Nondeterministic Bigraphical Reactive Systems for Markov Decision Processes*

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Abstract. In this paper we introduce two extensions of bigraphical reactive systems (BRSs) to handle probabilistic and nondeterministic behaviour, allowing BRSs to be used in discrete-time Markov chain and Markov decision process (MDP) generation. We extend the implementation of an open-source tool for BRSs BigraphER with new syntax for MDP actions and rewards, better visualisation capabilities, and data export options. Furthermore, we introduce a new front-end for working with BRSs based on Jupyter notebooks, allowing the user to quickly visualise bigraphs and reaction rules, simulate execution, or produce a complete transition system, all within the notebook interface. Finally, we demonstrate the modelling capabilities of the new system with a case study on the movement and behaviour of autonomous agents, exploring ideas such as collecting objects, tracking visited locations, hierarchies of spaces, and uncertainty in the topology of space.

Keywords: Bigraphs \cdot Bigraphical reactive systems \cdot Probabilistic semantics \cdot Markov decision process \cdot Spatial models.

1 Introduction

1.1 Markov Decision Processes

Definition 1 (Markov decision process). For any finite set X, let Dist(X) denote the set of discrete probability distributions over X. A Markov Decision Process is a tuple $(S, \overline{s}, A, P, L)$, where: S is a finite set of states and $\overline{s} \in S$ is the initial state; A is a finite set of actions; $P: S \times A \to Dist(S)$ is a (partial) probabilistic transition function, mapping state-action pairs to probability distributions over S; $L: S \to 2^P$ is a labelling with atomic propositions.

Definition 2 (reward structure). A reward structure for an MDP $(S, \overline{s}, A, P, L)$ is a pair (ρ, ι) , where $\rho: S \to \mathbb{R}$ is the state reward function, and $\iota: S \times A \to \mathbb{R}$ is the transition reward function.

^{*} Supported by organization x.

1.2 Bigraphs

We begin by introducing some notation used in the definitions. Natural numbers are sometimes interpreted as finite ordinals, i.e., $m = \{0, \ldots, m-1\}$. We write $S \uplus T$ to indicate the union of sets known or assumed to be disjoint. For two functions f and g with disjoint domains S and T we write $f \uplus g$ for the function with domain $S \uplus T$ such that $(f \uplus g)|_S = f$ and $(f \uplus g)|_T = g$. The identity function on the set S is written as Id_S . Given a binary relation $\operatorname{rel} \subseteq A \times B$, we denote the domain restriction of rel over $S \subseteq A$ by $S \lhd \operatorname{rel}$. Similarly, we write $\operatorname{rel} \rhd S$ for the range restriction of rel over $S \subseteq B$. We also denote the transitive closure of rel by rel^+ . In defining bigraphs we assume that names, node identifiers, edge identifiers, and controls are drawn from four infinite pairwise disjoint sets, respectively $\mathcal{X}, \mathcal{V}, \mathcal{E}$, and \mathcal{K} .

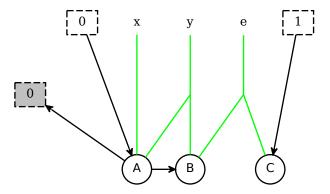


Fig. 1. An example bigraph.

Informally, a bigraph can be thought of as a graph, where a node can be inside another node¹. However, the pictures tend to become much clearer if we instead visualise the nesting relations between nodes as just a different type of edges. Consider the bigraph in Fig. 1. The white ellipses are the *nodes*. Each node has a type, called *control*, and denoted here by the labels A, B, and C. The green edges are actually hyperedges (can connect any positive number of nodes), and are called *links*. In this example they are given names x, y, and e. Each node has an ordered list of *ports* (not pictured in the diagram) that can be thought of as sockets for links. Nodes of the same control have the same number of ports. The number of ports a node of control K can have is called the *arity* of K and is denoted by ar(K). The two white dashed rectangles represent *regions*, and in this representation are the parents to any otherwise parentless nodes. Grey dashed rectangles are called *sites*, and encode parts of the model that have been

¹ To be specific, in this paper we extend the ideas behind *bigraphs with sharing* [7,8], where a node can be contained in the intersection of several other nodes.

abstracted away. Nodes, sites, and regions are also known as *places*. The directed black arrows between places show the containment relations, e.g., the A node has two children (inner nodes): B and the only site in this bigraph, labelled by the number 0.

Another, more formal way to think about a bigraph is as a pair of graphs: one for describing the placement of nodes, and one for the links between them. This is the approach that will be featured in our formal definitions. We also focus on *concrete* bigraphs, where nodes and links have distinct identifiers.

Definition 3 (concrete place graph with sharing). A concrete place graph with sharing

$$F = (V_F, ctrl_F, prnt_F) : m \to n$$

is a triple having an inner interface m and an outer interface n. These index the sites and regions of the place graph, respectively. F has a finite set $V_F \subset \mathcal{V}$ of nodes, a control map $ctrl_F: V_F \to \mathcal{K}$, and a parent relation

$$prnt_F \subseteq (m \uplus V_F) \times (V_F \uplus n)$$

that is acyclic, i.e., $(v,v) \notin prnt_F^+$ for any $v \in V_F$.

We define several operations on place graphs that will be used later.

Definition 4 (composition for place graphs with sharing). If $F: k \to m$ and $G: m \to n$ are two concrete place graphs with sharing with $V_F \cap V_G = \emptyset$, their composite

$$G \circ F = (V, ctrl, prnt) : k \to n$$

has nodes $V := V_F \uplus V_G$ and a control map $ctrl := ctrl_F \uplus ctrl_G$. Its parent relation $prnt \subseteq (k \uplus V) \times (V \uplus n)$ is given by:

$$prnt := prnt_G^{\triangleleft} \uplus prnt_{\circ} \uplus prnt_F^{\triangleright},$$

where

$$\begin{split} prnt_F^{\rhd} &= prnt_F \rhd V_F, \\ prnt_G^{\lhd} &= V_G \lhd prnt_G, \\ prnt_{\circ} &= (m \lhd prnt_G) \circ (prnt_F \rhd m). \end{split}$$

Definition 5 (tensor product for place graphs). If

$$G_0 = (V_{G_0}, ctrl_{G_0}, prnt_{G_0}) : m_0 \to n_0$$

and

$$G_1 = (V_{G_1}, ctrl_{G_1}, prnt_{G_1}) : m_1 \to n_1$$

are two concrete place graphs with sharing with $V_{G_0} \cap V_{G_1} = \emptyset$, their tensor product

$$G_0 \otimes G_1 = (V, ctrl, prnt) : m_0 + m_1 \rightarrow n_0 + n_1$$

has nodes $V := V_{G_0} \uplus V_{G_1}$ and a control map $ctrl := ctrl_{G_0} \uplus ctrl_{G_1}$. Its parent relation $prnt \subseteq [(m_0 + m_1) \uplus V] \times [V \uplus (n_0 + n_1)]$ is defined as

$$prnt_{G_0} \uplus prnt_{G_1}^{(m_0,n_0)},$$

where

Definition 6 (concrete link graph). A concrete link graph

$$F = (V_F, E_F, ctrl_F, link_F) : X \to Y$$

is a quadruple having an inner face X and an outer face Y, both finite subsets of \mathcal{X} , called respectively the inner and outer names of the link graph. F has finite sets $V_F \subset \mathcal{V}$ of nodes and $E_F \subset \mathcal{E}$ of edges, a control map $ctrl_F : V_F \to \mathcal{K}$ and a link map

$$link_F: X \uplus P_F \to E_F \uplus Y$$
,

where $P_F := \{(v,i) \mid i \in ar(ctrl_F(v))\}$ is the set of ports of F. Thus (v,i) is the ith port of node v. We shall call $X \uplus P_F$ the points of F, and $E_F \uplus Y$ its links.

Two places with the same parent, or two points with the same link are called siblings. A link is idle if it has no points. Identities over (inner or outer) name sets are defined by $id_X = (\emptyset, \emptyset, \emptyset, Id_X) : X \to X$.

Definition 7 (concrete bigraph with sharing). A concrete bigraph

$$F = (V_F, E_F, ctrl_F, prnt_F, link_F) : \langle k, X \rangle \rightarrow \langle m, Y \rangle$$

consists of a concrete place graph with sharing $F^P = (V_F, ctrl_F, prnt_F) : k \to m$ and a concrete link graph $F^L = (V_F, E_F, ctrl_F, link_F) : X \to Y$. We write the concrete bigraph with sharing as $F = (F^P, F^L)$.

We use ϵ as a shorthand for interface $\langle 0, \emptyset \rangle$. A bigraph with inner face ϵ is called ground.

Definition 8 (solid bigraph). A bigraph is solid if these conditions hold:

- 1. no regions or outer names are idle (a region is called idle if it is empty);
- 2. no two sites or inner names are siblings;
- 3. the parent of every site is a node;
- 4. no outer name is linked to an inner name.

Definition 9 (reaction rule). A reaction rule is a pair

$$R = (R: m \to J, R': m \to J),$$

sometimes written as $R \longrightarrow R'$, where R is the redex and R' the reactum, and R is solid. The rule generates all the ground reaction rules (r, r'), where $r = (R \otimes \operatorname{id}_Y) \circ d$ and $r' = (R' \otimes \operatorname{id}_Y) \circ d$ for some discrete ground parameter $d : \epsilon \to \langle m, Y \rangle$. The reaction relation $\longrightarrow_{\mathsf{R}}$ over ground bigraphs is defined by

$$g \longrightarrow_{\mathsf{R}} g' \text{ iff } g = D \circ r \text{ and } g' = D \circ r'$$

for some bigraph D and some ground reaction rule (r, r') generated from R.

Definition 10 (bigraphical reactive system (BRS)). A bigraphical reactive system consists of a pair $(\mathcal{B}, \mathcal{R})$, where \mathcal{B} is a set of ground bigraphs and \mathcal{R} is a set of reaction rules defined over \mathcal{B} . It has a reaction relation

which will be written \longrightarrow when \mathcal{R} is understood.

Given a set of reaction rules, we refer to the configurations that a system may adopt as *states*. *Rule priorities* [1] impose a partial ordering on the reaction rules, where a rule of lower priority can be applied only if no rule of higher priority is applicable. Priorities are implemented by assigning each reaction rule to a *priority class*, where all classes are strictly ordered, and reaction rules in the same class have the same priority.

Predicates can be used to generate labels for states (as in Definition 1) or to ease understanding and debugging of the system. They are expressed in a subset of a logic called BiLog [3], allowing any predicate to be encoded as a bigraph [2, 7]. Therefore, we can check each state for any satisfied predicates by solving a bigraph matching problem.

2 Extensions to BRSs

We propose two new BRSs: one adds probabilities to reaction rules, the other also groups the rules into actions and supports rewards based on both reaction rules and predicates. The main technicality in both the definitions and the implementation is that after generating a list of applicable ground reaction rules (for a particular action, if appropriate), the probabilities need to be normalised so that they add to 1, as a single reaction rule can sometimes be applied in multiple places, or some rules in an action may not be applicable.

2.1 Probabilistic BRS

Definition 11 (probabilistic reaction rule). A probabilistic reaction rule R is a triple (R, R', p), sometimes written $R \stackrel{p}{\longrightarrow} R'$, where (R, R') is a reaction rule and $p \in (0, 1]$ is a probability. Similarly to Definition 9, it generates a set of ground reaction rules of the form (r, r', p).

Definition 12 (probabilistic bigraphical reactive system (PBRS)). A probabilistic bigraphical reactive system consists of a pair $(\mathcal{B}, \mathcal{R})$, where \mathcal{B} is a set of ground bigraphs and \mathcal{R} is a set of probabilistic reaction rules defined over \mathcal{B} .

Let g, g' be ground bigraphs, and $\{(r_i, r'_i, p_i)\}_{i=1}^n$ a set of ground probabilistic reaction rules, where for each r_i , there exists a bigraph D_i such that $g = D_i \circ r_i$. Let $S = \{(r_i, r'_i, p_i) \mid g' = D_i \circ r'_i\}$ (for the same D_i), and

$$s = \sum_{i=1}^{n} p_i.$$

Then the reaction relation is defined as

$$g \stackrel{p}{\longrightarrow} \rhd_{\mathcal{R}} g' \text{ iff } S \neq \emptyset,$$

where

$$p = \frac{1}{s} \sum_{(r,r',p') \in S} p'.$$

2.2 Nondeterministic BRS

Definition 13 (nondeterministic reaction rule). Let A be a set of actions. A nondeterministic reaction rule R is a tuple (R, R', a, p), where (R, R', p) is a probabilistic reaction rule, and $a \in A$ is an action. We also define a reaction reward function $r: A \to \mathbb{R}$ that assigns a reward/cost to each action.

Definition 14 (nondeterministic bigraphical reactive system (NBRS)). A nondeterministic bigraphical reactive system consists of a pair $(\mathcal{B}, \mathcal{R})$, where \mathcal{B} is a set of ground bigraphs and \mathcal{R} is a set of nondeterministic reaction rules defined over \mathcal{B} .

Let g, g' be ground bigraphs, $a \in A$ an action, and $\{(r_i, r'_i, a, p_i)\}_{i=1}^n$ a set of ground nondeterministic reaction rules with action a, where for each r_i , there exists a bigraph D_i such that $g = D_i \circ r_i$. Let $S = \{(r_i, r'_i, a, p_i) \mid g' = D_i \circ r'_i\}$ (for the same D_i), and

$$s = \sum_{i=1}^{n} p_i.$$

Then the reaction relation for action a is defined as

$$g \xrightarrow[r(a)]{p} g' \text{ iff } S \neq \emptyset,$$

where

$$p = \frac{1}{s} \sum_{(r,r',a,p') \in S} p'.$$

Finally, we can associate a reward with each predicate, allowing us to assign rewards to states in a flexible and semantically meaningful way: the reward of a state is simply the sum of the rewards of all matching predicates (and 0 in case there are none).

3 Implementation

TODO: introduce BigraphER [9]

Example 1. Consider an MDP $(S, \overline{s}, A, P, L)$, where $S = \{s_0, s_1, s_2, s_3\}$, $\overline{s} = s_0$, $A = \{a, b, c\}$, and P, L defined as follows:

$$P(s_0, a) = [s_1 \mapsto 1],$$

$$P(s_1, b) = [s_0 \mapsto 0.7, s_1 \mapsto 0.3],$$

$$P(s_1, c) = [s_2 \mapsto 0.5, s_3 \mapsto 0.5],$$

$$P(s_2, a) = [s_2 \mapsto 1],$$

$$L(s_0) = \{initial\},$$

$$L(s_1) = \emptyset,$$

$$L(s_2) = \{heads\},$$

$$L(s_3) = \{tails\}.$$

Furthermore, equip it with a reward structure (ρ, ι) , where $\rho(s_2) = 3$, $\iota(s_1, b) = 1$, and both functions are zero everywhere else.

The MDP can be represented as an NBRS with BigraphER code in Listing 1.1. Furthermore, the BigraphER's visualisation of the (automatically generated) transition system is in Fig. 2.

More specifically, reaction rule probabilities are represented as floating-point numbers inside the arrows (e.g. -[0.7]->), each action encompasses its reaction rules with action actionName and end. Rewards can be added to actions simply by inserting an integer enclosed in squared brackets after the action name (e.g. action b[1]). Lastly, predicate rewards have the same format, but are listed on the preds line of the begin nbrs—end block (e.g. preds = {heads[3]};).

Visually, each state is represented by a white ellipse (the starting state is highlighted with a thicker border), and each action (per state) is a grey ellipse. Text inside state ellipses lists all predicates that are satisfied by that state. Edges from actions to states are labelled with both the probability and the name of the reaction rule. Transitions from either a fully generated transition system or a simulation, labels from predicates, and state/transition rewards can be exported to the various plain text formats of a probabilistic model checker PRISM [6] for further quantitative analysis.

TODO: introduce jupyter (with more/new screenshots)

4 A Case Study in Autonomous Agents

TODO: Akin to [4] and maybe [5].

4.1 Collecting Objects in a Grid

TODO: explain the syntax of the language

Our first scenario is concerned with collecting objects in a grid-like twodimensional world. The agent starts in the southwest corner of the map (which we call *home*), explores the environment until it collects a specified number

```
atomic ctrl S0 = 0;
atomic ctrl S1 = 0;
atomic ctrl S2 = 0;
atomic ctrl S3 = 0;
                                                        initial
big initial = S0;
big heads = S2;
big tails = S3;
action a
  react toTemp = S0 -[1.]-> S1;
  react loopH = S2 -[1.]-> S2;
                                                      toTemp
                                                            ♠.7, toInit
  react loopT = S3 -[1.]-> S3;
                                                  1
action b[1]
  react toInit = S1 -[0.7]-> S0
                                                      3, loop
  react loop = S1 -[0.3]-> S1;
end
action c
  react toH = S1 - [0.5] -> S2;
                                           0.5, toH
                                                 0.5, to⊤
  react toT = S1 -[0.5] -> S3;
end
                                     heads
                                                  tails
begin nbrs
                                       +3
  init initial;
  rules = [ {toTemp, toInit,
    loop, toH, toT,
                                          loopH
                                                    , loopT
    loopH, loopT} ];
  preds = { initial, heads[3],
    tails };
{\tt end}
```

Listing 1.1. BigraphER code.

Fig. 2. The full transition system.

Fig. 3. The BigraphER Jupyter interface with syntax highlighting.

of objects, and returns home (while possibly collecting more objects along the way). The model tracks which cells of the grid have been explored, and only the unexplored cells have a constant probability p of rewarding the agent with an object. The desired number of objects can be easily changed, and the map can be customised to make some transitions unavailable (i.e. add walls), as long as the agent can still return home.

```
ctrl Cell = 1;
ctrl Agent = 0;
ctrl Directions = 0;
ctrl Object = 0;

atomic ctrl North = 1;
atomic ctrl East = 1;
atomic ctrl West = 1;
atomic ctrl South = 1;
atomic ctrl Visited = 1;
atomic ctrl Unvisited = 1;
Listing 1.2. Controls.
```

We begin with a list of controls in Listing 1.2. On the right hand side of each equal sign is the number of ports that each node of that control has, while the **atomic** keyword specifies the nodes that cannot contain other nodes. Cells represents discrete locations. They are linked to either Visited or Unvisited (initially, all except one cells are unvisited). Cells always contain Directions, and one Cell contains an Agent. The Agent may contain an Object. In order to represent multiple objects, it is convenient to put the second Object inside the

first, and so on. Directions contains a subset of the four nodes, corresponding

to the four possible directions of travel: North, East, West, South. We connect neighbouring Cells by linking the West node of one Cell to the East node of another (or North to South).

```
big home = Cell{v}.(Directions.(North{v1})
                                | East{v2})
                   | Agent);
big goal = Agent.Object;
big initial = Visited{v}
           || Unvisited{u}
            # bottom left
            || Cell{v}.(Directions.(North{a} | East{b})
                      | Agent.1)
            # top left
            || Cell{u}.Directions.(East{c} | South{a})
            # bottom right
            || Cell{u}.Directions.(North{d} | West{b})
            # top right
            || Cell{u}.Directions.(West{c} | South{d});
              Listing 1.3. Predicates and the initial state.
```

Next, we define two predicates: home and goal (see Listing 1.3). The former states that the agent is home if it is in the southwest corner of the grid, while the latter specifies a goal of collecting one object. A larger number of objects can be set by nesting Objects inside each other, e.g., Agent.Object.Object. Additionally, Listing 1.3 defines the initial 2×2 grid, as described previously.

```
action north
 react goNorthToVisited =
      Visited{v}
    || Cell{v}.(Directions.(North{b} | id) | Agent)
    | | Cell{v}.Directions.(South{b} | id)
    -[1.0]->
       Visited{v}
    || Cell{v}.Directions.(North{b} | id)
    || Cell{v}.(Directions.(South{b} | id) | Agent)
      @ [0, 2, 1];
 react goNorthToUnvisited =
       Visited{v} || Unvisited{u}
    || Cell{v}.(Directions.(North{b} | id) | Agent)
    || Cell{u}.Directions.(South{b} | id)
    -[1.0-p]->
       Visited{v} || Unvisited{u}
    || Cell{v}.Directions.(North{b} | id)
    | Cell{v}.(Directions.(South{b} | id) | Agent)
      @ [0, 2, 1];
 react northObject =
```

Listing 1.4. The action of going north.

Then, we have an action for each of the four directions, each with three reaction rules: one for going to an already visited cell, one for visiting a cell for the first time, and one for visiting a cell for the first time and finding an object. We will use the rules for going north in Listing 1.4 as an example. When going to an already visited cell, the Agent simply traverses a link between the North node of its previous Cell and the corresponding South node. As this is the only reaction rule for this situation, its probability is 1. When going to an unvisited node, we collect an Object with probability p, or make the transition without getting an Object with probability 1-p. Regardless, the Cell changes its link from Unvisited to Visited.

```
react stayHome =
   Cell{v}.(Directions.(North{v1} | East{v2}) | goal)
   -[1.0]->
   Cell{v}.(Directions.(North{v1} | East{v2}) | goal);
```

Listing 1.5. Reaction rule to stay home if a required number of objects is acquired.

Lastly, we have three actions for when the agent acquires the specified number of objects: two for going home west or south (since these are the two directions necessary and sufficient to reach home), and one for staying home. The two actions for going home have two reaction rules each, depending on whether the destination cell is visited or not. They are just like the rules in Listing 1.4, except every mention of Agent is replaced with goal. The rule to stay home simply says that if the agent has achieved the goal, and is home (characterised by a Cell with only two directions: North and East), then nothing should change (see Listing 1.5). To make the agent start heading home as soon as enough objects are collected, all five reaction rules related to going and staying home are in a higher priority class.

4.2 Layers of Abstraction

TODO: Should mention [10].

Our second model explores two important concepts in modelling the topology of space: nesting and uncertainty. Nesting is the idea that spaces can be inside other spaces: rooms inside floors, floors inside buildings, buildings inside streets, etc. A key characteristic of nested spaces is that there is a separate action (or several) taking you from the outer space to the inner space or vice versa (e.g., using a door to enter or leave a room). In the rest of the paper, whenever we discuss one space inside another, we will refer to the inner space as the *room*.

The other concept explored in this section is uncertainty about the topology itself, i.e., is the door open or closed? The implementation of this idea is tied to that of rooms: the inside of a room is not defined until the outer cell encompassing the room (i.e., the area where the door is located) is entered. Then, the contents of the room are chosen from a finite set of possibilities with a set probability distribution. After that, the room stays fixed. This design decision supports the reality of missions with a reasonably short duration: if the agent finds a locked door, it should not keep revisiting the door hoping for it to suddenly be open.

As the ideas of collecting objects and tracking visited cells have been explored already, they are dropped from this model to simplify the implementation. The goal can simply be defined as moving from one location to another. We add one extra control, Node. While Cell can be imagined as a discrete patch of space (e.g., $1\,\mathrm{m}^2$), Node can be thought of as a room. While we stick to two layers (corridors and rooms) in this example, rooms can be freely nested: a room can contain any number of rooms, each of which can contain more rooms.

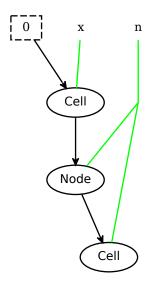


Fig. 4. The bigraphical structure behind nesting.

```
| Node{n}.1)
|| Cell{z}.Directions.West{b});
```

Listing 1.6. The initial map: three consecutive cells, the midmost of which has an inner room.

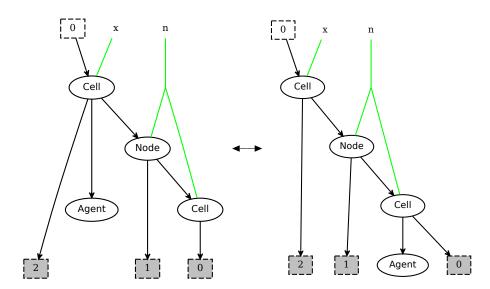


Fig. 5. Left to right: the reaction rule to go inside; right to left: the reaction rule to go outside.

In order to impement nesting, a Cell contains a Node, which can have any number of Cells of its own. Since after entering a room, you end up in a specific place inside the room (next to the door), we also need to link up the Node with one of the Cells it contains (see Fig. 4). Both Cells and Nodes support having a single link for exactly this purpose. Thus, most of the Cell ports are not linked to anything, which is implemented via closures (e.g. /x Cell $\{x\}$). The initial map in Listing 1.6 has a corridor of three cells, with the agent on the west-most cell, and a door to a room in the middle cell. One moves in and out of a room via a pair of actions with reaction rules pictured in Fig. 5.

```
action room
react openDoor =
   Cell{w}.(Agent | Node{n}.1 | id)
   -[p]-> /x /y /z /a /b /c /d
   Cell{w}.(Agent | Node{n}.(
        # top left
        Cell{n}.Directions.(East{a} | South{b})
        # top right
        | Cell{x}.Directions.(West{a} | South{c})
```

```
# bottom left
      | Cell{y}.Directions.(North{b} | East{d})
      # bottom right
      | Cell{z}.Directions.(North{c} | West{d}))
      | id);
  react closedDoor =
    Cell{w}.(Agent | Node{n}.1 | id)
    -[1.0-p] -> /x /y /z /b /c /d
    Cell{w}.(Agent | Node{n}.(
      # top left
        Cell{n}.Directions.(South{b})
      # top right
      | Cell{x}.Directions.(South{c})
      # bottom left
      | Cell{y}. Directions.(North{b} | East{d})
      # bottom right
      | Cell{z}.Directions.(North{c} | West{d}))
      | id);
end
```

Listing 1.7. Two ways to generate a room, leaving one door either open or closed.

Listing 1.8. The action of going north.

The room is a 2×2 grid of Cells, which is generated when the Agent gets to a Cell with an empty Node (represented by Node $\{n\}.1)^2$. In order for this to happen before any other action can take place, the two reaction rules in Listing 1.7 are put on a higher priority class. With probability p, the generated room has each Cell linked to its two neighbours; and with probability 1-p, the top two Cells are not linked, possibly forcing the agent to choose a different path. Finally, we have four actions, each with a single reaction rule with probability 1, corresponding to the four possible directions of travel. The reaction rules are very similar to those of our previous example, see Listing 1.8.

² Multiple different rooms can be implemented by having each previously empty Node contain a node of different control. This way a different probability distribution (with different possible outcomes) can be assigned to each room.

5 Conclusion

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