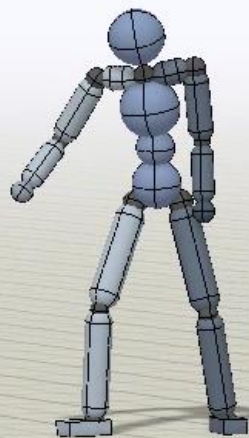


Towards a Virtual Stuntman



Xue Bin (Jason) Peng

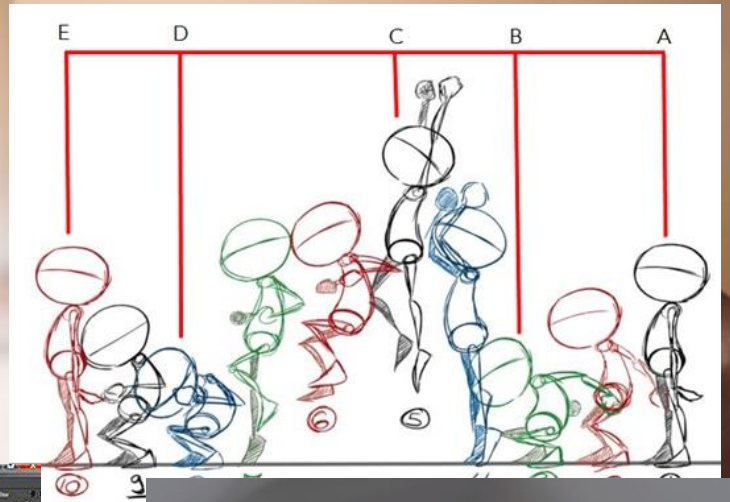
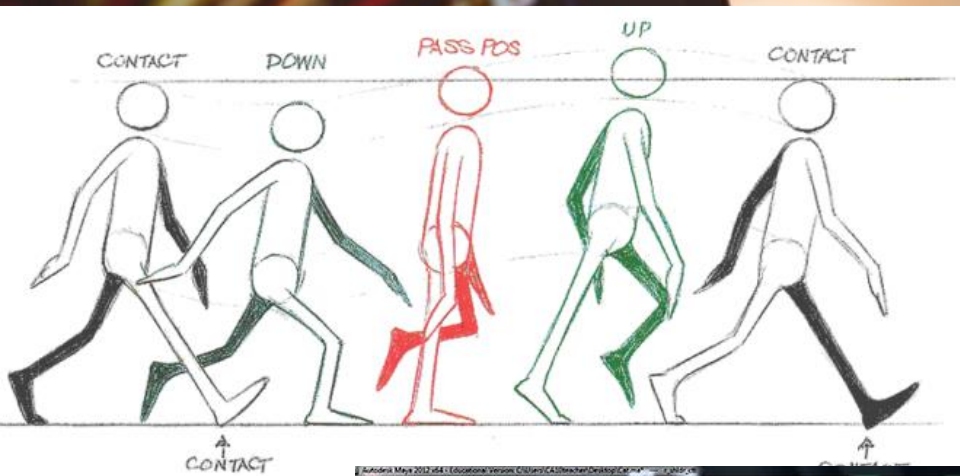
UC Berkeley



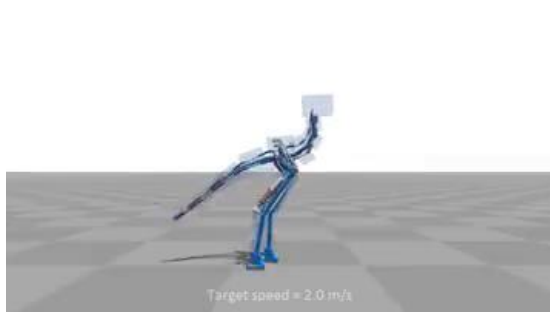
Animation



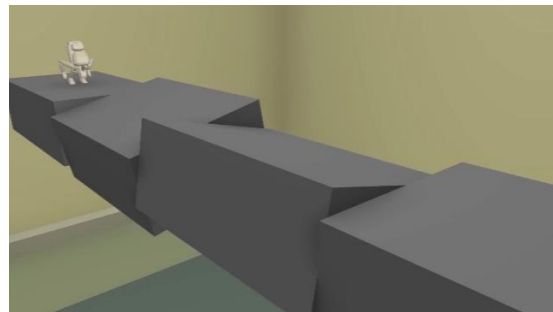
Animation



Computer Animation



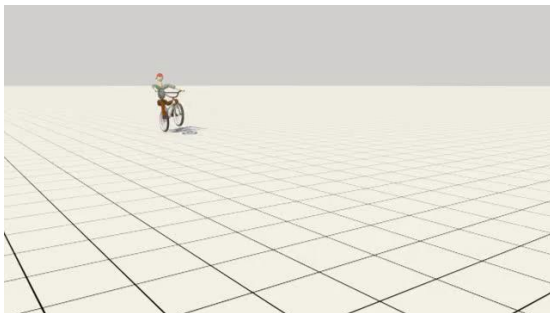
[Geijtenbeek et al. 2013]



[Brown et al. 2013]



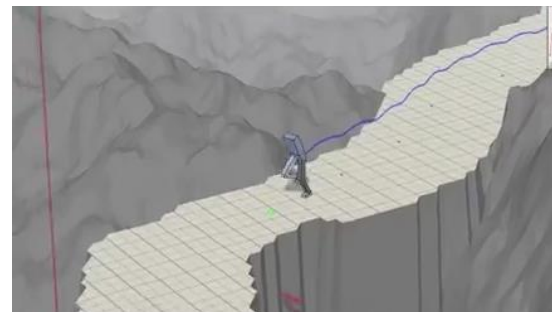
[Ju et al. 2013]



[Tan et al. 2014]



[Kwon and Hodgins 2017]

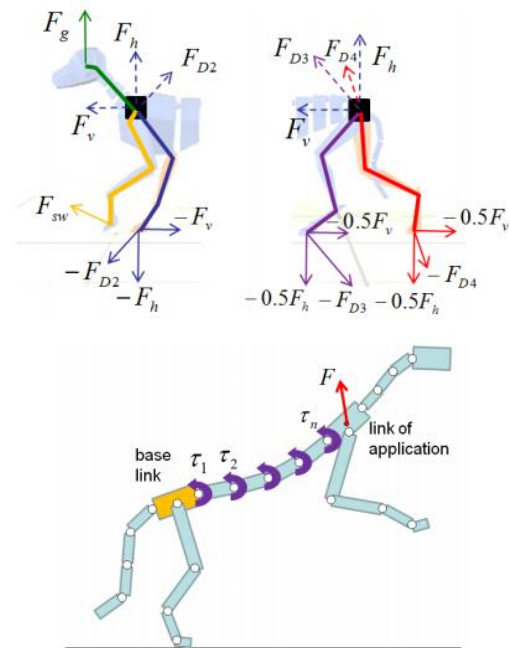


[Peng et al. 2018]

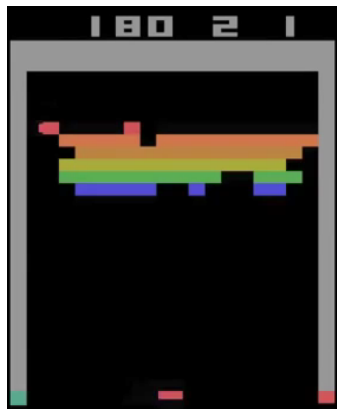
Physics-Based Animation



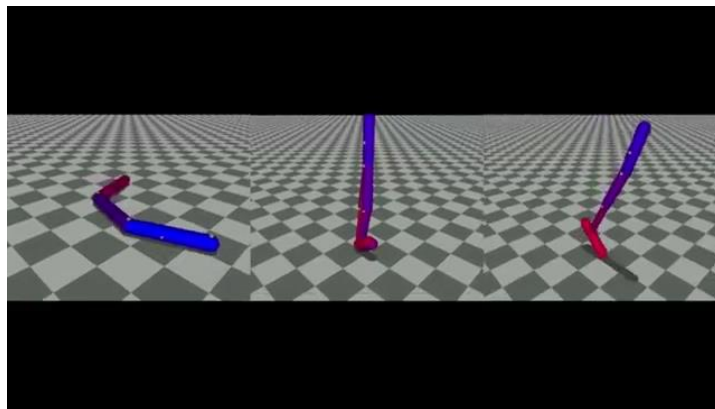
[Coros et al. 2011]



Deep RL



[Mnih et al. 2015]

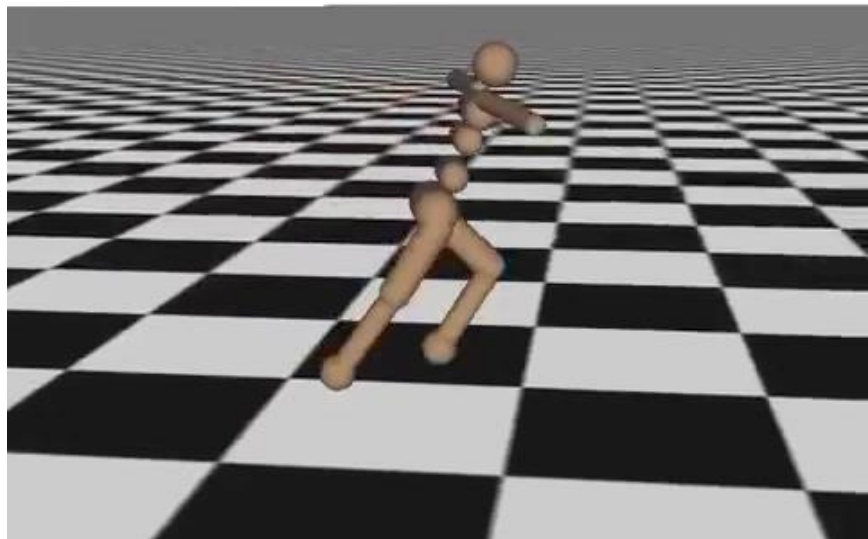


[Schulman et al. 2016]



[Chebotar et al. 2017]

Motion Quality

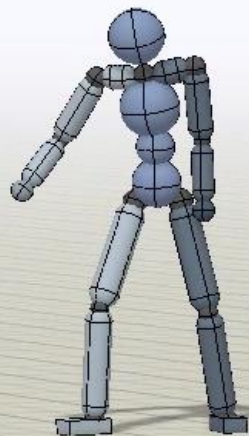


[Schulman et al. 2016]

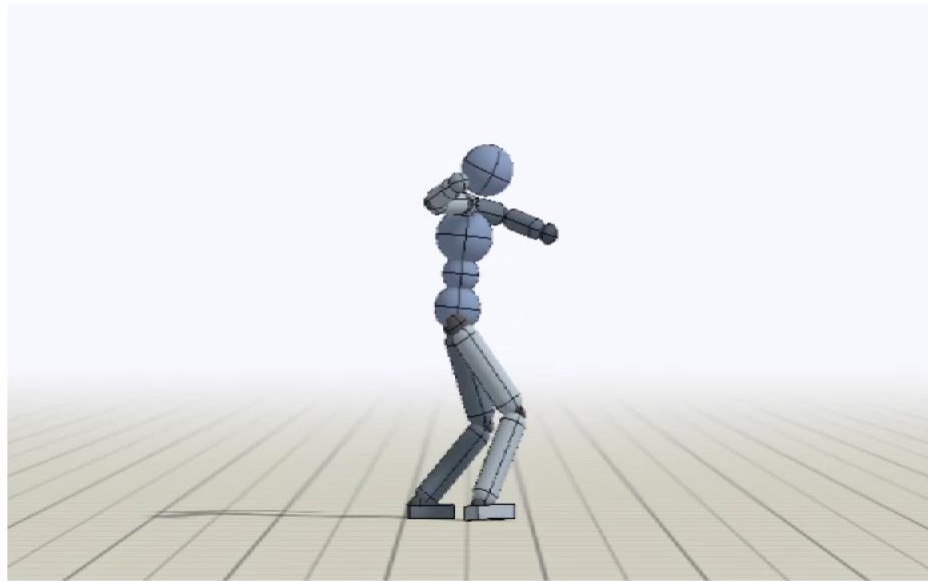
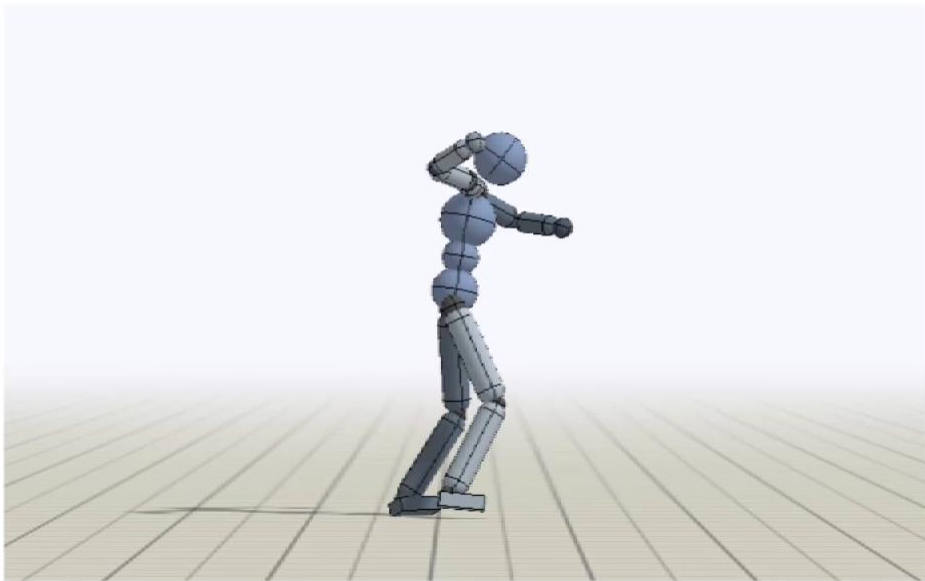


[Heess et al. 2017]

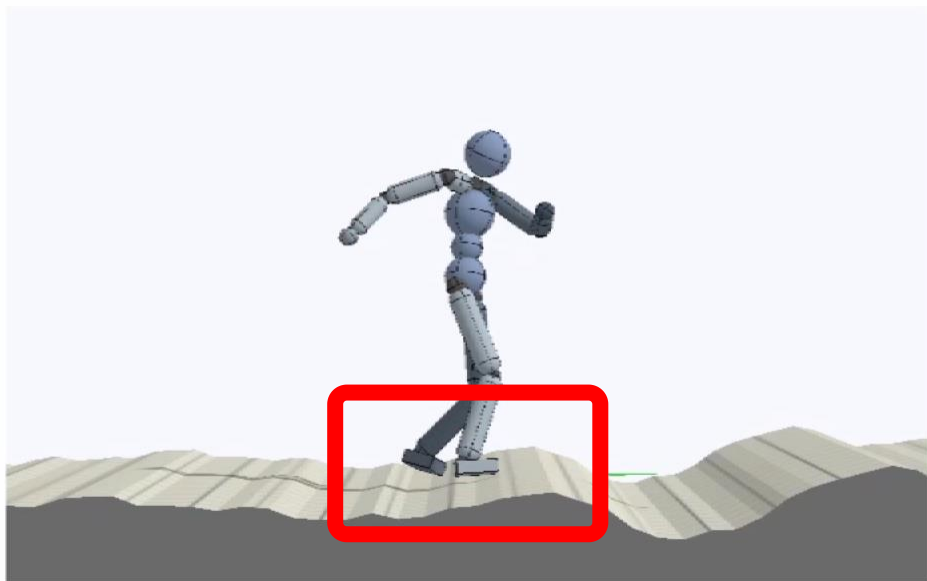
Motivation



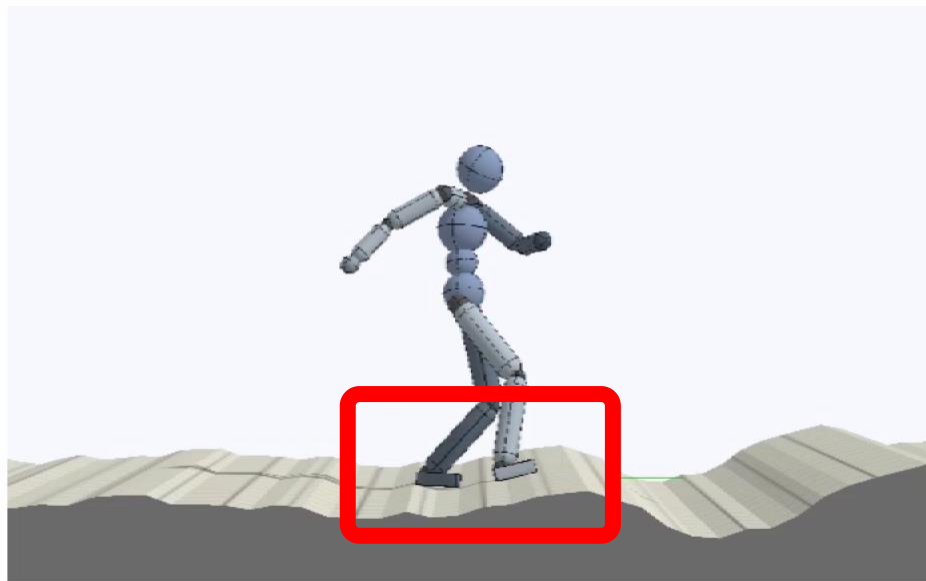
Which is Mocap?



Which is Mocap?

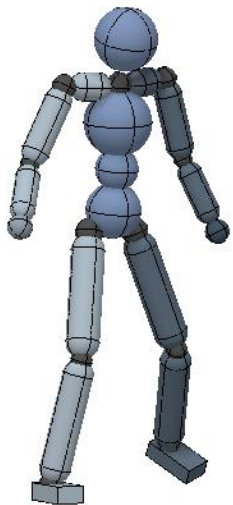


Mocap

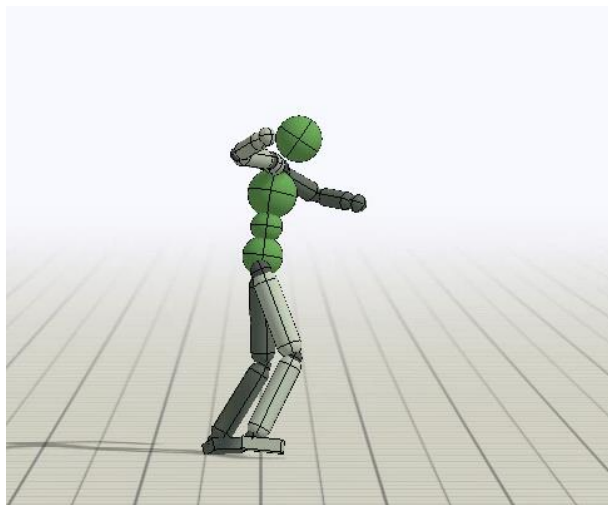


Simulation

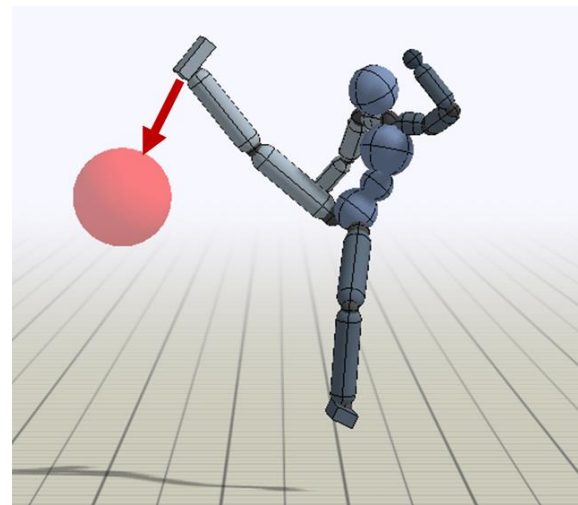
Overview



+



+

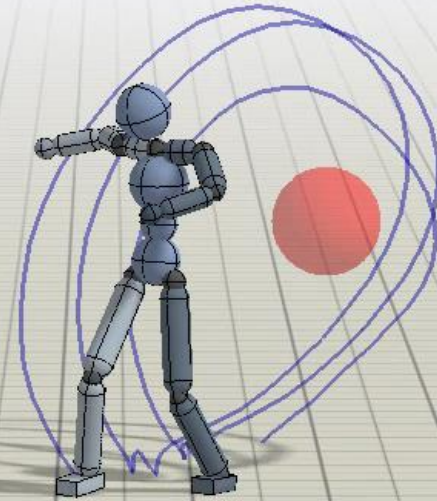


Character

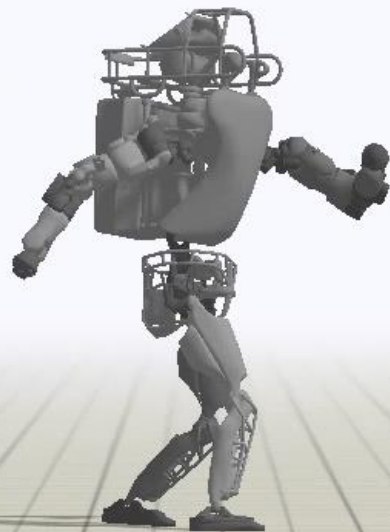
Reference Motion

Task: Hit Target

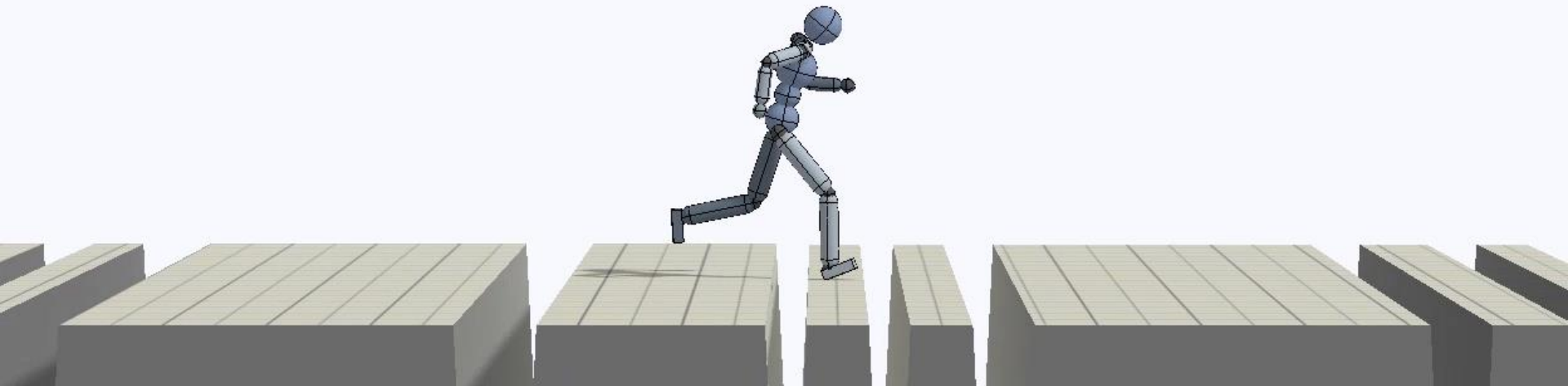
Overview



Overview

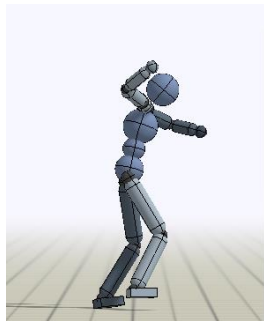


Overview

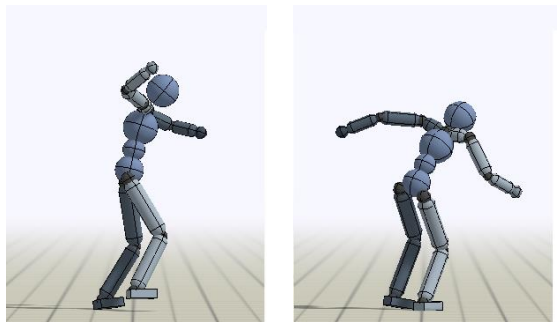


Reference Motion

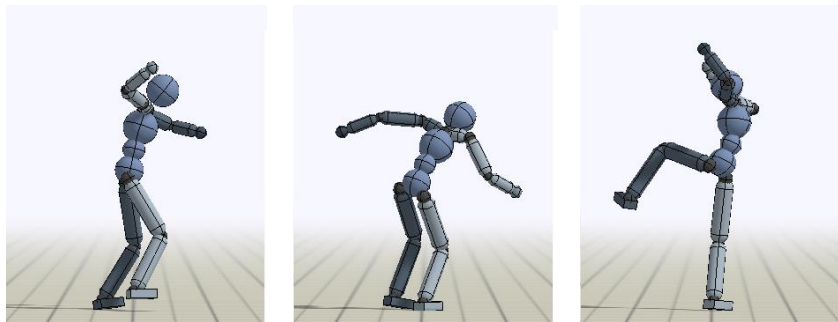
Reference Motion



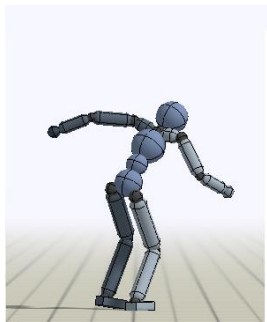
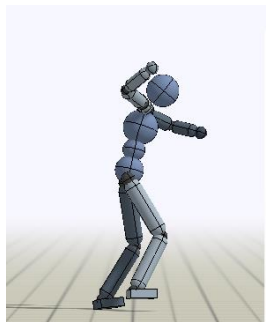
Reference Motion



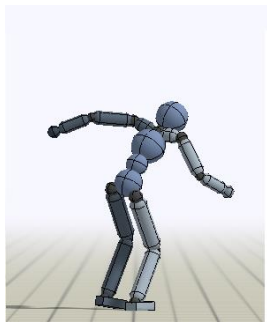
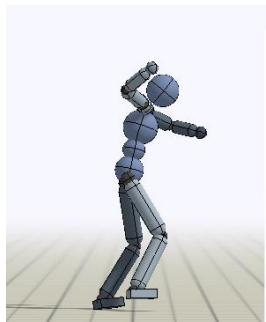
Reference Motion



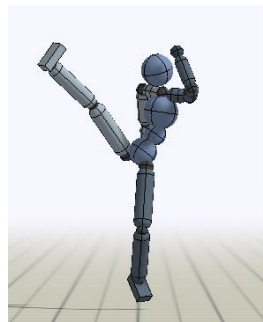
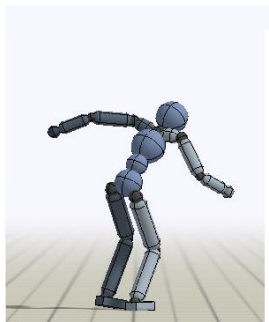
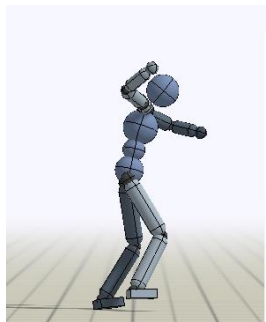
Reference Motion



Reference Motion



Reference Motion



a_0

a_1

a_2

a_3

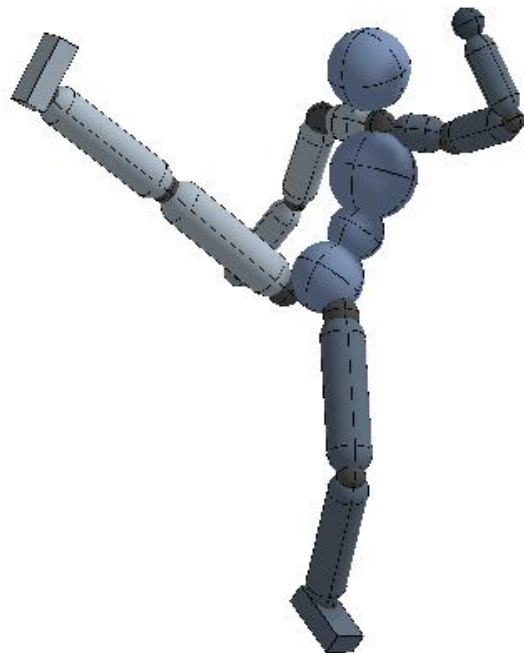
a_4



State + Action

State:

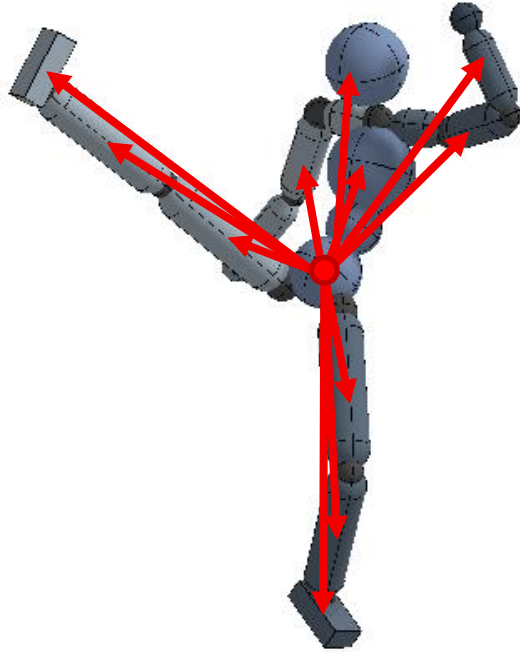
- link positions
- link velocities



State + Action

State:

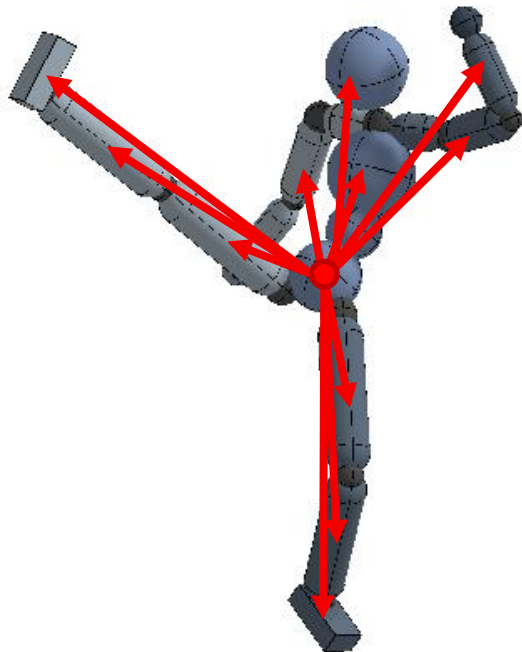
- link positions
- link velocities



State + Action

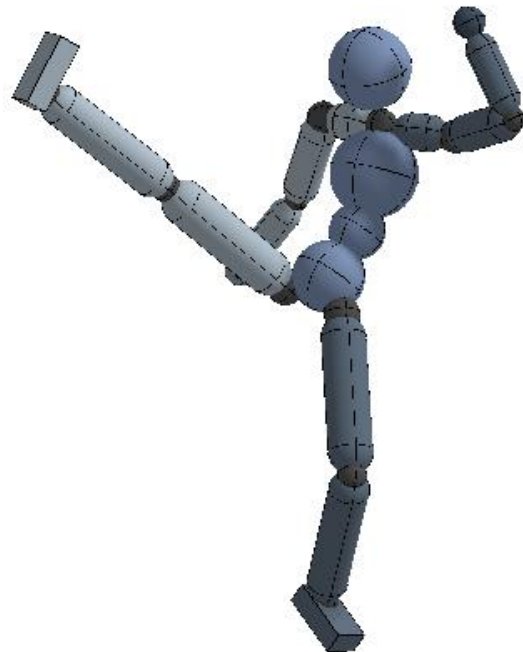
State:

- link positions
- link velocities



Action:

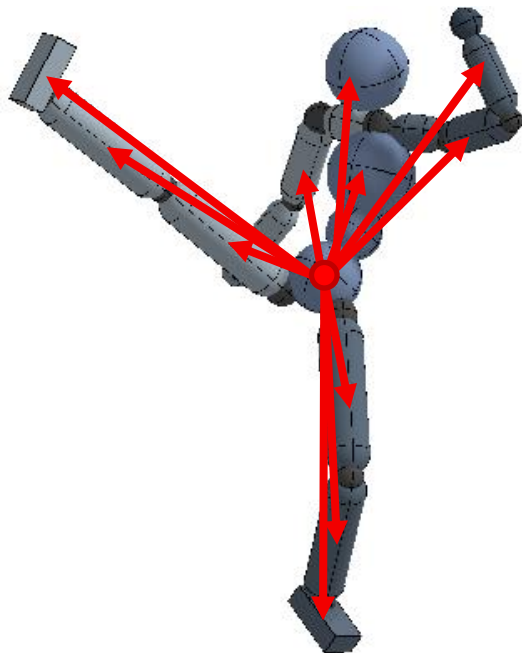
- PD targets



State + Action

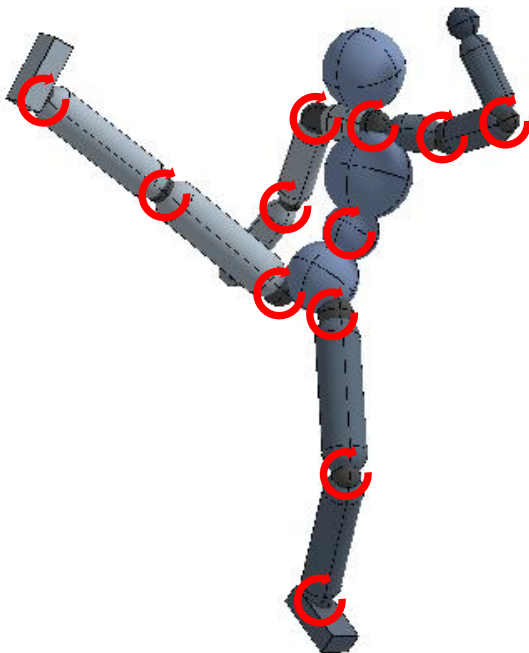
State:

- link positions
- link velocities



Action:

- PD targets



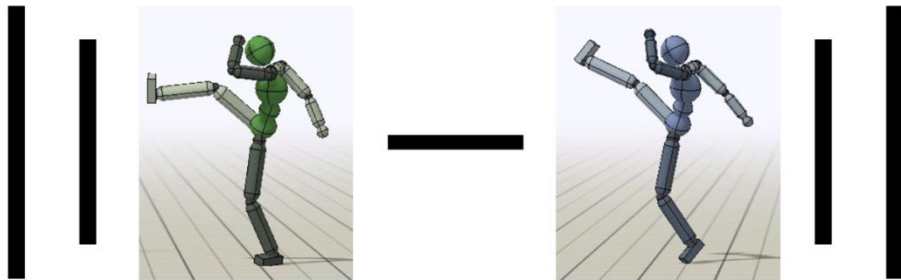
Reward

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

Reward

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

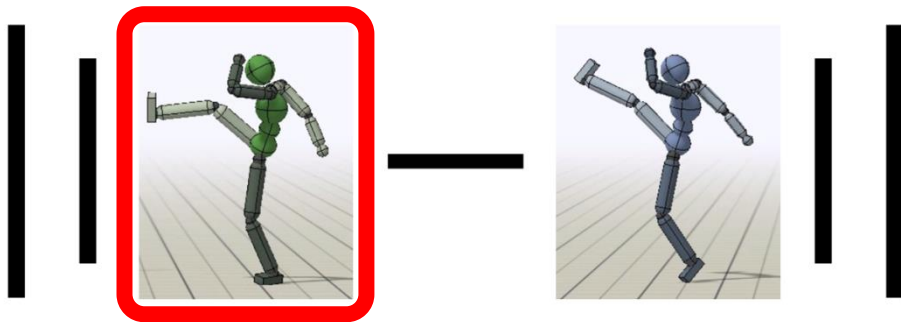
Imitation Objective



Reward

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

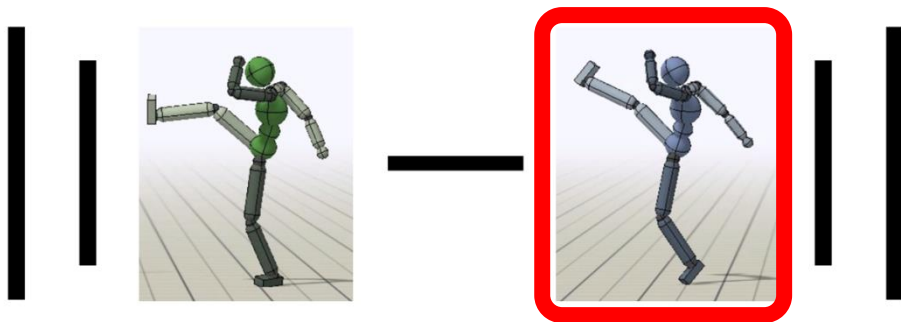
Imitation Objective



Reward

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

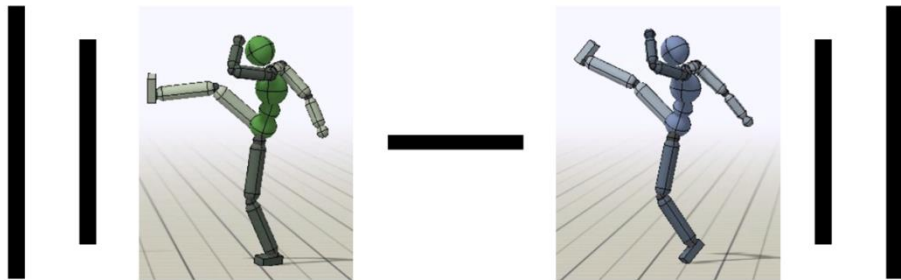
Imitation Objective



Reward

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

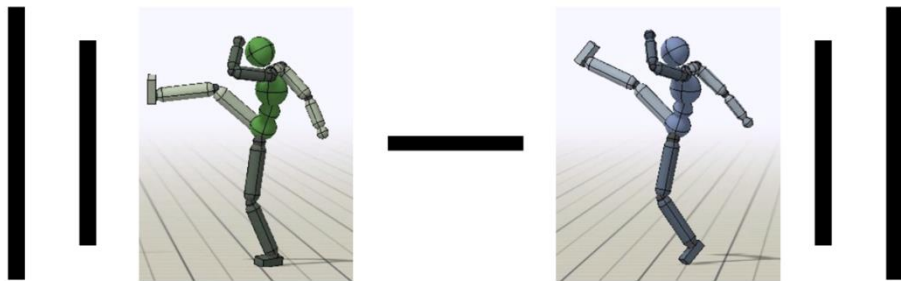
Imitation Objective



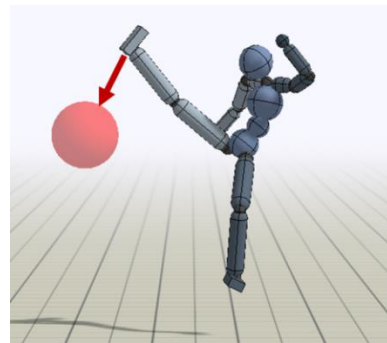
Reward

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

Imitation Objective



Task Objective



Proximal Policy Optimization (PPO)

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

[Schulman et al. 2017]

Proximal Policy Optimization (PPO)

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

[Schulman et al. 2017]

Proximal Policy Optimization (PPO)

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

[Schulman et al. 2017]

Proximal Policy Optimization (PPO)

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

[Schulman et al. 2017]

Proximal Policy Optimization (PPO)

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

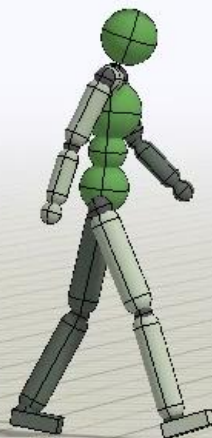
[Schulman et al. 2017]

Humanoid: Walk

Simulation

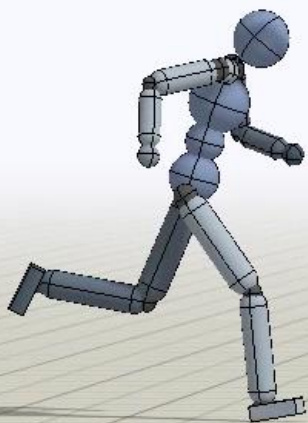


Reference

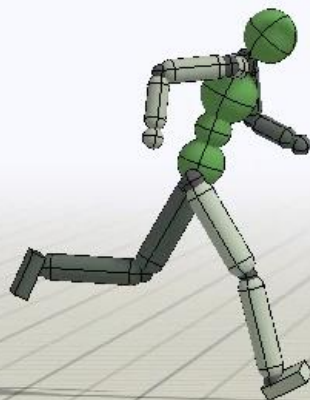


Humanoid: Run

Simulation



Reference



Comparison

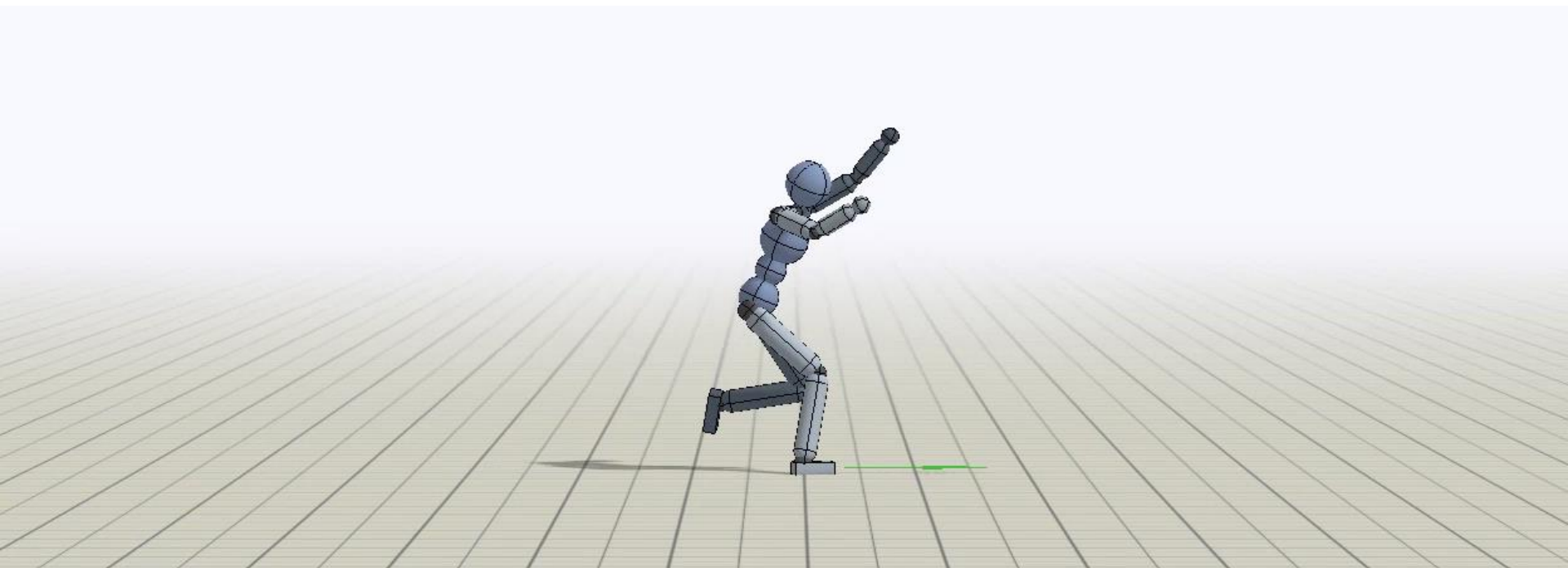


Ours



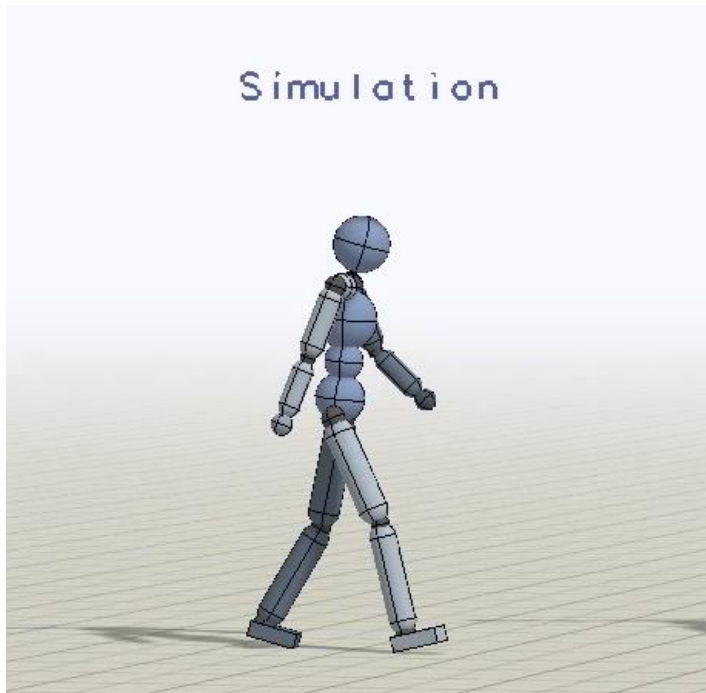
[Merel et al. 2017]

No Reference Motion



Locomotion

Simulation

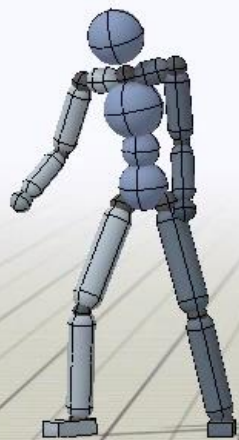


Simulation

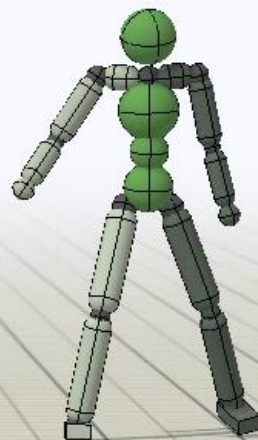


Humanoid: Cartwheel

Simulation



Reference



Humanoid: Backflip

Simulation



Reference



Humanoid: Frontflip

Simulation



Reference



Humanoid: Roll

Simulation

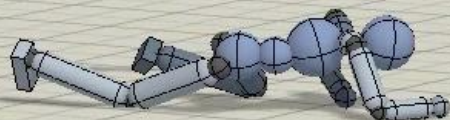


Reference



Humanoid: Crawl

Simulation



Reference



Humanoid: Dance A

Simulation



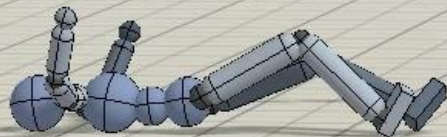
Reference



Humanoid: Kip-Up

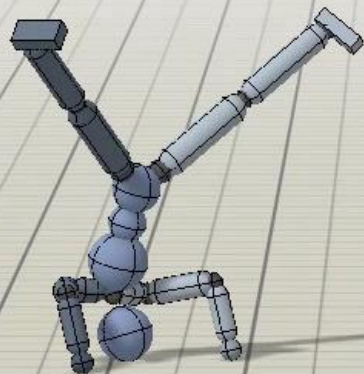
Simulation

Reference

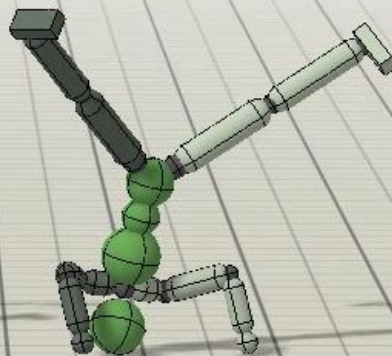


Humanoid: Headspin

Simulation



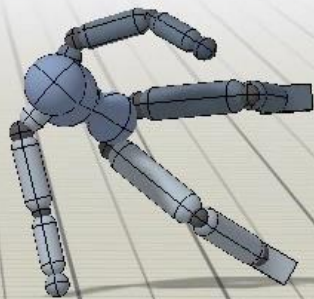
Reference



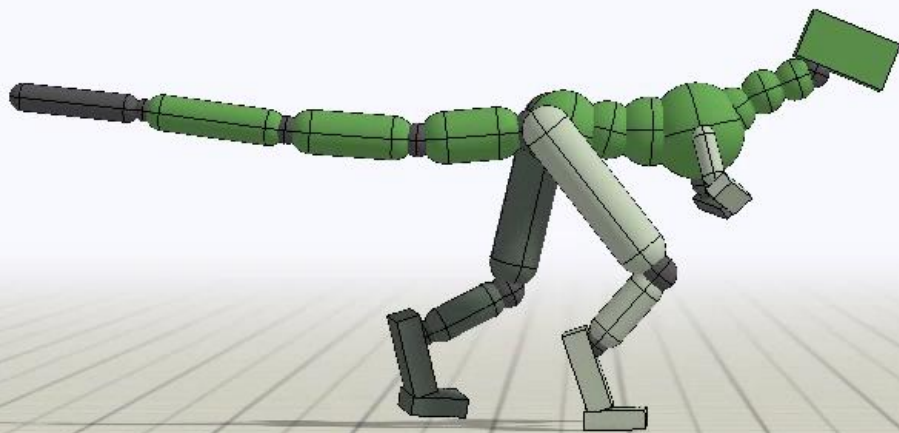
Humanoid: Vault 1-Handed



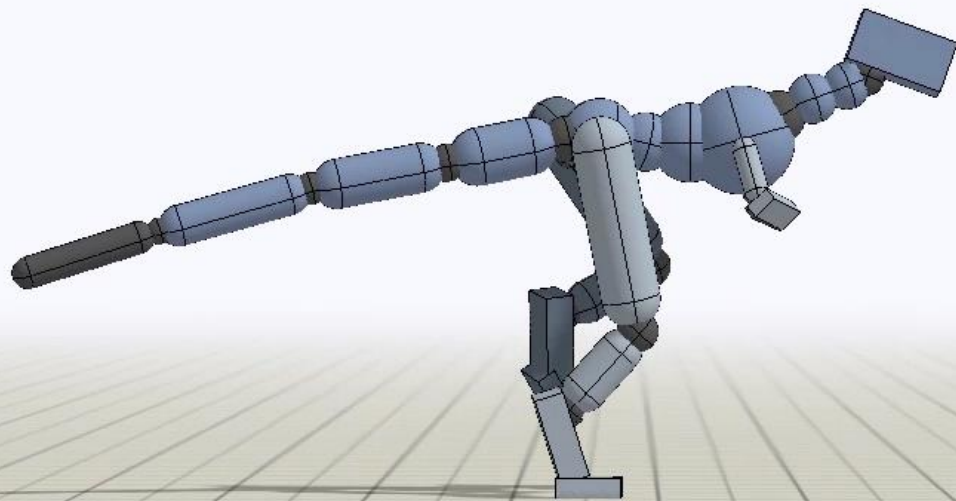
Humanoid: Flare



Keyframe Animation

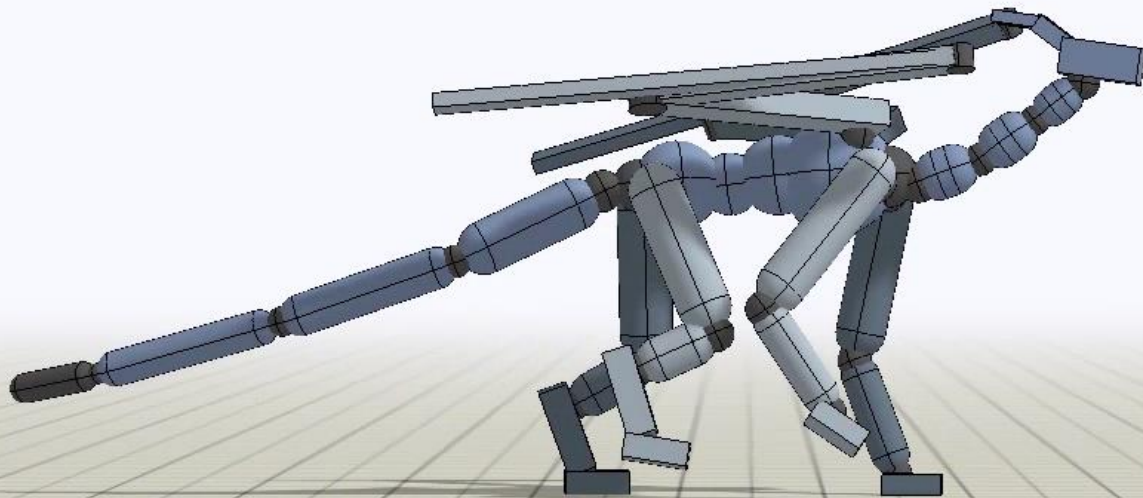


T-Rex: Walk



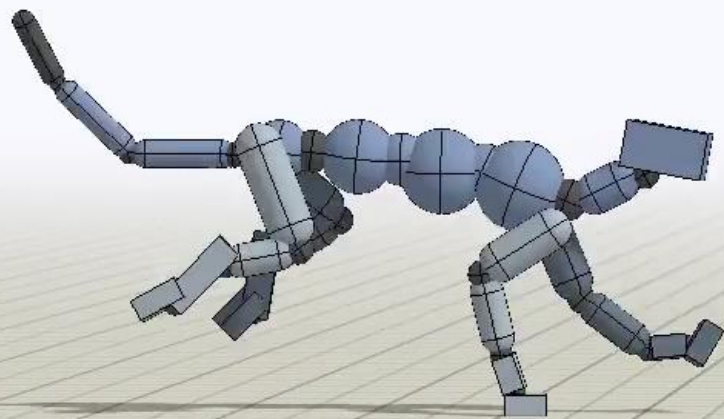
Simulated Character

Dragon: Walk



Simulated Character

Lion: Run



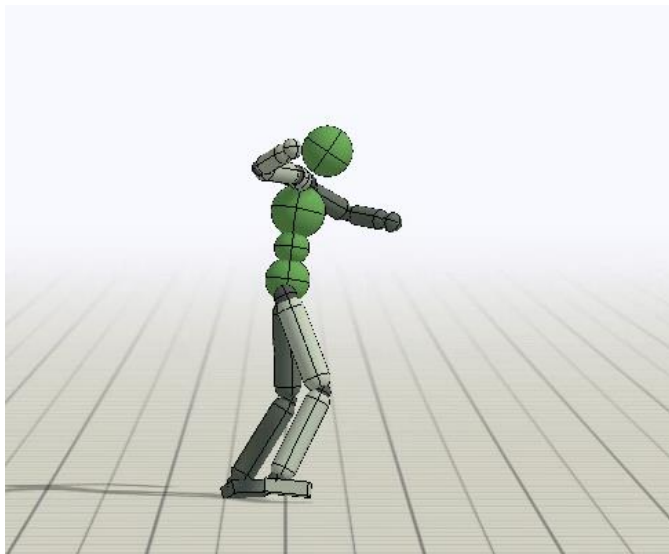
Simulated Character



ZIVA

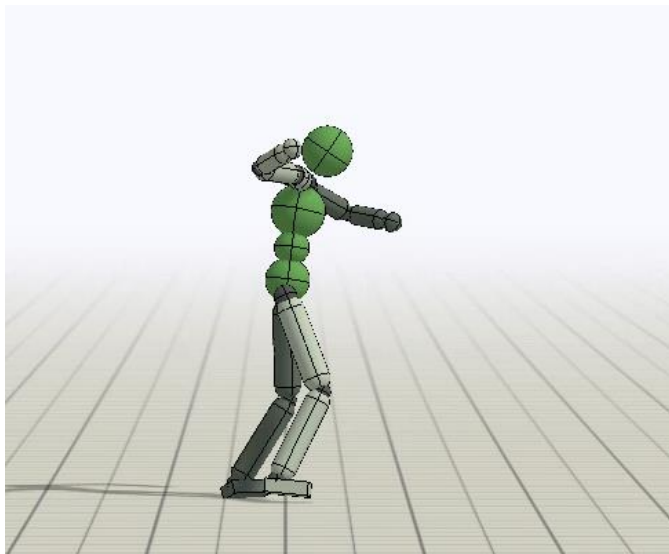
Tasks

Tasks



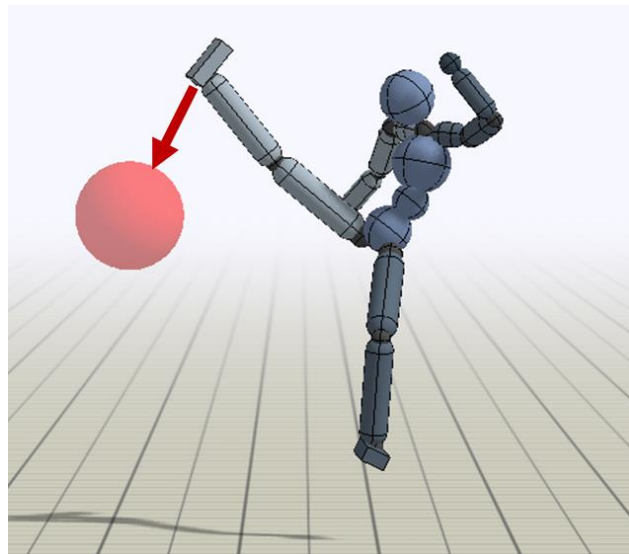
Reference Motion

Tasks



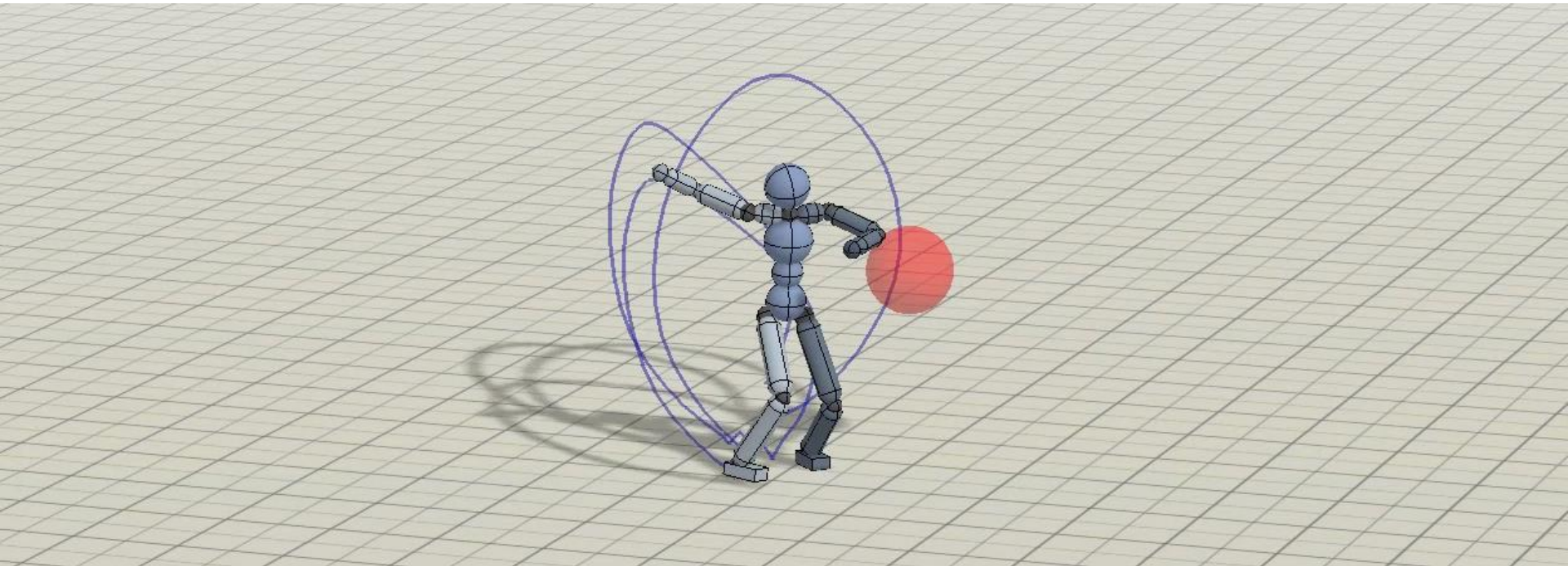
Reference Motion

+



Task

Humanoid: Spinkick - Strike



Humanoid: Baseball Pitch - Throw



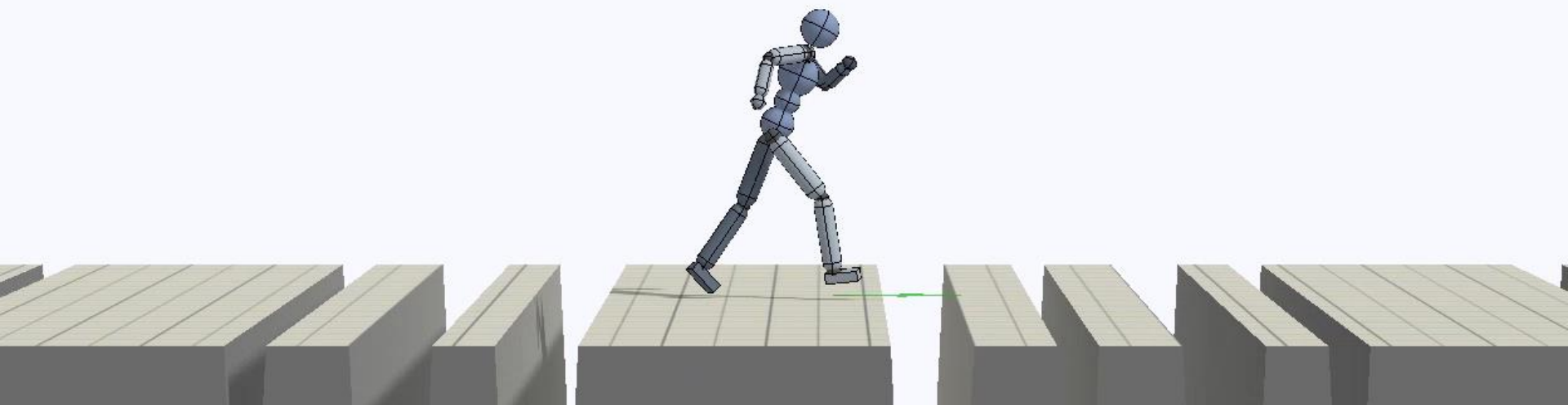
No Reference Motion



Humanoid: Balance Beam

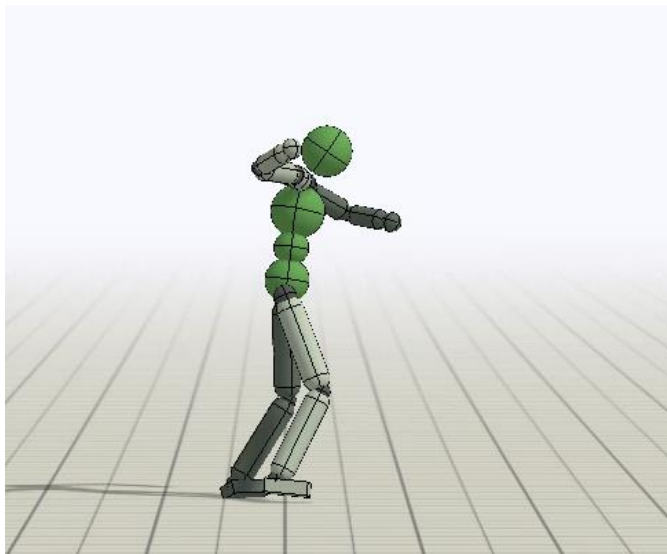


Humanoid: Run – Dense Gaps



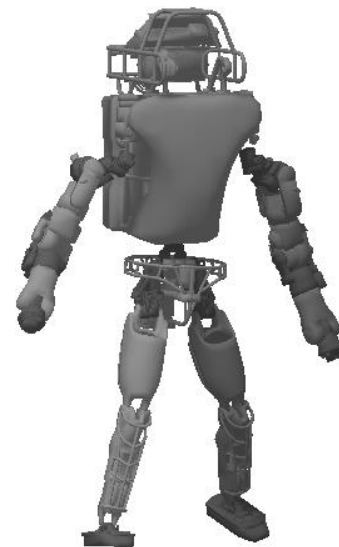
Retargeting

Character Retargeting



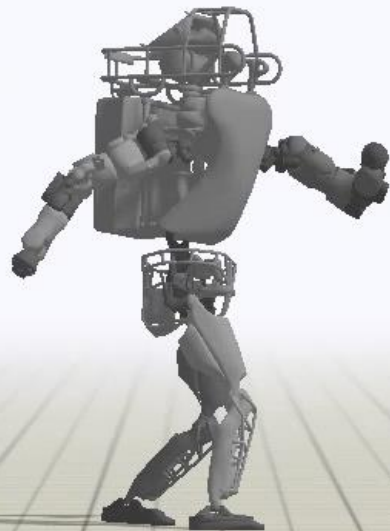
Reference Motion

+



Atlas

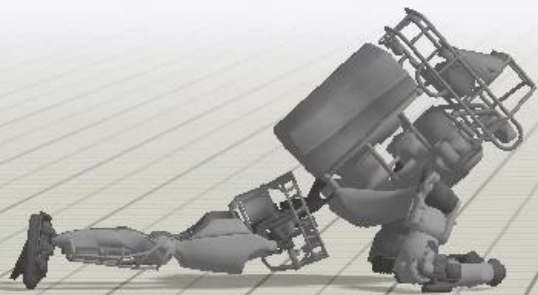
Atlas: Spinkick



Atlas: Run



Atlas: Getup-Facedown



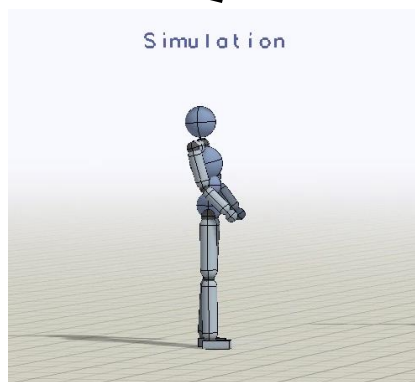
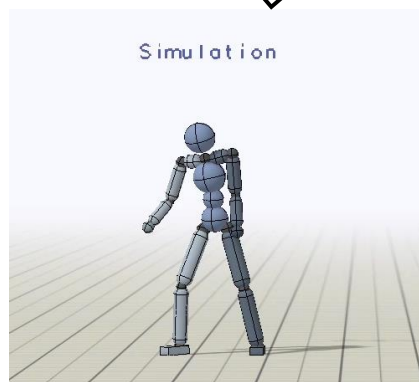
Atlas: Backflip



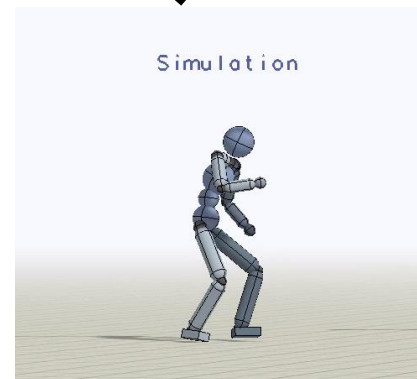
Multi-Clip Integration

Multi-Clip Integration

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

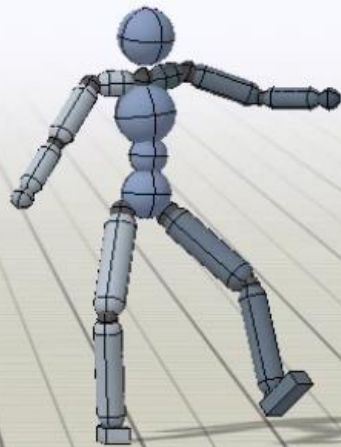


...



Multi-Clip Integration

Left Cartwheel



Mocap is a Hassle



[Holden 2018]

Skills From Videos

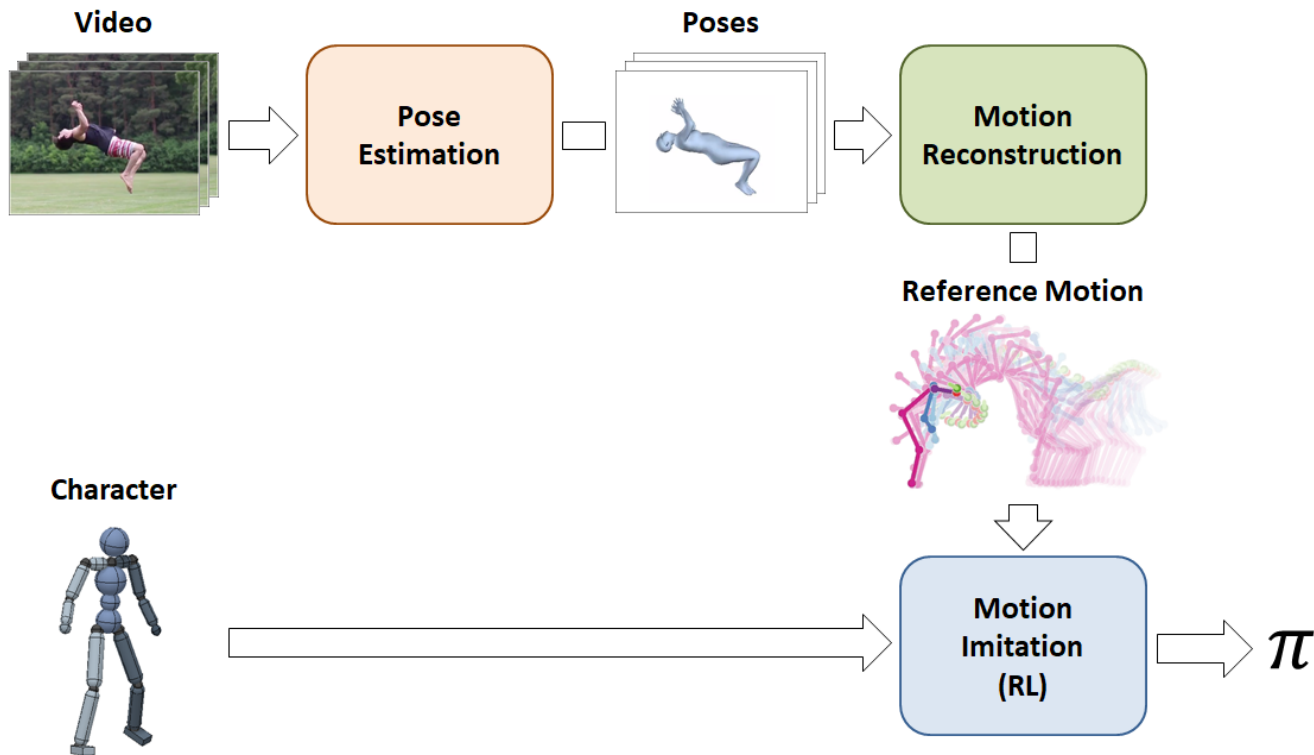


Learning from Videos

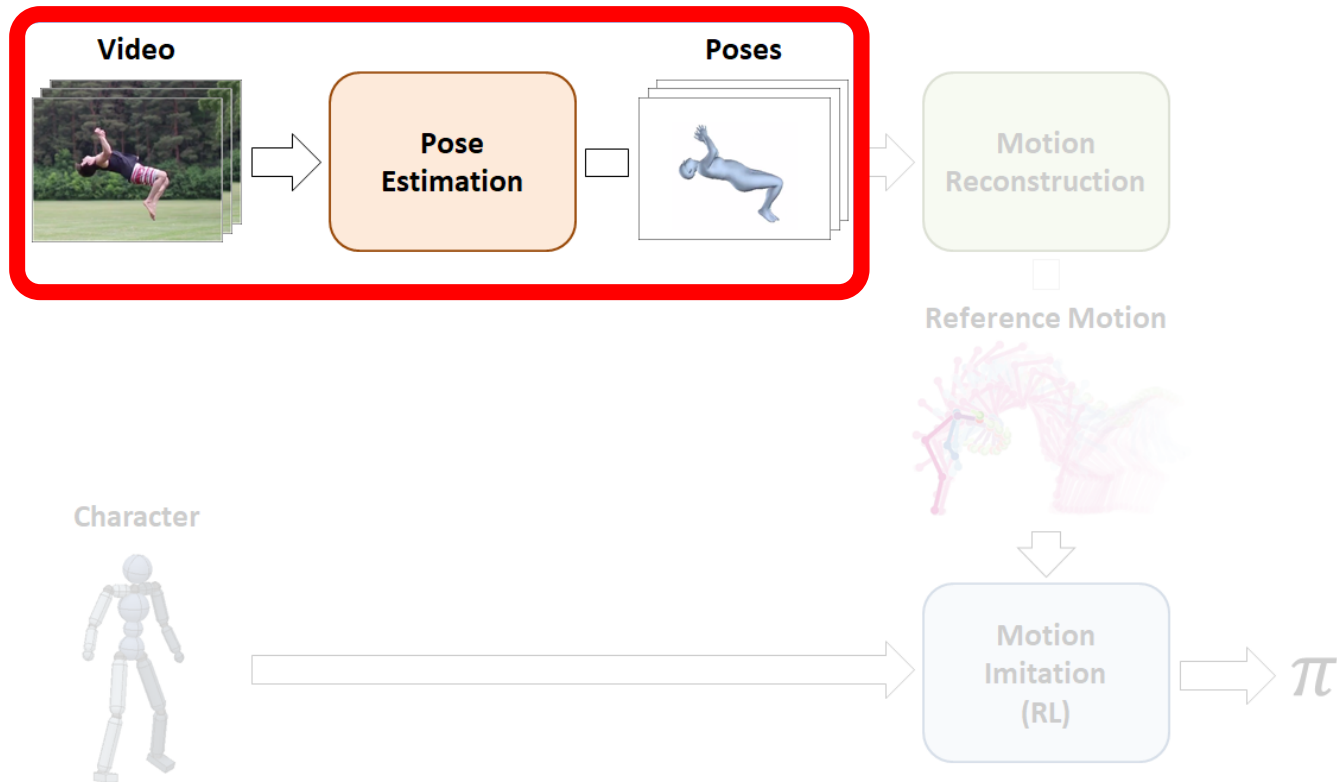


Video

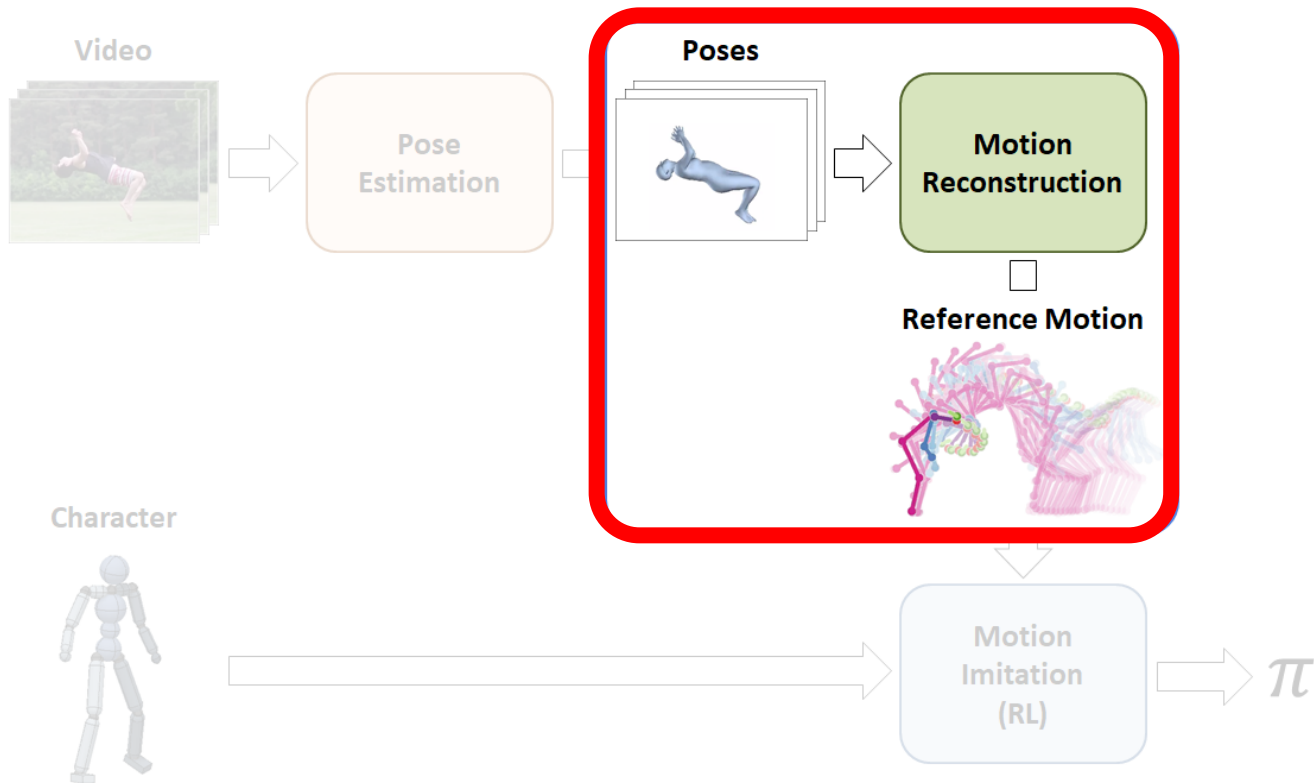
Overview



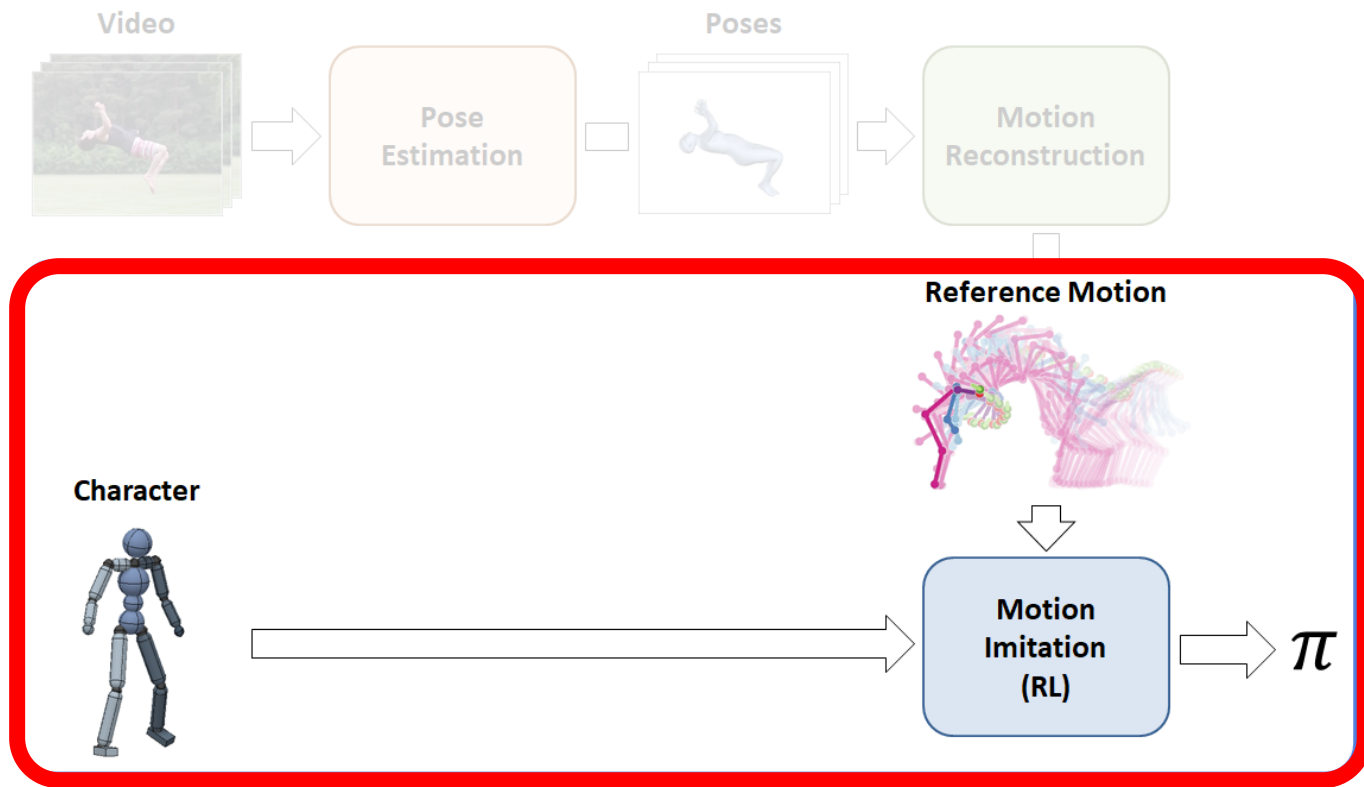
Overview



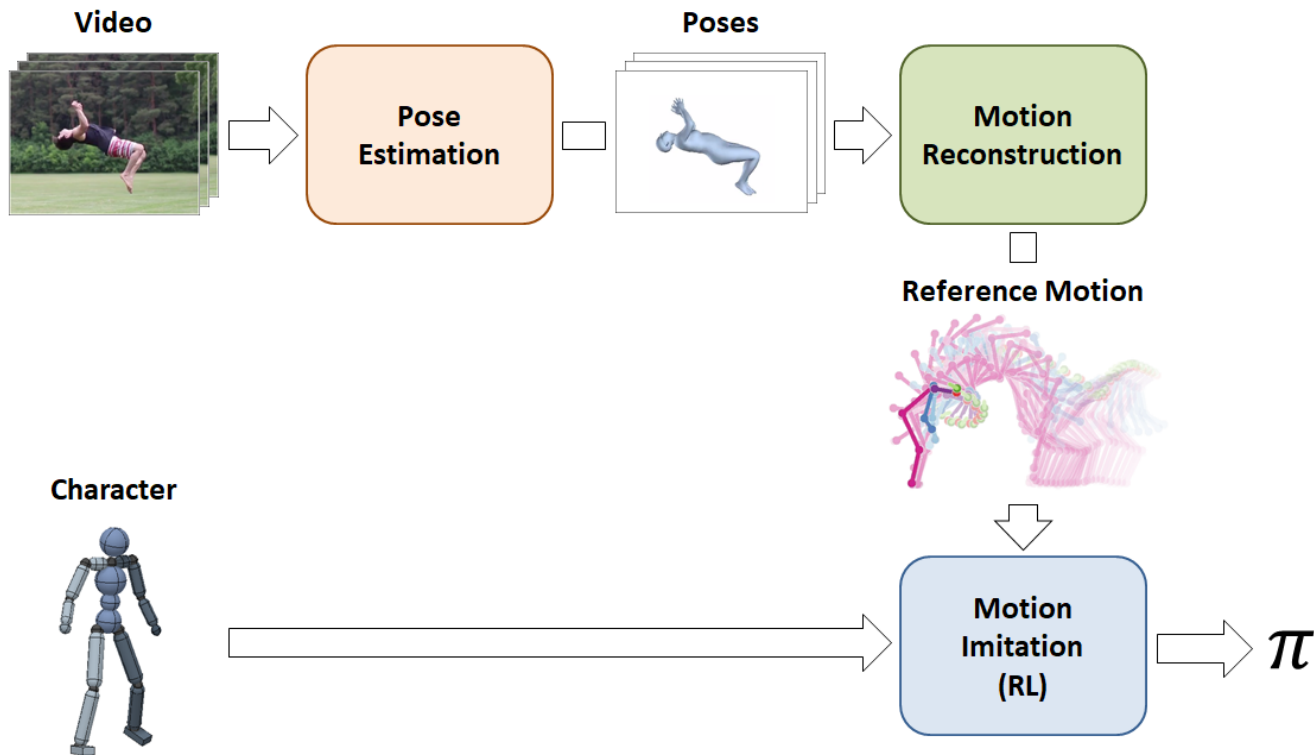
Overview



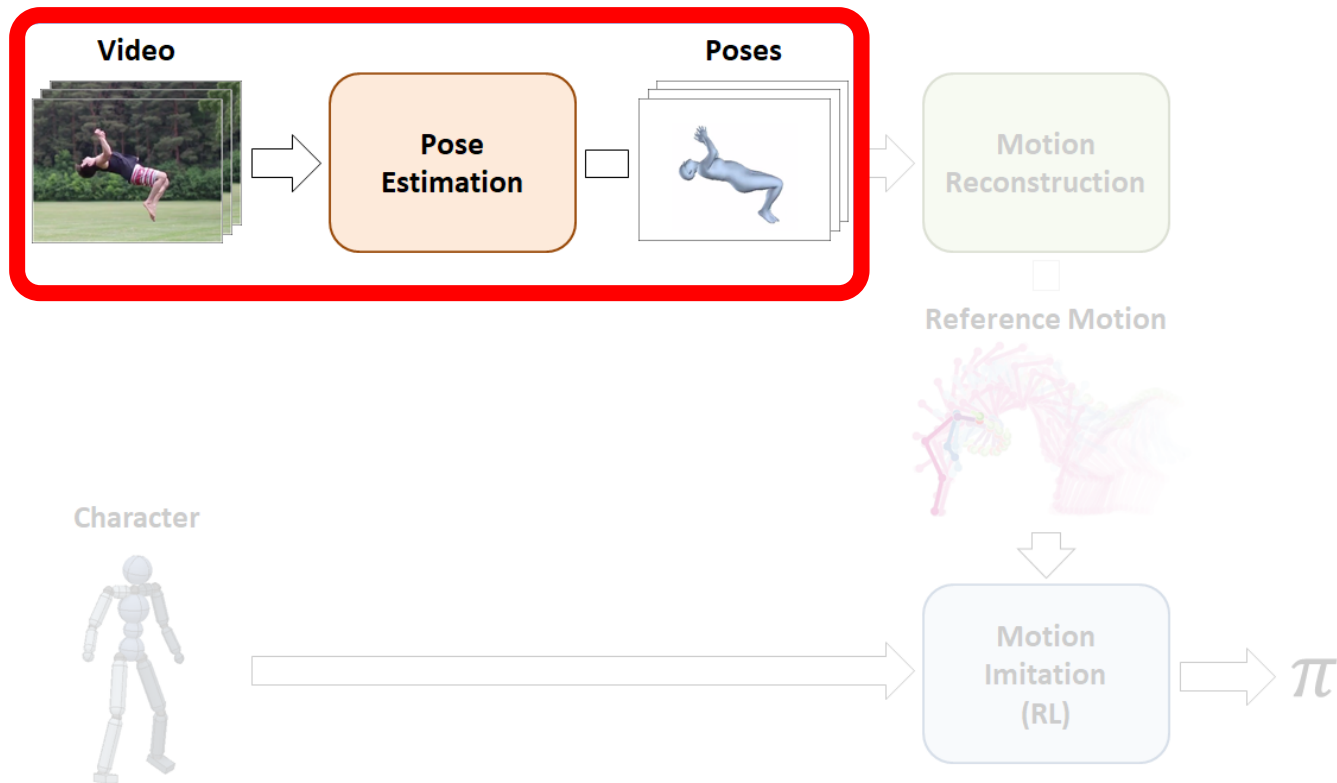
Overview



Overview



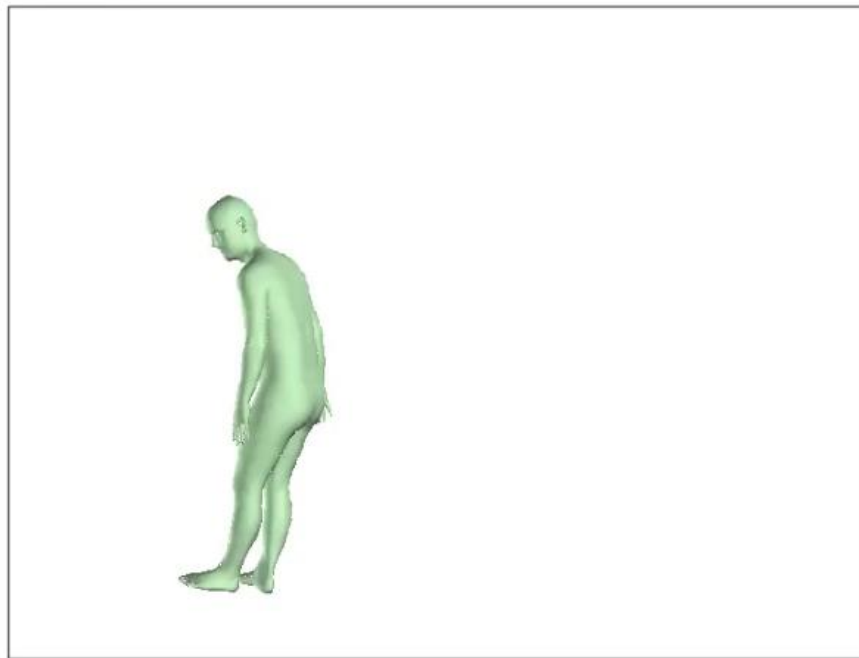
Pose Estimation



Pose Estimation



Video: Handspring A



Pose Prediction

Human Mesh Recovery (HMR)

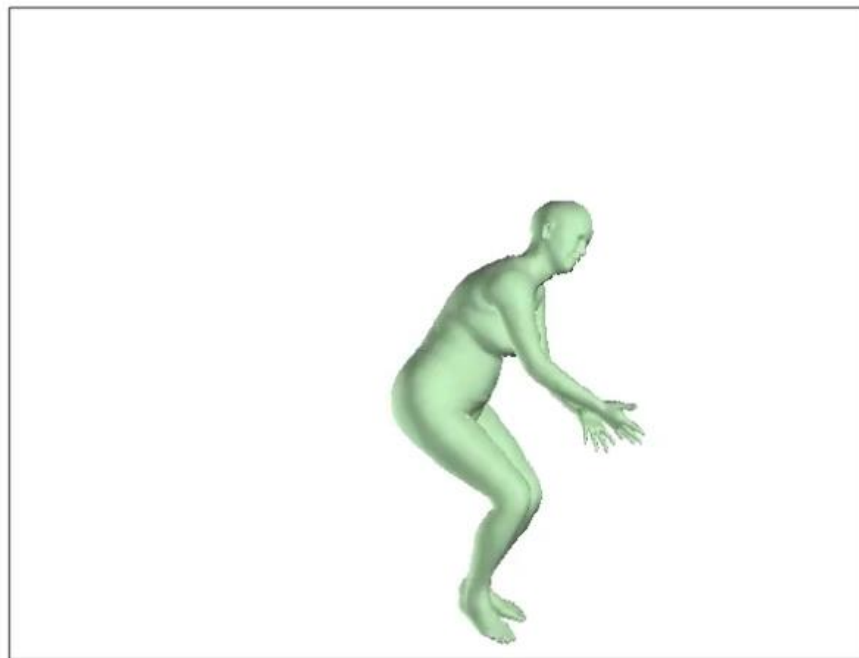


[Kanazawa et al., 2018]

Pose Estimation

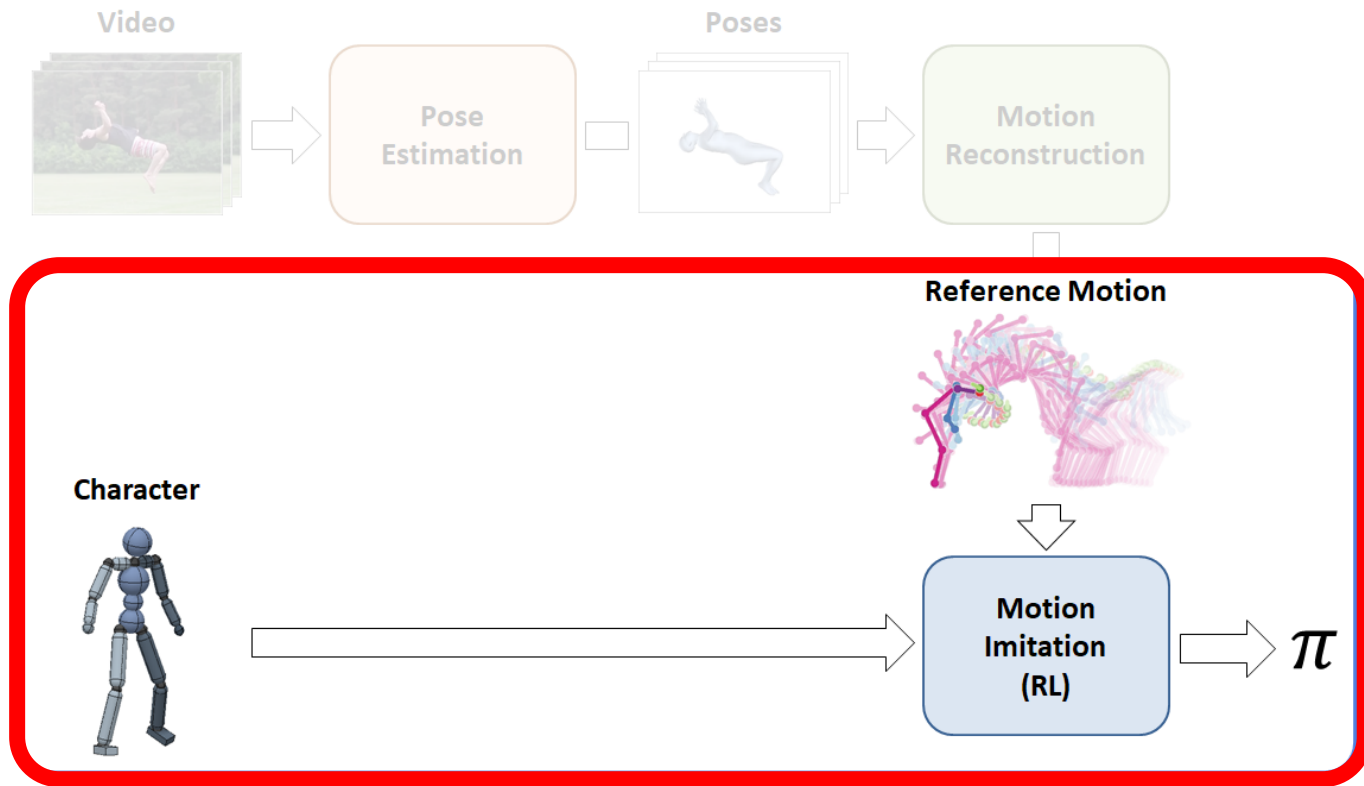


Video: Backflip A



Pose Prediction

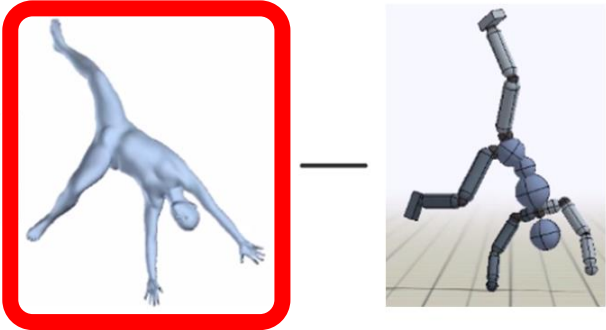
Motion Imitation



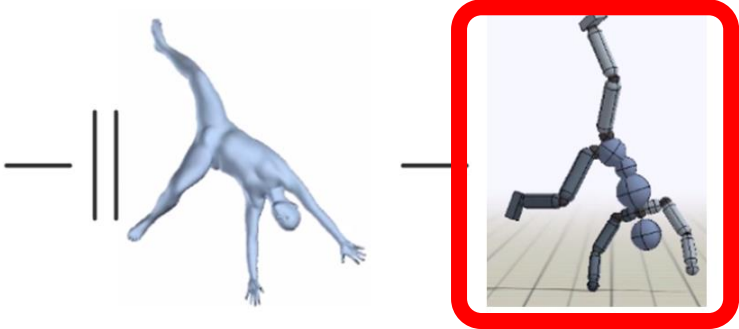
Motion Imitation

$$r_t = \exp \left(- \left\| \begin{array}{c} \text{Human} \\ \text{Robot} \end{array} \right\|^2 \right)$$

Motion Imitation

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


Motion Imitation

$$r_t = \exp \left(- \left\| \begin{array}{c} \text{Human} \\ \text{Robot} \end{array} \right\|^2 \right)$$


Humanoid: Cartwheel B



Video: Cartwheel B



Reference Motion

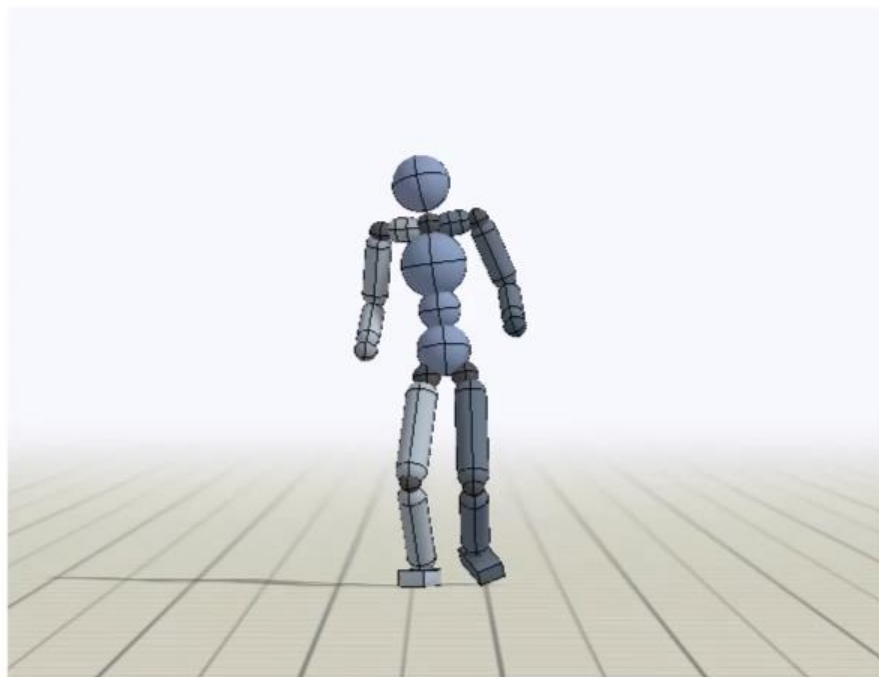


Policy

Humanoid: Jumping Jack



Video: Jumping Jack

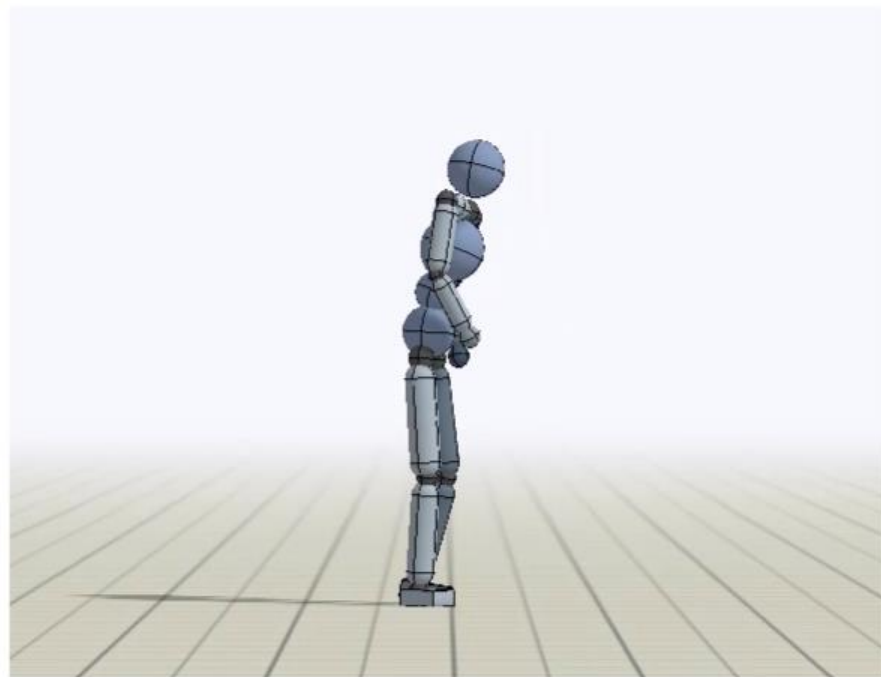


Policy

Humanoid: Backflip B



Video: Backflip B



Policy

Humanoid: Frontflip



Video: Frontflip

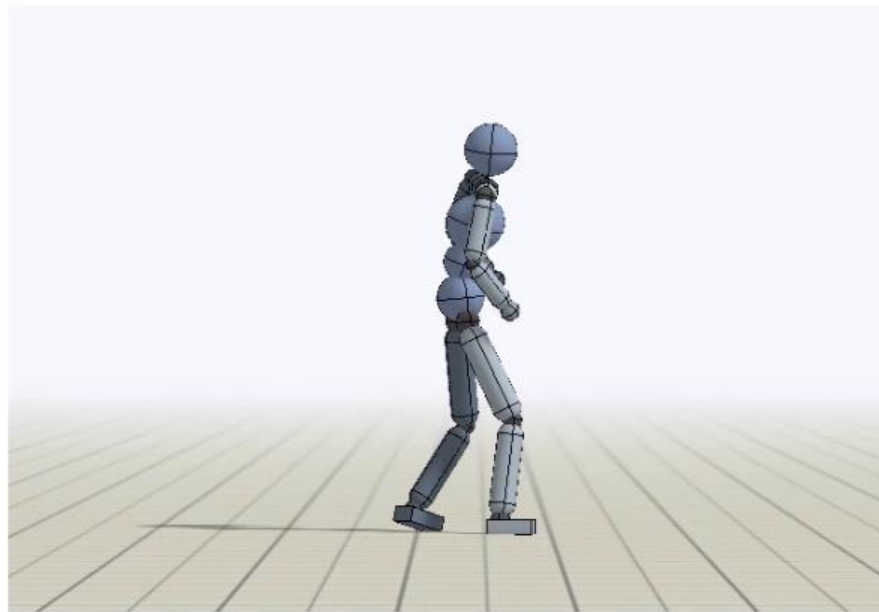


Policy

Humanoid: Roll



Video: Roll

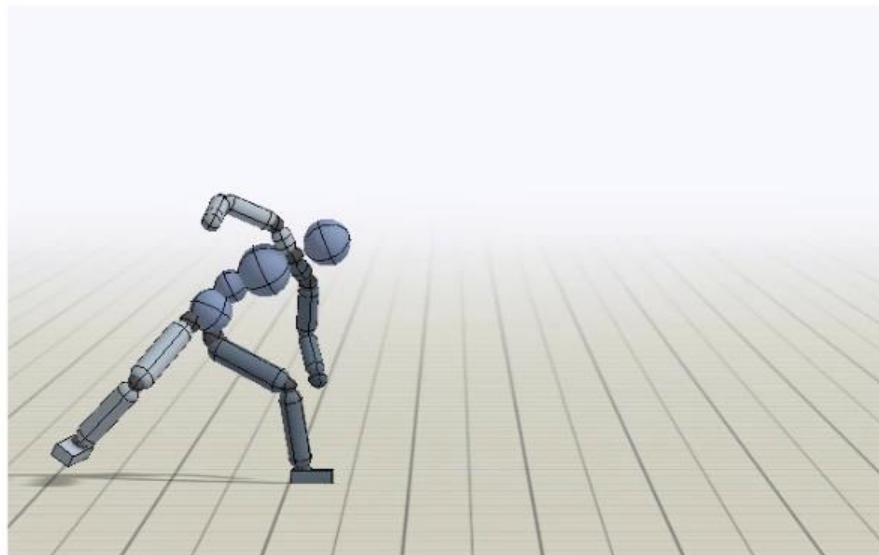


Policy

Humanoid: Spin



Video: Spin

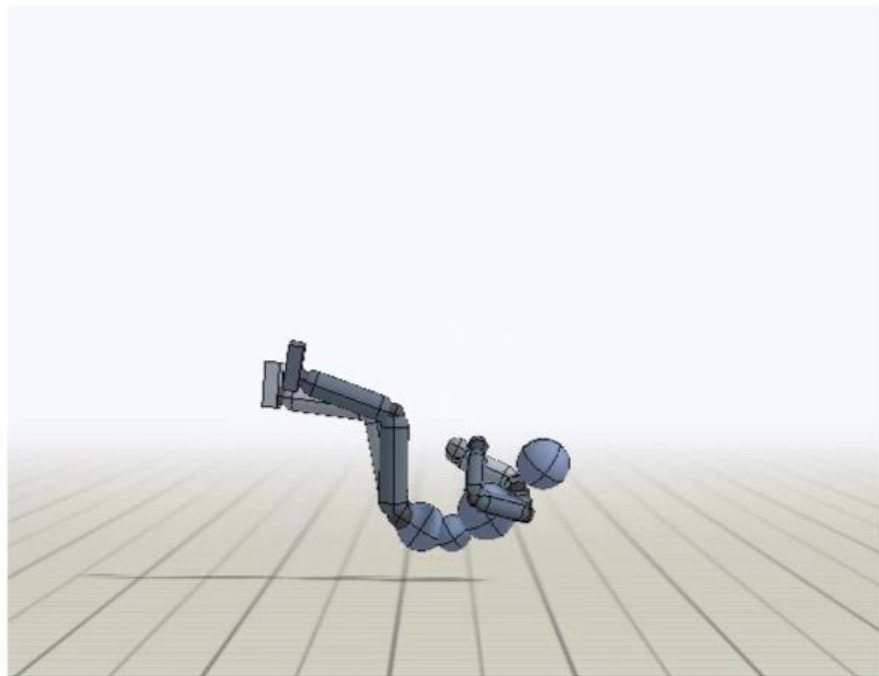


Policy

Humanoid: Kip-Up



Video: Kip-Up

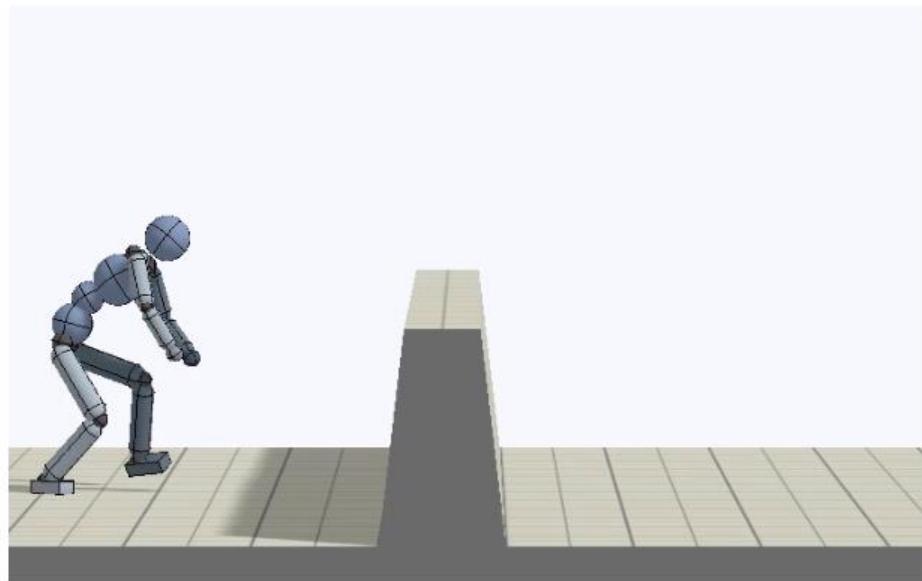


Policy

Humanoid: Vault



Video: Vault

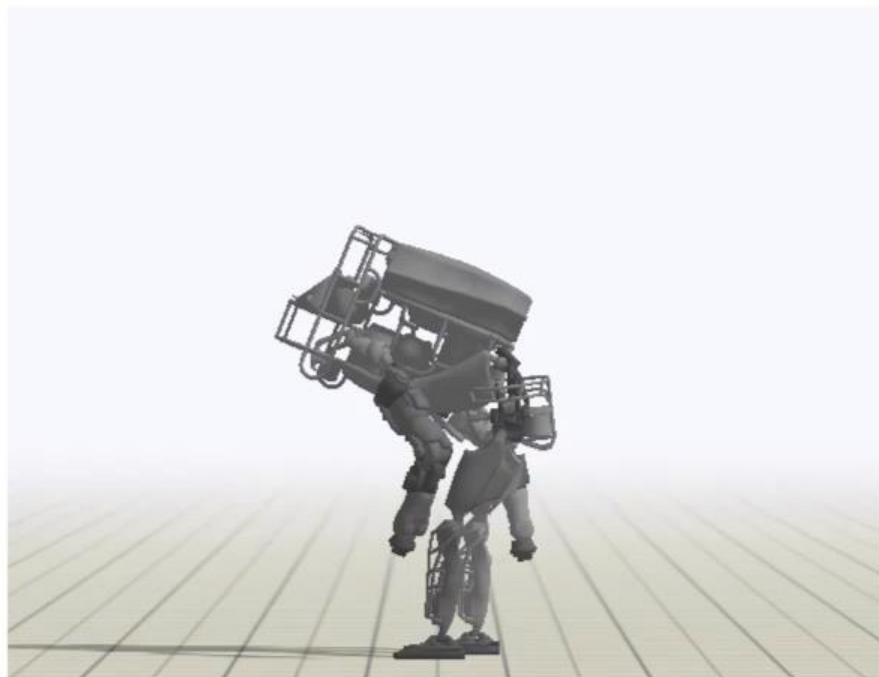


Policy

Atlas: Handspring A



Video: Handspring A

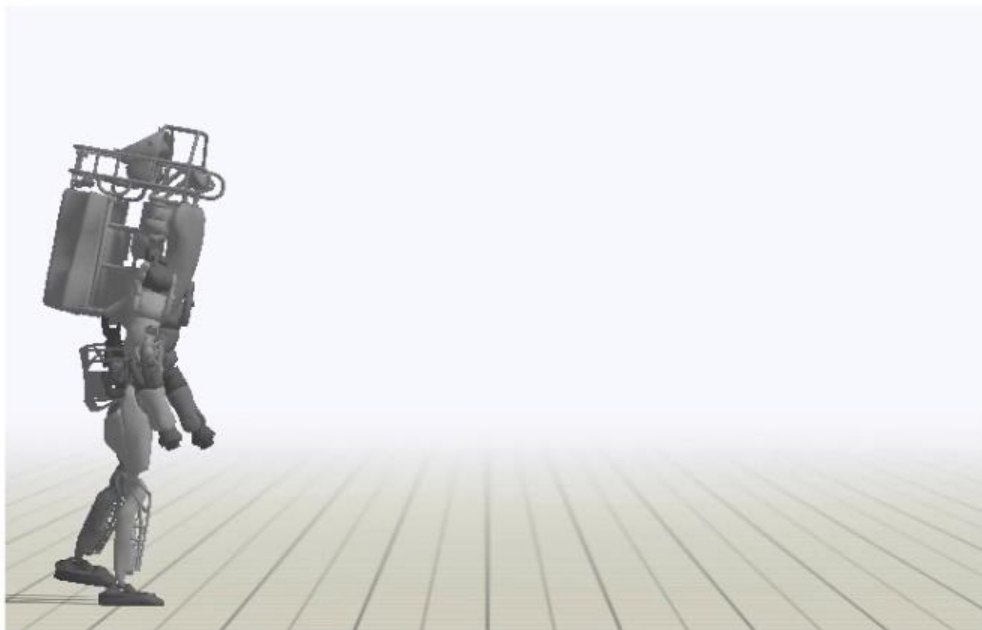


Policy

Atlas: Jump



Video: Jump

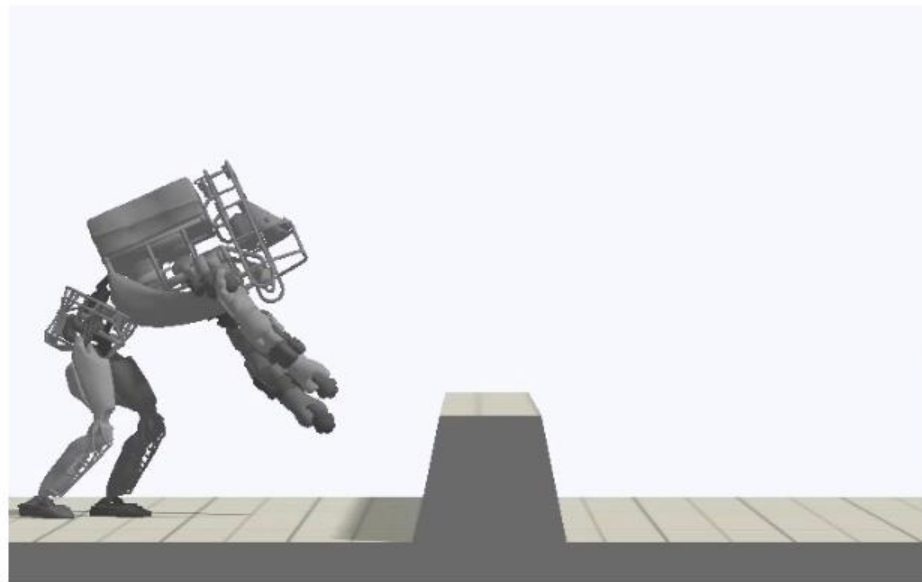


Policy

Atlas: Vault



Video: Vault

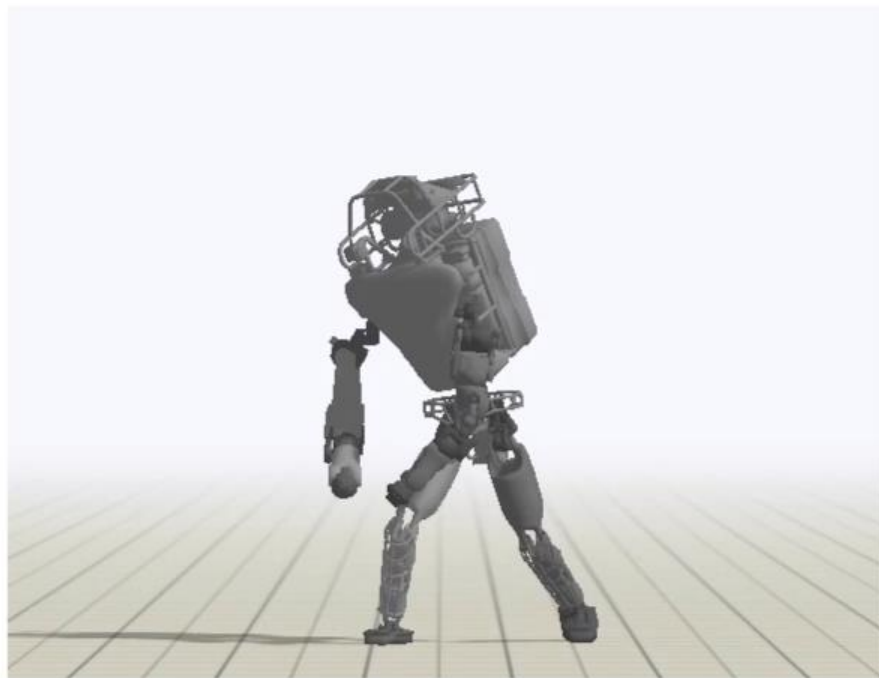


Policy

Atlas: Dance

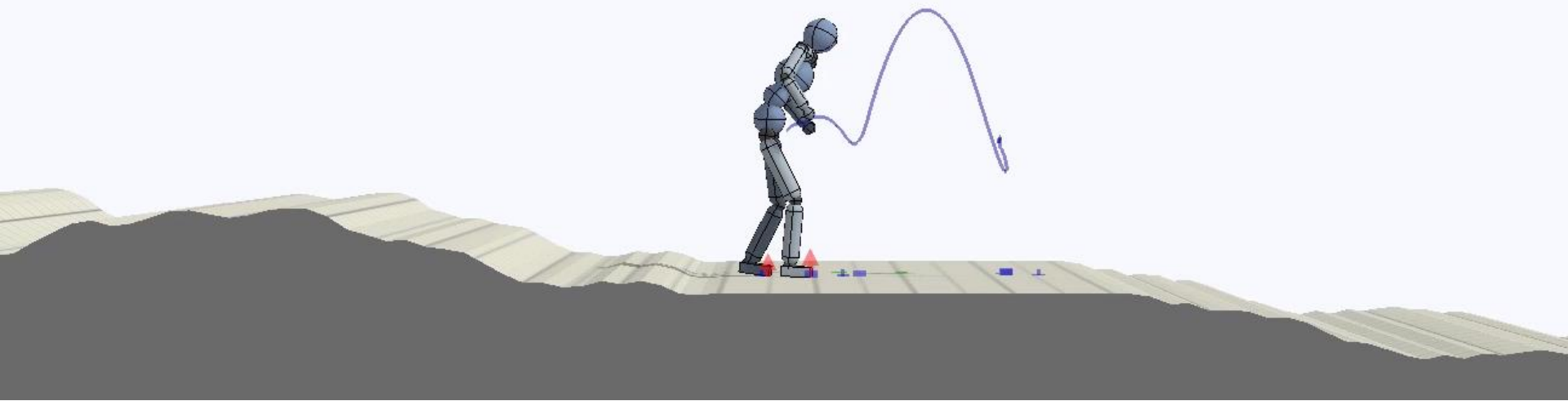


Video: Dance

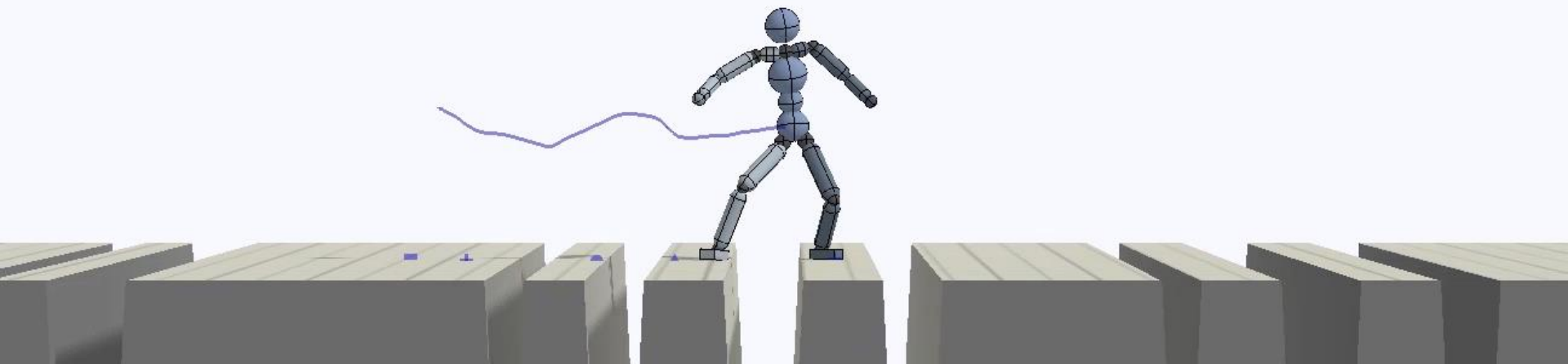


Policy

Environment Retargeting



Environment Retargeting



Failure Cases



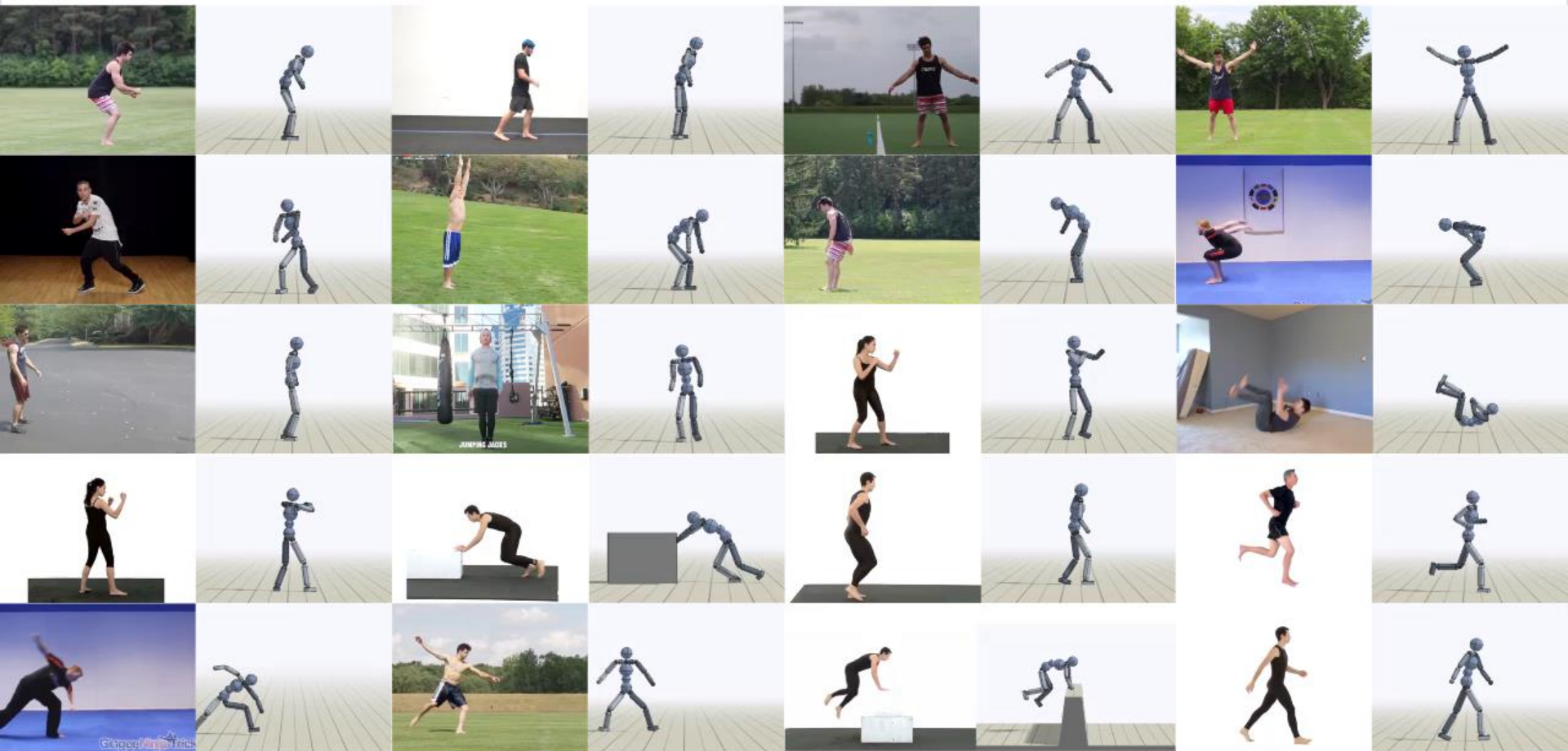
Video: Gangnam Style



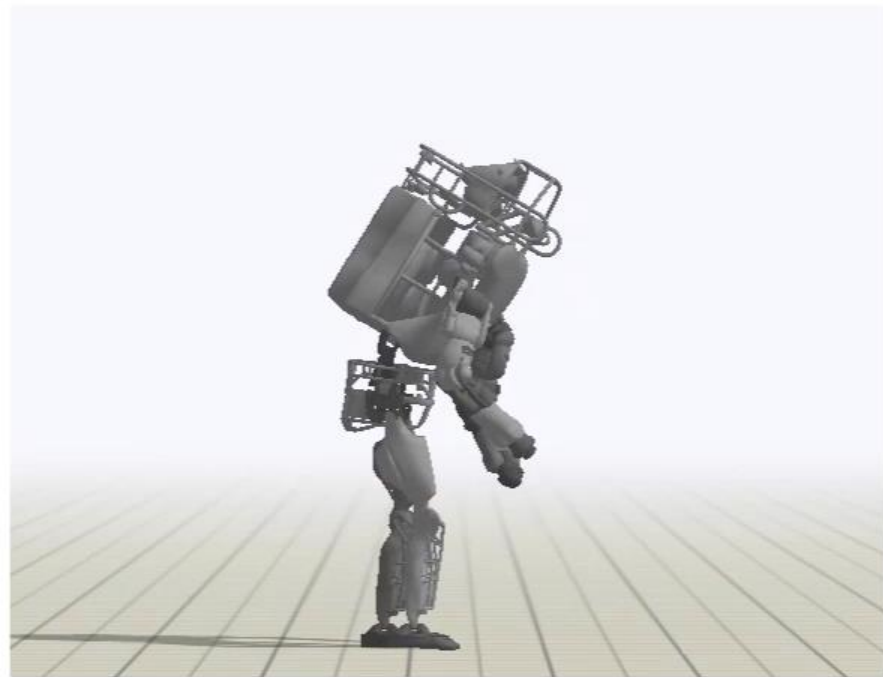
Reference Motion



Simulation



Skills from Videos



Policy

Concluding Remarks

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

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 - 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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Jitendra Malik



Michiel van de Panne

Questions?

