深度增强学习方向论文整理



一. 开山鼻祖DQN

- 1. Playing Atari with Deep Reinforcement Learning, V. Mnih et al., NIPS Workshop, 2013.
- 2. Human-level control through deep reinforcement learning, V. Mnih et al., Nature, 2015.

二. DQN的各种改进版本(侧重于算法上的改进)

- 1. Dueling Network Architectures for Deep Reinforcement Learning. Z. Wang et al., arXiv, 20 15.
- 2. Prioritized Experience Replay, T. Schaul et al., ICLR, 2016.
- 3. Deep Reinforcement Learning with Double Q-learning, H. van Hasselt et al., arXiv, 2015.
- 4. Increasing the Action Gap: New Operators for Reinforcement Learning, M. G. Bellemare et al., AAAI, 2016.
- 5. Dynamic Frame skip Deep Q Network, A. S. Lakshminarayanan et al., IJCAI Deep RL Workshop, 2016.
- 6. Deep Exploration via Bootstrapped DQN, I. Osband et al., arXiv, 2016.
- 7. How to Discount Deep Reinforcement Learning: Towards New Dynamic Strategies, V. Fran çois-Lavet et al., NIPS Workshop, 2015.
- 8. Learning functions across many orders of magnitudes, H Van Hasselt, A Guez, M Hess el, D Silver
- 9. Massively Parallel Methods for Deep Reinforcement Learning, A. Nair et al., ICML Worksh op, 2015.
- 10. State of the Art Control of Atari Games using shallow reinforcement learning
- 11. Learning to Play in a Day: Faster Deep Reinforcement Learning by Optimality Tightening (11.13更新)
- 12. Deep Reinforcement Learning with Averaged Target DQN (11.14更新)

三. DQN的各种改进版本(侧重于模型的改进)

- 1. Deep Recurrent Q-Learning for Partially Observable MDPs, M. Hausknecht and P. Stone, arXiv, 2015.
- 2. Deep Attention Recurrent Q-Network

- 3. Control of Memory, Active Perception, and Action in Minecraft, J. Oh et al., ICML, 2016.
- 4. Progressive Neural Networks
- 5. Language Understanding for Text-based Games Using Deep Reinforcement Learning
- 6. Learning to Communicate to Solve Riddles with Deep Distributed Recurrent Q-Networks
- 7. Hierarchical Deep Reinforcement Learning: Integrating Temporal Abstraction and Intrinsic Motivation
- 8. Recurrent Reinforcement Learning: A Hybrid Approach

四. 基于策略梯度的深度强化学习

深度策略梯度:

- 1. End-to-End Training of Deep Visuomotor Policies
- 2. Learning Deep Control Policies for Autonomous Aerial Vehicles with MPC-Guided Policy S earch
- 3. Trust Region Policy Optimization

深度行动者评论家算法:

- 1. Deterministic Policy Gradient Algorithms
- 2. Continuous control with deep reinforcement learning
- 3. High-Dimensional Continuous Control Using Using Generalized Advantage Estimation
- 4. Compatible Value Gradients for Reinforcement Learning of Continuous Deep Policies
- 5. Deep Reinforcement Learning in Parameterized Action Space
- 6. Memory-based control with recurrent neural networks
- 7. Terrain-adaptive locomotion skills using deep reinforcement learning
- 8. Compatible Value Gradients for Reinforcement Learning of Continuous Deep Policies
- 9. SAMPLE EFFICIENT ACTOR-CRITIC WITH EXPERIENCE REPLAY (11.13更新)

搜索与监督:

- 1. End-to-End Training of Deep Visuomotor Policies
- 2. Interactive Control of Diverse Complex Characters with Neural Networks

连续动作空间下探索改进:

1. Curiosity-driven Exploration in DRL via Bayesian Neuarl Networks

结合策略梯度和Q学习:

1. Q-PROP: SAMPLE-EFFICIENT POLICY GRADIENT WITH AN OFF-POLICY

2. PGQ: COMBINING POLICY GRADIENT AND Q-LEARNING (11.13更新)

其它策略梯度文章:

- 1. Gradient Estimation Using Stochastic Computation Graphs
- 2. Continuous Deep Q-Learning with Model-based Acceleration
- 3. Benchmarking Deep Reinforcement Learning for Continuous Control
- 4. Learning Continuous Control Policies by Stochastic Value Gradients

五. 分层DRL

- 1. Deep Successor Reinforcement Learning
- 2. Hierarchical Deep Reinforcement Learning: Integrating Temporal Abstraction and Intrinsic Motivation
- 3. Hierarchical Reinforcement Learning using Spatio-Temporal Abstractions and Deep Neural Networks
- 4. Stochastic Neural Networks for Hierarchical Reinforcement Learning Authors: Carlos Florensa, Yan Duan, Pieter Abbeel (11.14更新)

六. DRL中的多任务和迁移学习

- 1. ADAAPT: A Deep Architecture for Adaptive Policy Transfer from Multiple Sources
- 2. A Deep Hierarchical Approach to Lifelong Learning in Minecraft
- 3. Actor-Mimic: Deep Multitask and Transfer Reinforcement Learning
- 4. Policy Distillation
- 5. Progressive Neural Networks
- 6. Universal Value Function Approximators
- 7. Multi-task learning with deep model based reinforcement learning (11.14更新)
- 8. Modular Multitask Reinforcement Learning with Policy Sketches (11.14更新)

七. 基于外部记忆模块的DRL模型

- 1. Control of Memory, Active Perception, and Action in Minecraft
- 2. Model-Free Episodic Control

八. DRL中探索与利用问题

1. Action-Conditional Video Prediction using Deep Networks in Atari Games

- 2. Curiosity-driven Exploration in Deep Reinforcement Learning via Bayesian Neural Network
- 3. Deep Exploration via Bootstrapped DQN
- 4. Hierarchical Deep Reinforcement Learning: Integrating Temporal Abstraction and Intrinsic Motivation
- 5. Incentivizing Exploration In Reinforcement Learning With Deep Predictive Models
- 6. Unifying Count-Based Exploration and Intrinsic Motivation
- 7. #Exploration: A Study of Count-Based Exploration for Deep Reinforcemen Learning (11.14更新)
- 8. Surprise-Based Intrinsic Motivation for Deep Reinfo rcement Learning (11.14更新)

九. 多Agent的DRL

- 1. Learning to Communicate to Solve Riddles with Deep Distributed Recurrent Q-Networks
- 2. Multiagent Cooperation and Competition with Deep Reinforcement Learning

十. 逆向DRL

- 1. Guided Cost Learning: Deep Inverse Optimal Control via Policy Optimization
- 2. Maximum Entropy Deep Inverse Reinforcement Learning
- 3. Generalizing Skills with Semi-Supervised Reinforcement Learning (11.14更新)

十一. 探索+监督学习

- 1. Deep learning for real-time Atari game play using offline Monte-Carlo tree search planning
- 2. Better Computer Go Player with Neural Network and Long-term Prediction
- 3. Mastering the game of Go with deep neural networks and tree search, D. Silver et al., Natu re, 2016.

十二. 异步DRL

- 1. Asynchronous Methods for Deep Reinforcement Learning
- 2. Reinforcement Learning through Asynchronous Advantage Actor-Critic on a GPU (11.14 更新)

十三:适用于难度较大的游戏场景

- 1. Hierarchical Deep Reinforcement Learning: Integrating Temporal Abstraction and Intrinsic Motivation, T. D. Kulkarni et al., arXiv, 2016.
- 2. Strategic Attentive Writer for Learning Macro-Actions

3. Unifying Count-Based Exploration and Intrinsic Motivation

十四:单个网络玩多个游戏

- 1. Policy Distillation
- 2. Universal Value Function Approximators
- 3. Learning values across many orders of magnitude

十五: 德州poker

- 1. Deep Reinforcement Learning from Self-Play in Imperfect-Information Games
- 2. Fictitious Self-Play in Extensive-Form Games
- 3. Smooth UCT search in computer poker

十六: Doom游戏

- 1. ViZDoom: A Doom-based Al Research Platform for Visual Reinforcement Learning
- 2. Training Agent for First-Person Shooter Game with Actor-Critic Curriculum Learning
- 3. Playing FPS Games with Deep Reinforcement Learning
- 4. LEARNING TO ACT BY PREDICTING THE FUTURE (11.13更新)
- 5. Deep Reinforcement Learning From Raw Pixels in Doom (11.14更新)

十七: 大规模动作空间

1. Deep Reinforcement Learning in Large Discrete Action Spaces

十八:参数化连续动作空间

1. Deep Reinforcement Learning in Parameterized Action Space

十九: Deep Model

- 1. Learning Visual Predictive Models of Physics for Playing Billiards
- 2. J. Schmidhuber, On Learning to Think: Algorithmic Information Theory for Novel Combinations of Reinforcement Learning Controllers and Recurrent Neural World Models, arXiv, 2015. arXiv
- 3. Learning Continuous Control Policies by Stochastic Value Gradients
- 4.Data-Efficient Learning of Feedback Policies from Image Pixels using Deep Dynamical Mod els
- 5. Action-Conditional Video Prediction using Deep Networks in Atari Games
- 6. Incentivizing Exploration In Reinforcement Learning With Deep Predictive Models

二十: DRL应用

机器人领域:

- 1. Trust Region Policy Optimization
- 2. Towards Vision-Based Deep Reinforcement Learning for Robotic Motion Control
- 3. Path Integral Guided Policy Search
- 4. Memory-based control with recurrent neural networks
- 5. Learning Hand-Eye Coordination for Robotic Grasping with Deep Learning and Large-Scal e Data Collection
- 6. Learning Deep Neural Network Policies with Continuous Memory States
- 7. High-Dimensional Continuous Control Using Generalized Advantage Estimation
- 8. Guided Cost Learning: Deep Inverse Optimal Control via Policy Optimization
- 9. End-to-End Training of Deep Visuomotor Policies
- 10. DeepMPC: Learning Deep Latent Features for Model Predictive Control
- 11. Deep Visual Foresight for Planning Robot Motion
- 12. Deep Reinforcement Learning for Robotic Manipulation
- 13. Continuous Deep Q-Learning with Model-based Acceleration
- 14. Collective Robot Reinforcement Learning with Distributed Asynchronous Guided Policy S earch
- 15. Asynchronous Methods for Deep Reinforcement Learning
- 16. Learning Continuous Control Policies by Stochastic Value Gradients

机器翻译:

1. Simultaneous Machine Translation using Deep Reinforcement Learning

目标定位:

1. Active Object Localization with Deep Reinforcement Learning

目标驱动的视觉导航:

1. Target-driven Visual Navigation in Indoor Scenes using Deep Reinforcement Learning

自动调控参数:

1. Using Deep Q-Learning to Control Optimization Hyperparameters

人机对话:

1. Deep Reinforcement Learning for Dialogue Generation

- 2. SimpleDS: A Simple Deep Reinforcement Learning Dialogue System
- 3. Strategic Dialogue Management via Deep Reinforcement Learning
- 4. Towards End-to-End Learning for Dialog State Tracking and Management using Deep Rei nforcement Learning

视频预测:

1. Action-Conditional Video Prediction using Deep Networks in Atari Games

文本到语音:

1. WaveNet: A Generative Model for Raw Audio

文本生成:

1. Generating Text with Deep Reinforcement Learning

文本游戏:

1. Language Understanding for Text-based Games Using Deep Reinforcement Learning

无线电操控和信号监控:

Deep Reinforcement Learning Radio Control and Signal Detection with KeRLym, a Gym R
Agent

DRL来学习做物理实验:

1. LEARNING TO PERFORM PHYSICS EXPERIMENTS VIA DEEP REINFORCEMENT LEA RNING (11.13更新)

DRL加速收敛:

1. Deep Reinforcement Learning for Accelerating the Convergence Rate (11.14更新)

利用DRL来设计神经网络:

- 1. Designing Neural Network Architectures using Reinforcement Learning (11.14更新)
- 2. Tuning Recurrent Neural Networks with Reinforcement Learning (11.14更新)
- 3. Neural Architecture Search with Reinforcement Learning (11.14更新)

控制信号灯:

1. Using a Deep Reinforcement Learning Agent for Traffic Signal Control (11.14更新)

二十一: 其它方向

避免危险状态:

1. Combating Deep Reinforcement Learning's Sisyphean Curse with Intrinsic Fear (11.14更新)

DRL中On-Policy vs. Off-Policy 比较:

1. On-Policy vs. Off-Policy Updates for Deep Reinforcement Learning (11.14更新)

注1: 小伙伴们如果觉得论文一个个下载太麻烦,可以私信 我,我打包发给你。

注2: 欢迎大家及时补充新的或者我疏漏的文献。

深度学习(Deep Learning) 深度增强学习(Deep Reinforcement Learning)

机器学习

414

分享 鋼举报



16 条评论

写下你的评论



韩劼群

最近放出来许多2017ICLR的投稿,有不少是关于DRL的,我目前读过里面比较有意思的 有: 1. Learning to Play in a Day: Faster Deep Reinforcement Learning by Optimality Tightening, 2. PGQ: Combining policy gradient and Q-learning, 3. Q-Prop: Sample-Efficient Policy Gradient with An Off-Policy Critic, 4. Sample Efficient Actor-Critic with Experience Replay, 5. Learning to Act by Predicting the Future。1,2,4都应用在了Atari Games上, 3,4 应用在Robotics continuous control上, 5 在 Doom Full Deathmatch track 中赢得了第一名。

17 天前 6 赞



lin feng

赞

17 天前 1 赞



Alex-zhai (作者) 回复 韩劼群

查看对话

最近比较忙,还没来得及关注,我加一下。谢谢啦。

17 天前 2 赞



朱军

可以在github上建个repo

17 天前

吴江



计算机论文的学习不要脱离代码,在各位同学看论文时请尽量在github上撸点代码,有时候 他们的程序实现跟论文写得完全不是一回事。

17 天前 4 赞



张夏天

赞

17 天前



Alex-zhai (作者) 回复 吴江

谢谢老师的建议。

17 天前



天清

感谢

16 天前



韩劼群 回复 朱军

可以看一下这个 repo,muupan/deep-reinforcement-learning-papers ,不过作者最近好像 没啥更新

16 天前



机械键盘侠

mark

16 天前

1

2 下一页