

Literature Review: Machine Learning Applied to Dynamic Physical System.

August 2018

1 Abstract

2 Background

2.1 Modeling of physical systems

1. Traditional work in modeling physical systems
Automated Design of Complex Dynamic Systems
2. Data driven design
Theory-Guided Data Science: A New Paradigm for Scientific Discovery from Data
 - (a) Machine learning based approach
 - (b) Deep learning based approach
 - Towards a Hybrid Approach to Physical Process Modeling
 - Deep learning for universal linear embeddings of nonlinear dynamics
 - Nonlinear Systems Identification Using Deep Dynamic Neural Networks
 - Analyzing Inverse Problems with Invertible Neural Networks
 - Deep Hidden Physics Models: Deep Learning of Nonlinear Partial Differential Equations
 - How Can Physics Inform Deep Learning Methods in Scientific Problems?: Recent Progress and Future Prospects
 - Learning New Physics from a Machine
 - Nanophotonic Particle Simulation and Inverse Design Using Artificial Neural Networks
 - Particle Track Reconstruction with Deep Learning
 - Neural Message Passing for Jet Physics
 - Physics-guided Neural Networks (PGNN): An Application in Lake

Temperature Modeling

- (c) Reinforcement learning based approach
 - Large-Scale Study of Curiosity-Driven Learning
 - DeepMimic: Example-Guided Deep Reinforcement Learning of Physics-Based Character Skills
- (d) Adversarial learning based approach
 - Tips and Tricks for Training GANs with Physics Constraints
 - Adversarial learning to eliminate systematic errors: a case study in High Energy Physics

2.2 Solving PDEs

Solving differential equations with unknown constitutive relations as recurrent neural networks

2.3 Non-linear control

2.4 Motor control

2.5 Time series