

**2021 Southern African Universities Power Engineering  
Conference/Robotics and Mechatronics/Pattern Recognition  
Association of South Africa (SAUPEC/RobMech/PRASA),  
Potchefstroom, South Africa, 27-29 January 2021**

**Knowledge transfer using model-based deep reinforcement learning**

Boloka, Tlou J  
Council for Scientific and Industrial Research  
Pretoria, 0001, South Africa  
[tboloka@csir.co.za](mailto:tboloka@csir.co.za)

**Abstract**

Deep reinforcement learning has recently been adopted for robot behavior learning, where robot skills are acquired and adapted from data generated by the robot while interacting with its environment through a trial-and-error process. Despite this success, most model-free deep reinforcement learning algorithms learn a task-specific policy from a clean slate and thus suffer from high sample complexity (i.e., they require a significant amount of interaction with the environment to learn reasonable policies and even more to reach convergence). They also suffer from poor initial performance due to executing a randomly initialized policy in the early stages of learning to obtain experience used to train a policy or value function. Model based deep reinforcement learning mitigates these shortcomings. However, it suffers from poor asymptotic performance in contrast to a model-free approach. In this work, we investigate knowledge transfer from a model-based teacher to a task-specific model-free learner to alleviate executing a randomly initialized policy in the early stages of learning. Our experiments show that this approach results in better asymptotic performance, enhanced initial performance, improved safety, better action effectiveness, and reduced sample complexity.