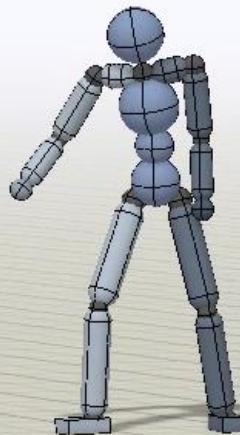


# Towards a Virtual Stuntman



Xue Bin (Jason) Peng

UC Berkeley

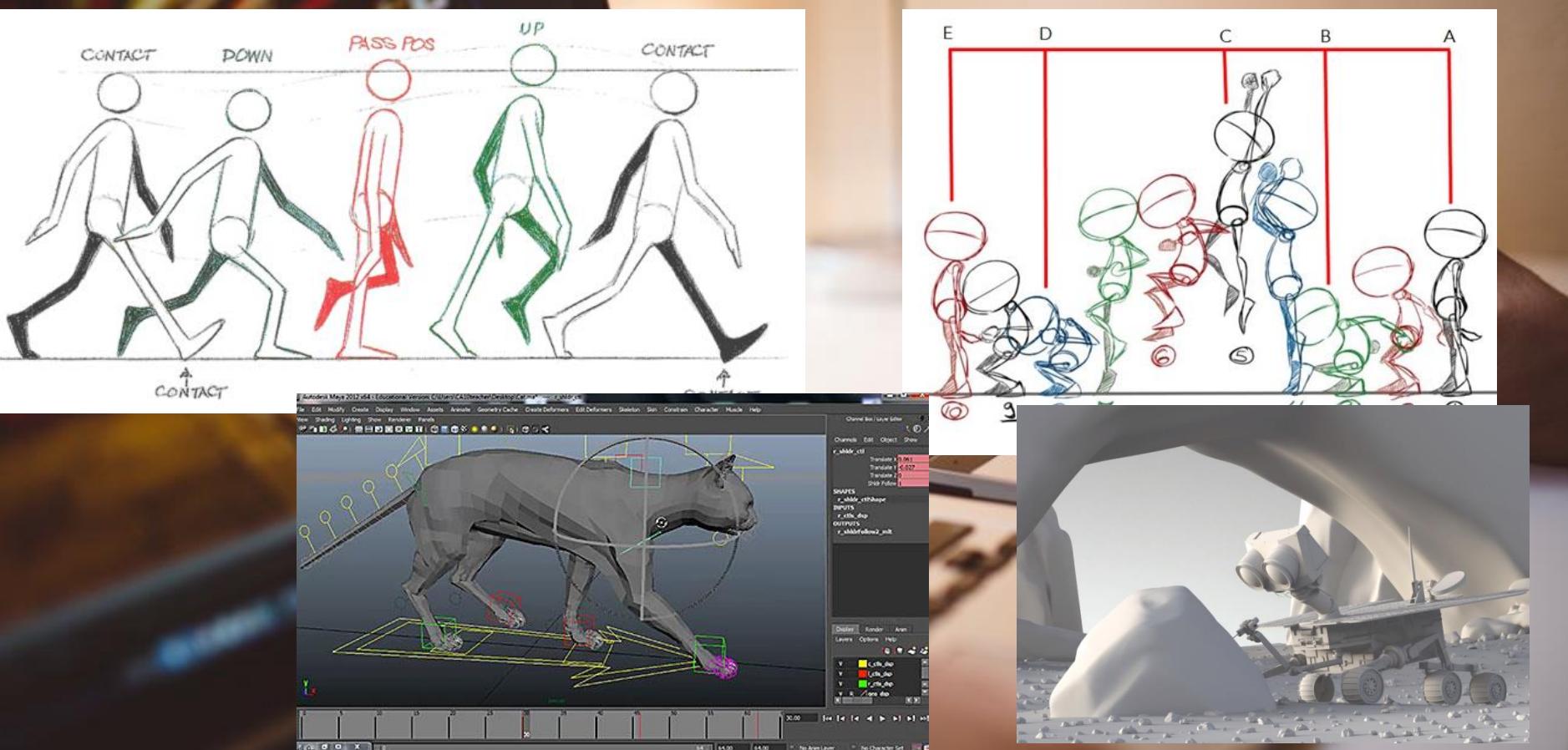


# Animation

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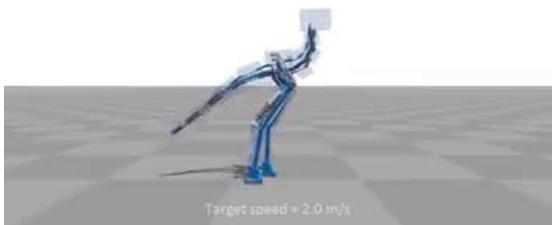


# Animation

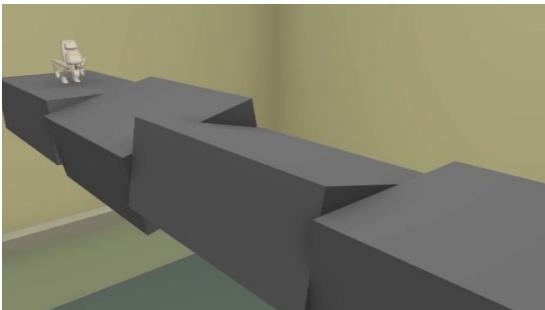


# Computer Animation

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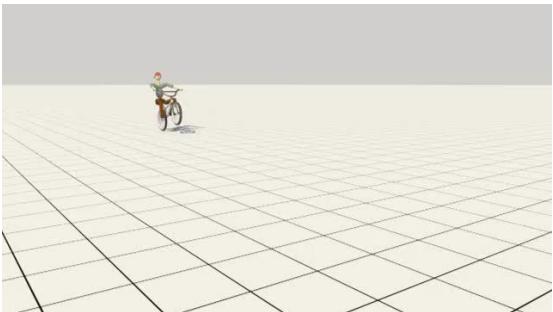
[Geijtenbeek et al. 2013]



[Brown et al. 2013]



[Ju et al. 2013]



[Tan et al. 2014]



[Kwon and Hodgins 2017]

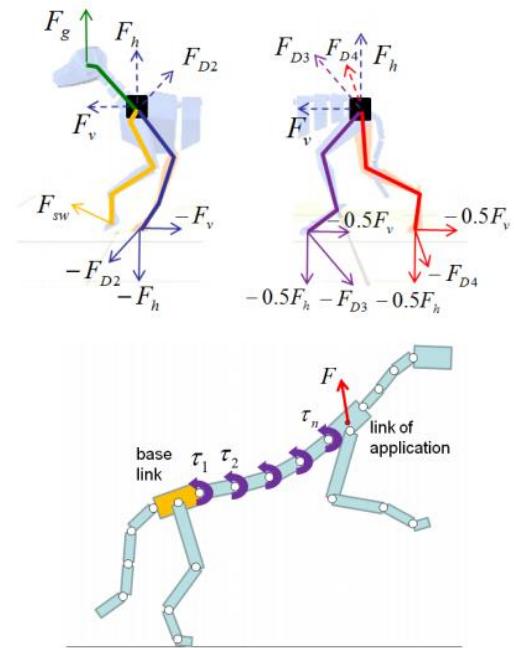


[Peng et al. 2018]

# Physics-Based Animation



[Coros et al. 2011]

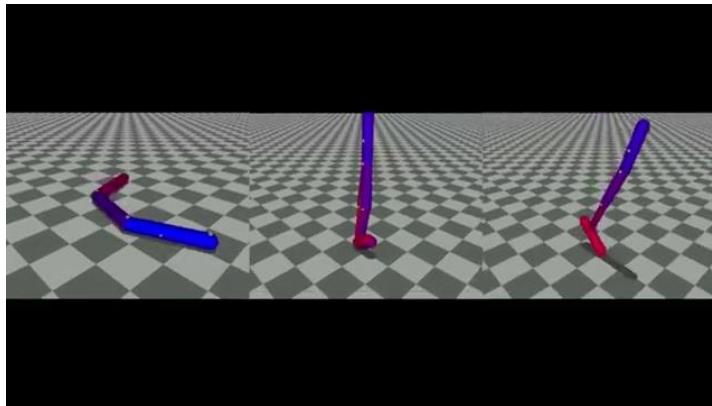


# Deep RL

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[Mnih et al. 2015]



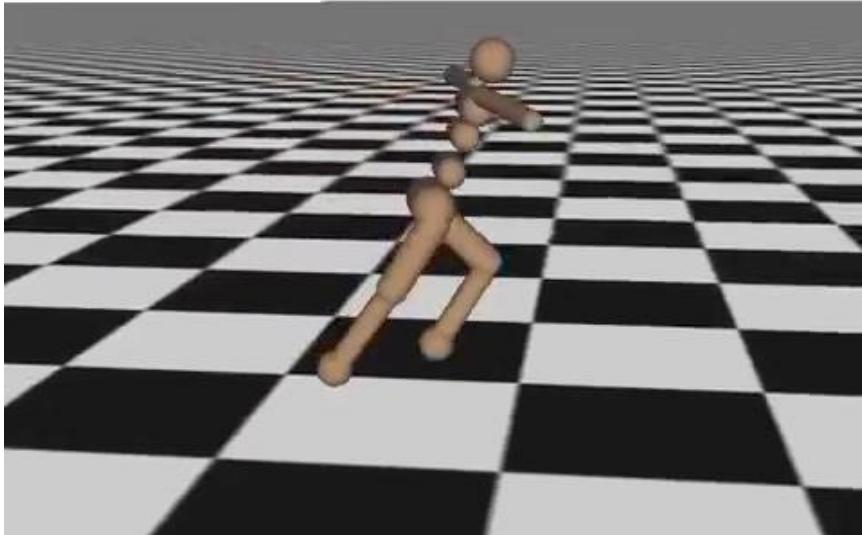
[Schulman et al. 2016]



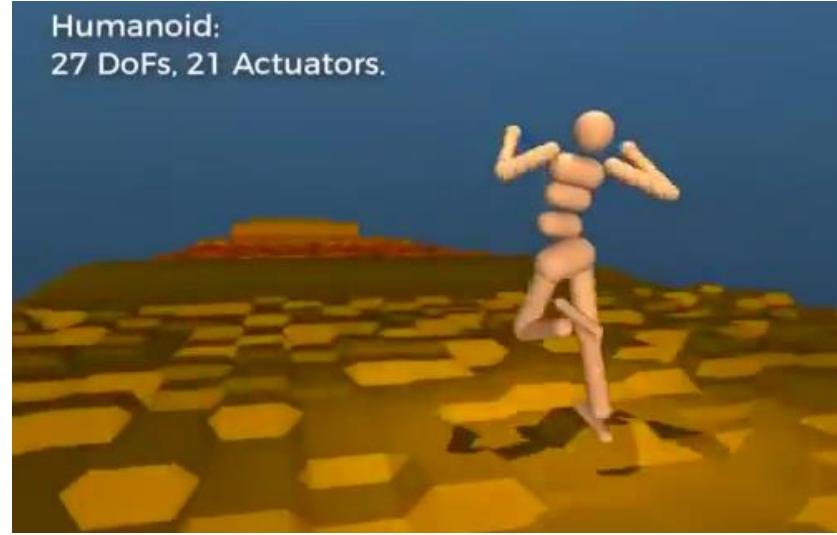
[Chebotar et al. 2017]

# Motion Quality

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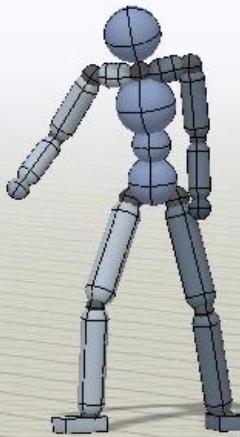
[Schulman et al. 2016]



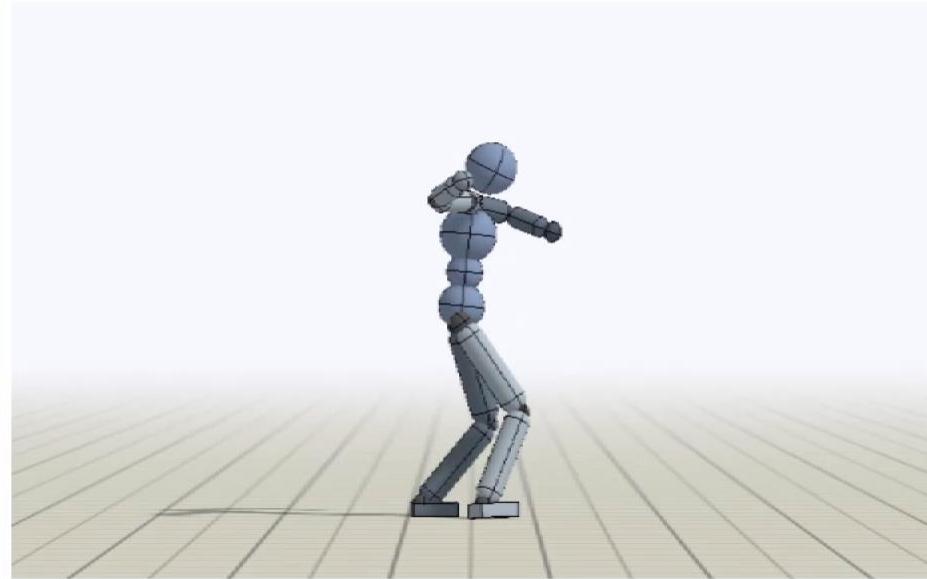
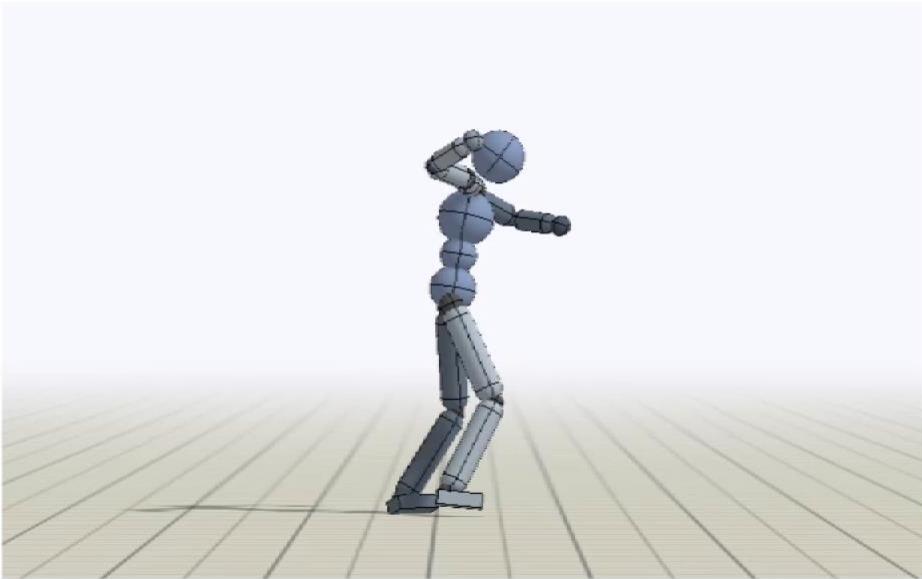
[Heess et al. 2017]

# Motivation

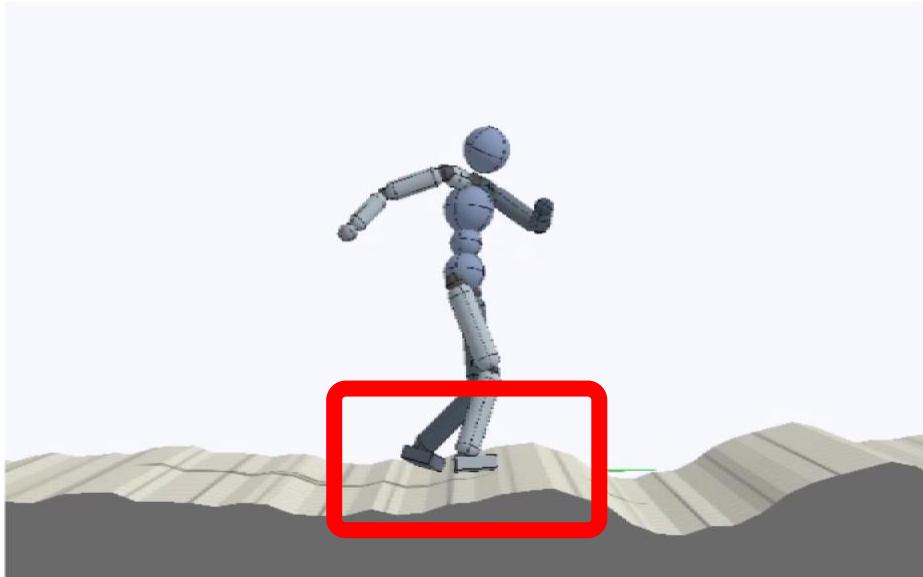
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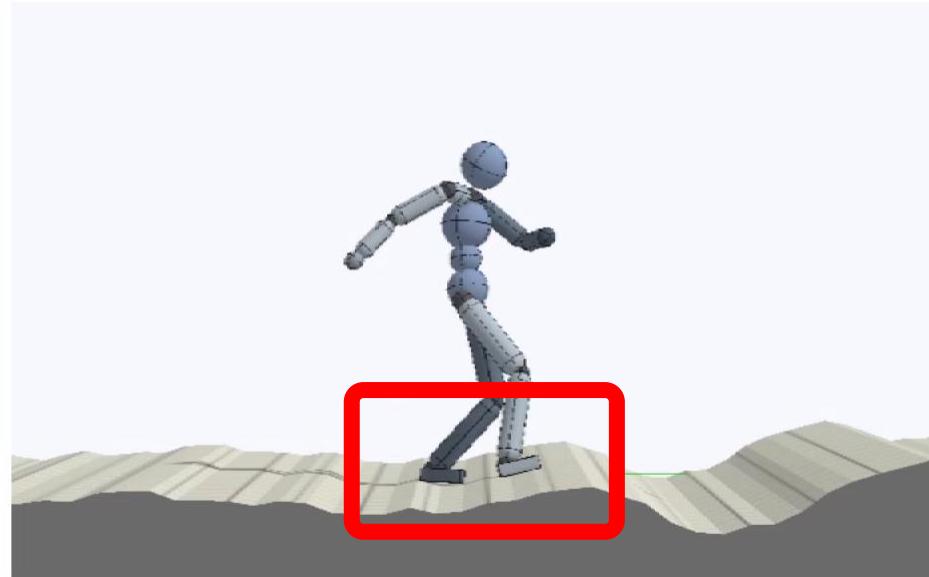
# Which is Mocap?



# Which is Mocap?



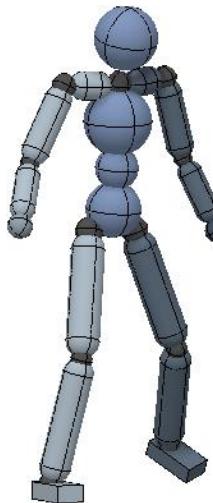
Mocap



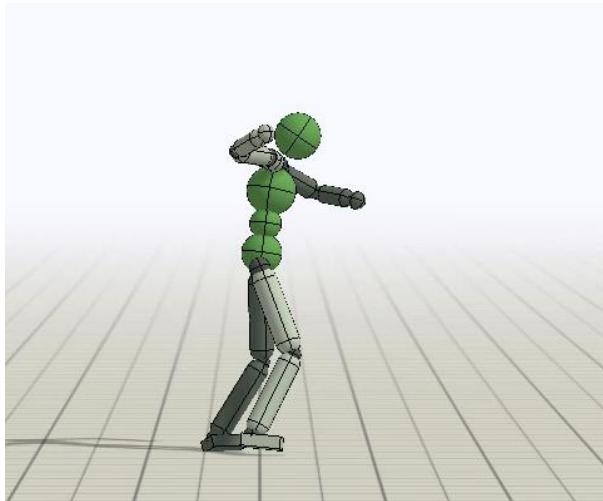
Simulation

# Overview

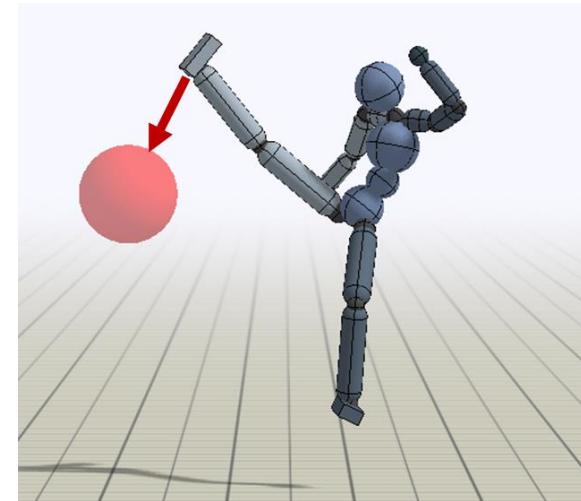
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+



+



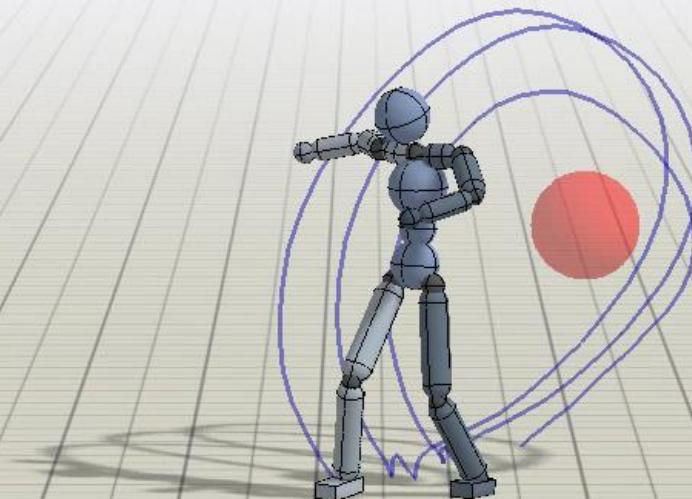
Character

Reference Motion

Task: Hit Target

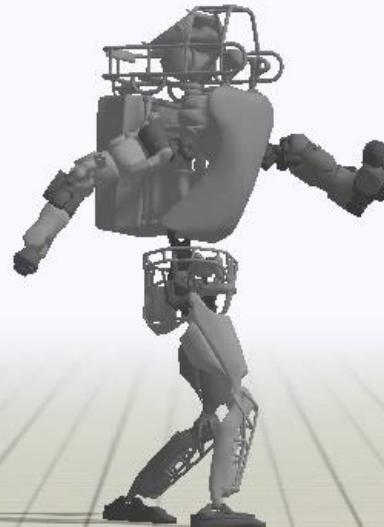
# Overview

---



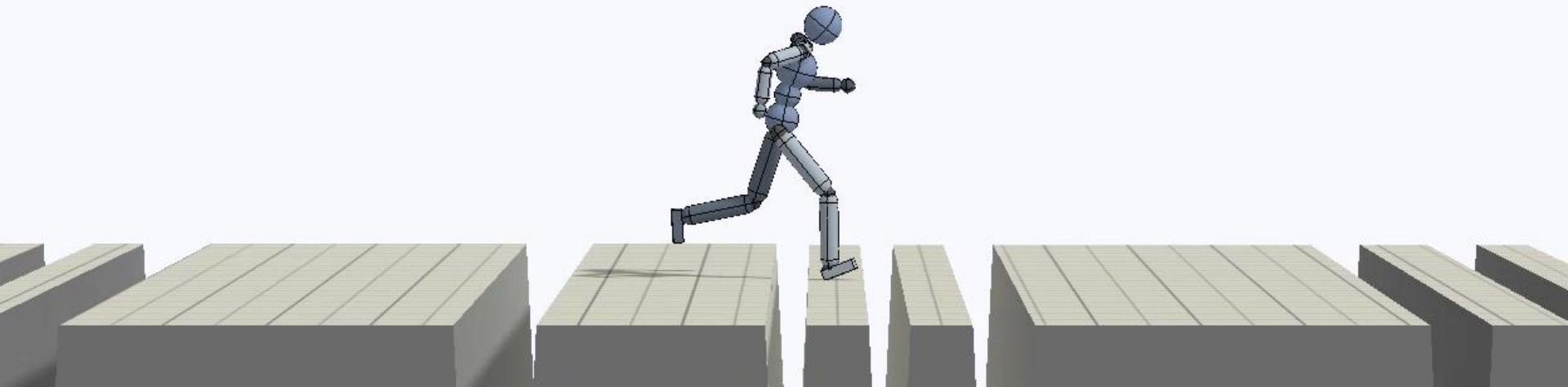
# Overview

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# Overview

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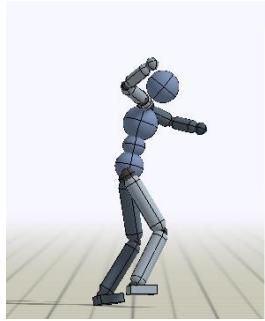


# Reference Motion

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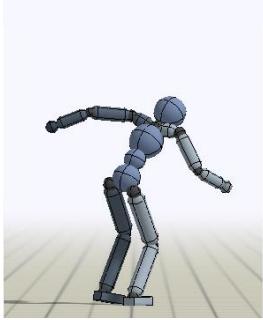
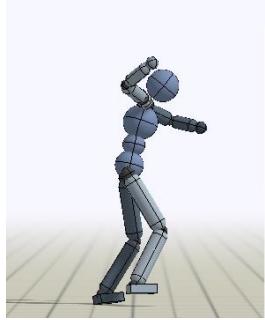
# Reference Motion

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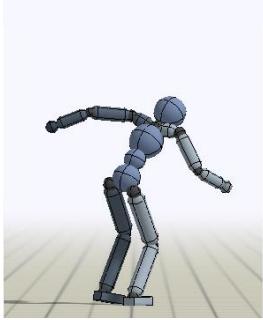
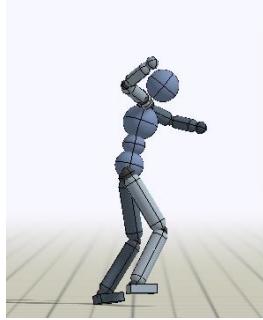
# Reference Motion

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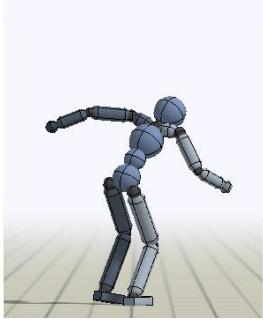
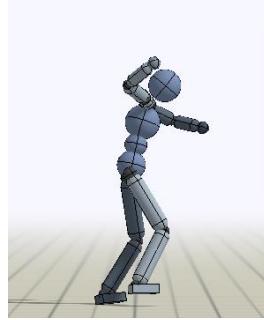
# Reference Motion

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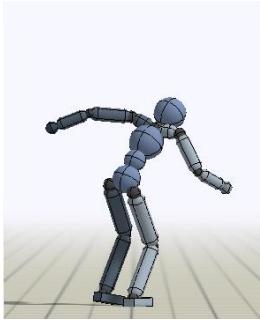
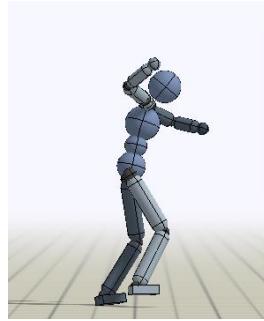
# Reference Motion

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# Reference Motion

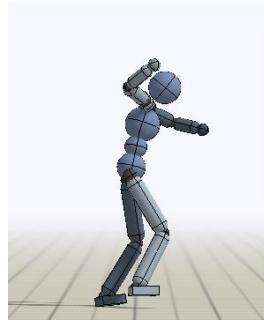
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...

# Reference Motion

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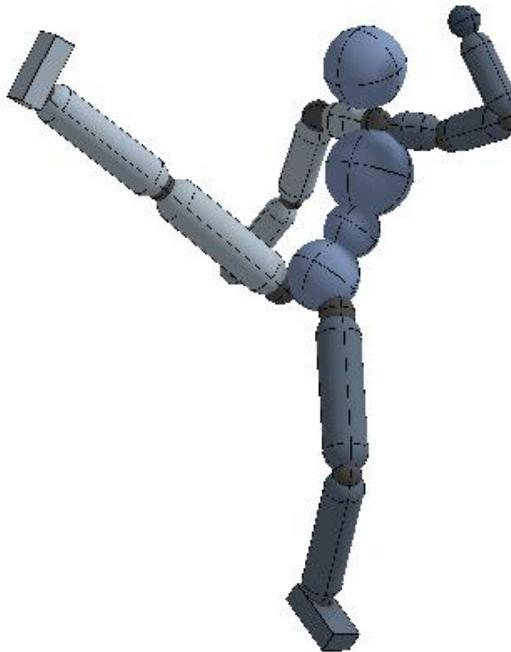
 $a_0$  $a_1$  $a_2$  $a_3$  $a_4$ ...  
......  
...

# State + Action

---

State:

- link positions
- link velocities

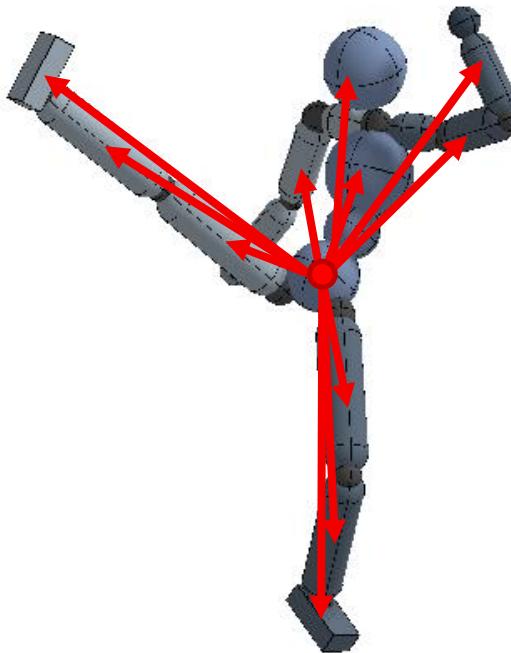


# State + Action

---

State:

- link positions
- link velocities

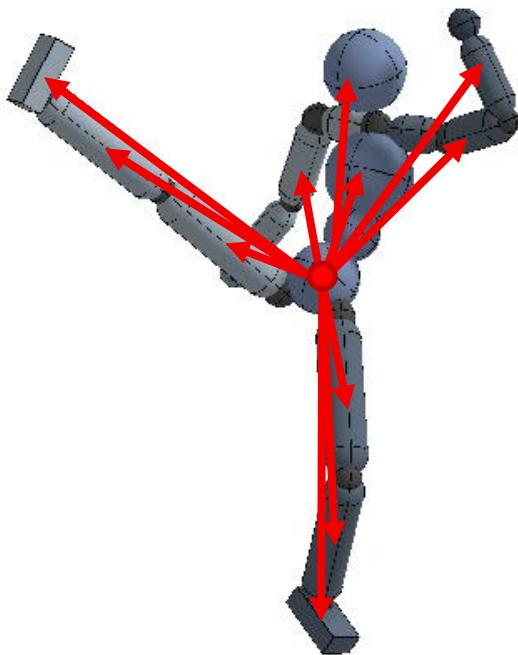


# State + Action

---

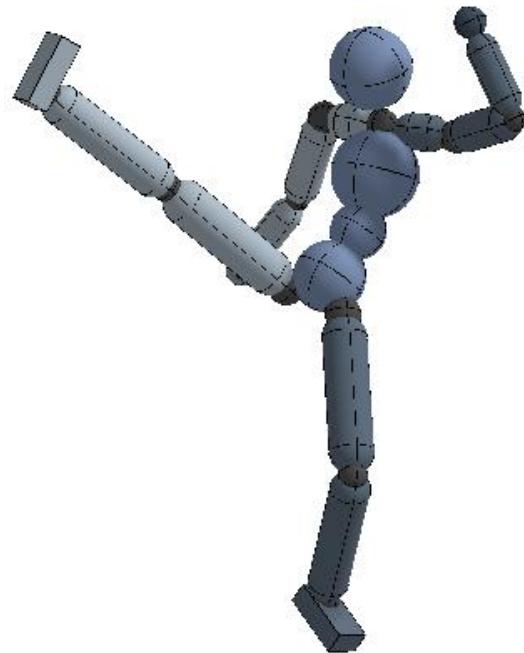
State:

- link positions
- link velocities



Action:

- PD targets

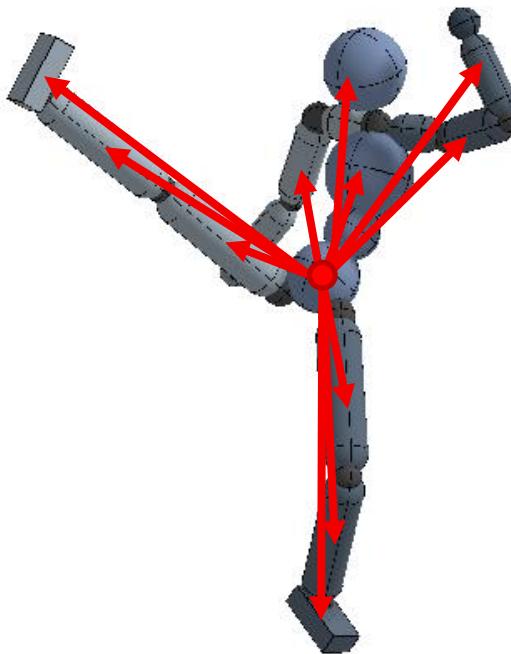


# State + Action

---

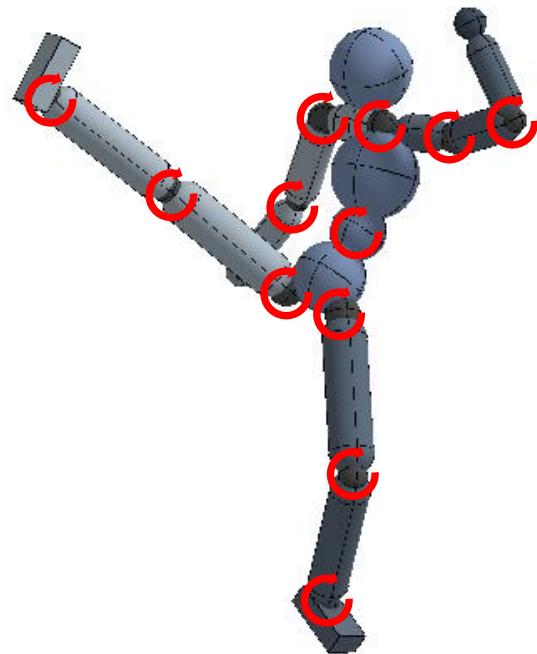
State:

- link positions
- link velocities



Action:

- PD targets



# Reward

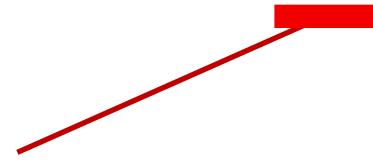
---

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

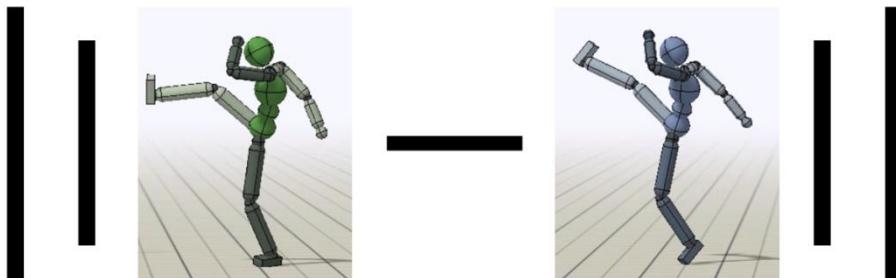
# Reward

---

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



Imitation Objective



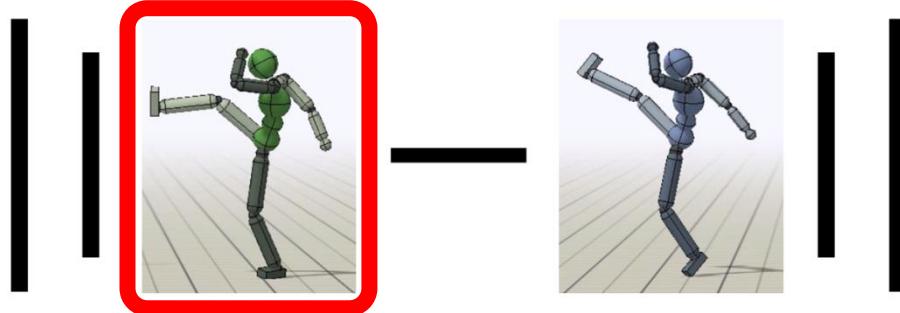
# Reward

---

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



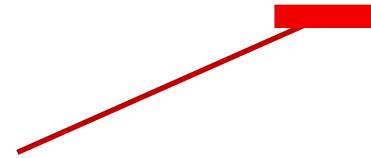
Imitation Objective



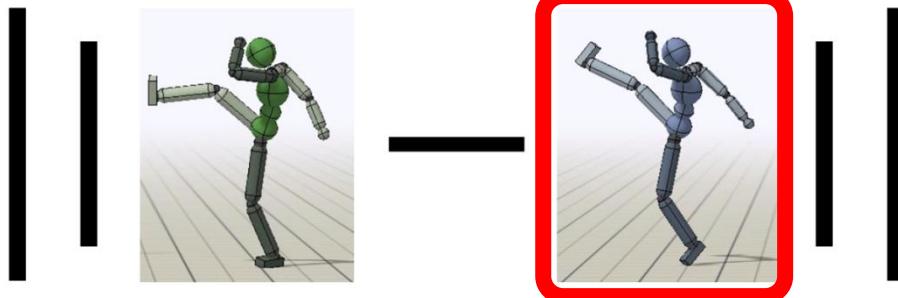
# Reward

---

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



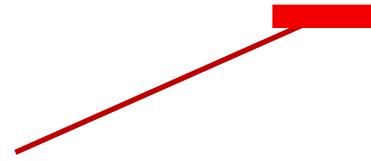
Imitation Objective



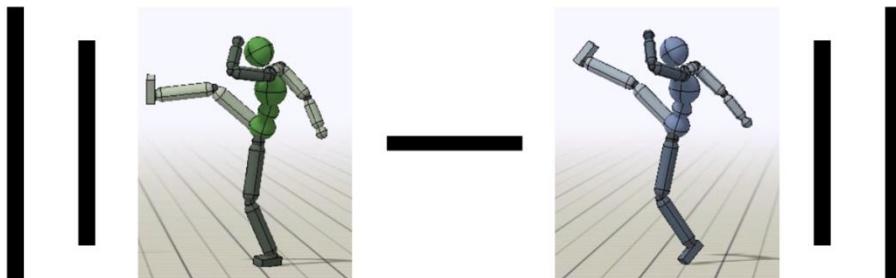
# Reward

---

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



Imitation Objective



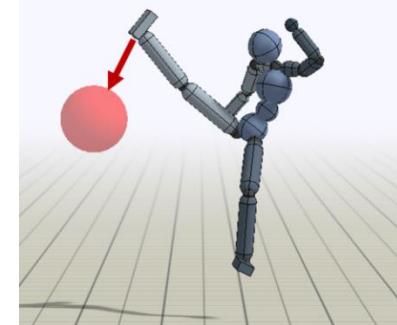
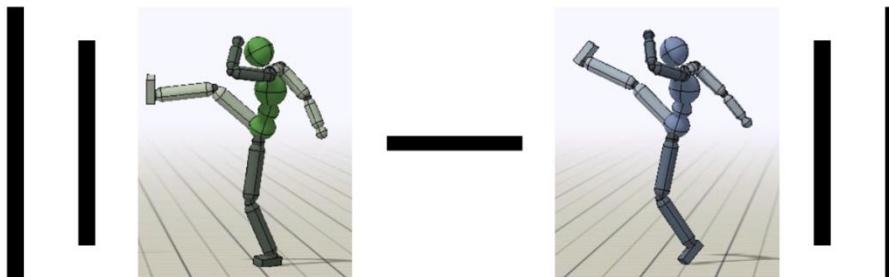
# Reward

---

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

Imitation Objective

Task Objective



# Proximal Policy Optimization (PPO)

$$\max_{\theta} \quad J(\theta)$$

[Schulman et al. 2017]

# Proximal Policy Optimization (PPO)

$$\begin{aligned} \max_{\theta} \quad & J(\theta) \\ \text{s.t.} \quad & \mathbb{E}_{s_t \sim d_{\theta}(s_t)} \left[ KL \left( \pi_{\theta_{old}}(\cdot | s_t) \middle| \pi_{\theta}(\cdot | s_t) \right) \right] \leq \delta_{KL} \end{aligned}$$

[Schulman et al. 2017]

# Proximal Policy Optimization (PPO)

$$\begin{aligned} \max_{\theta} \quad & J(\theta) \\ \text{s.t.} \quad & \mathbb{E}_{s_t \sim d_{\theta}(s_t)} \left[ KL \left( \pi_{\theta_{old}}(\cdot | s_t) \middle| \underline{\pi_{\theta}(\cdot | s_t)} \right) \right] \leq \delta_{KL} \end{aligned}$$

[Schulman et al. 2017]

# Proximal Policy Optimization (PPO)

$$\max_{\theta} \quad J(\theta)$$

$$\text{s.t.} \quad \mathbb{E}_{s_t \sim d_\theta(s_t)} \left[ KL \left( \pi_{\theta_{old}}(\cdot | s_t) \middle| \underline{\pi_\theta(\cdot | s_t)} \right) \right] \leq \delta_{KL}$$

[Schulman et al. 2017]

# Proximal Policy Optimization (PPO)

$$\begin{aligned} \max_{\theta} \quad & J(\theta) \\ \text{s.t.} \quad & \mathbb{E}_{s_t \sim d_{\theta}(s_t)} \left[ KL \left( \pi_{\theta_{old}}(\cdot | s_t) \middle| \pi_{\theta}(\cdot | s_t) \right) \right] \leq \delta_{KL} \end{aligned}$$

[Schulman et al. 2017]

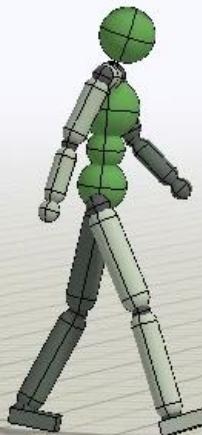
# Humanoid: Walk

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Simulation



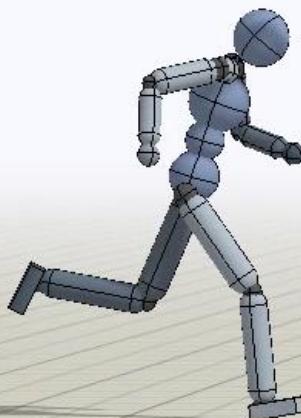
Reference



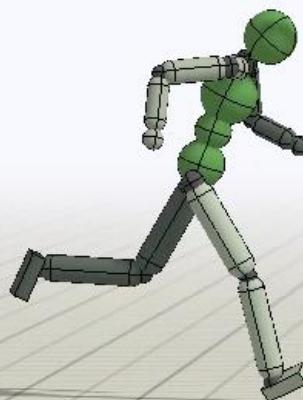
# Humanoid: Run

---

Simulation

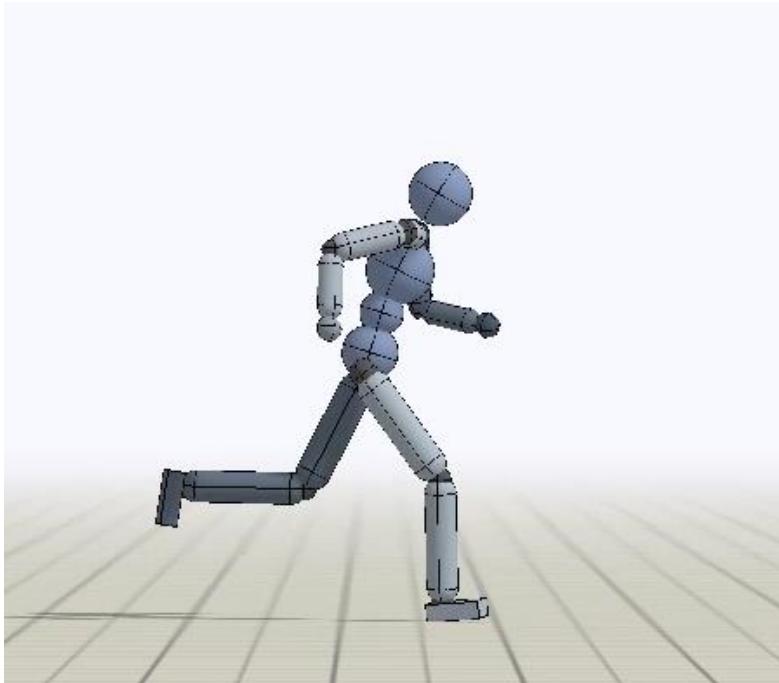


Reference

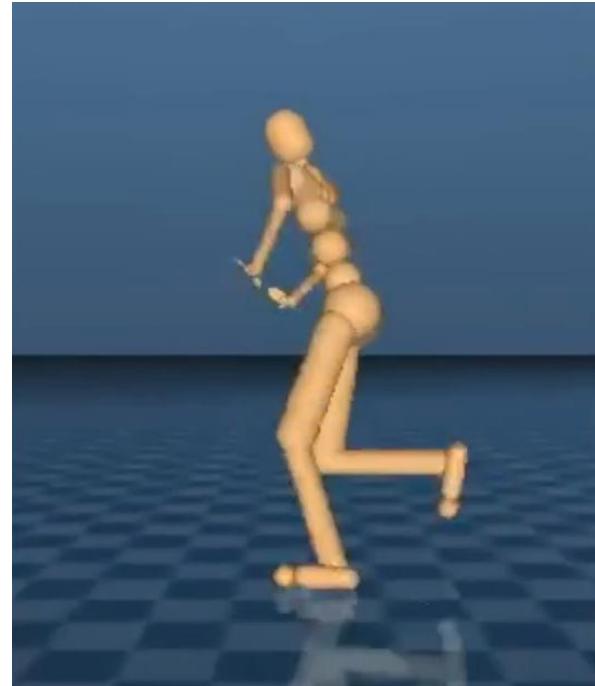


# Comparison

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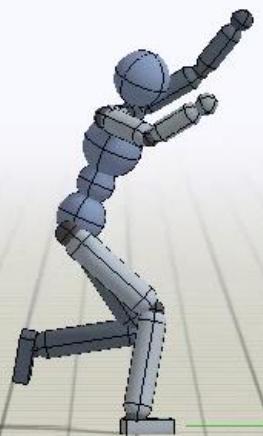
Ours



[Merel et al. 2017]

# No Reference Motion

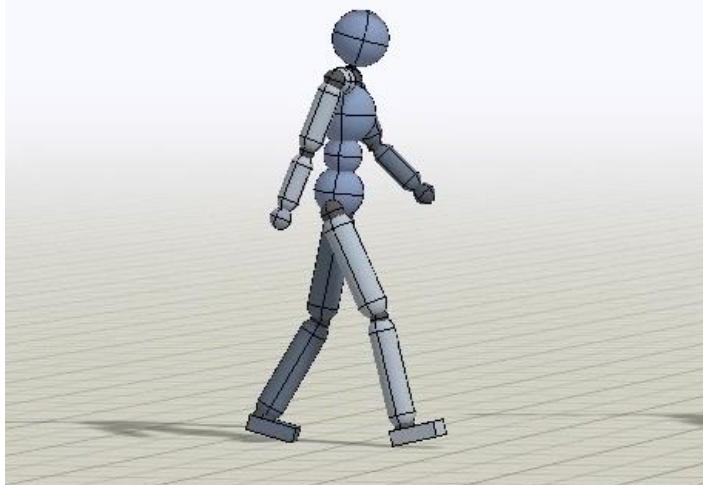
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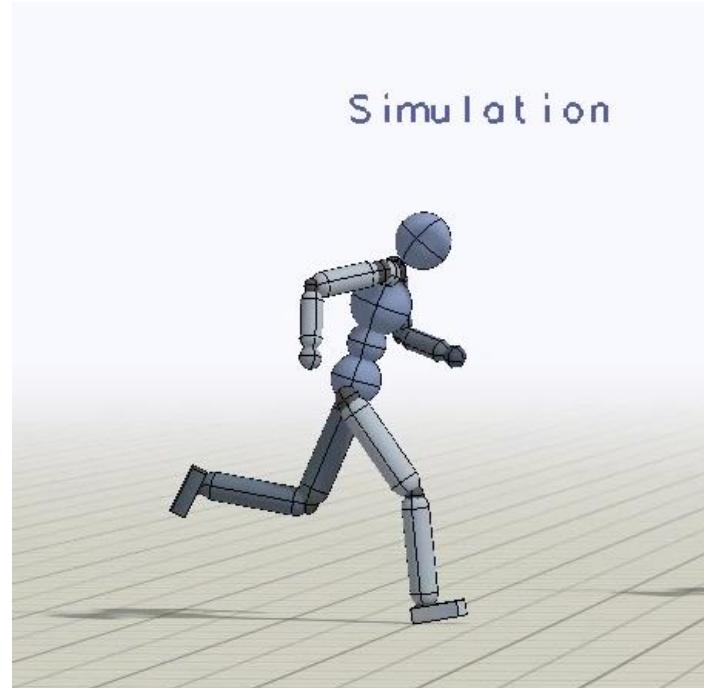
# Locomotion

---

Simulation



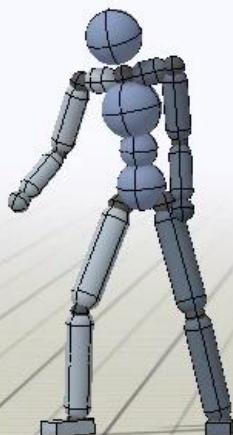
Simulation



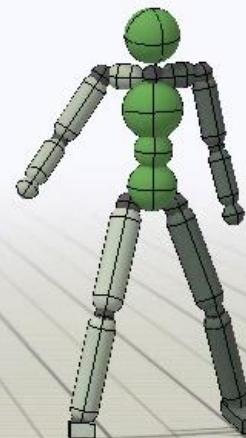
# Humanoid: Cartwheel

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Simulation



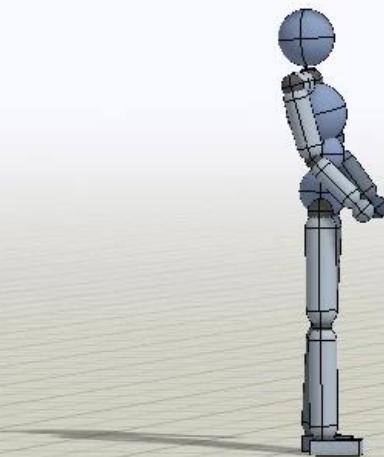
Reference



# Humanoid: Backflip

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Simulation



Reference



# Humanoid: Frontflip

Simulation



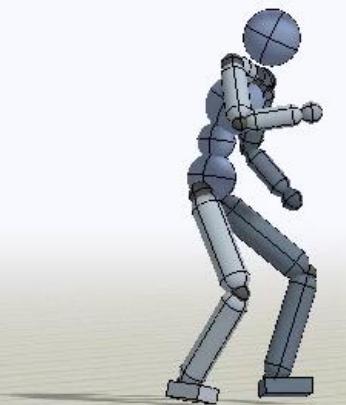
Reference



# Humanoid: Roll

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Simulation



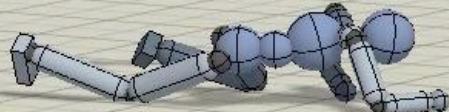
Reference



# Humanoid: Crawl

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Simulation



Reference



# Humanoid: Dance A

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Simulation

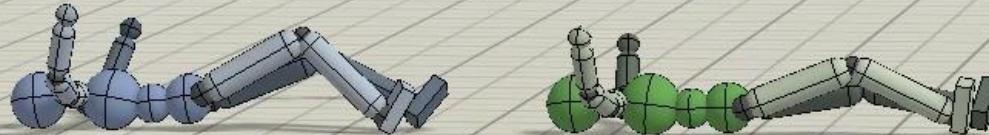


Reference

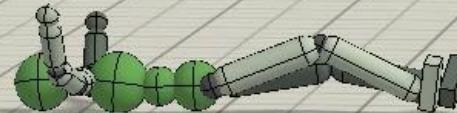


# Humanoid: Kip-Up

Simulation

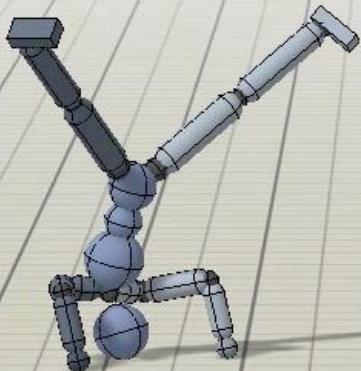


Reference

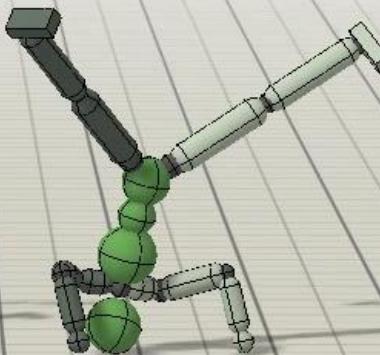


# Humanoid: Headspin

Simulation

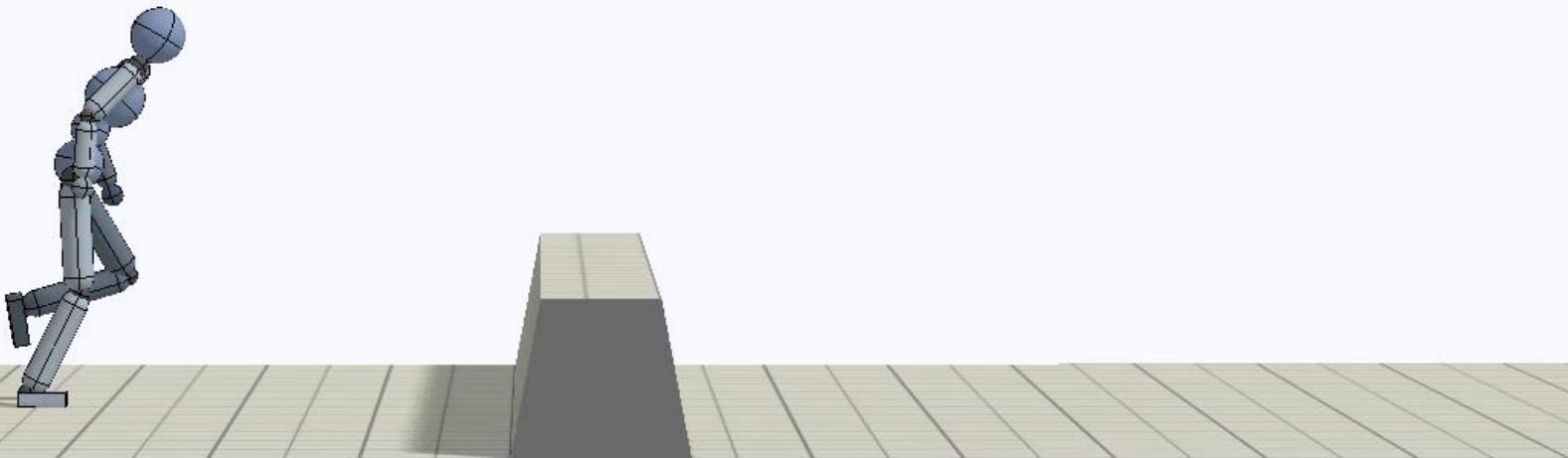


Reference



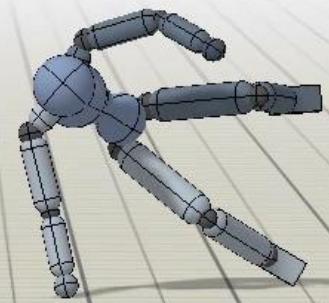
# Humanoid: Vault 1-Handed

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# Humanoid: Flare

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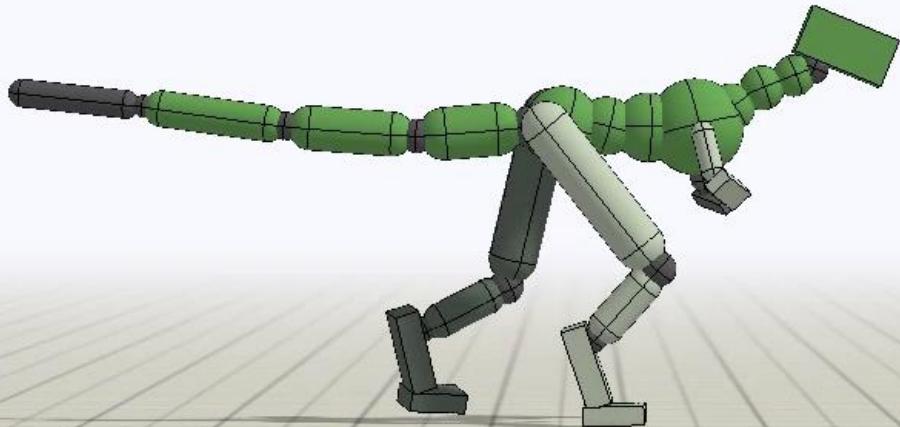


# 20+ Skills

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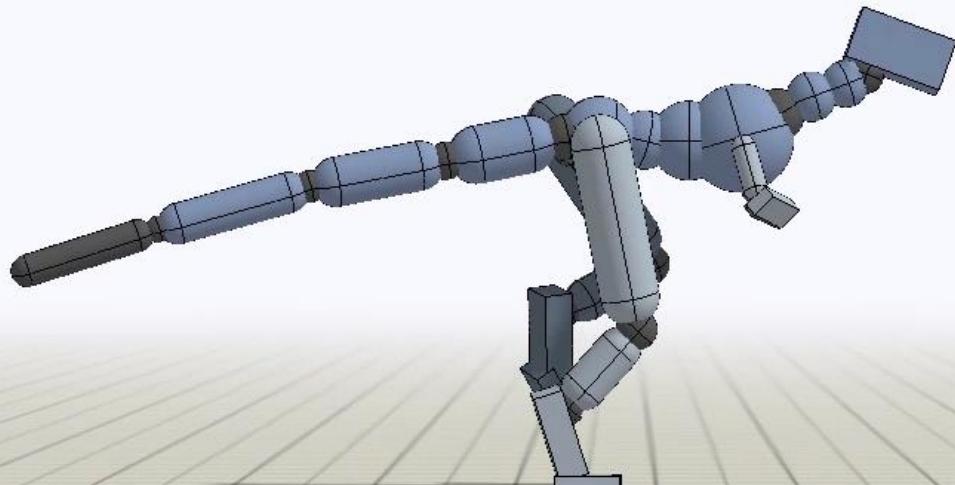


# Keyframe Animation



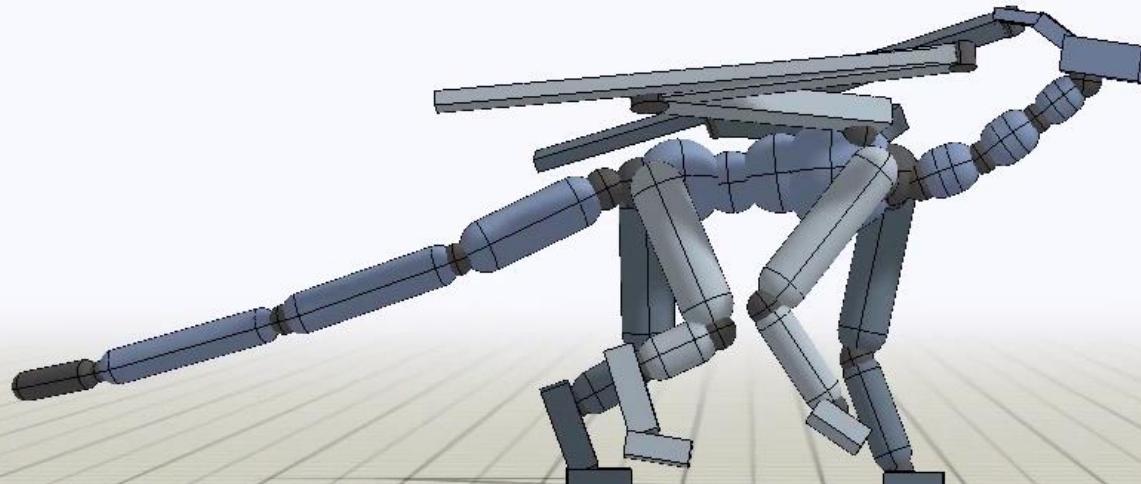
# T-Rex: Walk

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Simulated Character

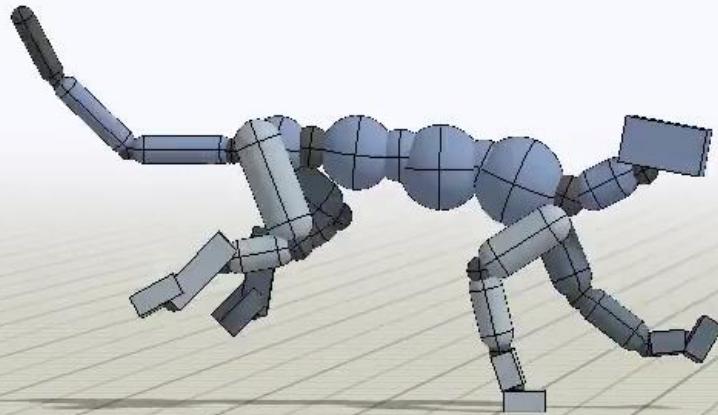
# Dragon: Walk



Simulated Character

# Lion: Run

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Simulated Character

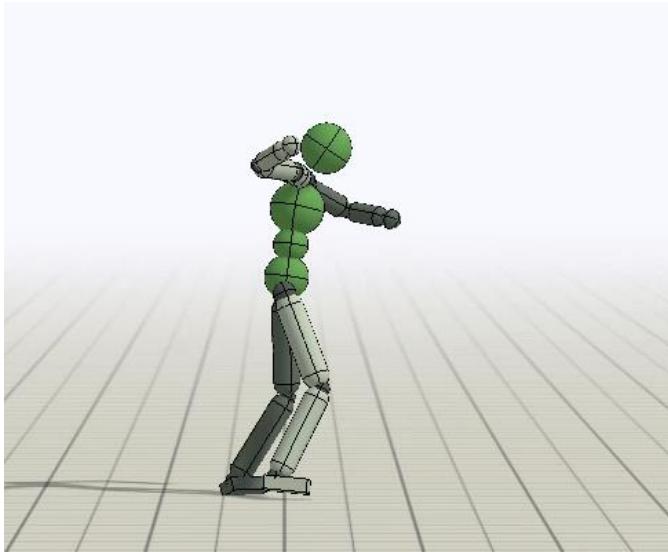
 ZIVA

# Tasks

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# Tasks

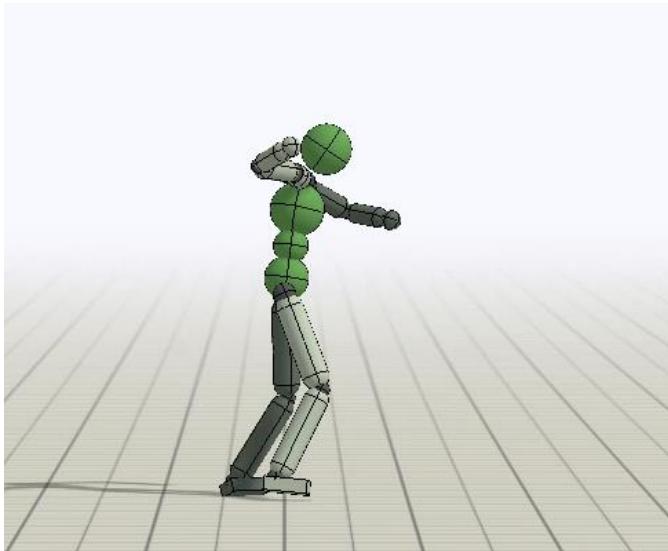
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Reference Motion

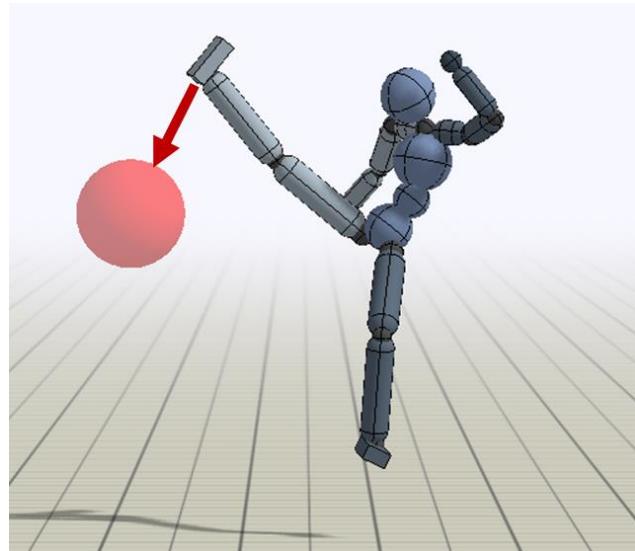
# Tasks

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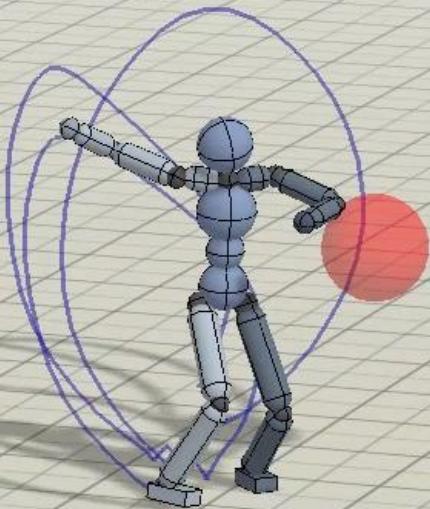
Reference Motion

+



Task

# Humanoid: Spinkick - Strike



# Humanoid: Baseball Pitch - Throw



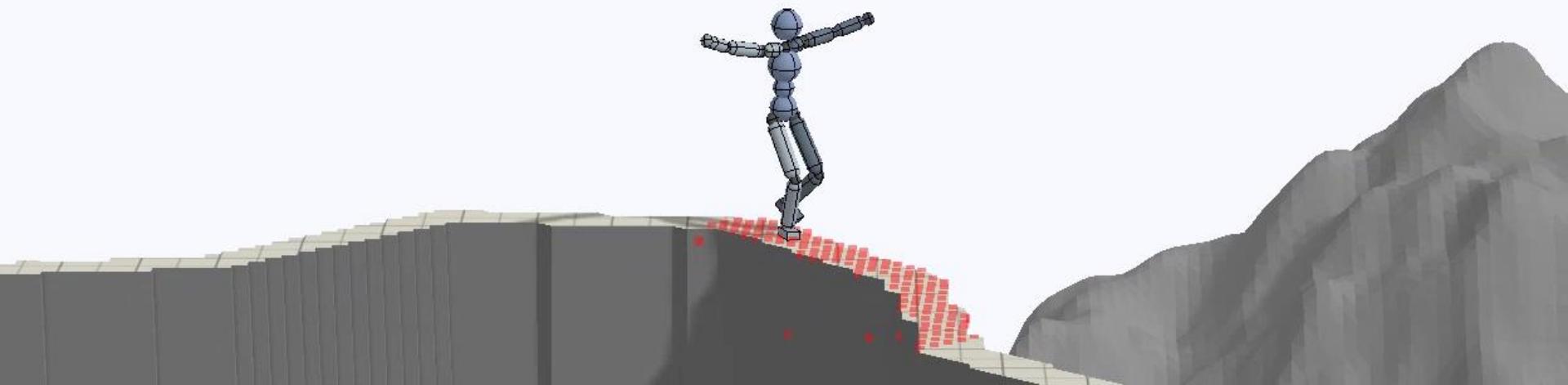
# No Reference Motion

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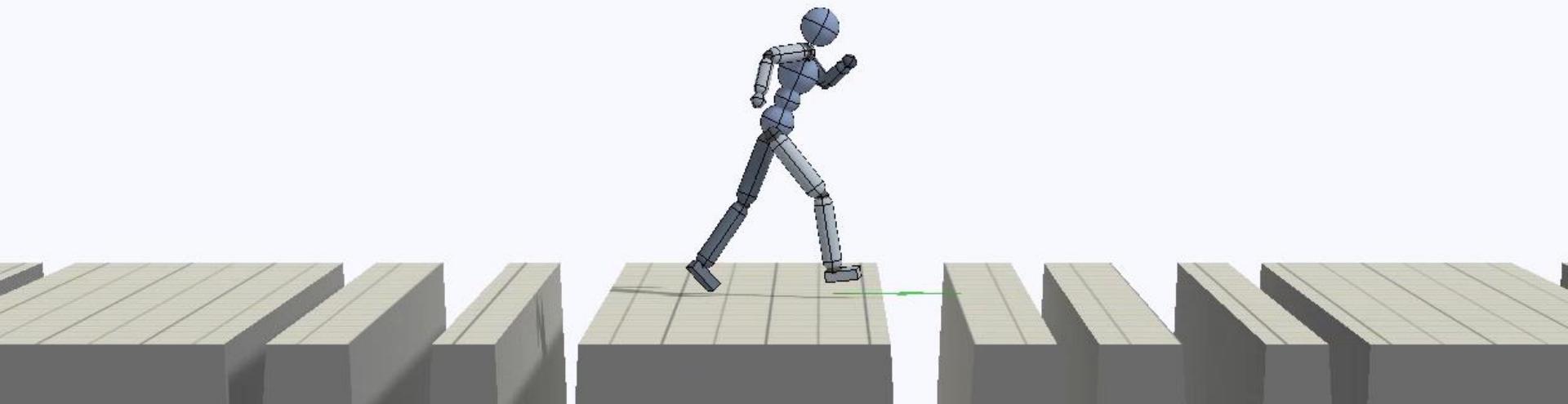


# Humanoid: Balance Beam

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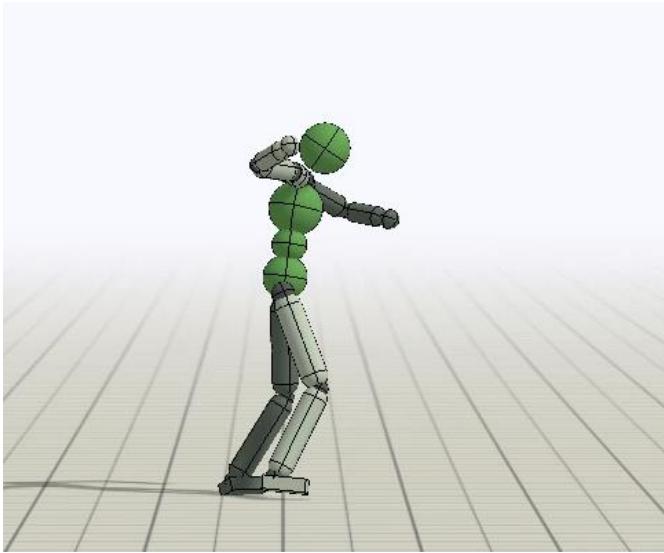
# Humanoid: Run – Dense Gaps



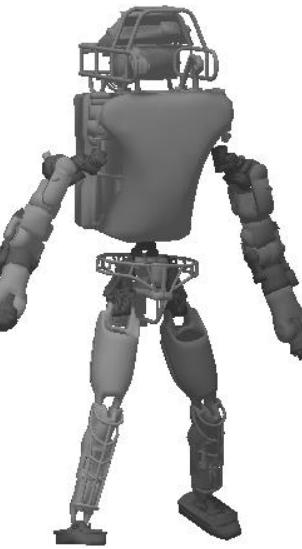
# Retargeting

# Character Retargeting

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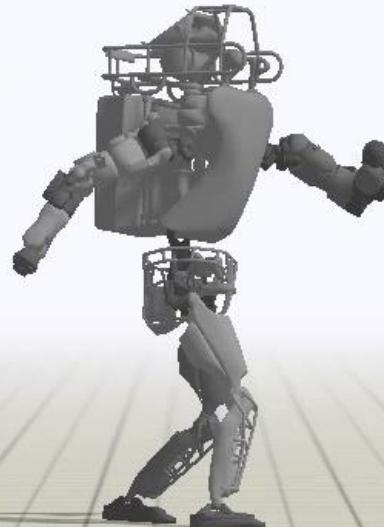
Reference Motion



Atlas

# Atlas: Spinkick

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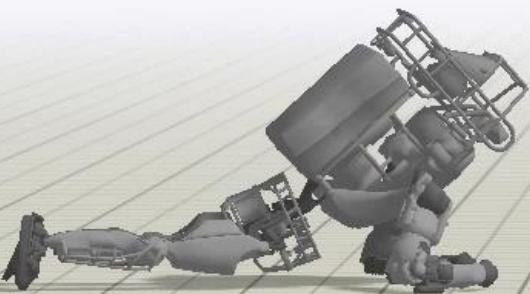


# Atlas: Run

---



# Atlas: Getup-Facedown



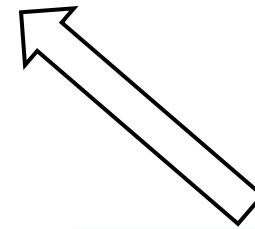
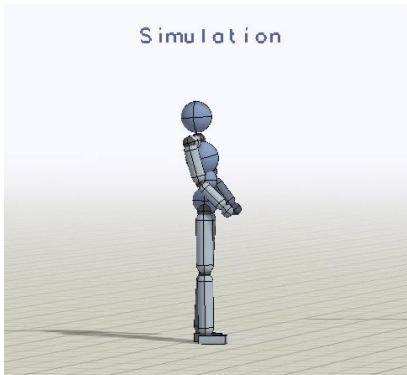
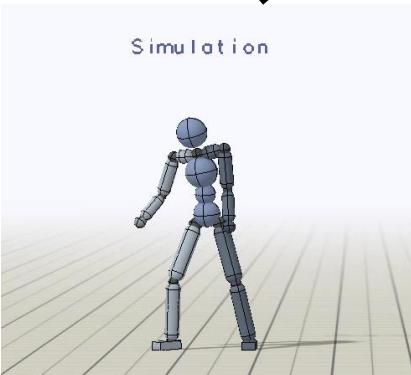
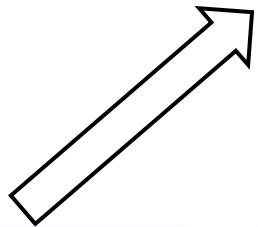
# Atlas: Backflip



# Multi-Clip Integration

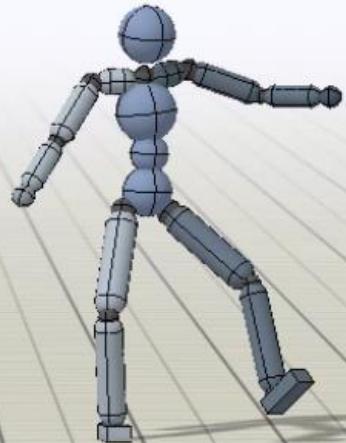
# Multi-Clip Integration

$$\Pi(a|s) = \sum_{i=1}^k p^i(s)\pi^i(a|s)$$



# Multi-Clip Integration

Left Cartwheel



# Learning from Mocap



# Mocap is a Hassle



[Holden 2018]

# Skills From Videos

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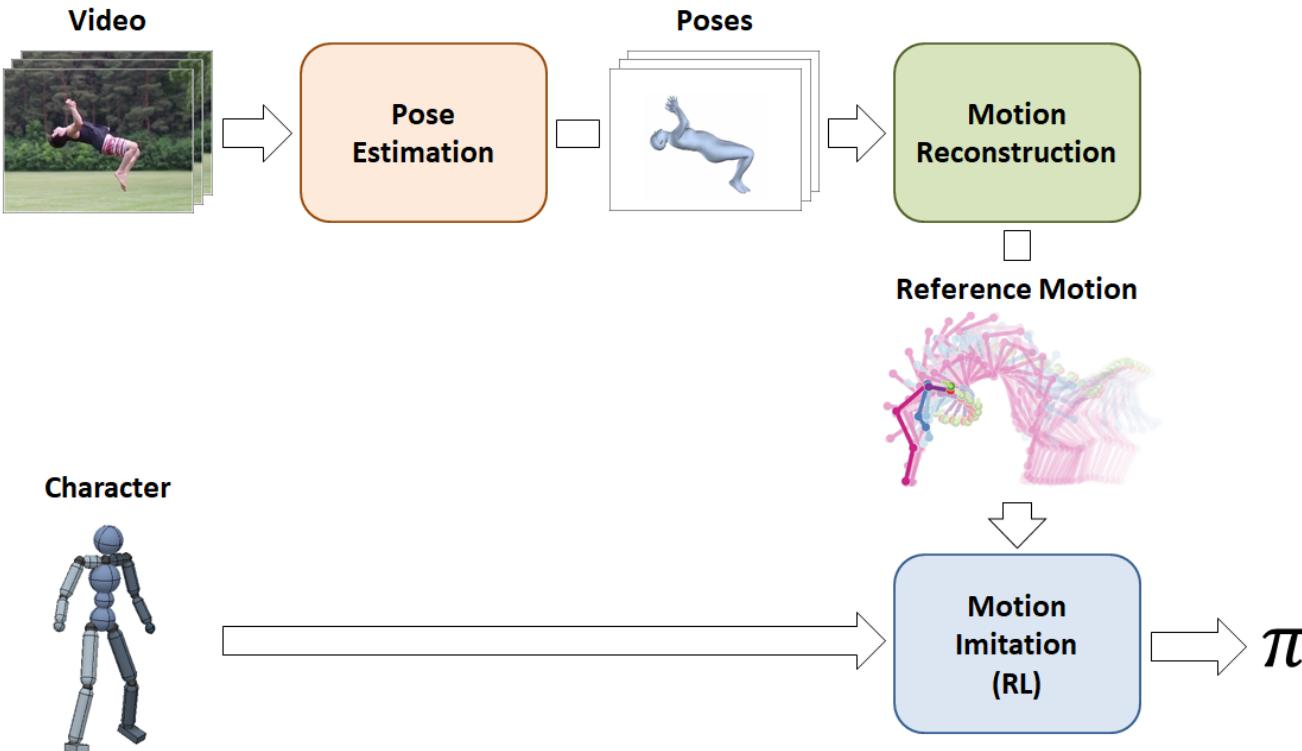
# Learning from Videos



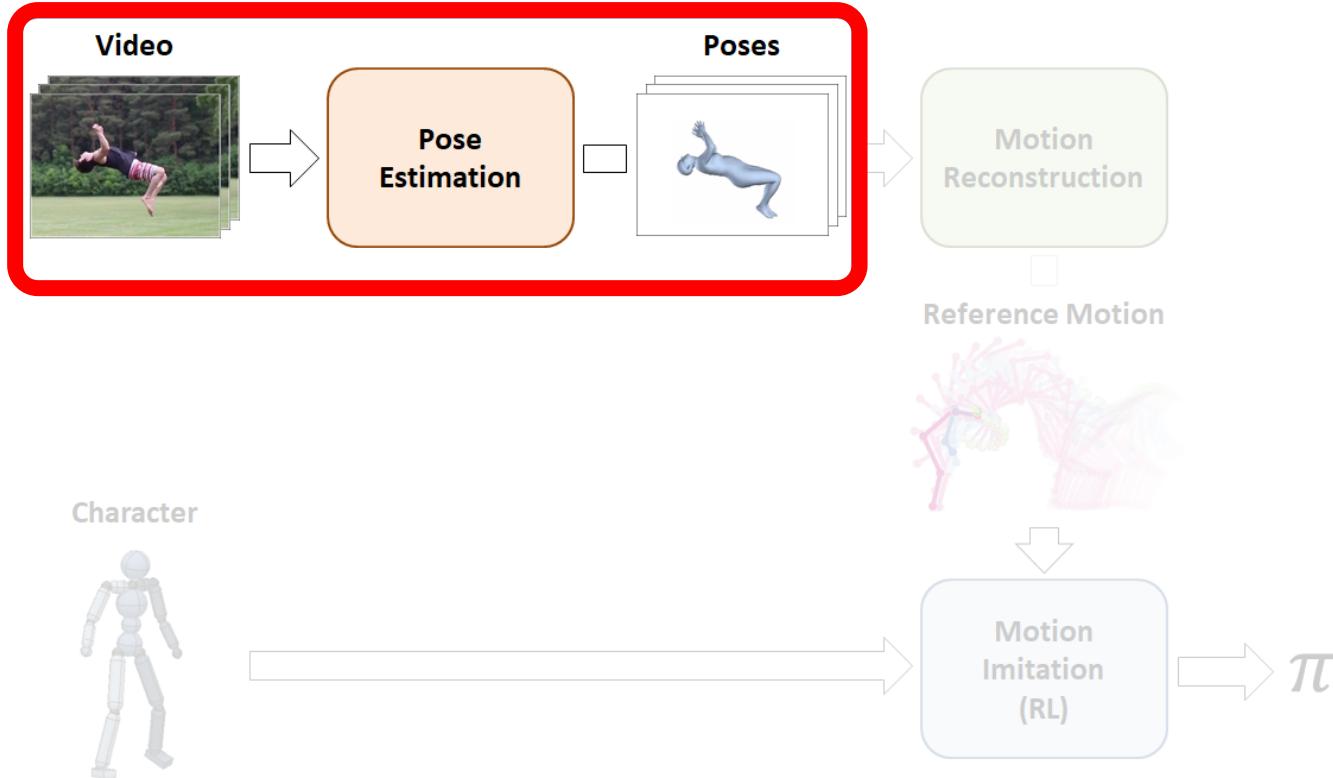
Video

# Overview

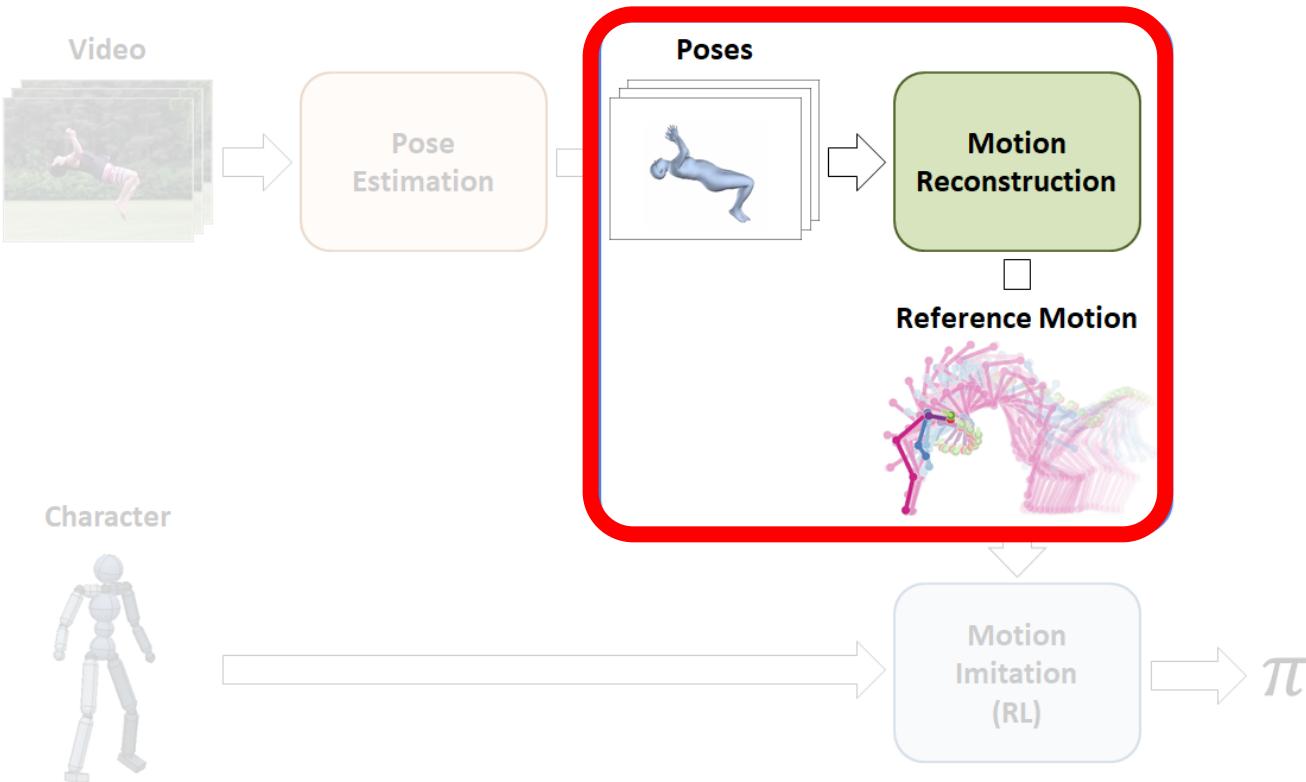
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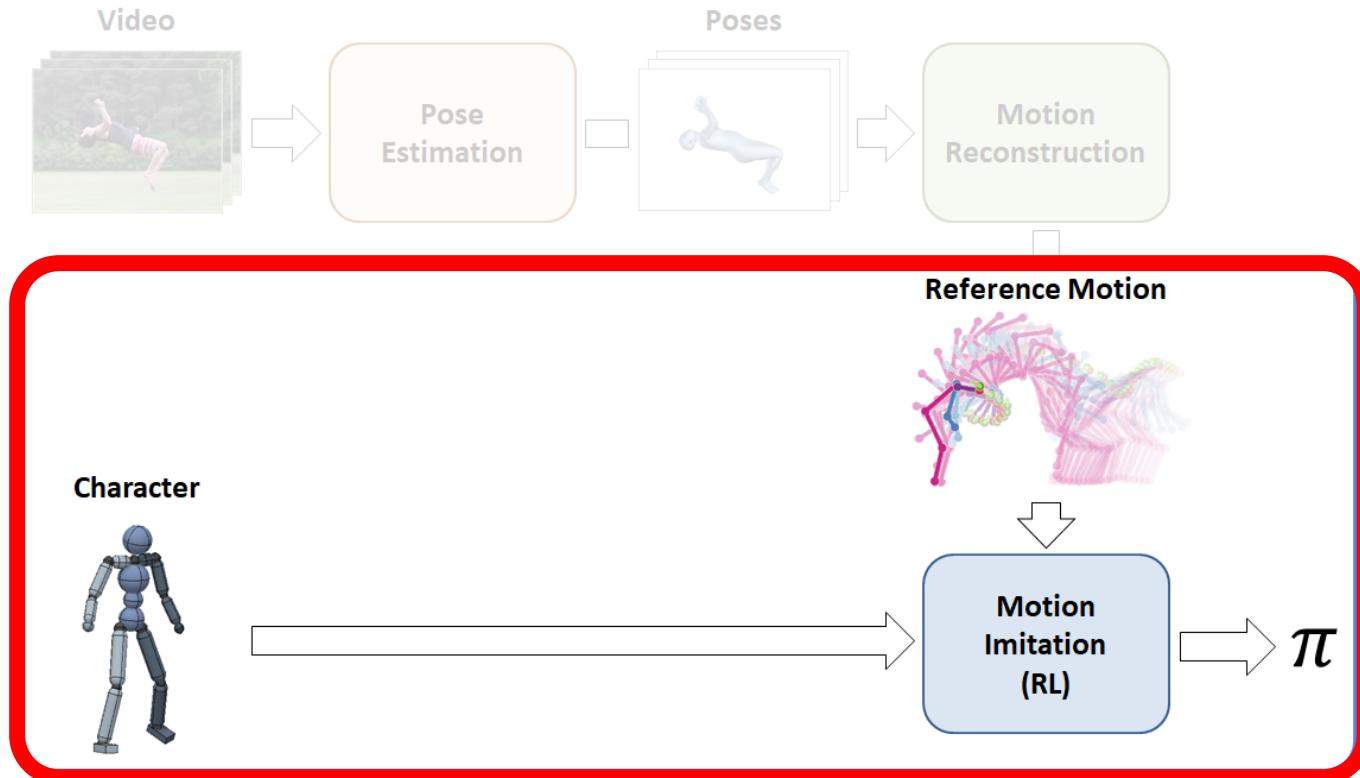
# Overview



# Overview

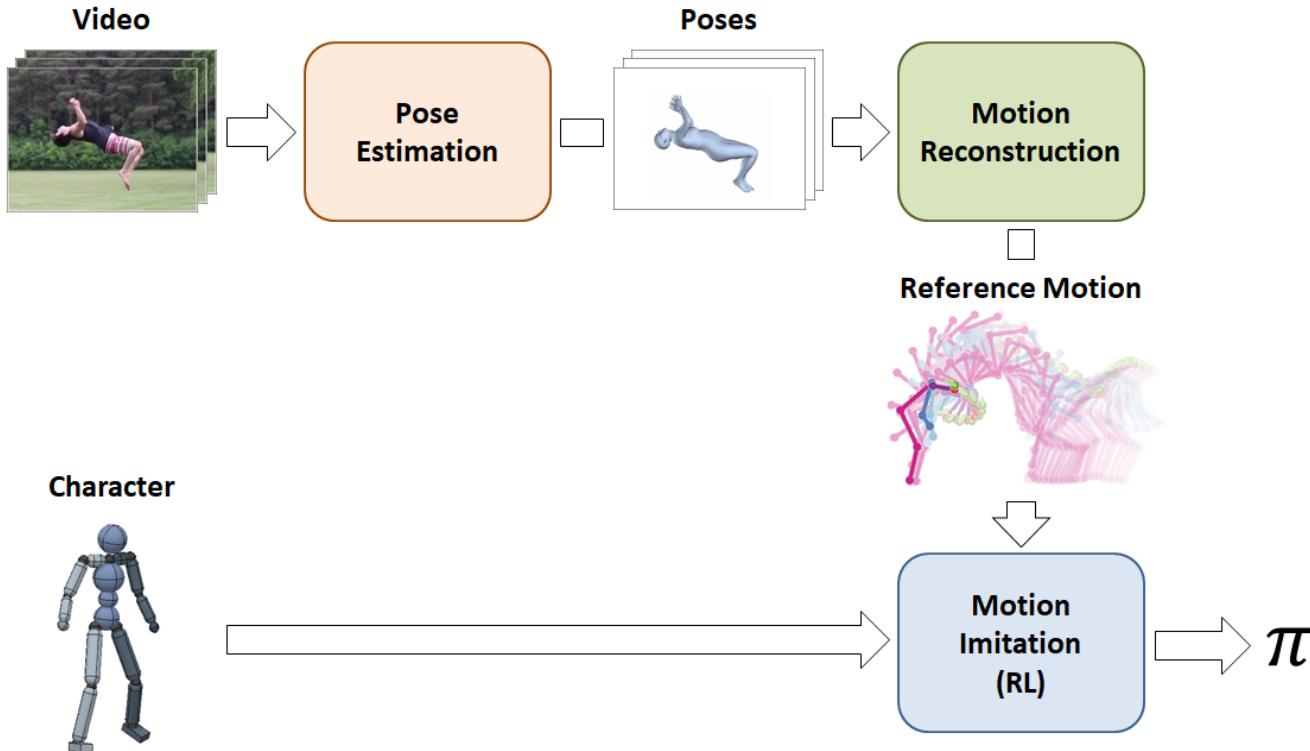


# Overview



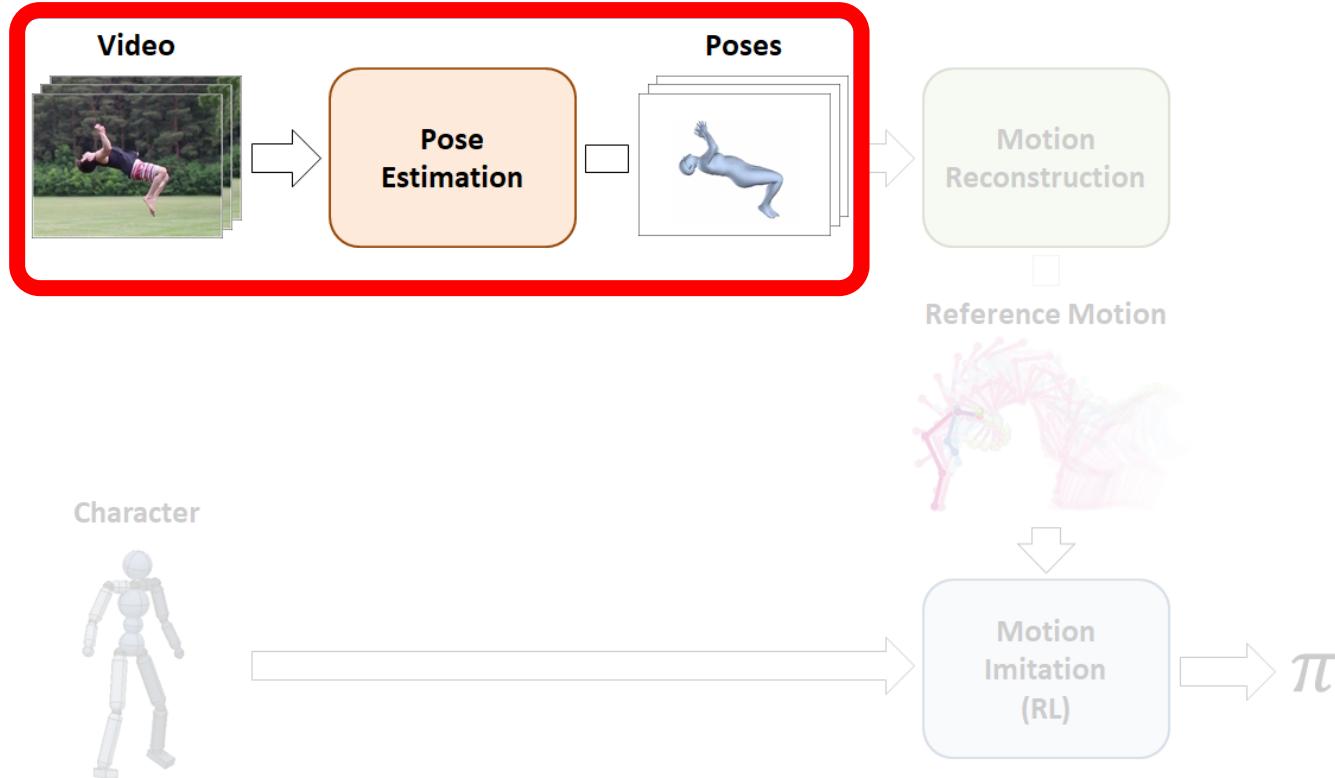
# Overview

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# Pose Estimation

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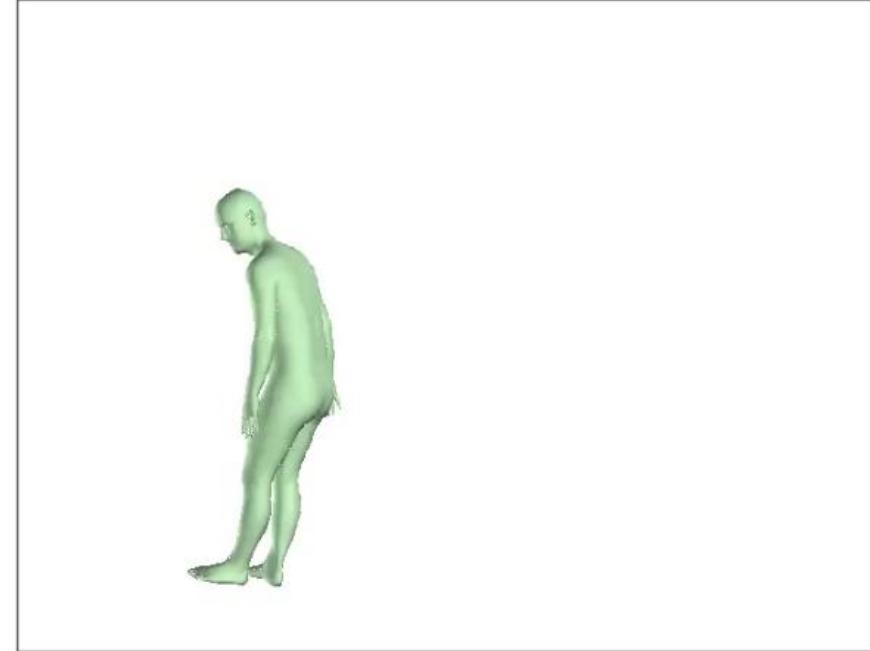


# Pose Estimation

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Video: Handspring A



Pose Prediction

# Human Mesh Recovery (HMR)



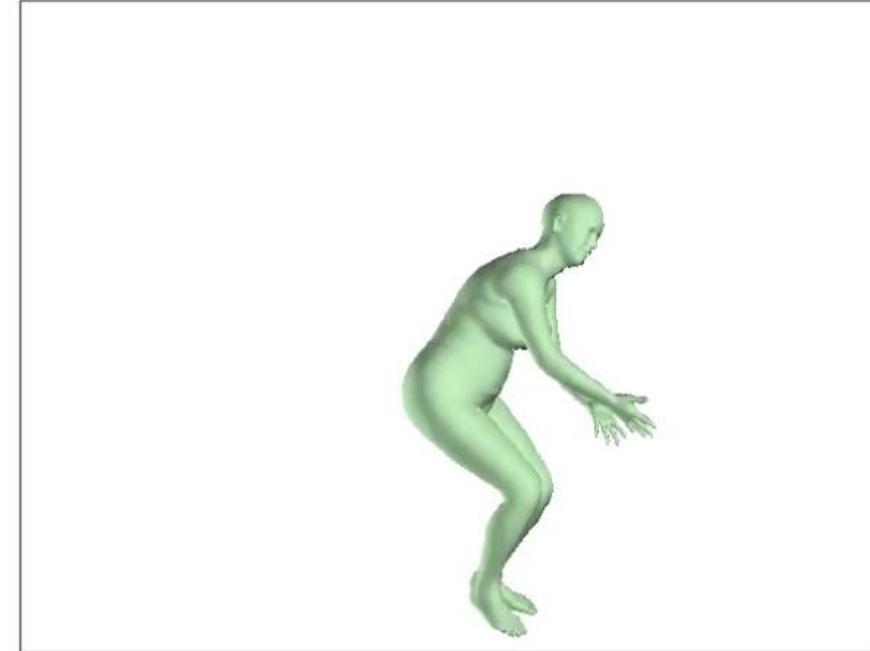
[Kanazawa et al., 2018]

# Pose Estimation

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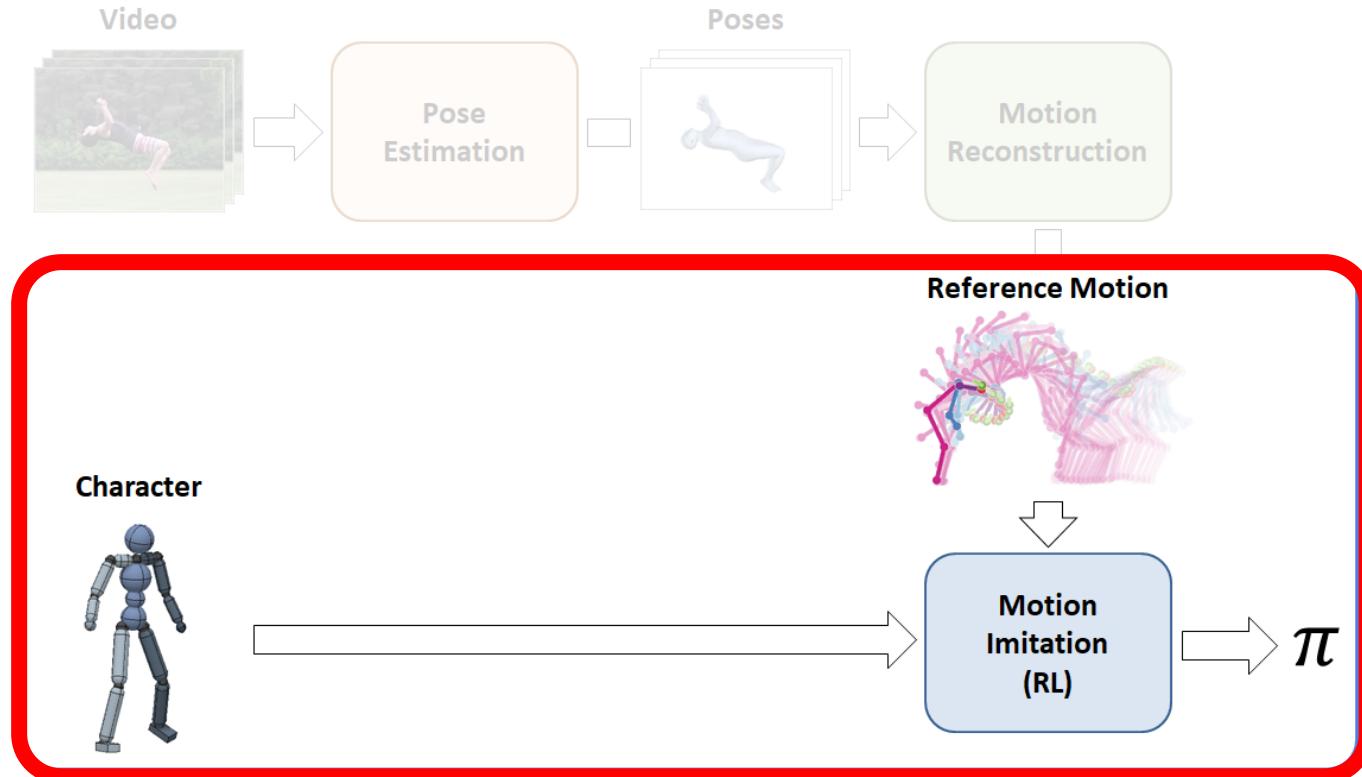


Video: Backflip A



Pose Prediction

# Motion Imitation



# Motion Imitation

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$$r_t = \exp \left( - \left\| \text{Humanoid Pose} - \text{Robot Pose} \right\|^2 \right)$$



# Motion Imitation

---

$$r_t = \exp \left( - \| \text{Human Pose} - \text{Robot Pose} \|^2 \right)$$

# Motion Imitation

---

$$r_t = \exp \left( - \left\| \begin{array}{c} \text{Humanoid in a handstand pose} \\ \xrightarrow{\hspace{1cm}} \\ \text{Robot in a handstand pose} \end{array} \right\|^2 \right)$$

# Humanoid: Cartwheel B

---



Video: Cartwheel B



Reference Motion

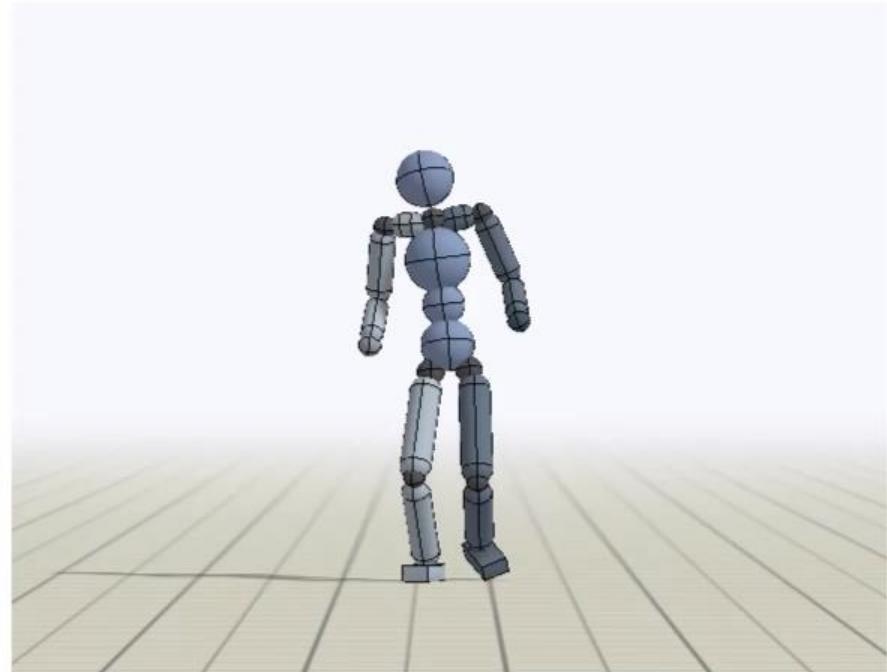


Policy

# Humanoid: Jumping Jack



Video: Jumping Jack

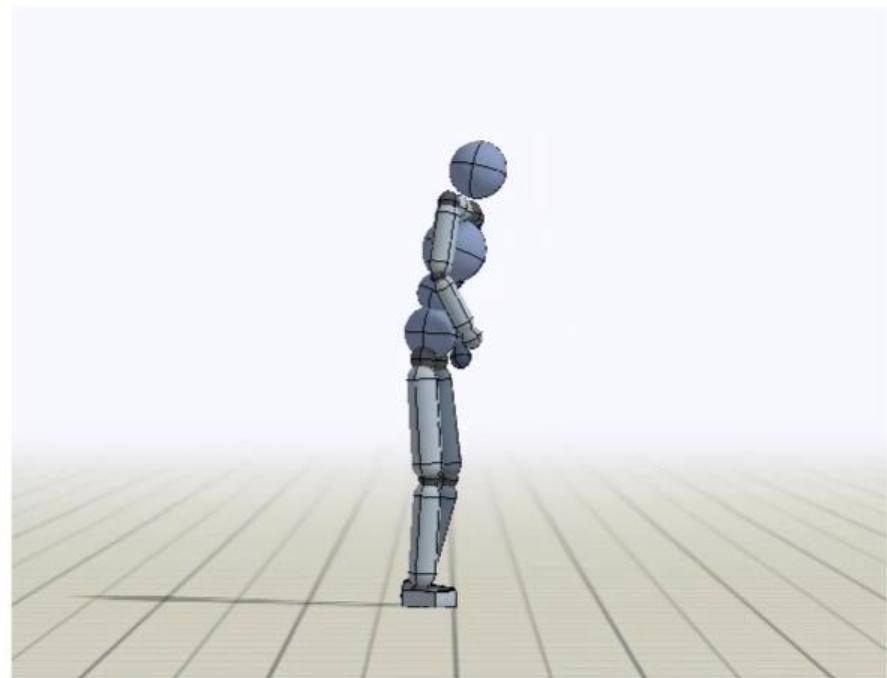


Policy

# Humanoid: Backflip B



Video: Backflip B



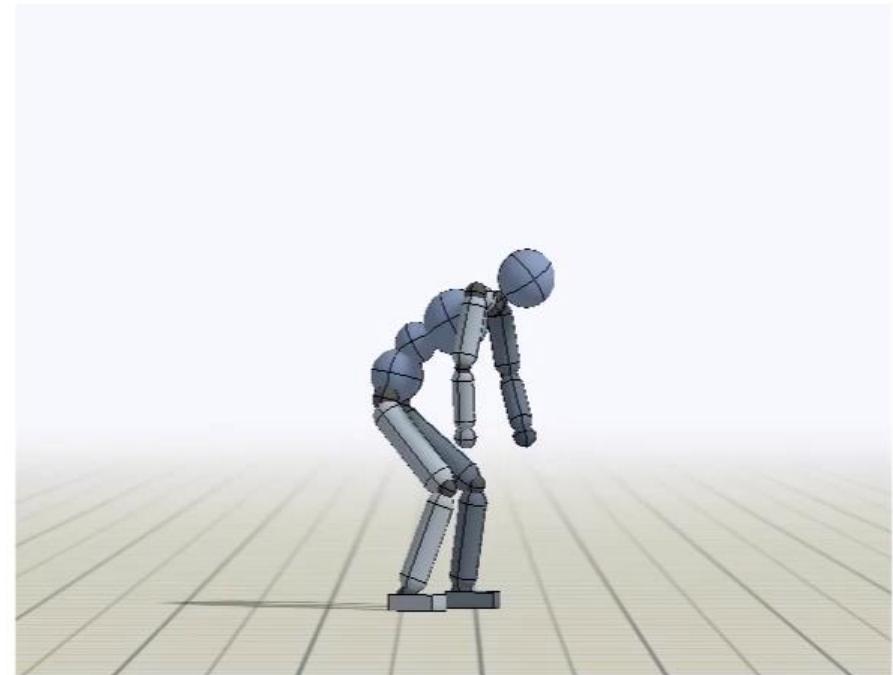
Policy

# Humanoid: Frontflip

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Video: Frontflip



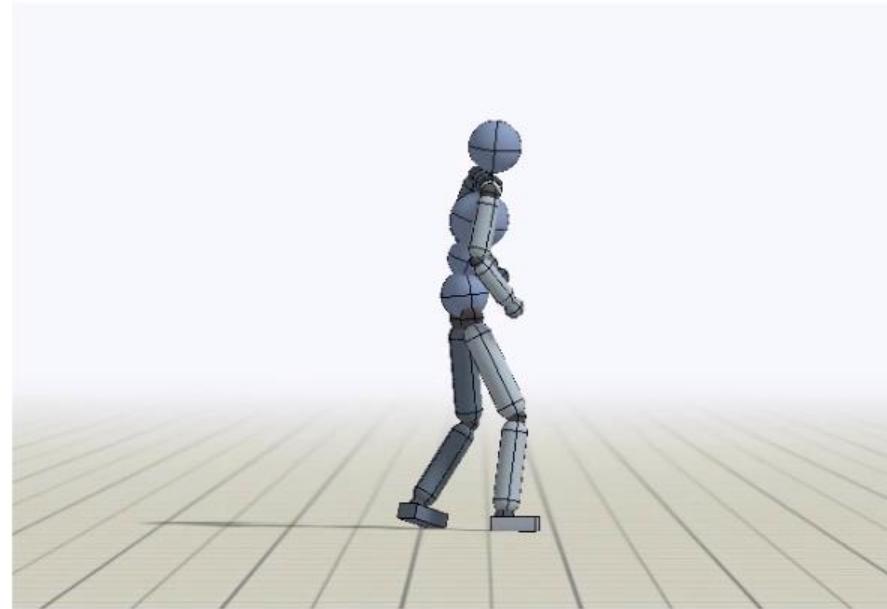
Policy

# Humanoid: Roll

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Video: Roll

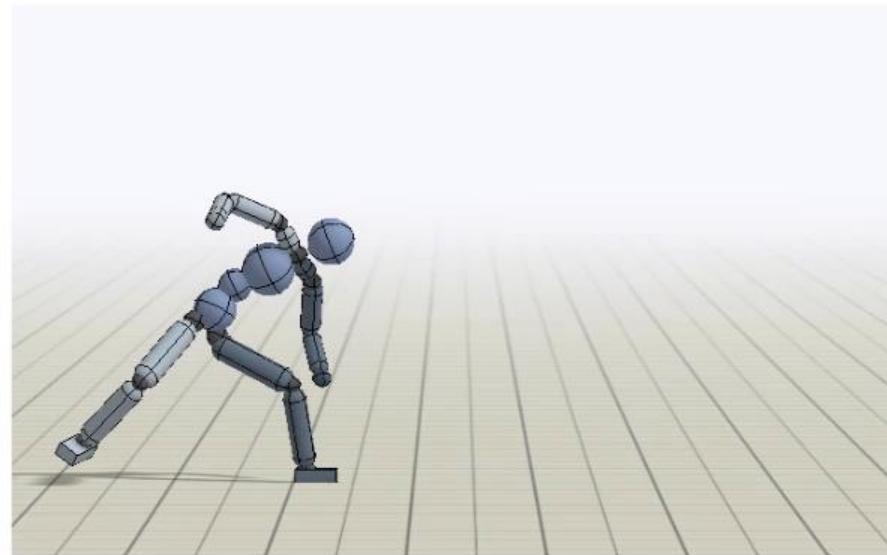


Policy

# Humanoid: Spin



Video: Spin

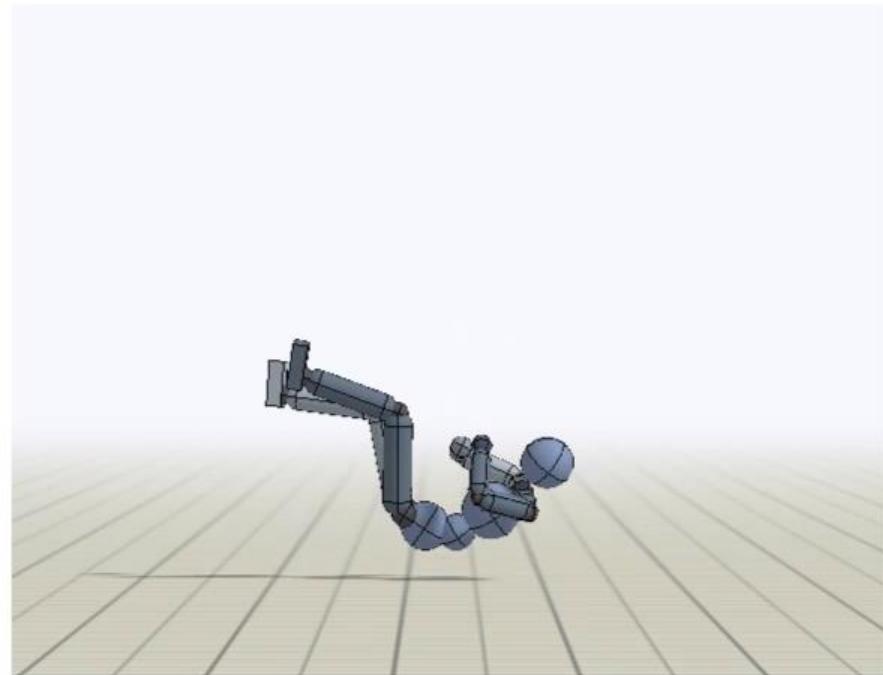


Policy

# Humanoid: Kip-Up



Video: Kip-Up



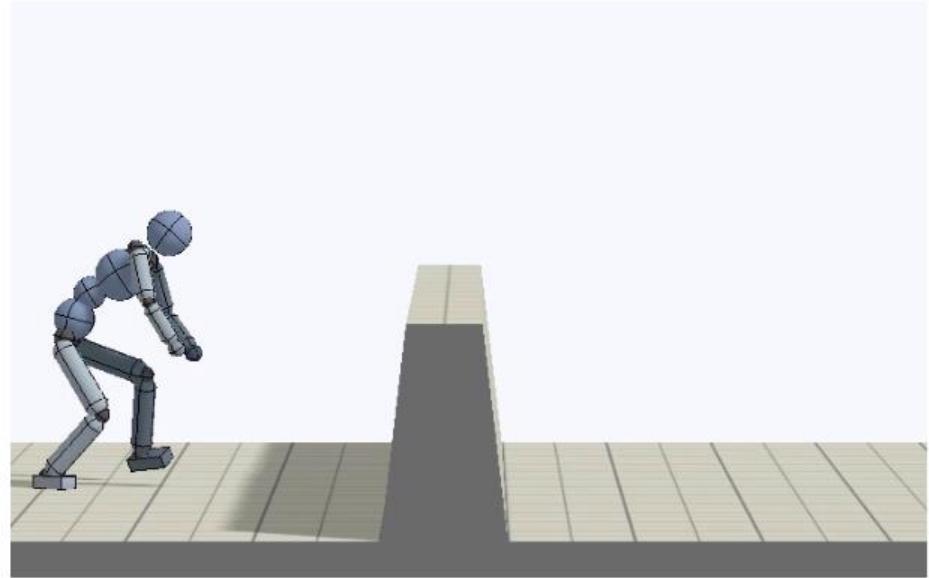
Policy

# Humanoid: Vault

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Video: Vault

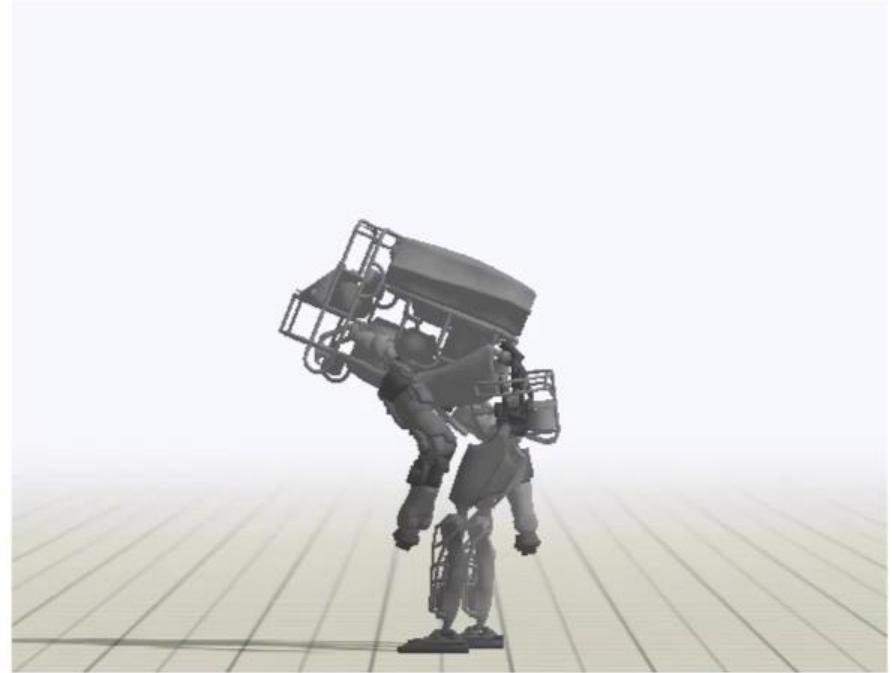


Policy

# Atlas: Handspring A



Video: Handspring A

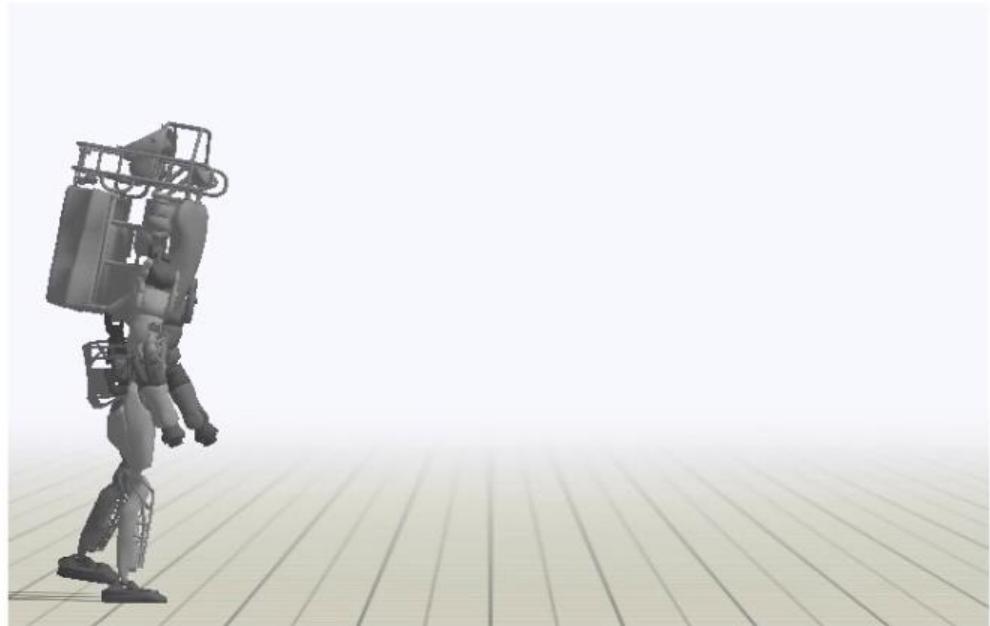


Policy

# Atlas: Jump



Video: Jump



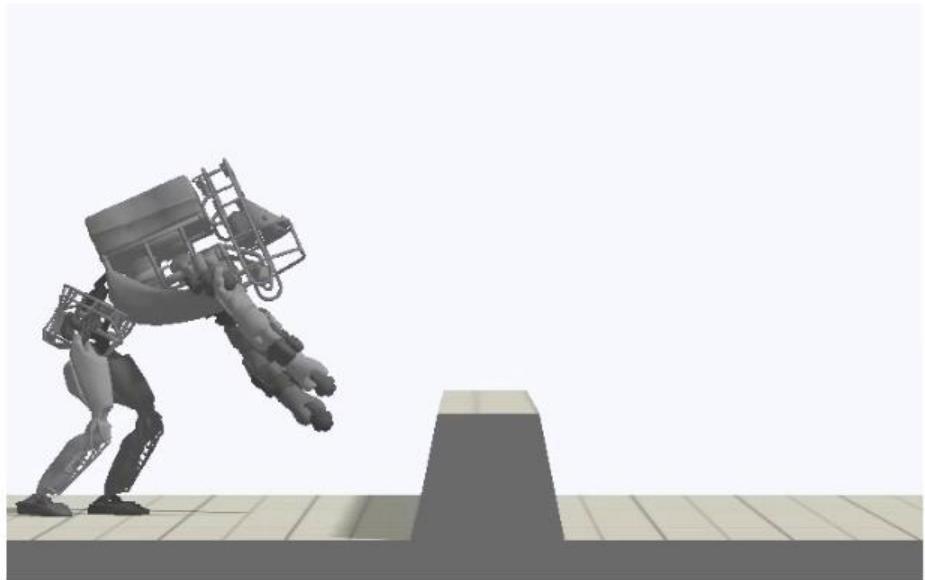
Policy

# Atlas: Vault

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Video: Vault



Policy

# Atlas: Dance

---

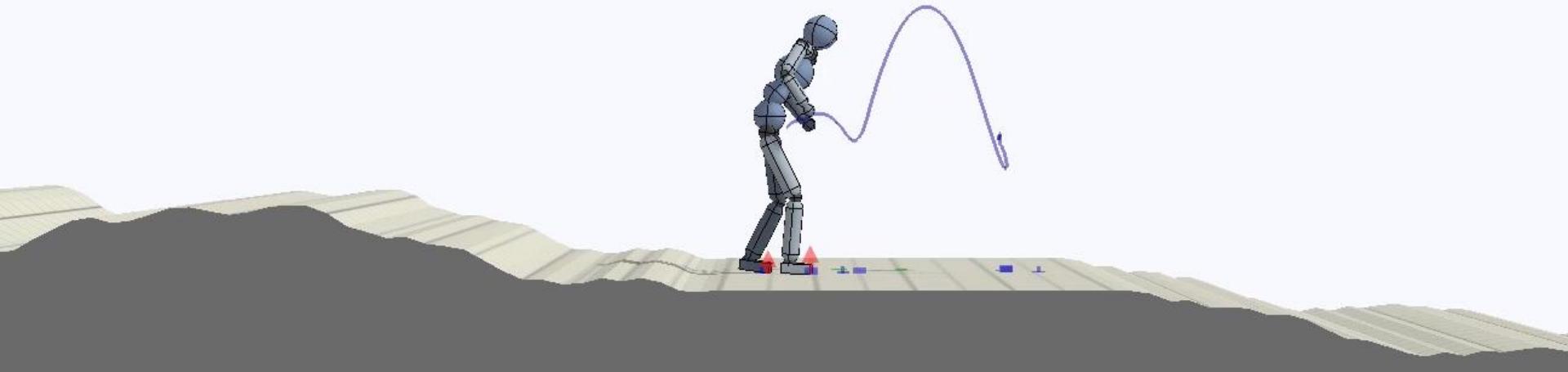


Video: Dance

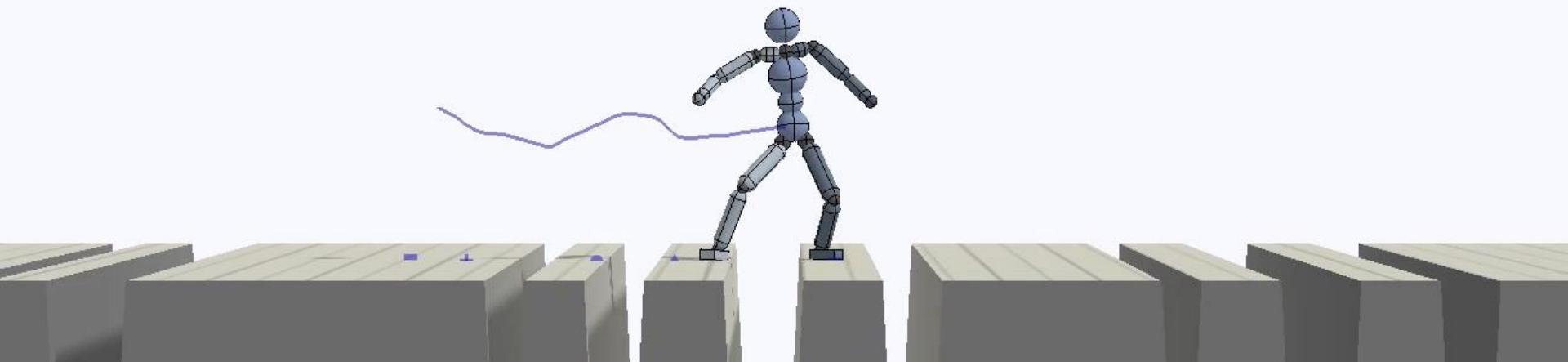


Policy

# Environment Retargeting



# Environment Retargeting

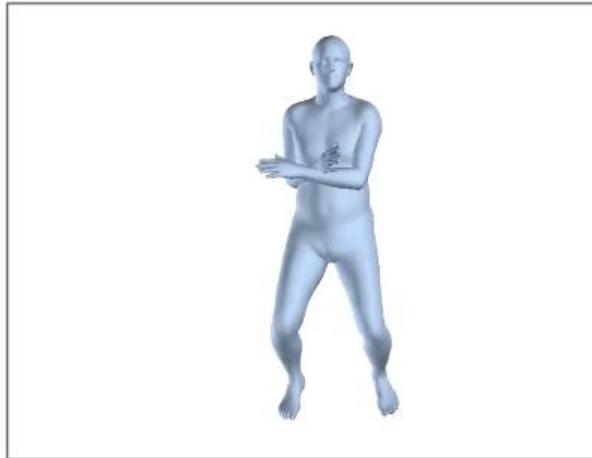


# Failure Cases

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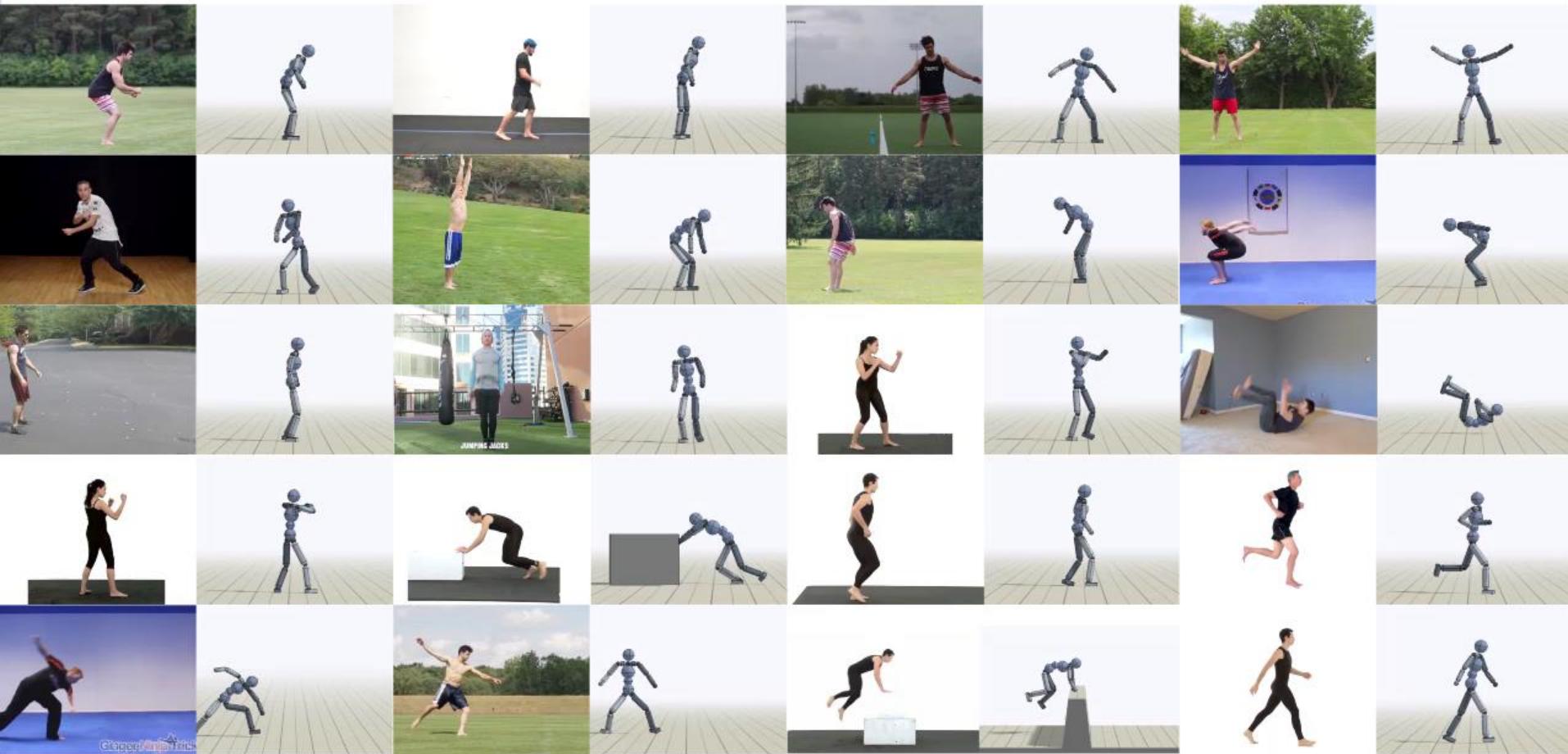
Video: Gangnam Style



Reference Motion

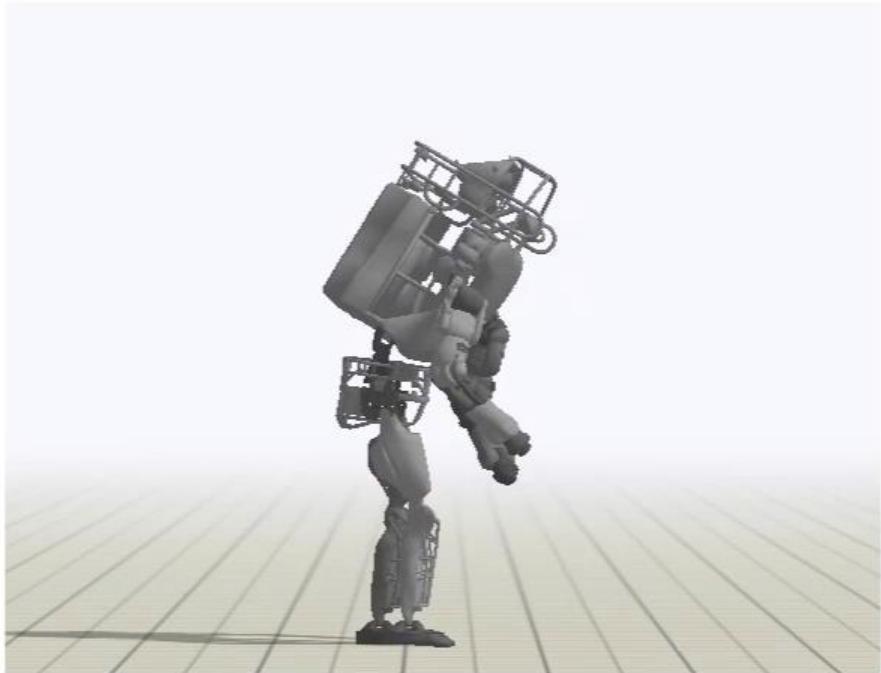


Simulation



# Skills from Videos

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Policy

# Concluding Remarks

- Simple method can learn a large repertoire of skills
- Minimizing tracking error works (surprisingly) well

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- Simple method can learn a large repertoire of skills
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- A lot of room for improvement for video imitation
  - More end-to-end approach
  - Outdoor sports
  - Multiple actors

# Concluding Remarks

- Simple method can learn a large repertoire of skills
- Minimizing tracking error works (surprisingly) well
- A lot of room for improvement for video imitation
  - More end-to-end approach
  - Outdoor sports
  - Multiple actors
- Code: <https://github.com/xbpeng/DeepMimic>

# Collaborators

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Pieter Abbeel



Angjoo Kanazawa



Sergey Levine



Jitendra Malik



Michiel van de Panne

# Questions?

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