

2.4 Problem Formulation and Examples

Example ① Reaching a goal in minimum expected time

$$\lambda = 1, \quad g(i, u, j) = 1, \quad \forall i \neq 0, j, \quad u \in U(i).$$

Here, the optimal cost-to-go $J^*(i)$ is the min. expected time to termination starting from i .

$$J^*(i) = \min_{u \in U(i)} \left[1 + \sum_{j=1}^n p_{ij}(u) J^*(j) \right], \quad i = 1, 2, \dots, n$$

Example ② Problems with uncontrollable state components

In many shortest path problems the evolution of a state (i, y) may not be affected by control u .

For the next state (j, z) ,

j is generated according $p_{ij}(u, y)$.

z is generated according to conditional probabilities $p(z|j)$ that depend on j .

Consider the cost of transition $g(i, y, u, j)$ and does not depend on uncontrollable component z .

If g depends on z , $\hat{g}(i, y, u, j) = \sum_z p(z|j) g(i, y, u, j, z)$.

Such algorithms provide a smaller state space.
To see this,

$$(T\hat{J})(i, y) = \min_{u \in U(i)} \sum_{j=0}^n p_{ij}(u, y) (\hat{g}(i, y, u, j) + \sum_z p(z|j) J(j, z)).$$

$$(T_J)(i, y) = \sum_{j=0}^n p_{ij}(u(i, y), y) (g(i, y, u(i, y), j) + \sum_z p(z|j) J(j, z))$$

for each J ,

$$\hat{J}(j) = \sum_z p(z|j) J(j, z).$$

To compute the optimal policy, it is sufficient to know $\hat{f}^*(j)$,

$$\hat{f}^*(j) = \sum_z p(z|j) J^*(j, z).$$

$\underbrace{\quad}_{\text{reduced optimal cost-to-go vector}}$

This leads to simplified versions,

$$(\hat{T}\hat{f})(i) = \sum_y p(y|i) (\hat{T}\hat{J})(i, y) = \sum_y p(y|i) \left(\min_{u \in U(i, y)} \sum_{j=0}^n p_{ij}(u, y) (g(i, y, u, y, j) + \hat{f}^*(j)) \right)$$

$$\text{and } (\hat{T}_u \hat{f})(i) = \sum_y p(y|i) (T_u J)(i, y) = \sum_y p(y|i) \left(\sum_{j=0}^n p_{ij}(u, y) (g(i, y, u, y, j) + \hat{f}^*(j)) \right).$$

(a) **Policy Evaluation** - Compute unique $\hat{f}^{u_k}(i)$ with $\hat{f}^{u_k} = \hat{T}_{u_k} \hat{f}^{u_k}$

$$\hat{f}^{u_k}(i) = \sum_y p(y|i) \left(\sum_{j=0}^n p_{ij}(u_k(i, y)) (g(i, y, u_k(i, y), j) + \hat{f}^{u_k}(j)) \right).$$

(b) **Policy Improvement** - Compute $\mu_{k+1}(i, y)$ from $\hat{T}_{\mu_{k+1}} \hat{f}^{u_k} = \hat{T} \hat{f}^{u_k}$,

$$\mu_{k+1}(i, y) = \arg \min_{a \in U(i, y)} \sum_{j=0}^n p_{ij}(a, y) (g(i, y, a, j) + \hat{f}^{u_k}(j)).$$

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