Algorithms in Reinforcement Learning

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Algorithm 1 Actor-Critic with Eligibility Traces (episodic), for estimating \pi_{\theta} \approx \pi_*
Input: a differentiable policy parameterization \hat{v}(s, \mathbf{w})
Input: a differentiable state-value function parameterization \hat{v}(s, \mathbf{w})
Algorithm parameters: trace-decay rate \lambda^{\theta} \in [0,1], \lambda^{\mathbf{w}} \in [0,1]; step sizes \alpha^{\theta} > 0, \alpha^{\mathbf{w}} > 0
Initialize policy parameter \boldsymbol{\theta} \in \mathbb{R}^{d'} and state-value weights \mathbf{w} \in \mathbb{R}^{d} (e.g., to 0)
loop forever (for each episode):
       initialize S (first state of episode)
       \mathbf{z}^{\boldsymbol{\theta}} \leftarrow \mathbf{0} \ (d'-component eligibility trace vector)
       \mathbf{z}^{\mathbf{w}} \leftarrow \mathbf{0} \ (d'-component eligibility trace vector)
       loop while S is not terminal (for each time step):
             A \in \pi(\cdot|S, \boldsymbol{\theta})
             Take action A, observe S', R
                                                                       (if S' is terminal, then \hat{v}(S', \mathbf{w}))
             \delta \leftarrow \gamma \hat{v}(S', \mathbf{w}) - \hat{v}(S, \mathbf{w})
             \mathbf{z}^{\mathbf{w}} \leftarrow \gamma \lambda^{\mathbf{w}} \mathbf{z}^{\mathbf{w}} + I \nabla \hat{v}(s, \mathbf{w})
             \mathbf{z}^{\boldsymbol{\theta}} \leftarrow \gamma \lambda^{\boldsymbol{\theta}} \mathbf{z}^{\boldsymbol{\theta}} + I \nabla \hat{v}(s, \boldsymbol{\theta})
             \mathbf{w} \leftarrow \mathbf{w} + \alpha^{\mathbf{w}\delta\mathbf{z}^{\mathbf{w}}}
             \boldsymbol{\theta} \leftarrow \boldsymbol{\theta} + \alpha^{\theta} \delta \mathbf{z}^{\boldsymbol{\theta}}
             \mathbf{I} \leftarrow \gamma \mathbf{I}
             S \leftarrow S'
       end loop
end loop
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Algorithm 2 Actor-Critic with Eligibility Traces (continuing), for estimating \pi_{\theta} \approx \pi_*
Input: a differentiable policy parameterization \pi(a|s, \theta)
Input: a differentiable state-value function parameterization \hat{v}(s, \mathbf{w})
Algorithm parameters: trace-decay rates \lambda^{\theta} \in [0,1], \lambda^{\mathbf{w}} \in [0,1]; step sizes \alpha^{\theta} > 0, \alpha^{\mathbf{w}} > 0
0, \eta > 0
Initialize \bar{R} \in \mathbb{R} (e.g., to 0)
Initialize policy parameter \boldsymbol{\theta} \in \mathbb{R}^{d'} and state-value weights \mathbf{w} \in \mathbb{R}^{d} (e.g., to \mathbf{0}) \doteq 0
loop forever (for each time step):
       A \sim \pi(\cdot|S, \boldsymbol{\theta})
       Take Action A, observe S', R
       \delta \leftarrow R - \bar{R} + \hat{v}(S', \mathbf{w}) - \hat{v}(S, \mathbf{w})
       \bar{R} \leftarrow \bar{R} + \eta \delta
       \mathbf{z}^{\mathbf{w}} \leftarrow \lambda^{\mathbf{w}} \mathbf{z}^{\dot{\mathbf{w}}} + \nabla \hat{v}(S, \mathbf{w})
       \mathbf{z}^{\boldsymbol{\theta}} \leftarrow \lambda^{\boldsymbol{\theta}} \mathbf{z}^{\boldsymbol{\theta}} + \nabla \ln \widehat{\pi}(A|S, \boldsymbol{\theta})
       \mathbf{w} \leftarrow \mathbf{w} + \alpha^{\mathbf{w}} \delta \mathbf{z}^{\mathbf{w}}
       \mathbf{w} \leftarrow \boldsymbol{\theta} + \alpha^{\boldsymbol{\theta}} \delta \mathbf{z}^{\boldsymbol{\theta}}
       S \leftarrow S^{'}
end loop
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