



December 4, 2017

MASTER'S THESIS
for
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Task-Constrained Optimization of Multiple Demonstrations using Gaussian Mixture Models

Problem description:

In the context of programming by demonstration, Gaussian mixture models have the advantage of being able to explicitly represent the variance in multiple demonstrations for the same task [4]. Recently, GMMs have been extended so that they are able to adapt to different variations of a task [1]. In another line of research, model-free reinforcement learning and black-box optimization have been used to optimize robotic policies represented by, for instance, dynamical movement primitives [2] or probabilistic movement primitives [3]. The aim of this Thesis is to apply black-box optimization to (task-parameterized) Gaussian mixture Models. The novelty of the project is to analyze the variance in the set of multiple demonstrations in order to determine task constraints which should not be violated during optimization. The parameters of the Gaussians, which are not constrained, are optimized, i.e. to minimize the quadratic command cost along the trajectory.

Tasks:

- Understand task-parameterized Gaussian mixture models (TP-GMM) and their implementation
- Understand black-box optimization of robotic policies and trajectories
- Apply black-box optimization to TP-GMMs for a toy task in Matlab
- Conduct experiments on a KUKA LBR 4+ (Matlab Simulink)
- Write a conference paper (ICRA'18) and Master's thesis

Supervisor: Univ.-Prof. Dr.-Ing Donheui Lee
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(D. Lee)
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Bibliography:

- [1] S. Calinon. A tutorial on task-parameterized movement learning and retrieval. *Intelligent Service Robotics*, 9:1–29, 2016.
- [2] O. Sigaud F. Stulp. Robot skill learning: From reinforcement learning to evolution strategies. *Paladyn. Journal of Behavioral Robotics*, 4:49–61, 2013.
- [3] A. Paraschos. Probabilistic movement primitives. in advances in neural information processing systems (nips). *MIT Press*, 2013.
- [4] A. Billard S. Calinon, F. Guenter. On learning, representing and generalizing a task in a humanoid robot. *IEEE Transactions on Systems, Man and Cybernetics, Part B. Special issue on robot learning by observation, demonstration and imitation*, 37:26–28, 2007.