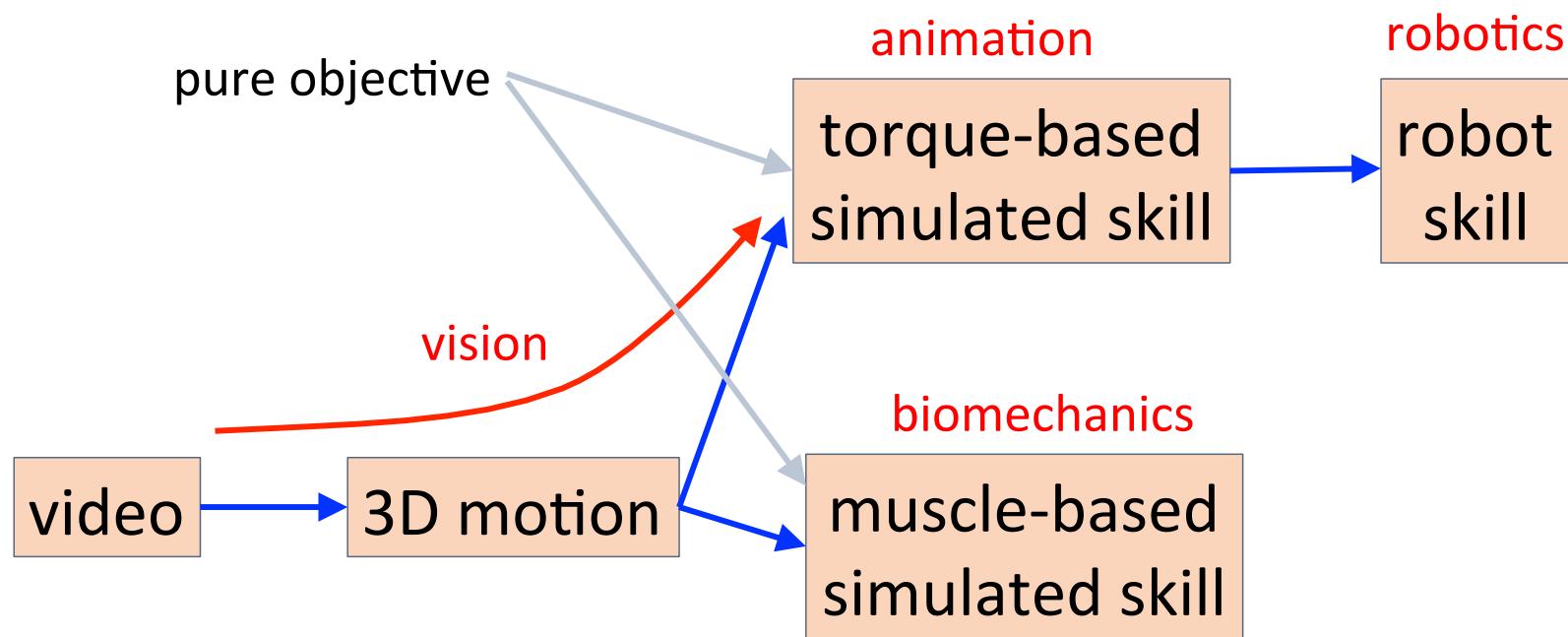


Scalable Muscle-Actuated Human Simulation and Control



Seunghwan Lee⁽¹⁾, Kyoungmin Lee⁽²⁾, Moonseok Park⁽²⁾, and Jehee Lee⁽¹⁾
Seoul National University⁽¹⁾, Seoul National University Bundang Hospital⁽²⁾

OVERVIEW



MUCH TO BE DONE...

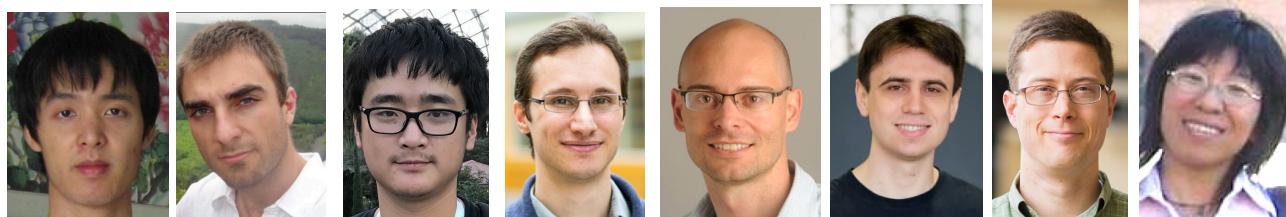
- multi-skilled digital humans & animals
 - richer perception
 - interaction with the world
 - collaborative & interacting characters
- behavior from video
- motion planning

CONCLUSIONS

- “the best way to predict the future is to create it”
- physics-based human movement:
 rapid advances, many uses

ACKNOWLEDGEMENTS

Xue Bin Peng, Glen Berseth, Xinyi Zhang, Zhaoming Xie,
Patrick Clary, Jonathan Hurst, Pieter Abbeel, Sergey Levine,
Kangkang Yin, Hung Yu Ling



University of British Columbia

UC Berkeley

Oregon State University

Agility Robotics

Questions?

