

# Reinforcement Learning

## ■ Q - Learning

- Off-Policy Temporal-Difference Control
  - differentiate behavior-policy from learning-policy
  - SARSA (on policy)
    - ▶  $\langle s, a, r, s', a' \leftarrow \pi(s) \rangle \Rightarrow$  Learning
  - Q-Learning
    - ▶  $\langle s, a \leftarrow \pi(s), r, s' \rangle \Rightarrow$  Learning
  - Update rule

$$Q(S_t, A_t)_{new} = Q(S_t, A_t)_{old} + \alpha [R_{t+1} + \gamma \max_{a \in A} Q(S_{t+1}, a') - Q(S_t, A_t)_{old}]$$

# Reinforcement Learning

## ■ Function Approximation

- Why function approximation?
  - Problem with large state spaces
    - ▶ Large memory for large table task
    - ▶ data should be accurate
  - Generalization
    - ▶ to generalize from previous encounters with different states that are in some sense similar to the current ones
- Generalization => function approximation
  - to generalize desired functions (e.g value function, q function etc.)
  - utilize *supervised learning*