

Axiomatizations for probabilistic finite-state behaviors

Yuxin Deng^{1*} and Catuscia Palamidessi^{2**}

¹ INRIA Sophia-Antipolis and Université Paris 7

² INRIA Futurs and LIX, École Polytechnique

Abstract. We study a process calculus which combines both nondeterministic and probabilistic behavior in the style of Segala and Lynch's probabilistic automata. We consider various strong and weak behavioral equivalences, and we provide complete axiomatizations for finite state processes, restricted to guarded definitions in case of the weak equivalences. We conjecture that in the general case of unguarded recursion the “natural” weak equivalences are undecidable.

This is the first work, to our knowledge, that provides a complete axiomatization for weak equivalences in the presence of recursion and both nondeterministic and probabilistic choice.

1 Introduction

The last decade has witnessed increasing interest in the area of formal methods for the specification and analysis of probabilistic systems [16, 5, 3, 15, 19, 7]. In [20] van Glabbeek et al. classify probabilistic models into *reactive*, *generative* and *stratified*. In reactive models, each labeled transition is associated with a probability, and for each state the sum of the probabilities with the same label is 1. Generative models differ from reactive ones in that for each state the sum of the probabilities of all the outgoing transitions is 1. Stratified models have more structure and for each state either there is exactly one outgoing labeled transition or there are only unlabeled transitions and the sum of their probabilities is 1.

In [16] Segala pointed out that neither reactive nor generative nor stratified models capture real nondeterminism, an essential notion for modeling scheduling freedom, implementation freedom, the external environment and incomplete information. He then introduced a model, the *probabilistic automata* (PA), where both probability and nondeterminism are taken into account. Probabilistic choice is expressed by the notion of

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transition, which, in PA, leads to a probabilistic distribution over pairs (action, state) and deadlock. Nondeterministic choice, on the other hand, is expressed by the possibility of choosing different transitions. Segala proposed also a simplified version of PA called *simple probabilistic automata* (SPA), which are like ordinary automata except that a labeled transition leads to a probabilistic distribution over a set of states instead of a single state.

Figure 1 exemplifies the probabilistic models discussed above. In models where both probability and nondeterminism are present, like those of diagrams (4) and (5), a transition is usually represented as a bundle of arrows linked by a small arc. [17] provides a detailed comparison between the various models, and argues that PA subsume all other models above except for the stratified ones.

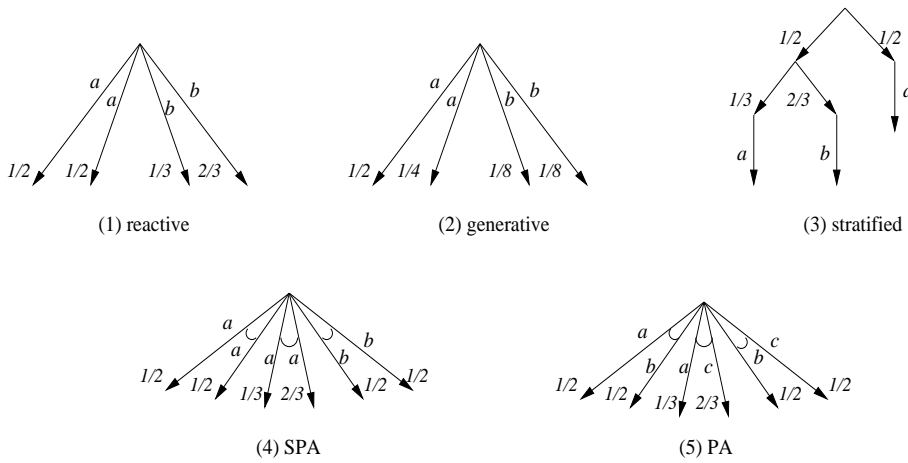


Fig. 1. Probabilistic models

In this paper we are interested in investigating axiom systems for a process calculus based on PA, in the sense that the operational semantics of each expression of the language is a probabilistic automaton³. Axiom systems are important both at the theoretical level, as they help gaining insight of the calculus and establishing its foundations, and at the practical level, as tools for systems specification and verification. Our calculus is

³ Except for the case of deadlock, which is treated slightly differently: following the tradition of process calculi, in our case deadlock is a state, while in PA it is one of the possible components of a transition.

basically a probabilistic version of the calculus used by Milner to express finite-state behaviors [9, 11].

We shall consider the two strong and the weak behavioral equivalences common in literature, plus one novel notion of weak equivalence having the advantage of being sensitive to divergency. For recursion-free expressions we provide complete axiomatizations of all the four equivalences. For the strong equivalences we also give complete axiomatizations for all expressions, while for the weak equivalences we achieve this result only for guarded expressions.

The reason why we are interested in studying a model which expresses both nondeterministic and probabilistic behavior, and an equivalence sensitive to divergency, is that one of the long-term goals of this line of research is to develop a theory which will allow us to reason about probabilistic algorithms used in distributed computing. In that domain it is important to ensure that an algorithm will work under any scheduler, and under other unknown or uncontrollable factors. The nondeterministic component of the calculus allows coping with these conditions in a uniform and elegant way. Furthermore, in many distributed computing applications it is important to ensure livelock-freedom (progress), and therefore we will need a semantics which does not simply ignore divergencies.

We are interested, in particular, in developing a fully distributed implementation of the (synchronous) π -calculus (π) using a probabilistic asynchronous π -calculus (π_{pa}) as an intermediate language. The reason why we need a probabilistic calculus is that it has been shown impossible to implement certain mechanisms of the pi-calculus without using randomization [12]. We need also the nondeterministic dimension for the usual reason: the implementation should be portable and in particular make no assumption about the scheduler. Some preliminary initial results of this project appeared in [13], but the part on implementation was very preliminary. We are now investigating a more realistic and efficient implementation.

We consider it important that an implementation does not introduce livelocks (or other kinds of unintended outcomes), hence the translation from π to π_{pa} should preserve livelock-freedom (see [14] for a discussion on the subject), and the semantics should be sensitive to divergency. For this reason, the second author choose (a probabilistic version of) testing semantics in [13]. However, it turned out that probabilistic testing semantics, at least the version invented in [13], was rather difficult to use. The correctness proofs were ad-hoc, by hand, and rather complicated.

For the realistic (and necessarily more sophisticated) implementation, we need proof methods feasible and (at least in part) automatic. For this reason, we are investigating here a divergency-sensitive *bisimulation-like* semantics. In the future, we plan to extend the achievements of this paper to π_{pa} .

2 Related work

In [9] and [11] Milner gave complete axiomatizations for strong bisimulation and observational equivalence, respectively, for a core *CCS* [10]. These two papers serve as our starting point: in several completeness proofs that involve recursion we adopt Milner’s *equational characterization theorem* and *unique solution theorem*. In Section 6.1 and Section 7.2 we extend [9] and [11] (for guarded expressions) respectively, to the setting of probabilistic process algebra.

In [18] Stark and Smolka gave a probabilistic version of the results of [9]. So, our paper extends [18] in that we consider also nondeterminism. Note that when nondeterministic choice is added, Stark and Smolka’s technique of proving soundness of axioms is no longer usable. (See the discussion at the beginning of Appendix A.) The same remark applies also to [1] which follows the approach of [18] but uses some axioms from iteration algebra to characterize recursion. In contrast, our probabilistic version of “bisimulation up to” technique works well when combined with the usual transition induction.

In [6] Bandini and Segala axiomatized both strong and weak behavioral equivalences for process calculi corresponding to SPA and to an alternated-model version of SPA. As their process algebra with non-alternating semantics corresponds to SPA, our results in Section 8 can be regarded as an extension of that work to PA.

For probabilistic process algebra of ACP-style, several complete axiom systems have appeared in the literature. However, in each of the systems either weak bisimulation is not investigated [4, 2] or nondeterministic choice is prohibited [4, 3].

Contribution of this work

The original contributions of this paper are:

- A complete axiomatization of a calculus which contains both non-deterministic and probabilistic choice, and recursion. We axiomatize

both strong and weak behavioral equivalences. It is the first time, as far as we know, that a complete axiomatization of weak behavioral equivalences is presented for a language of this kind.

- The development and the axiomatization of a (probabilistic) weak behavioral equivalence sensitive to livelock.

Plan of the paper

In the next section we briefly recall some basic concepts and definitions about probabilistic distributions. In Section 3 we introduce the calculus, with its syntax and operational semantics. In Section 4 we define the four behavioral equivalences we are interested in, and we extend the technique of “up-to” bisimulation to the probabilistic case. This technique is used extensively for the proofs of completeness, especially in the case of the weak equivalences. In Sections 5 and 6 we give complete axiomatizations for the strong equivalences and for the weak equivalences respectively, restricted to guarded expressions in the second case. Section 7 gives complete axiomatizations for the four equivalences in the case of the finite fragment of the language. The interest of this section is that we use different and much simpler proof techniques. Section 8 concludes and illustrates our research plans.

3 Preliminaries

Let S be a set. A function $\eta : S \mapsto [0, 1]$ is called a *discrete probability distribution*, or *distribution* for short, on S if the *support* of η , defined as $spt(\eta) = \{x \in S \mid \eta(x) > 0\}$, is finite or countably infinite and $\sum_{x \in S} \eta(x) = 1$. If η is a distribution with finite support and $V \subseteq spt(\eta)$ we use the set $\{(s_i : \eta(s_i))\}_{s_i \in V}$ to enumerate the probability associated with each element of V . To manipulate the set we introduce the operator \uplus defined as follows.

$$\{(s_i : p_i)\}_{i \in I} \uplus \{(s : p)\} = \begin{cases} \{(s_i : p_i)\}_{i \in I \setminus j} \cup \{s_j : (p_j + p)\} & \text{if } s = s_j \text{ for some } j \in I \\ \{(s_i : p_i)\}_{i \in I} \cup \{(s : p)\} & \text{otherwise.} \end{cases}$$

$$\{(s_i : p_i)\}_{i \in I} \uplus \{(t_j : p_j)\}_{j \in 1..n} = (\{(s_i : p_i)\}_{i \in I} \uplus \{(t_1 : p_1)\}) \uplus \{(t_j : p_j)\}_{j \in 2..n}$$

Given some distributions η_1, \dots, η_n on S and some real numbers $r_1, \dots, r_n \in [0, 1]$ with $\sum_{i \in 1..n} r_i = 1$, we define the *convex combination* $r_1\eta_1 + \dots + r_n\eta_n$ of η_1, \dots, η_n to be the distribution η such that $\eta(s) = \sum_{i \in 1..n} r_i\eta_i(s)$, for each $s \in S$.

4 Probabilistic process algebra

We use a countable set of variables, $Var = \{X, Y, \dots\}$, and a countable set of atomic actions, $Act = \{a, b, \dots\}$. Given a special action τ , we let u, v, \dots range over the set $Act_\tau = Act \cup \{\tau\}$, and let α, β, \dots range over the set $Var \cup Act_\tau$. The class of expressions \mathcal{E} is defined by the following syntax:

$$E, F ::= \bigoplus_{i \in 1..n} p_i u_i . E_i \mid \sum_{i \in 1..m} E_i \mid X \mid \mu_X E$$

Here $\bigoplus_{i \in 1..n} p_i u_i . E_i$ stands for a *probabilistic choice* operator, where the p_i 's represent positive probabilities, i.e., they satisfy $p_i \in (0, 1]$ and $\sum_{i \in 1..n} p_i = 1$. When $n = 0$ we abbreviate the probabilistic choice as $\mathbf{0}$; when $n = 1$ we abbreviate it as $u_1 . E_1$. Sometimes we are interested in certain branches of the probabilistic choice; in this case we write $\bigoplus_{i \in 1..n} p_i u_i . E_i$ as $p_1 u_1 . E_1 \oplus \dots \oplus p_n u_n . E_n$ or $(\bigoplus_{i \in 1..(n-1)} p_i u_i . E_i) \oplus p_n u_n . E_n$ where $\bigoplus_{i \in 1..(n-1)} p_i u_i . E_i$ abbreviates (with a slight abuse of notation) $p_1 u_1 . E_1 \oplus \dots \oplus p_{n-1} u_{n-1} . E_{n-1}$. The second construction $\sum_{i \in 1..m} E_i$ stands for *nondeterministic choice*, and occasionally we may write it as $E_1 + \dots + E_m$. The notation μ_X stands for a recursion which binds the variable X . We shall use $fv(E)$ for the set of free variables (i.e., not bound by any μ_X) in E . As usual we identify expressions which differ only by a change of bound variables. We shall write $E\{F_1, \dots, F_n / X_1, \dots, X_n\}$ or $E\{\tilde{F} / X\}$ for the result of simultaneously substituting F_i for each occurrence of X_i in E ($1 \leq i \leq n$), renaming bound variables if necessary.

Definition 1. *The variable X is weakly guarded (resp. guarded) in E if every free occurrence of X in E occurs within some subexpression $u.F$ (resp. $a.F$), otherwise X is weakly unguarded (resp. unguarded) in E .*

The operational semantics of an expression E is defined as a probabilistic automaton whose states are the expressions reachable from E and the transition relation is defined by the axioms and inference rules in Table 1, where $E \rightarrow \eta$ describes a transition that leaves from E and leads to a distribution η over $(Var \cup Act_\tau) \times \mathcal{E}$. We shall use $\vartheta(X)$ for the special distribution $\{(X, \mathbf{0} : 1)\}$. It is evident that $E \rightarrow \vartheta(X)$ iff X is weakly unguarded in E .

The behavior of each expression can be visualized by a transition graph. For instance, the expression $(\frac{1}{2}a \oplus \frac{1}{2}b) + (\frac{1}{3}a \oplus \frac{2}{3}c) + (\frac{1}{2}b \oplus \frac{1}{2}c)$ exhibits the behavior drawn in diagram (5) of Figure 1.

As in [6], we define the notion of *combined transition* as follows: $E \rightarrow_c \eta$ if there exists a collection $\{\eta_i, r_i\}_{i \in 1..n}$ of distributions and probabilities

var $X \rightarrow \vartheta(X)$	psum $\bigoplus_{i \in 1..n} p_i u_i . E_i \rightarrow \biguplus_{i \in 1..n} \{(u_i, E_i : p_i)\}$
rec $\frac{E\{\mu_X E/X\} \rightarrow \eta}{\mu_X E \rightarrow \eta}$	nsum $\frac{E_j \rightarrow \eta}{\sum_{i \in 1..m} E_i \rightarrow \eta}$ for some $j \in 1..m$

Table 1. Strong transitions

such that $\sum_{i \in 1..n} r_i = 1$, $\eta = r_1 \eta_1 + \dots + r_n \eta_n$ and $E \rightarrow \eta_i$, for each $i \in 1..n$.

We now introduce the notion of weak transitions. First we discuss the intuition behind it. Given an expression E , if we unfold its transition graph, we get a finitely branching tree. By cutting away all but one alternative in case of several nondeterministic candidates, we are left with a subtree with only probabilistic branches. A weak transition of E is a finite subtree of this kind, called *weak transition tree*, such that in any path from the root to a leaf there is at most one visible action. For example, let E be the expression $\mu_X(\frac{1}{2}a \oplus \frac{1}{2}\tau.X)$. It is represented by the transition graph displayed in Diagram (1) of Figure 2. After one unfolding, we get Diagram (2) which represents the weak transition $E \Rightarrow \eta$, where $\eta = \{(a, \mathbf{0} : \frac{3}{4}), (\tau, E : \frac{1}{4})\}$.

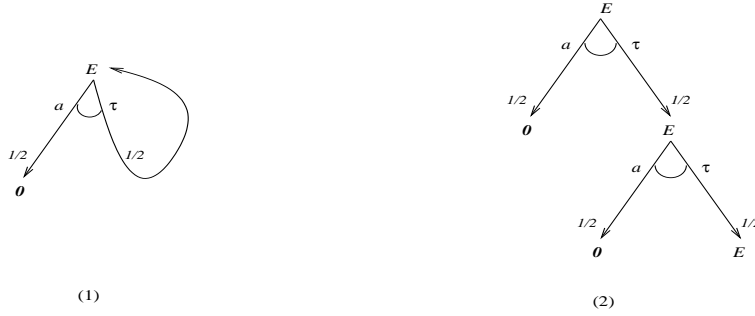


Fig. 2. A weak transition

Formally, weak transitions are defined by the rules in Table 2. Rule *weal* says that a weak transition tree starts from a bundle of labelled arrows derived from a strong transition. The meaning of Rule *wea2* is as follows. Given two expressions E, F and their weak transition trees $tr(E), tr(F)$, if F is a leaf of $tr(E)$ and there is no visible action in $tr(F)$,

then we can extend $tr(E)$ with $tr(F)$ at node F . If F_j is a leaf of $tr(F)$ then the probability of reaching F_j from E is pq_j , where p and q_j are the probabilities of reaching F from E , and F_j from F , respectively. Rule wea3 is similar to Rule wea2, with the difference that we can have visible actions in $tr(F)$, but not in the path from E to F . Rule wea4 allows to construct weak transitions to unguarded variables. Note that if $E \Rightarrow \vartheta(X)$ then X is unguarded in E .

wea1	$\frac{E \rightarrow \eta}{E \Rightarrow \eta}$
wea2	$\frac{E \Rightarrow \{(u_i, E_i : p_i)\}_i \uplus \{(u, F : p)\} \quad F \Rightarrow \{(\tau, F_j : q_j)\}_j}{E \Rightarrow \{(u_i, E_i : p_i)\}_i \uplus \{(u, F_j : pq_j)\}_j}$
wea3	$\frac{E \Rightarrow \{(u_i, E_i : p_i)\}_i \uplus \{(\tau, F : p)\} \quad F \Rightarrow \{(v_j, F_j : q_j)\}_j}{E \Rightarrow \{(u_i, E_i : p_i)\}_i \uplus \{(v_j, F_j : pq_j)\}_j}$
wea4	$\frac{E \Rightarrow \{(\tau, E_i : p_i)\}_i \quad \forall i : E_i \Rightarrow \vartheta(X)}{E \Rightarrow \vartheta(X)}$

Table 2. Weak transitions

For any expression E , we use $\delta(E)$ for the unique distribution $\{(\tau, E : 1)\}$, called the *virtual distribution* of E . For any expression E , we introduce a special weak transition, called *virtual transition*, denoted by $E \xrightarrow{\delta} \delta(E)$. We also define a *weak combined transition*: $E \xrightarrow{c} \eta$ if there exists a collection $\{\eta_i, r_i\}_{i \in 1..n}$ of distributions and probabilities such that $\sum_{i \in 1..n} r_i = 1$, $\eta = r_1\eta_1 + \dots + r_n\eta_n$ and for each $i \in 1..n$, either $E \Rightarrow \eta_i$ or $E \xrightarrow{\delta} \eta_i$. We write $E \Rightarrow_c \eta$ if every component is a “normal” (i.e., non-virtual) weak transition, namely, $E \Rightarrow \eta_i$ for all $i \leq n$.

5 Behavioral equivalences

In this section we define the behavioral equivalences that we mentioned in the Introduction, namely, strong bisimulation, strong probabilistic bisimulation, divergency-sensitive equivalence and observational equivalence. We also introduce a probabilistic version of “bisimulation up to” technique to show some interesting properties of the behavioral equivalences.

To define behavioral equivalences in probabilistic process algebra, it is customary to consider equivalence of distributions with respect to equivalence relations on processes.

5.1 Equivalence of distributions

If η is a distribution on $S \times T$, $s \in S$ and $V \subseteq T$, we write $\eta(s, V)$ for $\sum_{t \in V} \eta(s, t)$. We lift an equivalence relation on \mathcal{E} to a relation between distributions over $(Var \cup Act_\tau) \times \mathcal{E}$ in the following way.

Definition 2. *Given two distributions η_1 and η_2 over $(Var \cup Act_\tau) \times \mathcal{E}$, we say that they are equivalent w.r.t. an equivalence relation \mathcal{R} on \mathcal{E} , written $\eta_1 \equiv_{\mathcal{R}} \eta_2$, if*

$$\forall V \in \mathcal{E}/\mathcal{R}, \forall \alpha \in Var \cup Act_\tau : \eta_1(\alpha, V) = \eta_2(\alpha, V).$$

5.2 Behavioral equivalences

Strong bisimulation is defined by requiring equivalence of distributions at every step. Because of the way equivalence of distributions is defined, we need to restrict to bisimulations which are equivalence relations.

Definition 3. *An equivalence relation $\mathcal{R} \subseteq \mathcal{E} \times \mathcal{E}$ is a strong bisimulation if $E \mathcal{R} F$ implies:*

- whenever $E \rightarrow \eta_1$, there exists η_2 such that $F \rightarrow \eta_2$ and $\eta_1 \equiv_{\mathcal{R}} \eta_2$.

Two expressions E, F are strong bisimilar, written $E \sim F$, if there exists a strong bisimulation \mathcal{R} s.t. $E \mathcal{R} F$.

If we allow a strong transition to be matched by a strong combined transition, then we get a relation slightly weaker than strong bisimulation.

Definition 4. *An equivalence relation $\mathcal{R} \subseteq \mathcal{E} \times \mathcal{E}$ is a strong probabilistic bisimulation if $E \mathcal{R} F$ implies:*

- whenever $E \rightarrow \eta_1$, there exists η_2 such that $F \rightarrow_c \eta_2$ and $\eta_1 \equiv_{\mathcal{R}} \eta_2$.

We write $E \sim_c F$, if there exists a strong probabilistic bisimulation \mathcal{R} s.t. $E \mathcal{R} F$.

We now consider the case of the weak bisimulation. The definition of weak bisimulation for PA is not at all straightforward. In fact, the “natural” weak version of Definition 3 would be the following one.

Definition (Tentative). *An equivalence relation $\mathcal{R} \subseteq \mathcal{E} \times \mathcal{E}$ is a weak bisimulation if $E \mathcal{R} F$ implies:*

- whenever $E \rightarrow \eta_1$, then there exists η_2 such that either $F \Rightarrow \eta_2$ or $F \xrightarrow{\epsilon} \eta_2$, and $\eta_1 \equiv_{\mathcal{R}} \eta_2$.

E and F are weak bisimilar, written $E \asymp F$, whenever there exists a weak bisimulation \mathcal{R} s.t. $E \mathcal{R} F$.

Unfortunately the above definition is incorrect because it defines a relation which is not transitive. That is, there exist E , F and G with $E \asymp F$ and $F \asymp G$ but $E \not\asymp G$. For example, consider the following expressions and relations:

$$\begin{aligned}
E &\stackrel{\text{def}}{=} (\tfrac{1}{2}\tau.a \oplus \tfrac{1}{2}\tau.a) + (\tfrac{1}{2}\tau.a \oplus \tfrac{1}{2}a) \\
F &\stackrel{\text{def}}{=} \tfrac{1}{2}\tau.a \oplus \tfrac{1}{2}\tau.a \\
G &\stackrel{\text{def}}{=} \tau.a \\
\mathcal{R}_1 &\stackrel{\text{def}}{=} \{(E, F), (F, E), (E, E), (F, F), (a, a), (\mathbf{0}, \mathbf{0})\} \\
\mathcal{R}_2 &\stackrel{\text{def}}{=} \{(F, G), (G, F), (F, F), (G, G), (a, a), (\mathbf{0}, \mathbf{0})\}
\end{aligned}$$

It can be checked that \mathcal{R}_1 and \mathcal{R}_2 are weak bisimulations according to the tentative definition. However we have $E \not\asymp G$. To see this, consider the transition $E \rightarrow \eta$, where $\eta = \{(\tau, a : \frac{1}{2}), (a, \mathbf{0} : \frac{1}{2})\}$. There are only three possible weak transitions from G : $G \xrightarrow{\xi} \delta(G)$, $G \Rightarrow \eta_1$ and $G \Rightarrow \eta_2$ where $\eta_1 = \{(\tau, a : 1)\}$ and $\eta_2 = \{(a, \mathbf{0} : 1)\}$. Now, among the three distributions η_1, η_2 and $\delta(G)$, none is equivalent to η . Therefore, E and G are not bisimilar. Nevertheless, if we consider the weak combined transition: $G \Rightarrow_c \eta'$ where $\eta' = \frac{1}{2}\eta_1 + \frac{1}{2}\eta_2$, we observe that $\eta \equiv \eta'$.

The above example suggests that for a “good” definition of weak bisimulation it is necessary to use combined transitions. So we cannot give a weak variant of Definition 3, but only of Definition 4, called weak probabilistic bisimulation.

Definition 5. *An equivalence relation $\mathcal{R} \subseteq \mathcal{E} \times \mathcal{E}$ is a weak probabilistic bisimulation if $E \mathcal{R} F$ implies:*

- whenever $E \rightarrow \eta_1$, there exists η_2 such that $F \xrightarrow{\xi}_c \eta_2$ and $\eta_1 \equiv_{\mathcal{R}} \eta_2$.

We write $E \approx F$ whenever there exists a weak probabilistic bisimulation \mathcal{R} s.t. $E \mathcal{R} F$.

As usual, observational equivalence is defined in terms of weak probabilistic bisimulation.

Definition 6. *Two expressions E, F are observationally equivalent, written $E \simeq F$, if*

1. whenever $E \rightarrow \eta_1$, there exists η_2 such that $F \Rightarrow_c \eta_2$ and $\eta_1 \equiv_{\approx} \eta_2$.
2. whenever $F \rightarrow \eta_2$, there exists η_1 such that $E \Rightarrow_c \eta_1$ and $\eta_1 \equiv_{\approx} \eta_2$.

Often observational equivalence is criticised for being insensitive to divergency. We therefore introduce a variant which does not have this shortcoming.

Definition 7. *An equivalence relation $\mathcal{R} \subseteq \mathcal{E} \times \mathcal{E}$ is a divergency-sensitive equivalence if $E \mathcal{R} F$ implies:*

- whenever $E \rightarrow \eta_1$, there exists η_2 such that $F \Rightarrow_c \eta_2$ and $\eta_1 \equiv_{\mathcal{R}} \eta_2$.

We write $E \simeq F$ whenever there exists a divergency-sensitive equivalence \mathcal{R} s.t. $E \mathcal{R} F$.

It is easy to see that \simeq lies between \sim_c and \simeq . For example, we have that $\mu_X(\tau.X + a)$ and $\tau.a$ are related by \simeq but not by \simeq (this shows also that \simeq is sensitive to divergency), while $\tau.a$ and $\tau.a + a$ are related by \simeq but not by \sim_c .

One can check that all the relations defined above (except for \succ) are indeed equivalence relations and we have the inclusion ordering: $\sim \subsetneq \sim_c$
 $\subsetneq \simeq \subsetneq \simeq \approx$.

5.3 Probabilistic “bisimulation up to” technique

In the classical process algebra, the conventional approach to show $E \sim F$, for some expressions E, F , is to construct a binary relation \mathcal{R} which includes the pair (E, F) , and then to check that \mathcal{R} is a bisimulation. This approach can still be used in probabilistic process algebra, but things are more complicated because of the extra requirement that \mathcal{R} must be an equivalence relation. For example we cannot use some standard set-theoretic operators to construct \mathcal{R} , because, even if \mathcal{R}_1 and \mathcal{R}_2 are equivalences, $\mathcal{R}_1 \mathcal{R}_2$ and $\mathcal{R}_1 \cup \mathcal{R}_2$ may not be equivalences.

To avoid the restrictive condition and at the same time to reduce the size of the relation \mathcal{R} , we introduce the probabilistic version of “bisimulation up to” technique, whose usefulness will be exhibited in the next subsection.

In the following definitions, for a binary relation \mathcal{R} we denote the relation $(\mathcal{R} \cup \sim)^*$ by \mathcal{R}_{\sim} . Similar for other notations such as \mathcal{R}_{\approx} and \mathcal{R}_{\simeq} .

Definition 8. *A binary relation \mathcal{R} is a strong bisimulation up to \sim if $E \mathcal{R} F$ implies:*

1. whenever $E \rightarrow \eta_1$, there exists η_2 such that $F \rightarrow \eta_2$ and $\eta_1 \equiv_{\mathcal{R}_{\sim}} \eta_2$.
2. whenever $F \rightarrow \eta_2$, there exists η_1 such that $E \rightarrow \eta_1$ and $\eta_1 \equiv_{\mathcal{R}_{\sim}} \eta_2$.

A strong bisimulation up to \sim is not necessarily an equivalence relation. It is just an ordinary binary relation included in \sim , as shown by the next proposition.

Proposition 1. *If \mathcal{R} is a strong bisimulation up to \sim , then $\mathcal{R} \subseteq \sim$.*

One can also define a strong probabilistic bisimulation up to \sim_c relation and show that it is included in \sim_c . For weak probabilistic bisimulation, the “up to” relation can be defined as well, but we need to be careful.

Definition 9. *A binary relation \mathcal{R} is a weak probabilistic bisimulation up to \approx if $E \mathcal{R} F$ implies:*

1. *whenever $E \Rightarrow \eta_1$, there exists η_2 such that $F \xrightarrow{\xi}_c \eta_2$ and $\eta_1 \equiv_{\mathcal{R}\approx} \eta_2$.*
2. *whenever $F \Rightarrow \eta_2$, there exists η_1 such that $E \xrightarrow{\xi}_c \eta_1$ and $\eta_1 \equiv_{\mathcal{R}\approx} \eta_2$.*

In the above definition, we are not able to replace the first double arrow in each clause by a simple arrow. Otherwise, the resulting relation is not included in \approx .

Proposition 2. *If \mathcal{R} is a weak probabilistic bisimulation up to \approx , then $\mathcal{R} \subseteq \approx$.*

Definition 10. *A binary relation \mathcal{R} is an observational equivalence up to \simeq if $E \mathcal{R} F$ implies:*

1. *whenever $E \Rightarrow \eta_1$, there exists η_2 such that $F \Rightarrow_c \eta_2$ and $\eta_1 \equiv_{\mathcal{R}\simeq} \eta_2$.*
2. *whenever $F \Rightarrow \eta_2$, there exists η_1 such that $E \Rightarrow_c \eta_1$ and $\eta_1 \equiv_{\mathcal{R}\simeq} \eta_2$.*

As expected, observational equivalence up to \simeq is useful because of the following property.

Proposition 3. *If \mathcal{R} is an observational equivalence up to \simeq , then $\mathcal{R} \subseteq \simeq$.*

5.4 Some properties of behavioral equivalences

The “bisimulation up to” technique works well with Milner’s transition induction technique [10], and by combining them we obtain the following results.

Proposition 4 (Properties of \sim and \sim_c).

1. *\sim is a congruence relation.*
2. *$\mu_X E \sim E\{\mu_X E/X\}$.*
3. *$\mu_X(E + X) \sim \mu_X E$.*

4. If $E \sim F\{E/X\}$ and X weakly guarded in F , then $E \sim \mu_X F$.

Properties 1-4 are also valid for \sim_c .

Proposition 5 (Properties of \simeq and \simeq_c).

1. \simeq is a congruence relation.
2. If $\tau.E \simeq \tau.E + F$ and $\tau.F \simeq \tau.F + E$ then $\tau.E \simeq \tau.F$.
3. If $E \simeq F\{E/X\}$ and X is guarded in F then $E \simeq \mu_X F$.

Properties 1-3 hold for \simeq_c as well.

Each property above is shown by exhibiting an equivalence up to the corresponding bisimulation relation. For instance, in Clause 3 of Proposition 5 we prove that the relation $\mathcal{R} = \{(G\{E/X\}, G\{\mu_X F/X\}) \mid \text{for any } G \in \mathcal{E}\}$ is an observational equivalence up to \simeq by transition induction (see Appendix A for more details). We find it necessary to use the “bisimulation up to” technique particularly in the cases of Properties 1 and 3 of Proposition 5, since we are not able to directly construct an equivalence relation and prove that it is an observational equivalence. In all other cases the “up to” technique is optional.

6 Axiomatizations for all expressions

In this section we provide sound and complete inference systems for two strong behavioral equivalences: \sim and \sim_c . The class of expressions to be considered is \mathcal{E} .

6.1 Axiomatizing strong bisimulation

First we present the axiom system \mathcal{A}_r , which includes all axioms and rules displayed in Table 3. We assume the usual rules for equality (reflexivity, symmetry, transitivity and substitutivity), and the alpha-conversion of bound variables.

The notation $\mathcal{A}_r \vdash E = F$ (and $\mathcal{A}_r \vdash \tilde{E} = \tilde{F}$ for a finite sequence of equations) means that the equation $E = F$ is derivable by applying the axioms and rules from \mathcal{A}_r . The interest of \mathcal{A}_r is that it characterizes exactly strong bisimulation, as shown by the following theorem.

Theorem 1 (Soundness and completeness of \mathcal{A}_r). $E \sim E'$ iff $\mathcal{A}_r \vdash E = E'$.

S1 $E + \mathbf{0} = E$
S2 $E + E = E$
S3 $\sum_{i \in I} E_i = \sum_{i \in I} E_{\rho(i)}$ ρ is any permutation on I
S4 $\bigoplus_{i \in I} p_i u_i . E_i = \bigoplus_{i \in I} p_{\rho(i)} u_{\rho(i)} . E_{\rho(i)}$ ρ is any permutation on I
S5 $(\bigoplus_i p_i u_i . E_i) \oplus pu . E \oplus qu . E = (\bigoplus_i p_i u_i . E_i) \oplus (p + q)u . E$
R1 $\mu_X E = E\{\mu_X E/X\}$
R2 If $E = F\{E/X\}$, X weakly guarded in F , then $E = \mu_X F$
R3 $\mu_X(E + X) = \mu_X E$

Table 3. The axiom system \mathcal{A}_r

The soundness of \mathcal{A}_r is easy to prove: **R1-3** correspond to clauses 2-4 of Proposition 4; **S1-4** are obvious, and **S5** is a consequence of Definition 2. For the completeness we give a detailed proof in Appendix B. The basic points of the proof are: (1) if two expressions are bisimilar then we can construct an equation set in a certain format (standard format) that they both satisfy; (2) if two expressions satisfy the same standard equation set, then they can be proved equal by \mathcal{A}_r . This schema is inspired by [9, 18], but in our case the definition of standard format and the proof itself are more complicated due to the presence of both probabilistic and nondeterministic dimensions.

6.2 Axiomatizing strong probabilistic bisimulation

The difference between \sim and \sim_c is characterized by the following axiom:

$$\mathbf{C} \quad \sum_{i \in 1..n} \bigoplus_j p_{ij} u_{ij} . E_{ij} = \sum_{i \in 1..n} \bigoplus_j p_{ij} u_{ij} . E_{ij} + \bigoplus_{i \in 1..n} \bigoplus_j r_i p_{ij} u_{ij} . E_{ij}$$

where $\sum_{i \in 1..n} r_i = 1$. It is easy to show that the expressions on the left and right sides are strong probabilistic bisimilar. We denote $\mathcal{A}_r \cup \{\mathbf{C}\}$ by \mathcal{A}_{rc} .

Theorem 2 (Soundness and completeness of \mathcal{A}_{rc}). $E \sim_c E'$ iff $\mathcal{A}_{rc} \vdash E = E'$.

The soundness part follows immediately by the definition of \rightarrow_c . Concerning completeness, the idea is as follows. Given E, E' s.t. $E \sim_c E'$, we first construct two standard equation sets which are provably satisfied by E and E' respectively (this can always be done for any E, E'). Then we use axiom **C** to saturate the right hand side of each equation set so

as to transform them into expressions \widetilde{B} and \widetilde{B}' respectively, with the following property:

(*) For any $C_1, C_2 \in \widetilde{B} \cup \widetilde{B}'$ with $C_1 \sim_c C_2$, if $C_1 \rightarrow \eta_1$ then there exists some η_2 s.t. $C_2 \rightarrow \eta_2$ and $\eta_1 \equiv_{\sim_c} \eta_2$.

Thanks to this property, we can construct a single equation set (based on \widetilde{B} and \widetilde{B}'), which is provably satisfied by both E and E' . The rest of the proof is like the one for Theorem 1.

7 Axiomatizations for guarded expressions

Now we proceed with the axiomatizations of the two weak behavioral equivalences: \simeq and \simeq . We are not able to give a complete axiomatization for the whole set of expressions (and we conjecture that it is not possible, see Conclusion), so we restrict to the subset of \mathcal{E} consisting of *guarded expressions* only. An expression is guarded if for each of its subexpression of the form $\mu_X F$, the variable X is guarded in F (cf: Definition 1).

7.1 Axiomatizing divergency-sensitive equivalence

We first study the axiom system for \simeq . As a starting point, let us consider the system \mathcal{A}_{rc} . Clearly, **S1-5** are still valid for \simeq , as well as **R1**. **R3** turns out to be not needed in the restricted language we are considering. As for **R2**, we replace it with its (strongly) guarded version, which we shall denote as **R2'** (see Table 4). As in the standard process algebra, we need some τ -laws to abstract from invisible steps. For \simeq we use the probabilistic τ -laws **T1-3** shown in Table 4. Note that **T3** is the probabilistic extension of Milner's third τ -law ([11] page 231), and **T1** and **T2** together are equivalent, in the nonprobabilistic case, to Milner's second τ -law. However, Milner's first τ -law cannot be derived from **T1-3**, and it is actually unsound for \simeq . Below we let $\mathcal{A}_{gd} = \{\mathbf{R2}', \mathbf{T1-3}\} \cup \mathcal{A}_{rc} \setminus \{\mathbf{R2-3}\}$.

The rule **R2'** is shown to be sound in Proposition 5. The soundness of **T1-3**, and therefore of \mathcal{A}_{gd} , is evident. For the completeness proof, it is convenient to use the following saturation property, which relates operational semantics to term transformation, and which can be proved by transition induction, using the probabilistic τ -laws and the axiom **C**.

Lemma 1 (Saturation).

1. If $E \Rightarrow_c \eta$ with $\eta = \{(u_i, E_i : p_i)\}_i$, then $\mathcal{A}_{gd} \vdash E = E + \bigoplus_i p_i u_i . E_i$.
2. If $E \Rightarrow \vartheta(X)$ then $\mathcal{A}_{gd} \vdash E = E + X$.

R2'	If $E = F\{E/X\}$, X guarded in F , then $E = \mu_X F$
T1	$\bigoplus_i p_i \tau.(E_i + X) = X + \bigoplus_i p_i \tau.(E_i + X)$
T2	$(\bigoplus_i p_i u_i.E_i) \oplus p\tau.(F + \bigoplus_j q_j \beta_j.F_j) + (\bigoplus_i p_i u_i.E_i) \oplus (\bigoplus_j pq_j \beta_j.F_j)$ $= (\bigoplus_i p_i u_i.E_i) \oplus p\tau.(F + \bigoplus_j q_j \beta_j.F_j)$
T3	$(\bigoplus_i p_i u_i.E_i) \oplus pu.(F + \bigoplus_j q_j \tau.F_j) + (\bigoplus_i p_i u_i.E_i) \oplus (\bigoplus_j pq_j u.F_j)$ $= (\bigoplus_i p_i u_i.E_i) \oplus pu.(F + \bigoplus_j q_j \tau.F_j)$

Table 4. Some laws for the axiom system \mathcal{A}_{gd}

The completeness result can be proved in a similar way as Theorem 1. The main difference is that here the key role is played by equation sets which are not only in standard format, but also saturated. The transformation of a standard equation set into a saturated one is obtained by using Lemma 1.

Theorem 3 (Soundness and completeness of \mathcal{A}_{gd}). *If E and E' are guarded expressions then $E \simeq E'$ iff $\mathcal{A}_{gd} \vdash E = E'$.*

7.2 Axiomatizing observational equivalence

In this section we focus on the axiomatization of \simeq . In order to obtain completeness, we can follow the same schema as for Theorem 1, with the additional machinery required for dealing with observational equivalence, like in [11]. The crucial point of the proof is to show that, if $E \simeq F$, then we can construct an equation set in standard format which is satisfied by E and F . The construction of the equation is more complicated than in [11] because of the subtlety introduced by the probabilistic dimension (cf: Theorem 10 in Appendix C). Indeed, it turns out that the simple probabilistic extension of Milner's three τ -laws would not be sufficient, and we need an additional rule for the completeness proof to go through. We shall further comment on this rule at the end of Section 8).

The probabilistic extension of Milner's τ -laws are axioms **T1-4**, where **T1-3** are those introduced in previous section, and **T4**, defined in Table 5, takes the same form as Milner's first τ -law [11]. In the same table **T5** is the additional rule mentioned above. We let $\mathcal{A}_{go} = \mathcal{A}_{gd} \cup \{\mathbf{T4-5}\}$.

Rule **T5** is proved to be sound in Proposition 5. The soundness of **T4**, and therefore of \mathcal{A}_{go} , is straightforward. The completeness of \mathcal{A}_{go} is shown in Appendix C.

Theorem 4 (Soundness and completeness of \mathcal{A}_{go}). *If E and F are guarded expressions then $E \simeq F$ iff $\mathcal{A}_{go} \vdash E = F$.*

T4 $u.\tau.E = u.E$
T5 If $\tau.E = \tau.E + F$ and $\tau.F = \tau.F + E$ then $\tau.E = \tau.F$.

Table 5. Two τ -laws for the axiom system \mathcal{A}_{go}

8 Axiomatizations for finite expressions

In this section we consider the recursion-free fragment of \mathcal{E} , that is the class \mathcal{E}_f of all expressions which do not contain constructs of the form $\mu_X F$. In other words all expressions in \mathcal{E}_f have the form: $\sum_i \bigoplus_j p_{ij} u_{ij}.E_{ij} + \sum_k X_k$.

We define four axiom systems for the four behavioral equivalences studied in this paper. Basically $\mathcal{A}_s, \mathcal{A}_{sc}, \mathcal{A}_{fd}, \mathcal{A}_{fo}$ are obtained from $\mathcal{A}_r, \mathcal{A}_{rc}, \mathcal{A}_{gd}, \mathcal{A}_{go}$ respectively, by cutting away all those axioms and rules that involve recursions.

$$\begin{aligned} \mathcal{A}_s &\stackrel{\text{def}}{=} \{\mathbf{S1-5}\} & \mathcal{A}_{sc} &\stackrel{\text{def}}{=} \mathcal{A}_s \cup \{\mathbf{C}\} \\ \mathcal{A}_{fd} &\stackrel{\text{def}}{=} \mathcal{A}_{sc} \cup \{\mathbf{T1-3}\} & \mathcal{A}_{fo} &\stackrel{\text{def}}{=} \mathcal{A}_{fd} \cup \{\mathbf{T4-5}\} \end{aligned}$$

Theorem 5 (Soundness and completeness). *For any $E, F \in \mathcal{E}_f$,*

1. $E \sim F$ iff $\mathcal{A}_s \vdash E = F$;
2. $E \sim_c F$ iff $\mathcal{A}_{sc} \vdash E = F$;
3. $E \simeq F$ iff $\mathcal{A}_{fd} \vdash E = F$;
4. $E \simeq F$ iff $\mathcal{A}_{fo} \vdash E = F$.

The soundness part is obvious. The completeness can be shown by following the lines of previous sections. However, since there is no recursion here, we have a much simpler proof which does not use the equational characterization theorem and the unique solution theorem. Roughly speaking, all the clauses are proved by induction on the depth of the expressions. The completeness proof of \mathcal{A}_{fo} is a bit tricky. In the classical process algebra the proof can be carried out directly by using Hennessy Lemma [10], which says that if $E \approx F$ then either $\tau.E \simeq F$ or $E \simeq F$ or $E \simeq \tau.F$. In the probabilistic case, however, Hennessy's Lemma does not hold. For example, let

$$E \stackrel{\text{def}}{=} a \quad \text{and} \quad F \stackrel{\text{def}}{=} a + \left(\frac{1}{2}\tau.a \oplus \frac{1}{2}a\right).$$

We can check that: (1) $\tau.E \not\approx F$, (2) $E \not\approx F$, (3) $E \not\approx \tau.F$. In (1) the distribution $\{(\tau, E : 1)\}$ cannot be simulated by any distribution from

F . In (2) the distribution $\{(\tau, a : \frac{1}{2}), (a, \mathbf{0} : \frac{1}{2})\}$ cannot be simulated by any distribution from E . In (3) the distribution $\{(\tau, F : 1)\}$ cannot be simulated by any distribution from E .

Fortunately, to prove the completeness of \mathcal{A}_{fo} , it is sufficient to use the following weaker property.

Lemma 2 (Promotion). *For any $E, F \in \mathcal{E}_f$, if $E \approx F$ then $\mathcal{A}_{fo} \vdash \tau.E = \tau.F$.*

The promotion lemma is inspired by [8], where a similar result is proved for a language of mobile processes.

It is worth noticing that rule **T5** is necessary to prove Lemma 2. Consider the following two expressions: $\tau.a$ and $\tau.(a + (\frac{1}{2}\tau.a \oplus \frac{1}{2}a))$. It is easy to see that they are observational equivalent. However, we cannot prove their equality if rule **T5** is excluded from the inference system \mathcal{A}_{fo} . In fact, by using only the other rules and axioms it is impossible to transform $\tau.(a + (\frac{1}{2}\tau.a \oplus \frac{1}{2}a))$ into an expression without a probabilistic branch $p\tau.a$ occurring in any subexpression, for some p with $0 < p < 1$. So it is not provably equal to $\tau.a$, which has no probabilistic choice.

9 Concluding remarks

In this work we have proposed a probabilistic process calculus which corresponds to Segala's probabilistic automata. We have presented strong bisimulation, strong probabilistic bisimulation, divergency-sensitive equivalence and observational equivalence. Sound and complete inference systems for the four behavioral equivalences are summarized in Table 7.

Note that we have axiomatized divergency-sensitive equivalence and observational equivalence only for guarded expressions. For unguarded expressions whose transition graphs include τ -loops, we conjecture that the two behavioral equivalences are undecidable and therefore not finitely axiomatizable. Note that in [7] the authors give a decision algorithm for a weak probabilistic bisimulation in SPA, but our case is different because our weak probabilistic bisimulation is different, and also because we consider PA instead of SPA.

In the future it might be interesting to see how to refine our process algebra to allow for parallel composition. To do that it seems necessary to add some syntactic constraints, because parallel composition is hard to define for PA, as discussed in [16]. Another interesting research direction is to develop some automated verification tool by exploiting the axioms

and inference rules in Table 6, and then to do case-study for some practical examples in which probabilistic algorithms are shown to be quite useful. Our long term goal, as explained in the introduction, is to develop verification techniques for the asynchronous probabilistic π -calculus and to apply them to the verification of distributed algorithms.

S1	$E + \mathbf{0} = E$
S2	$E + E = E$
S3	$\sum_{i \in I} E_i = \sum_{i \in I} E_{\rho(i)}$ ρ is any permutation on I
S4	$\bigoplus_{i \in I} p_i u_i . E_i = \bigoplus_{i \in I} p_{\rho(i)} u_{\rho(i)} . E_{\rho(i)}$ ρ is any permutation on I
S5	$(\bigoplus_i p_i u_i . E_i) \oplus p u . E \oplus q u . E = (\bigoplus_i p_i u_i . E_i) \oplus (p + q) u . E$
C	$\sum_{i \in 1..n} \bigoplus_j p_{ij} u_{ij} . E_{ij} = \sum_{i \in 1..n} \bigoplus_j p_{ij} u_{ij} . E_{ij} + \bigoplus_{i \in 1..n} \boxplus_j r_i p_{ij} u_{ij} . E_{ij}$
T1	$\bigoplus_i p_i \tau . (E_i + X) = X + \bigoplus_i p_i \tau . (E_i + X)$
T2	$(\bigoplus_i p_i u_i . E_i) \oplus p \tau . (F + \bigoplus_j q_j \beta_j . F_j) + (\bigoplus_i p_i u_i . E_i) \oplus (\bigoplus_j p q_j \beta_j . F_j)$ $= (\bigoplus_i p_i u_i . E_i) \oplus p \tau . (F + \bigoplus_j q_j \beta_j . F_j)$
T3	$(\bigoplus_i p_i u_i . E_i) \oplus p u . (F + \bigoplus_j q_j \tau . F_j) + (\bigoplus_i p_i u_i . E_i) \oplus (\bigoplus_j p q_j u . F_j)$ $= (\bigoplus_i p_i u_i . E_i) \oplus p u . (F + \bigoplus_j q_j \tau . F_j)$
T4	$u . \tau . E = u . E$
T5	If $\tau . E = \tau . E + F$ and $\tau . F = \tau . F + E$ then $\tau . E = \tau . F$.
R1	$\mu_X E = E\{\mu_X E/X\}$
R2	If $E = F\{E/X\}$, X weakly guarded in F , then $E = \mu_X F$
R2'	If $E = F\{E/X\}$, X guarded in F , then $E = \mu_X F$
R3	$\mu_X (E + X) = \mu_X E$

In **C**, there is a side condition $\sum_{i \in 1..n} r_i = 1$.

Table 6. All the axioms and rules

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strong equivalences	finite processes	all expressions
\sim	\mathcal{A}_s : S1-5	\mathcal{A}_r : S1-5,R1-3
\sim_c	\mathcal{A}_{sc} : S1-5,C	\mathcal{A}_{rc} : S1-5,R1-3,C

weak equivalences	finite processes	guarded expressions
\simeq	\mathcal{A}_{fd} : S1-5,C,T1-3	\mathcal{A}_{gd} : S1-5,C,T1-3,R1,R2'
\approx	\mathcal{A}_{fo} : S1-5,C,T1-5	\mathcal{A}_{go} : S1-5,C,T1-5,R1,R2'

Table 7. All the inference systems

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Appendix

A Proofs from Section 5.4

In [18] Stark and Smolka use a special function f that associates a probability to a nonprobabilistic transition so as to form a probabilistic transition. For example, let $E \equiv \frac{1}{3}a \oplus \frac{2}{3}b$, then $f(E \xrightarrow{a} \mathbf{0}) = \frac{1}{3}$ and $f(E \xrightarrow{b} \mathbf{0}) = \frac{2}{3}$. The function f can be characterized as $f = \sup_{i \geq 0} f_i$ for some functions f_0, f_1, \dots that take nonprobabilistic transitions to probabilities and respect some ordering. Therefore in the soundness proofs of some axioms, to show that $f(E \xrightarrow{a} E') \leq p$, it suffices to prove by induction on i that $f_i(E \xrightarrow{a} E') \leq p$ for all $i \geq 0$. In the presence of nondeterministic choice, however, this technique becomes unusable because now the probability with which an expression performs an action and evolves into another expression is not deterministic any more. For example, let $E \equiv (\frac{1}{3}a \oplus \frac{2}{3}b) + (\frac{1}{2}a \oplus \frac{1}{2}c)$, then what is the value of $f(E \xrightarrow{a} \mathbf{0})$? Should it be $\frac{1}{3}$, $\frac{1}{2}$, or some value between them? Now the meaning of the function f is unclear because it depends on how the nondeterminism is resolved. Nevertheless, our “bisimulation up to” technique works well with Milner’s transition induction technique, as can be seen in the proof of Proposition 5(3) below.

Lemma 3. *If $\eta_1 \equiv_{\mathcal{R}_1} \eta_2$ and $\mathcal{R}_1 \subseteq \mathcal{R}_2$ then $\eta_1 \equiv_{\mathcal{R}_2} \eta_2$.*

Proof. Let $V \in \mathcal{E}/\mathcal{R}_2$. Since \mathcal{R}_1 is contained in \mathcal{R}_2 , we know that V is the disjoint union of V_1, \dots, V_n , for some $n \geq 1$ and $V_i \in \mathcal{E}/\mathcal{R}_1$ with $i \leq n$. It follows from $\eta_1 \equiv_{\mathcal{R}_1} \eta_2$ that

$$\forall i \leq n, \forall \alpha \in \text{Var} \cup \text{Act}_\tau : \eta_1(\alpha, V_i) = \eta_2(\alpha, V_i).$$

Therefore we have

$$\eta_1(\alpha, V) = \sum_{i \in 1..n} \eta_1(\alpha, V_i) = \sum_{i \in 1..n} \eta_2(\alpha, V_i) = \eta_2(\alpha, V).$$

□

Lemma 4. *Let $\eta = r_1\eta_1 + \dots + r_n\eta_n$ and $\eta' = r_1\eta'_1 + \dots + r_n\eta'_n$ with $\sum_{i \in 1..n} r_i = 1$. If $\eta_i \equiv_{\mathcal{R}} \eta'_i$ for each $i \leq n$, then $\eta \equiv_{\mathcal{R}} \eta'$.*

Proof. For any $V \in \mathcal{E}/\mathcal{R}$ and $\alpha \in \text{Var} \cup \text{Act}_\tau$, we have

$$\eta(\alpha, V) = \sum_{i \in 1..n} r_i \eta_i(\alpha, V) = \sum_{i \in 1..n} r_i \eta'_i(\alpha, V) = \eta'(\alpha, V).$$

Therefore $\eta \equiv_{\mathcal{R}} \eta'$ by definition. \square

Lemma 5. *Suppose $E \simeq F$. If $E \Rightarrow_c \eta$ then there exists η' s.t. $F \Rightarrow_c \eta'$ and $\eta \equiv_{\approx} \eta'$.*

Proof. By transition induction. \square

We use a measure $d_X(E)$ to count the depth of guardedness of the free variable X in expression E .

$$\begin{aligned} d_X(X) &= 0 \\ d_X(Y) &= 0 \\ d_X(a.E) &= d_X(E) + 1 \\ d_X(\tau.E) &= d_X(E) \\ d_X(\bigoplus_i p_i u_i.E_i) &= \min\{d_X(u_i.E_i)\}_i \\ d_X(\sum_i E_i) &= \min\{d_X(E_i)\}_i \\ d_X(\mu_Y E) &= d_X(E) \end{aligned}$$

If $d_X(E) > 0$ then X is guarded in E .

Lemma 6. *Let $d_X(G) = n$ and $\eta = \{(u_i, G_i : p_i)\}_{i \in I}$. Suppose $G\{E/X\} \Rightarrow \eta$. For all $i \in I$, it holds that*

1. *If $n > 0$ and $u_i = \tau$ then $G_i = G'_i\{E/X\}$ and $d_X(G'_i) \geq n$;*
2. *If $n > 1$ and $u_i \neq \tau$ then $G_i = G'_i\{E/X\}$ and $d_X(G'_i) \geq n - 1$.*

Proof. By induction on the depth of the inference of $G\{E/X\} \Rightarrow \eta$. There are three cases, depending on the last rule used in the inference. A typical case is for Rule wea3. In this case $\eta = \{(u_i, G_i : p_i)\}_{i \in I} \uplus \{(v_j, H_j : q_j)\}_{j \in J}$ and $G\{E/X\} \Rightarrow \eta$ is derived from the shorter inferences of $G\{E/X\} \Rightarrow \{(u_i, G_i : p_i)\}_{i \in I} \uplus \{(\tau, G_0 : p_0)\}$ and $G_0 \Rightarrow \{(v_j, H_j : q_j)\}_{j \in J}$. By induction hypothesis, for each $i \in I \cup \{0\}$, it holds that

1. *If $n > 0$ and $u_i = \tau$ then $G_i = G'_i\{E/X\}$ and $d_X(G'_i) \geq n$;*
2. *If $n > 1$ and $u_i \neq \tau$ then $G_i = G'_i\{E/X\}$ and $d_X(G'_i) \geq n - 1$.*

Particularly for G_0 we have $G_0 = G'_0\{E/X\}$ and $d_X(G'_0) \geq n > 0$. By induction hypothesis on the transition of $G'_0\{E/X\}$, it follows that for each $j \in J$

1. if $v_j = \tau$ then $H_j = H'_j\{E/X\}$ and $d_X(H'_j) \geq d_X(G'_0) \geq n$ for each $j \in J$;
2. $n > 1$ and $v_j \neq \tau$ then $H_j = H'_j\{E/X\}$ and $d_X(H'_j) \geq d_X(G'_0) - 1 \geq n - 1$. \square

Lemma 7. *Suppose $d_X(G) > 1$, $\eta = \{(u_i, G_i : p_i)\}_{i \in I}$ and $G\{E/X\} \Rightarrow \eta$. Then $G_i = G'_i\{E/X\}$ for each $i \in I$. Moreover, $G\{F/X\} \Rightarrow \eta'$ and $\eta \equiv_{\mathcal{R}^*} \eta'$, where $\eta' = \{(u_i, G'_i\{F/X\} : p_i)\}_{i \in I}$ and*

$$\mathcal{R} = \{(G\{E/X\}, G\{F/X\}) \mid \text{for any } G \in \mathcal{E}\}.$$

Proof. A direct consequence of Lemma 6. \square

Lemma 8. *Let $d_X(G) > 1$. If $G\{E/X\} \Rightarrow_c \eta$ then $G\{F/X\} \Rightarrow_c \eta'$ such that $\eta \equiv_{\mathcal{R}^*} \eta'$ where $\mathcal{R} = \{(G\{E/X\}, G\{F/X\}) \mid \text{for any } G \in \mathcal{E}\}$.*

Proof. Let $\eta = r_1\eta_1 + \dots + r_n\eta_n$ and $G\{E/X\} \Rightarrow \eta_i$ for each $i \leq n$. By Lemma 7, for each $i \leq n$, there exists η'_i s.t. $G\{F/X\} \Rightarrow \eta'_i$ and $\eta_i \equiv_{\mathcal{R}^*} \eta'_i$. Now let $\eta' = r_1\eta'_1 + \dots + r_n\eta'_n$, thus $G\{F/X\} \Rightarrow_c \eta'$. By lemma 4 it follows that $\eta \equiv_{\mathcal{R}^*} \eta'$. \square

Proof of Proposition 5(3). We show that the relation

$$\mathcal{R} = \{(G\{E/X\}, G\{\mu_X F/X\}) \mid \text{for any } G \in \mathcal{E}\}$$

is an observational equivalence up to \simeq . That is, we need to show the following assertions:

1. if $G\{E/X\} \Rightarrow \eta$ then there exists η' s.t. $G\{\mu_X F/X\} \Rightarrow_c \eta'$ and $\eta \equiv_{\mathcal{R}^*} \eta'$;
2. if $G\{\mu_X F/X\} \Rightarrow \eta'$ then there exists η s.t. $G\{E/X\} \Rightarrow_c \eta$ and $\eta \equiv_{\mathcal{R}^*} \eta'$;

We concentrate on the first clause as the second one is similar. The proof is carried out by induction on the depth of the inference of $G\{E/X\} \Rightarrow \eta$. There are several cases depending on the structure of G . As an example, here we consider the case that $G \equiv X$.

We write $G(E)$ for $G\{E/X\}$ and $G^2(E)$ for $G(G(E))$. Since $E \simeq F(E)$, we have $E \simeq F^2(E)$ since \simeq is a congruence relation by Proposition 5. If $E \Rightarrow \eta$ then by Lemma 5 there exists θ_1 s.t. $F^2(E) \Rightarrow_c \theta_1$ and $\eta \equiv_{\simeq} \theta_1$. Since X is guarded in F , i.e., $d_X(F) > 0$, then it follows that $d_X(F^2(X)) > 1$. By Lemma 8, there exists θ_2 s.t. $F^2(\mu_X F) \Rightarrow_c \theta_2$ and $\theta_1 \equiv_{\mathcal{R}^*} \theta_2$. From Proposition 4 we have $\mu_X F \sim F^2(\mu_X F)$, thus $\mu_X F \simeq F^2(\mu_X F)$. By Lemma 5 there exists η' s.t. $\mu_X F \Rightarrow_c \eta'$ and $\theta_2 \equiv_{\simeq} \eta'$. From Lemma 3 and the transitivity of $\equiv_{\mathcal{R}^*}$ it follows that $\eta \equiv_{\mathcal{R}^*} \eta'$. \square

B Proofs from Section 6

Definition 11. Let $\tilde{X} = \{X_1, \dots, X_m\}$ and $\tilde{W} = \{W_1, W_2, \dots\}$ be disjoint sets of variables. Let $\tilde{H} = \{H_1, \dots, H_m\}$ be expressions with free variables in $\tilde{X} \cup \tilde{W}$. In the equation set $S : \tilde{X} = \tilde{H}$, we call \tilde{X} formal variables and \tilde{W} free variables. We say S is standard if each H_i takes the form $\sum_j E_{f(i,j)} + \sum_l W_{h(i,l)}$ where $E_{f(i,j)} = \bigoplus_k P_{f(i,j,k)} u_{f(i,j,k)} \cdot X_{g(i,j,k)}$. We call S weakly guarded if there is no H_i s.t. $H_i \rightarrow \vartheta(X_i)$. We say that E provably satisfies S if there are expressions $\tilde{E} = \{E_1, \dots, E_m\}$, with $E_1 \equiv E$ and $fv(\tilde{E}) \subseteq \tilde{W}$, such that $\mathcal{A}_r \vdash \tilde{E} = \tilde{H}\{\tilde{E}/\tilde{X}\}$.

We first recall the theorem of unique solution of equations originally appeared in [9]. Adding probabilistic choice does not affect the validity of this theorem.

Theorem 6 (Unique solution of equations I). *If S is a weakly guarded equation set with free variables in \tilde{W} , then there is an expression E which provably satisfies S . Moreover, if F provably satisfies S and has free variables in \tilde{W} , then $\mathcal{A}_r \vdash E = F$.*

Proof. Exactly as in [9].

Below we give an extension of Milner's equational characterization theorem by accommodating probabilistic choice.

Theorem 7 (Equational characterization I). *For any expression E , with free variables in \tilde{W} , there exist some expressions $\tilde{E} = \{E_1, \dots, E_m\}$, with $E_1 \equiv E$ and $fv(\tilde{E}) \subseteq \tilde{W}$, satisfying m equations*

$$\mathcal{A}_r \vdash E_i = \sum_{j \in 1..n(i)} E_{f(i,j)} + \sum_{j \in 1..l(i)} W_{h(i,j)} \quad (i \leq m)$$

where $E_{f(i,j)} \equiv \bigoplus_{k \in [1..o(i,j)]} P_{f(i,j,k)} u_{f(i,j,k)} \cdot E_{g(i,j,k)}$.

Proof. By induction on the structure of E , similar to the proof in [9].

The following completeness proof is closely analogous to that of [18]. It is complicated somewhat by the presence of nondeterministic choice. For example, to construct the formal equations, we need to consider a more refined relation $L_{ij'j'}$ underneath the usual relation $K_{i'}$ while in [9, 18] it is sufficient to just use $K_{i'}$.

Proof of Theorem 1 (Completeness). Let E and E' have free variables in \widetilde{W} . By Theorem 7 there are provable equations such that $E \equiv E_1$, $E' \equiv E'_1$ and

$$\mathcal{A}_r \vdash E_i = \sum_{j \in 1..n(i)} E_{f(i,j)} + \sum_{j \in 1..l(i)} W_{h(i,j)} \quad (i \leq m)$$

$$\mathcal{A}_r \vdash E'_{i'} = \sum_{j' \in 1..n'(i')} E'_{f'(i',j')} + \sum_{j' \in 1..l'(i')} W_{h'(i',j')} \quad (i' \leq m')$$

with

$$E_{f(i,j)} \equiv \bigoplus_{k \in [1, \dots, o(i,j)]} p_{f(i,j,k)} u_{f(i,j,k)} \cdot E_{g(i,j,k)}$$

$$E'_{f'(i',j')} \equiv \bigoplus_{k' \in [1, \dots, o'(i',j')]} p'_{f'(i',j',k')} u'_{f'(i',j',k')} \cdot E'_{g'(i',j',k')}$$

Let $I = \{\langle i, i' \rangle \mid E_i \sim E'_{i'}\}$. By hypothesis we have $E_1 \sim E'_1$, so $\langle 1, 1 \rangle \in I$. Moreover, for each $\langle i, i' \rangle \in I$, the following holds, by the definition of strong bisimilarity:

1. There exists a total surjective relation $K_{ii'}$ between $\{1, \dots, n(i)\}$ and $\{1, \dots, n'(i')\}$, given by

$$K_{ii'} = \{\langle j, j' \rangle \mid \langle f(i, j), f'(i', j') \rangle \in I\}.$$

Furthermore, for each $\langle j, j' \rangle \in K_{ii'}$ there exists a total surjective relation $L_{ijj'j'}$ between $\{1, \dots, o(i, j)\}$ and $\{1, \dots, o'(i', j')\}$, given by

$$L_{ijj'j'} = \{\langle k, k' \rangle \mid u_{f(i,j,k)} = u'_{f'(i',j',k')} \text{ and } \langle g(i, j, k), g'(i', j', k') \rangle \in I\}.$$

2. $\vdash \sum_{j \in 1..l(i)} W_{h(i,j)} = \sum_{j' \in 1..l'(i')} W_{h'(i',j')}$.

Now, let $L_{ijj'j'}(k)$ denote the image of $k \in \{1, \dots, o(i, j)\}$ under $L_{ijj'j'}$ and $L_{ijj'j'}^{-1}(k')$ the preimage of $k' \in \{1, \dots, o'(i', j')\}$ under $L_{ijj'j'}$. We write $[k]_{ijj'j'}$ for the set $L_{ijj'j'}^{-1}(L_{ijj'j'}(k))$ and $[k']_{ijj'j'}$ for $L_{ijj'j'}(L_{ijj'j'}^{-1}(k'))$. It follows from the definitions that

1. If $\langle i, i'_1 \rangle \in I$, $\langle i, i'_2 \rangle \in I$, $\langle j, j'_1 \rangle \in K_{ii'_1}$ and $\langle j, j'_2 \rangle \in K_{ii'_2}$, then $[k]_{ijj'_1j'_1} = [k]_{ijj'_2j'_2}$.
2. If $q_1 \in [k]_{ijj'j'}$ and $q_2 \in [k]_{ijj'j'}$, then $u_{f(i,j,q_1)} = u_{f(i,j,q_2)}$ and $E_{g(i,j,q_1)} \sim E_{g(i,j,q_2)}$.

Define $\nu_{ijk} = \sum_{q \in [k]_{ij'j'}}$ $p_{f(i,j,q)}$ for any i', j' such that $\langle i, i' \rangle \in I$ and $\langle j, j' \rangle \in K_{ii'}$; define $\nu'_{i'j'k'} = \sum_{q' \in [k']_{i'j'j'}}$ $p'_{f'(i',j',q')}$ for any i, j such that $\langle i, i' \rangle \in I$ and $\langle j, j' \rangle \in K_{ii'}$. It is easy to see that whenever $\langle i, i' \rangle \in I$, $\langle j, j' \rangle \in K_{ii'}$ and $\langle k, k' \rangle \in L_{ij'j'}$ then $\nu_{ijk} = \nu'_{i'j'k'}$.

We now consider the formal equations, one for each $\langle i, i' \rangle \in I$:

$$X_{ii'} = \sum_{\langle j, j' \rangle \in K_{ii'}} H_{f(i,j),f'(i',j')} + \sum_{j \in 1..l(i)} W_{h(i,j)}$$

where

$$H_{f(i,j),f'(i',j')} \equiv \bigoplus_{\langle k, k' \rangle \in L_{ij'j'}} \left(\frac{p_{f(i,j,k)} p'_{f'(i',j',k')}}{\nu_{ijk}} \right) u_{f(i,j,k)} \cdot X_{g(i,j,k),g'(i',j',k')}.$$

These equations are provably satisfied when each $X_{ii'}$ is instantiated to E_i , since $K_{ii'}$ and $L_{ij'j'}$ are total and the right-hand side differs at most by repeated summands from that of the already proved equation for E_i . Note that each probabilistic branch $p_{f(i,j,k)} u_{f(i,j,k)} \cdot E_{g(i,j,k)}$ in E_i becomes the probabilistic summation of several branches like

$$\bigoplus_{q' \in [k']_{i'j'j'}} \left(\frac{p_{f(i,j,k)} p'_{f'(i',j',q')}}{\nu_{ijk}} \right) u_{f(i,j,k)} \cdot E_{g(i,j,k)}$$

in $H_{f(i,j),f'(i',j')} \{E_i / X_{ii'}\}_i$, where $\langle i, i' \rangle \in I$ and $\langle j, j' \rangle \in K_{ii'}$. But they are provably equal because

$$\begin{aligned} \sum_{q' \in [k']_{i'j'j'}} \left(\frac{p_{f(i,j,k)} p'_{f'(i',j',q')}}{\nu_{ijk}} \right) &= \frac{p_{f(i,j,k)}}{\nu_{ijk}} \cdot \sum_{q' \in [k']_{i'j'j'}} p'_{f'(i',j',q')} \\ &= \frac{p_{f(i,j,k)}}{\nu_{ijk}} \cdot \nu'_{i'j'k'} = p_{f(i,j,k)} \end{aligned}$$

and then the axiom **S5** can be used. Symmetrically, the equations are provably satisfied when each $X_{ii'}$ is instantiated to E'_i ; this depends on the surjectivity of $K_{ii'}$ and $J_{ij'j'}$.

Finally, we note that each $X_{ii'}$ is weakly guarded in the right-hand sides of the formal equations. It follows from Theorem 6 that $\vdash E_i = E'_i$ for each $\langle i, i' \rangle \in I$, and hence $\vdash E = E'$. \square

Proof of Theorem 2 (Completeness). Let E and E' have free variables in \widetilde{W} . By Theorem 7 there are provable equations such that $E \equiv E_1$, $E' \equiv E'_1$ and

$$\mathcal{A}_{rc} \vdash E_i = A_i \quad (i \leq m)$$

$$\mathcal{A}_{rc} \vdash E'_{i'} = A'_{i'} \quad (i' \leq m')$$

where $A_i \equiv \sum_{j \in 1..n(i)} E_{f(i,j)} + \sum_{j \in 1..l(i)} W_{h(i,j)}$ and

$$E_{f(i,j)} \equiv \bigoplus_{k \in [1, \dots, o(i,j)]} p_{f(i,j,k)} u_{f(i,j,k)} \cdot E_{g(i,j,k)}$$

Similar for the form of $A'_{i'}$.

Next we shall use axiom **C** to saturate the right hand side of each equation with some summands so as to transform each A_i (resp. $A'_{i'}$) into a provably equal expression B_i (resp. $B'_{i'}$) which satisfies the following property:

(*) For any $C_1, C_2 \in \widetilde{B} \cup \widetilde{B}'$ with $C_1 \sim_c C_2$, if $C_1 \rightarrow \eta_1$ then there exists some η_2 s.t. $C_2 \rightarrow \eta_2$ and $\eta_1 \equiv_{\sim_c} \eta_2$.

Initially we set $\widetilde{B} = \widetilde{A}$ and $\widetilde{B}' = \widetilde{A}'$. Let $S = \{(C_1, C_2) \mid C_1 \sim_c C_2 \text{ and } C_1, C_2 \in \widetilde{A} \cup \widetilde{A}'\}$. Clearly the set S is finite because there are finitely many expressions in $\widetilde{A} \cup \widetilde{A}'$. Without loss of generality, we take a pair (C_1, C_2) from S such that $C_1 \equiv A'_{i'} \in \widetilde{A}'$ and $C_2 \equiv A_i \in \widetilde{A}$ (we do similar manipulations for other three cases, namely (i) $C_1, C_2 \in \widetilde{A}$; (ii) $C_1, C_2 \in \widetilde{A}'$; (iii) $C_1 \in \widetilde{A}$ and $C_2 \in \widetilde{A}'$). If $A'_{i'} \rightarrow \eta'$ then for some η we have $A_i \rightarrow_c \eta$ and $\eta \equiv_{\sim_c} \eta'$, by the definition of bisimilarity. If $A_i \rightarrow \eta$ (obviously we are in this case if $\eta = \vartheta(X)$) we do nothing but go on to pick another pair from S to do the analysis. Otherwise η is a convex combination $\eta = r_1 \eta_1 + \dots + r_n \eta_n$ and $A_i \rightarrow \eta_j$ for each $j \leq n$. Hence each η_j must be in the form $\{(u_{f(i,j,k)}, E_{f(i,j,k)} : p_{f(i,j,k)})\}_k$ and $E_{f(i,j)}$ is a summand of A_i (so it is also a summand of B_i). By axiom **C** we have

$$\mathcal{A}_{rc} \vdash B_i = B_i + \bigoplus_{j \in 1..n} \bigoplus_k r_j p_{f(i,j,k)} u_{f(i,j,k)} \cdot E_{f(i,j,k)}.$$

Now we update B_i to be to the expression on the right hand side of last equation. To this point we have finished the analysis to the pair (C_1, C_2) . We need to pick a different pair from S to iterate the above procedure. When all the pairs in S are exhausted, we end up with \widetilde{B} and \widetilde{B}' which are easy to be verified to satisfy property (*). Observe that only axiom **C** is involved when updating B_i , so we have the following results:

$$\mathcal{A}_{rc} \vdash E_i = B_i \quad (i \leq m)$$

$$\mathcal{A}_{rc} \vdash E'_{i'} = B'_{i'} \quad (i' \leq m')$$

From now on, by using the above equations as our starting point, the subsequent arguments are nearly the same as those in the proof of Theorem 1, so we omit them. \square

C Proofs from Section 7

Given a standard equation set $S : \tilde{X} = \tilde{H}$, which has free variables \tilde{W} , we define the relations $\rightarrow_S \subseteq \tilde{X} \times \mathcal{P}((\text{Var} \cup \text{Act}_\tau) \times \tilde{X})$ (the notation $\mathcal{P}(V)$ represents all distributions on V) as $X_i \rightarrow_S \eta$ iff $H_i \rightarrow \eta$. From \rightarrow_S we can define the weak transition \Rightarrow_S in the same way as in Section 4. We write $X_i \rightsquigarrow_S X_k$ iff $X_i \Rightarrow_S \eta$, with $\eta = \{(u_j, X_j : p_j)\}_{j \in J}$, $k \in J$ and $u_k = \tau$. We shall call S *guarded* if there is no X_i s.t. $X_i \rightsquigarrow_S X_i$. We call S *saturated* if for all $X \in \tilde{X}$, $X \Rightarrow_S \eta$ implies $X \rightarrow_S \eta$. The variable W is *guarded* in S if it is not the case that $X_1 \rightarrow_S \vartheta(W)$ or $X_1 \rightsquigarrow_S \rightarrow_S \vartheta(W)$.

For guarded expressions, the equational characterisation theorem and the unique solution theorem given in last section can now be refined, as done in [11].

Theorem 8 (Equational characterisation II). *Every guarded expression E with free variables \tilde{W} provably satisfies a standard guarded equation set S with free variables in \tilde{W} . Moreover, if W is guarded in E then W is guarded in S .*

Proof. By induction on the structure of E . Consider the case that $E \equiv \bigoplus_{i \in I} p_i u_i . E_i$. For each $i \in I$, let X_i be the distinguished variable of the equation set S_i for E_i . We can define S as $\{X = \bigoplus_{i \in I} p_i u_i . X_i\} \cup \bigcup_{i \in I} S_i$, with the new variable X distinguished. All other cases are the same as in [11]. \square

Lemma 9. *Let E provably satisfies the standard guarded equation set S . Then there is a saturated, standard, and guarded equation set S' provably satisfied by E .*

Proof. By using Lemma 1, we show that if $X_i \Rightarrow \eta$ then $\mathcal{A}_{gd} \vdash E_i = E_i + \bigoplus_j p_j u_j . E_j$ when $\eta \equiv \{(u_j, X_j : p_j)\}_j$, and $\mathcal{A}_{gd} \vdash E_i = E_i + X$ when $\eta \equiv \vartheta(\tilde{X})$. Repeat this procedure for all weak transitions of E_i , at last we get $\mathcal{A}_{gd} \vdash E_i = H'_i \{\tilde{E}/\tilde{X}\}$. Hence we can take S' to be the equation set $\tilde{X} = \tilde{H}'$. \square

Theorem 9 (Unique solution of equations II). *If S is a guarded equation set with free variables in \tilde{W} , then there is an expression E which provably satisfies S . Moreover, if F provably satisfies S and has free variables in \tilde{W} , then $\mathcal{A}_{gd} \vdash E = F$.*

Proof. Nearly the same as the proof of Theorem 6, just replacing the recursion rule **R2** with **R2'**. \square

Proof of Theorem 3 (Completeness). By Theorem 8 there are provable equations such that $E \equiv E_1$, $E' \equiv E'_1$ and

$$\begin{aligned} \mathcal{A}_{gd} \vdash E_i &= A_i & (i \leq m) \\ \mathcal{A}_{gd} \vdash E'_{i'} &= A'_{i'} & (i' \leq m') \end{aligned}$$

For any $C \in \widetilde{A} \cup \widetilde{A}'$, we assume by Lemma 9 that C is saturated. Therefore it is easy to show that $C \Rightarrow_c \eta$ implies $C \rightarrow_c \eta$. Let $C' \in \widetilde{A} \cup \widetilde{A}'$. We note the interesting property that if $C \simeq C'$ and $C \rightarrow \eta$ then there exists η' s.t. $C' \rightarrow_c \eta'$ and $\eta \equiv_{\pm} \eta'$. Thanks to this property the remaining arguments are quite similar to that in Theorem 2, thus are omitted. \square

Lemma 10. 1. If $E \xrightarrow{\xi}_c \eta$ then $\tau.E \Rightarrow_c \eta$;
2. If $E \xrightarrow{\xi}_c \vartheta(X)$ then $E \Rightarrow \vartheta(X)$.

Proof. The first clause is easy to show. Let us consider the second one. If $\vartheta(X)$ is a convex combination of η_1, \dots, η_n and $E \Rightarrow \eta_i$ for all $i \in 1..n$, then each η_i must assign probability 1 to $(X, \mathbf{0})$, thus $\eta_i = \vartheta(X)$. \square

Lemma 11. If $E \xrightarrow{\xi}_c \eta$ with $\eta = \{(u_i, E_i : p_i)\}_i$ then $\mathcal{A}_{gd} \vdash \tau.E = \tau.E + \bigoplus_i p_i u_i . E_i$.

Proof. It follows from Lemma 10 and Lemma 1. \square

Theorem 10. Let E provably satisfy S and F provably satisfy T , where both S and T are standard, guarded equation sets, and let $E \simeq F$. Then there is a standard, guarded equation set U satisfied by both E and F .

Proof. Suppose that $\widetilde{X} = \{X_1, \dots, X_m\}$, $\widetilde{Y} = \{Y_1, \dots, Y_n\}$ and $\widetilde{W} = \{W_1, W_2, \dots\}$ are disjoint sets of variables. Let

$$\begin{aligned} S : \widetilde{X} &= \widetilde{H} \\ T : \widetilde{Y} &= \widetilde{J} \end{aligned}$$

with $fv(\widetilde{H}) \subseteq \widetilde{X} \cup \widetilde{W}$, $fv(\widetilde{J}) \subseteq \widetilde{Y} \cup \widetilde{W}$, and that there are expressions $\widetilde{E} = \{E_1, \dots, E_m\}$ and $\widetilde{F} = \{F_1, \dots, F_n\}$ with $E_1 \equiv E$, $F_1 \equiv F$, and $fv(\widetilde{E}) \cup fv(\widetilde{F}) \subseteq \widetilde{W}$, so that

$$\begin{aligned} \mathcal{A}_{go} \vdash \widetilde{E} &= \widetilde{H}\{\widetilde{E}/\widetilde{X}\} \\ \mathcal{A}_{go} \vdash \widetilde{F} &= \widetilde{J}\{\widetilde{F}/\widetilde{Y}\}. \end{aligned}$$

Consider the least equivalence relation $\mathcal{R} \subseteq (\widetilde{X} \cup \widetilde{Y}) \times (\widetilde{X} \cup \widetilde{Y})$ such that

1. whenever $(Z, Z') \in \mathcal{R}$ and $Z \rightarrow \eta$, then there exists η' s.t. $Z' \xrightarrow{c} \eta'$ and $\eta \equiv_{\mathcal{R}} \eta'$;
2. $(X_1, Y_1) \in \mathcal{R}$ and if $X_1 \rightarrow \eta$ then there exists η' s.t. $Y_1 \Rightarrow_c \eta'$ and $\eta \equiv_{\mathcal{R}} \eta'$.

Clearly \mathcal{R} is a weak probabilistic bisimulation on the transition system over $\tilde{X} \cup \tilde{Y}$, determined by $\rightarrow \stackrel{\text{def}}{=} \rightarrow_S \cup \rightarrow_T$. Now for two given distributions $\eta = \{(u_i, X_i : p_i)\}_{i \in I}$, $\eta' = \{(v_j, Y_j : q_j)\}_{j \in J}$, with $\eta \equiv_{\mathcal{R}} \eta'$, we introduce the following notations:

$$\begin{aligned} K_{\eta, \eta'} &= \{(i, j) \mid i \in I, j \in J, u_i = v_j \text{ and } (X_i, Y_j) \in \mathcal{R}\} \\ \nu_i &= \sum \{p_{i'} \mid i' \in I, u_{i'} = u_i, \text{ and } (X_i, X_{i'}) \in \mathcal{R}\} \quad \text{for } i \in I \\ \nu_j &= \sum \{p_{j'} \mid j' \in J, v_{j'} = v_j, \text{ and } (Y_j, Y_{j'}) \in \mathcal{R}\} \quad \text{for } j \in J \end{aligned}$$

Since $\eta \equiv_{\mathcal{R}} \eta'$ it follows by definition that if $(i, j) \in K_{\eta, \eta'}$, for some η, η' , then $\nu_i = \nu_j$. Thus we can define the expression

$$G_{\eta, \eta'} \stackrel{\text{def}}{=} \bigoplus_{(i, j) \in K_{\eta, \eta'}} \frac{p_i q_j}{\nu_i} u_i \cdot Z_{ij}$$

which will play the same role as the expression $H_{f(i, j), f'(i', j')}$ in the proof of Theorem 1. On the other hand, if $\eta = \eta' = \vartheta(X)$ we simply define the expression $G_{\eta, \eta'} \stackrel{\text{def}}{=} X$.

Based on the above \mathcal{R} we choose a new set of variables \tilde{Z} such that

$$\tilde{Z} = \{Z_{ij} \mid X_i \in \tilde{X}, Y_j \in \tilde{Y} \text{ and } (X_i, Y_j) \in \mathcal{R}\}.$$

Furthermore, for each $Z_{ij} \in \tilde{Z}$ we construct three auxiliary finite sets of expressions, denoted by A_{ij} , B_{ij} and C_{ij} , by the following procedure.

1. Initially the three sets are empty.
2. For each η with $X_i \rightarrow \eta$, arbitrarily choose one (and only one — the same principle applies in other cases too) η' (if it exists) satisfying $\eta \equiv_{\mathcal{R}} \eta'$ and $Y_j \Rightarrow_c \eta'$, construct the expression $G_{\eta, \eta'}$ and update A_{ij} to be $A_{ij} \cup \{G_{\eta, \eta'}\}$; Similarly for each η' with $Y_j \rightarrow \eta'$, arbitrarily choose one η (if it exists) satisfying $\eta \equiv_{\mathcal{R}} \eta'$ and $X_i \Rightarrow_c \eta$, construct $G_{\eta, \eta'}$ and update A_{ij} to be $A_{ij} \cup \{G_{\eta, \eta'}\}$.
3. For each η with $X_i \rightarrow \eta$, arbitrarily choose one η' (if it exists) satisfying $\eta \equiv_{\mathcal{R}} \eta'$, $Y_j \xrightarrow{c} \eta'$ but not $Y_j \Rightarrow_c \eta'$, construct the expression $G_{\eta, \eta'}$ and update B_{ij} to be $B_{ij} \cup \{G_{\eta, \eta'}\}$.
4. For each η' with $Y_j \rightarrow \eta'$, arbitrarily choose one η (if it exists) satisfying $\eta \equiv_{\mathcal{R}} \eta'$, $X_i \xrightarrow{c} \eta$ but not $X_i \Rightarrow_c \eta$, construct $G_{\eta, \eta'}$ and update C_{ij} to be $C_{ij} \cup \{G_{\eta, \eta'}\}$.

Clearly the three sets constructed in this way are finite. Now we build a new equation set

$$U : \tilde{Z} = \tilde{L}$$

where U_{11} is the distinguished variable and

$$L_{ij} = \begin{cases} \sum_{G \in A_{ij}} G & \text{if } B_{ij} \cup C_{ij} = \emptyset \\ \tau.(\sum_{G \in A_{ij} \cup B_{ij} \cup C_{ij}} G) & \text{otherwise.} \end{cases}$$

We assert that E provably satisfies the equation set U . To see this, we choose expressions

$$G_{ij} = \begin{cases} E_i & \text{if } B_{ij} \cup C_{ij} = \emptyset \\ \tau.E_i & \text{otherwise} \end{cases}$$

and verify that $\mathcal{A}_{g_o} \vdash G_{ij} = L_{ij}\{\tilde{G}/\tilde{Z}\}$.

In the case that $B_{ij} \cup C_{ij} = \emptyset$, all those summands of $L_{ij}\{\tilde{G}/\tilde{Z}\}$ which are not variables are of the forms:

$$\bigoplus_{(i,j) \in K_{\eta,\eta'}} \frac{p_i q_j}{\nu_i} u_i . E_i \quad \text{or} \quad \bigoplus_{(i,j) \in K_{\eta,\eta'}} \frac{p_i q_j}{\nu_i} u_i . \tau . E_i.$$

By **T4** we can transform the second form into the first one. Then by some arguments similar to those in Theorem 1, together with Lemma 1, we can show that

$$\mathcal{A}_{g_o} \vdash L_{ij}\{\tilde{G}/\tilde{Z}\} = H_i\{\tilde{E}/\tilde{X}\} = E_i.$$

On the other hand, if $B_{ij} \cup C_{ij} \neq \emptyset$, we let $C_{ij} = \{D_1, \dots, D_o\}$ ($C_{ij} = \emptyset$ is a special case of the following argument) and $D = \sum_{l \in 1..o} D_l\{\tilde{G}/\tilde{Z}\}$. As in last case we can show that

$$\mathcal{A}_{g_o} \vdash L_{ij}\{\tilde{G}/\tilde{Z}\} = \tau.(H_i\{\tilde{E}/\tilde{X}\} + D).$$

For any l with $1 \leq l \leq o$, let $D_l\{\tilde{G}/\tilde{Z}\} = \bigoplus_k p_k u_k . E_k$. It is easy to see that $E_i \xrightarrow{c} \eta$ with $\eta = \{(u_k, E_k : p_k)\}_k$. So by Lemma 11 it holds that

$$\mathcal{A}_{g_o} \vdash \tau.E_i = \tau.E_i + D_l\{\tilde{G}/\tilde{Z}\}.$$

As a result we can infer

$$\mathcal{A}_{g_o} \vdash \tau.E_i = \tau.E_i + D = \tau.E_i + (E_i + D).$$

by Lemma 1. Similarly,

$$\mathcal{A}_{g_o} \vdash \tau.(E_i + D) = \tau.(E_i + D) + E_i.$$

Consequently it follows from **T5** that

$$\mathcal{A}_{g_0} \vdash \tau.E_i = \tau.(E_i + D) = \tau.(H_i\{\tilde{E}/\tilde{X}\} + D) = L_{ij}\{\tilde{G}/\tilde{Z}\}.$$

In the same way we can show that F provably satisfies U . At last U is guarded because S and T are guarded. \square

To help understanding of the above theorem, we illustrate the construction of the equation set U by a simple example. Consider the equation sets S and T as follows.

$$\begin{array}{ll} S: & X_1 = a.X_2 \\ & X_2 = a.X_2 + \frac{1}{2}a.X_2 \oplus \frac{1}{2}\tau.X_1 \\ T: & Y_1 = \frac{1}{2}a.Y_2 \oplus \frac{1}{2}a.Y_3 \\ & Y_2 = a.Y_3 + \tau.Y_3 \\ & Y_3 = a.Y_2 \end{array}$$

Note that if E_1, E_2, E_3 provably satisfy S , and F_1, F_2 provably satisfy T , then $E_1 \simeq F_1 \simeq \mu_Z(a.Z)$.

Let \mathcal{R} be the equivalence relation that has a unique equivalence class $\{X_1, X_2, X_3, Y_1, Y_2\}$. It is easy to check that \mathcal{R} is a weak bisimulation on the transition system over $\tilde{X} \cup \tilde{Y}$. Now we take new variables $\{Z_{ij} \mid 1 \leq i \leq 2, 1 \leq j \leq 3\}$ and form the sets A_{ij}, B_{ij} and C_{ij} for each variable Z_{ij} , as displayed in Table 8, by using the procedure presented in the above proof. We construct the equation set U , based on all expressions shown

(i, j)	A_{ij}	B_{ij}	C_{ij}
(1, 1)	$\{\frac{1}{2}a.Z_{22} \oplus \frac{1}{2}a.Z_{23}\}$	\emptyset	\emptyset
(1, 2)	$\{a.Z_{23}\}$	\emptyset	$\{\tau.Z_{13}\}$
(1, 3)	$\{a.Z_{22}\}$	\emptyset	\emptyset
(2, 1)	$\{\frac{1}{2}a.Z_{22} \oplus \frac{1}{2}a.Z_{23}\}$	$\{\frac{1}{4}a.Z_{22} \oplus \frac{1}{4}a.Z_{23} \oplus \frac{1}{2}\tau.Z_{11}\}$	\emptyset
(2, 2)	$\{a.Z_{23}, \frac{1}{2}a.Z_{23} \oplus \frac{1}{2}\tau.Z_{13}\}$	\emptyset	$\{\tau.Z_{23}\}$
(2, 3)	$\{a.Z_{22}\}$	$\{\frac{1}{2}a.Z_{22} \oplus \frac{1}{2}\tau.Z_{13}\}$	\emptyset

Table 8. The construction of sets A_{ij}, B_{ij}, C_{ij}

in Table 8.

$$\begin{array}{l} U: \quad Z_{11} = \frac{1}{2}a.Z_{22} \oplus \frac{1}{2}a.Z_{23} \\ \quad Z_{12} = \tau.(a.Z_{23} + \tau.Z_{13}) \\ \quad Z_{13} = a.Z_{22} \\ \quad Z_{21} = \tau.(\frac{1}{2}a.Z_{22} \oplus \frac{1}{2}a.Z_{23} + \frac{1}{4}a.Z_{22} \oplus \frac{1}{4}a.Z_{23} \oplus \frac{1}{2}\tau.Z_{11}) \\ \quad Z_{22} = \tau.(a.Z_{23} + \frac{1}{2}a.Z_{23} \oplus \frac{1}{2}\tau.Z_{13} + \tau.Z_{23}) \\ \quad Z_{23} = \tau.(a.Z_{22} + \frac{1}{2}a.Z_{22} \oplus \frac{1}{2}\tau.Z_{13}) \end{array}$$

We can see that E_1 provably satisfies U by substituting $E_1, \tau.E_1, E_1, \tau.E_2, \tau.E_2, \tau.E_2$ for $Z_{11}, Z_{12}, Z_{13}, Z_{21}, Z_{22}, Z_{23}$; similarly F_1 provably satisfies U by substituting $F_1, \tau.F_2, F_3, \tau.F_1, \tau.F_2, \tau.F_3$ for these variables.

At last the completeness part of Theorem 4 follows from Theorem 8, 10 and 9.

D Proofs from Section 8

The depth of a process, $d(E)$, is defined as follows.

$$\begin{aligned} d(\mathbf{0}) &= 0 \\ d(X) &= 1 \\ d(\bigoplus_i p_i u_i . E_i) &= 1 + \max\{d(E_i)\}_i \\ d(\sum_i E_i) &= \max\{d(E_i)\}_i \end{aligned}$$

Proof of Lemma 2. By induction on $d = d(E) + d(F)$. We consider the nontrivial case that $d > 0$.

If X is a nondeterministic summand of E , then $E \rightarrow \vartheta(X)$. Since $E \simeq F$ it holds that $F \xrightarrow{c} \vartheta(X)$. By Lemma 10 we have $F \Rightarrow \vartheta(X)$. It follows from (the recursion-free version of) Lemma 1 that $\mathcal{A}_{fd} \vdash F = F + X$.

Let $\bigoplus_{i \in I} p_i u_i . E_i$ be any summand of E . Then we have $E \rightarrow \eta$, with $\eta = \{(u_i, E_i : p_i)\}_{i \in I}$. Since $E \approx F$, there exists η' , with $\eta' = \{(v_j, F_j : q_j)\}_{j \in J}$ s.t. $F \xrightarrow{c} \eta'$ and $\eta \equiv_{\approx} \eta'$. For any $k, l \in I$ with $u_k = u_l$ and $E_k \approx E_l$, it follows from **T4** and induction hypothesis that $\mathcal{A}_{fo} \vdash u_k . E_k = u_k . \tau . E_k = u_l . \tau . E_l = u_l . E_l$. By **S5** we can derive that $\mathcal{A}_{fo} \vdash \bigoplus_{i \in I} p_i u_i . E_i = \bigoplus_{i' \in I'} p'_{i'} u'_{i'} . E'_{i'}$, where the process on the right hand side is “compact”, i.e., for any $k', l' \in I'$, if $u'_{k'} = u'_{l'}$ and $E'_{k'} \approx E'_{l'}$, then $k' = l'$. Similarly we can derive $\mathcal{A}_{fo} \vdash \bigoplus_{j \in J} q_j v_j . F_j = \bigoplus_{j' \in J'} q'_{j'} v'_{j'} . F'_{j'}$ with the process on the right hand side “compact”. From $\eta \equiv_{\approx} \eta'$ and the soundness of \mathcal{A}_{fd} , it is easy to prove that $\mathcal{A}_{fo} \vdash \bigoplus_{i' \in I'} p'_{i'} u'_{i'} . E'_{i'} = \bigoplus_{j' \in J'} q'_{j'} v'_{j'} . F'_{j'}$ since each probabilistic branch of one process is provably equal to a unique branch of the other process. It follows that $\mathcal{A}_{fo} \vdash \bigoplus_{i \in I} p_i u_i . E_i = \bigoplus_{j \in J} q_j v_j . F_j$. By (a recursion-free version of) Lemma 11 we infer $\mathcal{A}_{fo} \vdash \tau . F = \tau . F + \bigoplus_{j \in J} q_j v_j . F_j = \tau . F + \bigoplus_{i \in I} p_i u_i . E_i$.

In summary $\mathcal{A}_{fo} \vdash \tau . F = \tau . F + E$. Symmetrically $\mathcal{A}_{fo} \vdash \tau . E = \tau . E + F$. Therefore $\mathcal{A}_{fo} \vdash \tau . E = \tau . F$ by **T5**. \square

Proof of Theorem 5 (4). The soundness part is easy. The completeness proof is similar to the proof of Lemma 2. Note that for any $k, l \in I$ with $u_k = u_l$ and $E_k \approx E_l$, we derive $\mathcal{A}_{fo} \vdash u_k . E_k = u_l . E_l$ by using **T4** and the promotion lemma instead of using induction hypothesis. \square