

---

---

# Assignment 11

## Reinforcement Learning

Prof. B. Ravindran

---

---

- Using the MAXQ approach leads to solutions which are
  - hierarchically optimal
  - recursively optimal
  - flat optimal
- We saw that each sub-task has an associated pseudo-reward function. Are the rewards of the core MDP available to the agent while it is learning policies of individual sub-tasks or is the agent restricted to the corresponding sub-task's pseudo rewards?
  - only pseudo rewards are available
  - both pseudo rewards and core MDP rewards are available
- In the MAXQ framework, is termination in a sub-task deterministic or stochastic as in the options framework?
  - deterministic
  - stochastic
- Each sub-task  $M_i$  is an SMDP because
  - the state space of the sub-task is a subset of the state space of the core MDP
  - each sub-task has its own policy
  - actions in a sub-task can be temporally extended
  - the rewards received in sub-tasks depend not only on the state but also on the sub-task in which an action was executed
- The expected reward function  $\bar{R}(s, a)$  of the SMDP corresponding to sub-task  $M_i$  is equivalent to the projected value function  $V^{\pi_i}(a, s)$ . True or false?
  - false
  - true
- In the MAXQ approach to solving a problem, suppose that sub-task  $M_i$  invokes sub-task  $M_j$ . Do the pseudo rewards of  $M_j$  have any effect on sub-task  $M_i$ ?
  - no
  - yes