## Input: Expert trajectories τ<sub>E</sub> ~ π<sub>E</sub>, initial policy and discriminator parameters θ<sub>0</sub>, w<sub>0</sub> for i = 0, 1, 2, ... do Sample trajectories $\tau_i \sim \pi_{\theta_i}$ Update the discriminator parameters from $w_i$ to $w_{i+1}$ with the gradient 4: $\hat{\mathbb{E}}_{\tau_s}[\nabla_w \log(D_w(s, a))] + \hat{\mathbb{E}}_{\tau_E}[\nabla_w \log(1 - D_w(s, a))]$ (17)Take a policy step from $\theta_i$ to $\theta_{i+1}$ , using the TRPO rule with cost function $\log(D_{w_{i+1}}(s, a))$ . 5: Specifically, take a KL-constrained natural gradient step with

6: end for

 $\hat{\mathbb{E}}_{\tau}$ .  $[\nabla_{\theta} \log \pi_{\theta}(a|s)Q(s,a)] - \lambda \nabla_{\theta}H(\pi_{\theta})$ ,

$$\hat{\mathbb{E}}_{\tau_i} \left[ \nabla_{\theta} \log \pi_{\theta}(a|s) Q(s,a) \right] - \lambda \nabla_{\theta} H(\pi_{\theta}),$$

(18)

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$$\mathbb{D}_{T_i} \left[ V \theta \log \pi \theta(\alpha | \beta) Q(\beta, \alpha) \right] \times V \theta \Pi(\pi \theta),$$

where 
$$O(\bar{z}, \bar{z}) = \hat{\mathbb{R}} \left[ \log(D - \langle z, z \rangle) | z - \bar{z}, z - \bar{z} \right]$$

where 
$$O(\bar{s}, \bar{a}) = \hat{\mathbb{E}}_{-} [\log(D_{-}, (s, a)) | s_0 = \bar{s}, a_0 = \bar{a}]$$

where  $Q(\bar{s}, \bar{a}) = \hat{\mathbb{E}}_{\tau_s}[\log(D_{w_{s+1}}(s, a)) | s_0 = \bar{s}, a_0 = \bar{a}]$ 

Algorithm 1 Generative adversarial imitation learning