

ShapeVerse: Physics-based Characters with Varied Body Shapes

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PROBLEM

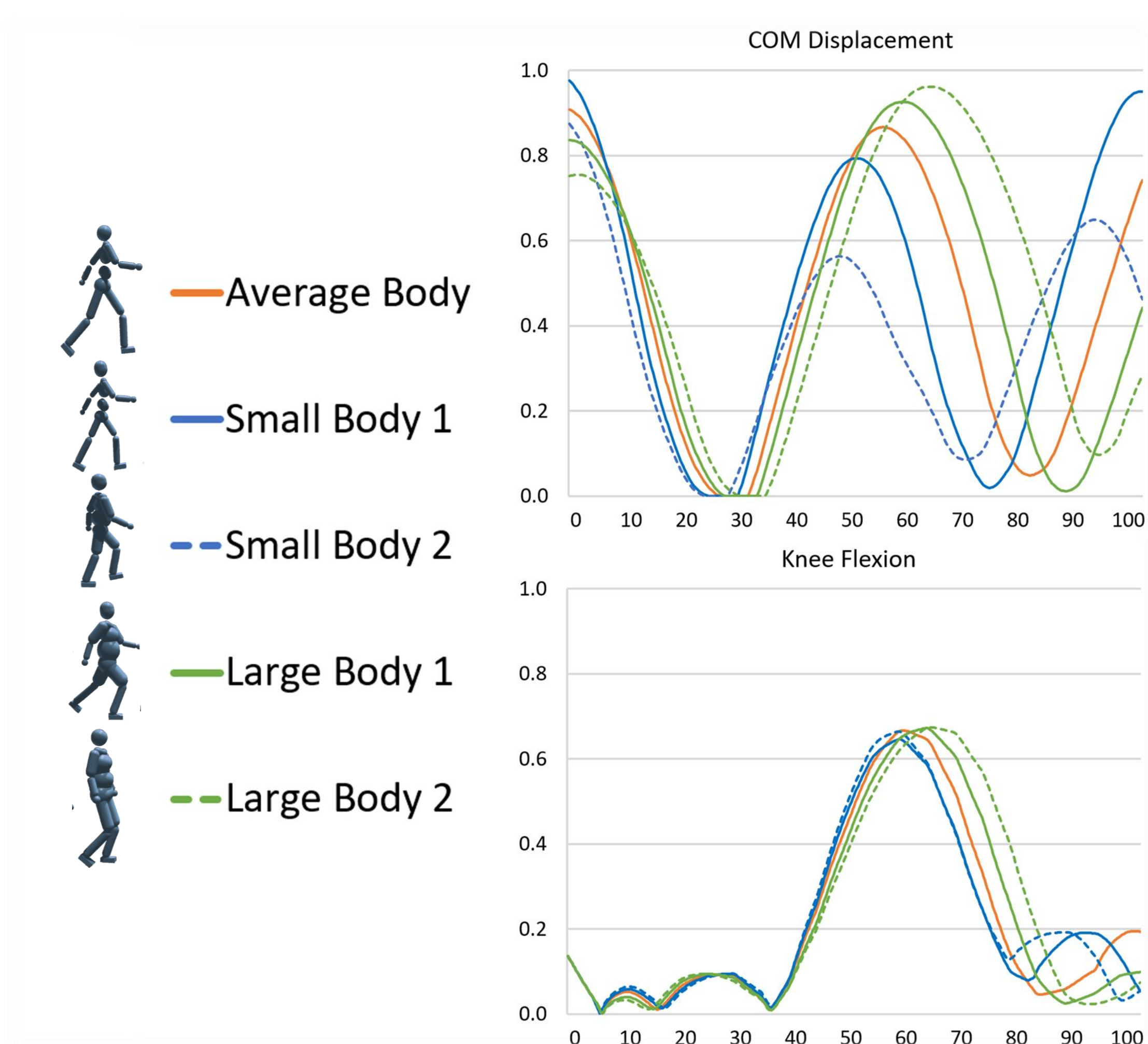
- Physics-based characters are not widely used although they create more **responsive and realistic motions**.
- Existing methods often lack diversity and do not account for the **effects of individual body shape parameters and mass** on character motion.
- Redundant motion capture** of multiple actors of different body shape.

SUMMARY

- This work proposed a **deep reinforcement learning (DRL) based framework** to generate physics-based characters with varying body shapes.
- During the **training process**, the characters were trained to **imitate the reference motion** based on their specific body shape parameters.
- The proposed approach enabled the generation of a diverse range of animations.

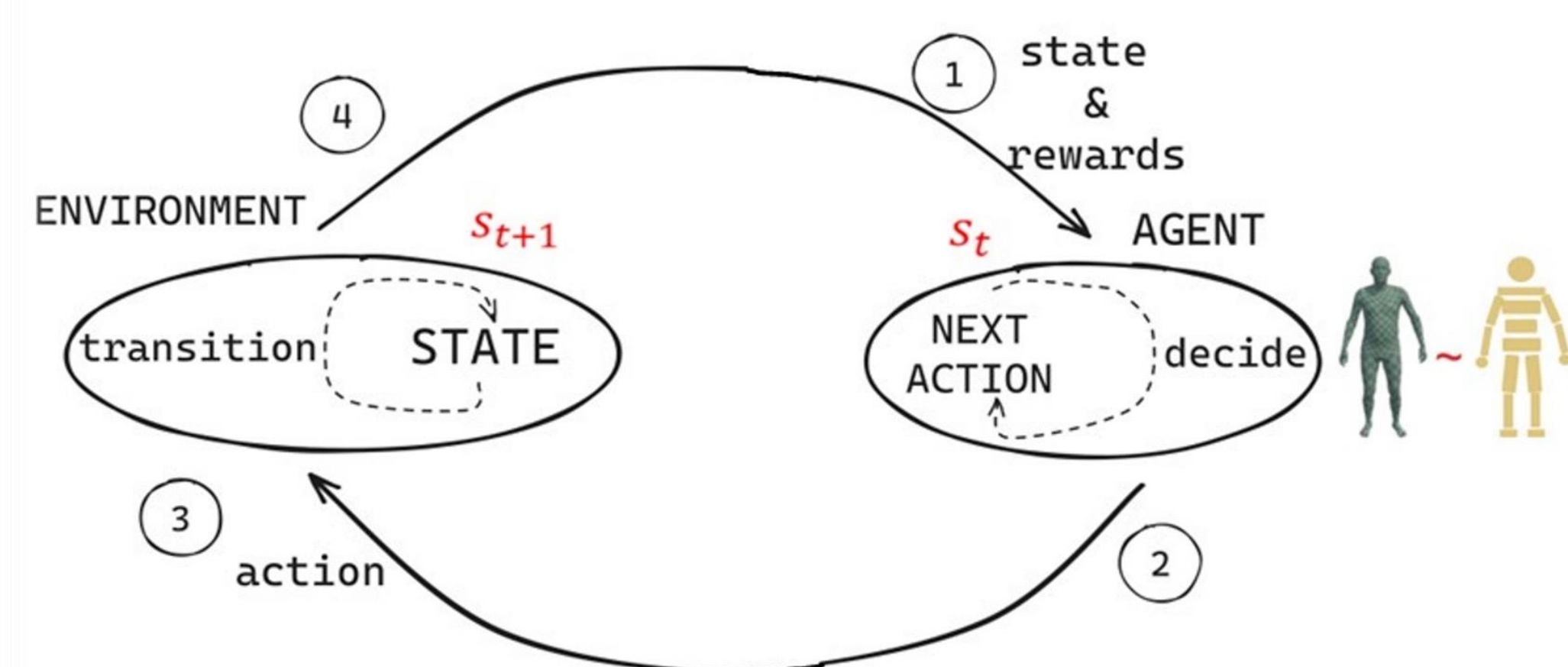
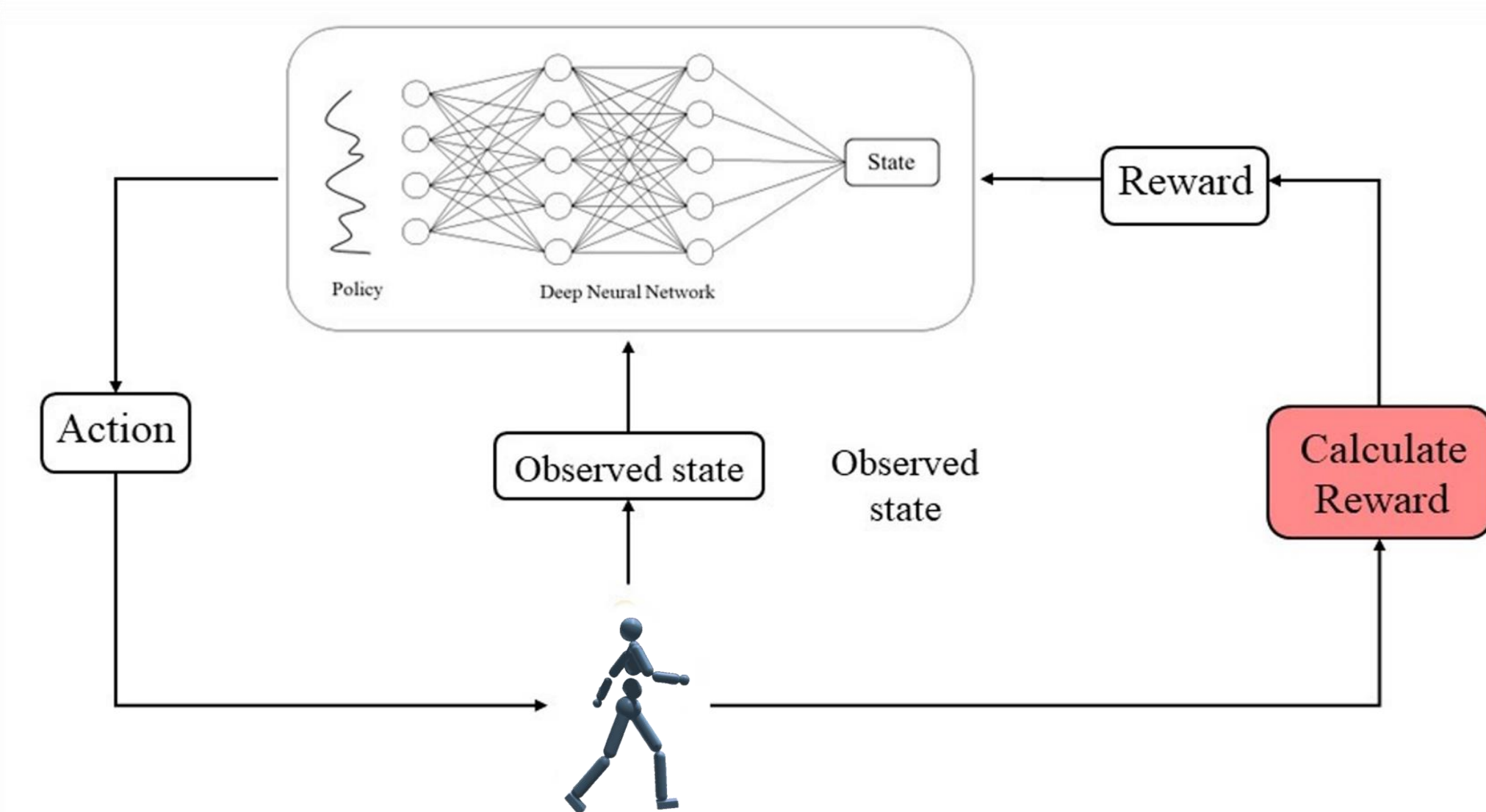
RESULTS

- Evaluation:**
 - Extracted beta parameters (β) using MoSH from a motion capture data. [5]
 - Compared the biological parameters from generated character motions with real human motion data. [6] [7]
 - The trained policy output a **distinct trajectory of the pelvis** for each character.
 - Knee flexion** values suggested that additional mass in characters can **generate lower leg motion diversity**



METHODOLOGY

- Proximal Policy Optimization (PPO) [1] was employed, a popular DRL algorithm.
- In each training episode,
 - Character was generated using the **SMPL body**[2] approximation.
 - Custom reward was calculated for each new character.
- Reward System:**
 - The total reward was a combination of the imitation (R_i)[3] and Energy rewards (R_e)[4].
 - Total Reward** = $(1-\theta) R_i + (\theta) R_e$
 $0 \leq \theta < 1$ (dependent on body parameters)



CONCLUSION

- Our work demonstrate the potential of imitation learning for human motion synthesis and the **role of body shape parameters** in generating **motion styles and patterns**.
- We show that certain parameters, especially **mass**, play a **crucial role in human movement**, and many variations in motion can be generated.
- In future work - the addition of physiological factors, such as **the center of mass of limbs**, can enhance the design of the reward system.

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