Overview of Dec-POMDP notation used

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Set of actions for agent i
A_i
         Local action for agent i
a_i
\vec{A}
         Set of joint actions, \vec{A} = \bigotimes_{i \in I} A_i
         Joint action, \vec{a} = \langle A_1, \dots, A_n \rangle
\vec{a}
C_{\Sigma}
         Cost of transmitting an atomic message
         Discount factor, \gamma \in [0, 1)
\gamma
         Local policy for agent i, mapping from \vec{\mathbf{o}}^i = o_{i_1} \cdots o_{i_t} to actions a_i
\pi_i
         Joint policy \pi = \langle \pi_1, \dots, \pi_n \rangle
E(\cdot)
         Expectation operator
         Planning horizon
         Action-observation history, \vec{\theta}_i^t = (a_i^0, o_i^1, \dots, a_i^{t-1}, o_i^t)
\vec{\theta}_i^t
         Set of joint types (Bayesian game), \Theta = \times_i \Theta_i
Θ
I
         Set of agents, indexed 1, \ldots, n
\Omega_i
         Set of observations for agent i
         Local observation for agent i
o_i
\vec{\Omega}
         Set of joint observation, \vec{\Omega} = \bigotimes_{i \in I} \Omega_i
\vec{o}
         Joint observation, \vec{o} = \langle o_1, \dots, o_n \rangle
         Number of agents
n
         Local policy tree for agent i
q_i
R
         Reward function (Dec-POMDP), R(\vec{a}, s')
R_i
         Individual reward function for POSG R_i(a_i, s')
S
         Set of states
         Initial state
s_0
         Local state (in factored Dec-MDP), \hat{s_i} \in S_i \times S_0
\hat{s_i}
\Sigma
         Alphabet of communication messages
         Atomic message sent by agent i
\sigma_i
         Joint message, \vec{\sigma} = \langle \sigma_1, \dots, \sigma_n \rangle
\vec{\sigma}
t
         Time step t = 0, \ldots, h-1
V
         Value function (expected discounted future reward)
\widehat{V}
         Heuristic value function
\varphi^t
         Partial joint policy (up to time step t), \varphi^t = (\delta^0, \delta^1, \dots, \delta^{t-1})
         Payoff function in Bayesian game
u_i
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