# Problem Solving in Computer Science - Week 1 Notes

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# 1 General Information for Course Attendants

- Course assessment will be done through project work, in groups of 3 to 4 people;
- Main course information hub is the course home page at: http://mtc.epfl.ch/courses/ProblemSolving-2008/;
- Teaching Assistant (TA): Ms. Verena Wolf (http://mtc.epfl.ch/~vwolf/);
- It is preferable that the decision to continue / discontinue attending the course be taken as soon as possible by the participants;
- Every week a volunteer will take notes for the whole class, turn them into a LATEX document, and make it available to everybody through the TA.

# 2 First Project - Probabilistic Graph Games

## 2.1 Description of the Problem (Informal)

**The problem:** the input to the problem is a directed graph, with 3 types of nodes. Type 1 nodes called MAX nodes, type 2 nodes called MIN nodes, and type 3 nodes called RAND nodes (from random). For simplicity, each node has out-degree 2, except for two nodes in the graph, one MAX node and one MIN node. These nodes without any descendants are called *sink nodes*. Two players, MAXP and MINP, and a token are also part of the problem.

Figure ?? shows an example of such a graph. Square nodes are MAX nodes, diamond nodes are MIN nodes, round nodes are RAND nodes. Doubled nodes are the sink-MAX and sink-MIN nodes. The graph game supposes the token is present at any given time at a given node in the graph, and can be moved from one node to another according to a set of rules and objectives. MAXP's objective is to take the token to the sink-MAX node, MINP's objective is to take the token to the sink-MIN node, point at which the game ends. The token moving rules are:

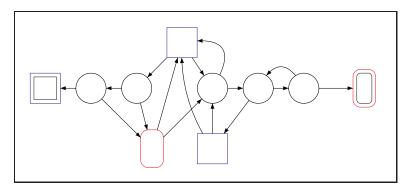


Figure 1: Example of a probabilistic graph. Square nodes are MAX nodes, roundish rectangular nodes are MIN nodes, and round nodes are RAND nodes. Sink-MAX and sink-MIN have doubled edges.

- 1. the token can be moved to a node if and only if a directed edge links the current node to the target node;
- 2. if the token is at a MAX node, it will be moved by the MAXP player;
- 3. if the token is at a MIN node, it will be moved by the MINP player;
- 4. if the token is at a RAND node, it will be moved with a random probability defined a priori for the RAND node (for our example, uniform probability  $\frac{1}{2}$ ).

**The question:** what is each player's probability to win, supposing they play optimally?

Additional simplifications:

- the full graph is known in advance by the two players;
- the probability to eventually reach one of the sink nodes is one.

### 2.2 Known Algorithms

**Observation:** in order to compute the winning probability for either player, we will need to compute partial at-node min/max winning probabilities. Without loss of generality, we will consider from here on only the winning probability of MAXP.

#### 2.2.1 Value Improvement Algorithm

Let us call the probability of MAXP to win the game if the token is at the current node i the value of the node, denoted by  $x_i$ . Under this notation, the value of sink-MAX for player MAXP is 1, and the value of sink-MIN for player MAXP is 0. In general, for MAXP we have the following rules for a node i's value with two descendants j and k, regardless of nodes' j and k types:

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• if i is a MAX node: x_i = \max(x_i, x_k);
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- if i is a MIN node:  $x_i = \min(x_i, x_k)$ ;
- if i is a RAND node:  $\frac{x_j + x_k}{2}$ .

The algorithm has the following steps:

Step 1 Naively assign node values as follows:

- all MAX nodes value: 1;
- all MIN nodes value: 0;
- all RAND nodes value:  $\frac{1}{2}$ .

Figure ?? shows this first step with the naive node values.

**Step 2:** pick a node for which the rules do not hold, update its value, and propagate this update through the entire graph. Repeat this step until all nodes' values satisfy the rules.

**Observation:** this algorithm is guaranteed to converge to a solution, however it may need exponential iterations (in the size of the graph).

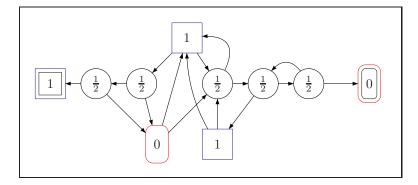


Figure 2: Naively assigning node values for MAXP

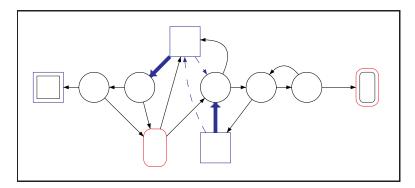


Figure 3: Fixing strategy  $\sigma$  for MAXP and removing unused edges

#### 2.2.2 Strategy Improvement Algorithm

**Observation:** it does not make sense to choose different edges for a different path history. Hence, we can talk about strategies for each of the players, denoted as  $\sigma$  for MAXP and  $\tau$  for MINP, that will select a direction (Left or Right) in each node:

- $\sigma: V_{max} \to \{\text{Left, Right}\};$
- $\tau$ :  $V_{min} \to \{\text{Left, Right}\}.$

Once a strategy is fixed, some edges become unused, and can thus be removed from the graph. Figure ?? exemplifies this: in green, the strategy, and in superimposed green dashes the removed edges.

Fixing a strategy for one player (say MAXP) renders the problem into a Markov Decision Process (MDP), with only the MINP being able to make choices. This MDP is solvable by linear programming, with the following constraints:

• for MIN nodes,  $x_i \ge min(x_j, x_k)$ ;

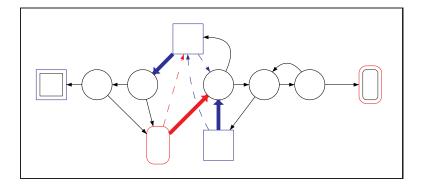


Figure 4: Fixing strategy  $\sigma$  for MAXP and removing unused edges, then determining optimal strategy  $\tau$  for MINP and removing unused edges

- for RAND nodes,  $x_i = \frac{x_j + x_k}{2}$ ;
- $x_i \geq 0$ .

The objective function in this case is  $O = \min \sum_{i} x_i$ 

The algorithm, assuming we are solving for MAXP, has the following steps: **Step 1:** arbitrarily guess a strategy  $\sigma$  for MAXP.

Step 2: determine optimal strategy  $\tau$  for MINP. This is exemplified in Figure ??.

Once both strategies are fixed, the problem becomes a Markov Chain (MC) with no choices except the RAND nodes. In this MC we can compute node values through linear equations on each node:

- for RAND nodes,  $x_i = \frac{x_j + x_k}{2}$ ;
- for sink-MAX,  $x_i = 1$ .
- for sink-MIN,  $x_i = 0$ ;
- MIN and MAX nodes (other than sinks) inherit the value of their (now unique) successor.

Step 3: once node values are available, improve if possible the strategy  $\sigma$  taking into account these values, and then continue from Step 2 until no more changes are available. This step is exemplified in Figure ??.

**Observation** This algorithm is not guaranteed to find a solution in polynomial number of iterations.

**Observation** The decision problem is in  $NP \cap coNP$ , strategies are polynomial size witnesses.

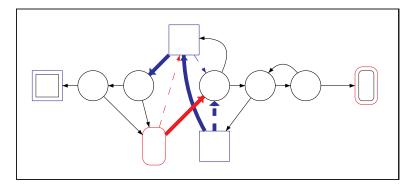


Figure 5: Re-evaluating strategy  $\sigma$  based on node values

# 2.3 TO DO

- $\bullet$  Form groups of 3 to 4 people, and come up with a formal definition of the problem;
- Test on a sample graph the two algorithms described above;
- Make decision on taking or dropping the course.