# **Know What You Stream: Generating Event Streams from CPN Models in ProM 6**

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**Abstract.** The field of process mining is concerned with supporting the analysis, improvement and understanding of business processes. In recent years the field has matured and several promising techniques have been proposed within different area's such as process discovery and conformance checking. However, some recent trends, originally stemming from the area of data mining, have only been little exposed in context of process mining. The incorporation of data stream mining techniques w.r.t. process mining has received little attention in particular. In this paper we present new developments that build on top of previous work related to the integration of data streams within the process mining framework ProM. We have developed means to use Coloured Petri Net (CPN) models as a basis for event stream generation. The newly introduced functionality greatly enhances the use of event streams in context of process mining as it allows us to be actively aware of the originating model of the event stream under analysis.

Keywords: Process mining, event streams, coloured Petri nets, ProM, CPN Tools

## 1 Introduction

We assume the reader to be familiar with the basics of process mining and refer to [1] for a detailed overview of the field.

Streams of events have only been a limited subject of study within the field of process mining. Given the current state of art in computer science we are able to store huge quantities of information related to business process execution. However, classical analysis techniques are not able to cope with these quantities of data, i.e. within the ProM Framework [2]<sup>1</sup> it is currently impossible to analyze an event log which size exceeds the computer's physical memory. Moreover, some business processes owners simply do not gain anything from static posteriori analysis. As an example, a chip manufacturer is interested in process deviation detection upon production of a batch of chips rather than after shipment of the batch to a customer. Thus, treating business process data as a dynamic sequence of events rather than a static event log is a natural next step.

In [4] we presented a standardized approach that extends the ProM framework with basic support for handling streaming data. The approach allows us to connect dynamic and volatile sources of data to the ProM framework. Although a preliminary implementation of a connection to an external source was provided, the framework has not

<sup>1</sup>http://www.promtools.org/

boosted the integration of stream based analysis within the process mining community. A potential cause of this could be explained by the lack of existing quality metrics for process models learned over streams. In this paper we present the integration of Coloured Petri nets [5], as implemented by CPN Tools [6]<sup>2</sup>, with the existing stream framework within ProM. The connection to CPN Tools provides great flexibility w.r.t. generating streams, and, allows us to generate streams of which the actual process model is known. This in turn can greatly help in further development of established quality metrics for stream based process mining.

The contents of this paper are organized as follows. In Section 2 we briefly touch upon the architecture and implementation of the newly developed integration. In Section 3 we demonstrate the use of the integration by means of an explanatory case study. Section 4 concludes the paper and provides pointers to interesting topics for future work. Additionally, we have recorded a screen-cast in which we discuss the integration in more detail, based on the case study <sup>3</sup>.

# 2 Architecture and Implementation

The core of the integration of CPN Tools with the ProM stream framework is an *author entity* [4] that generates events and emits these onto a designated stream. The underlying connection to CPN Tools, i.e. for the purpose of simulation of a CPN Model, is handled by the Access/CPN framework [7,8].

In order to generate events, the author entity needs a *CPN Model*, an *initial marking* of the CPN Model, a *Simulator Object* and a *Parameters Object*. The Parameters Object specifies certain parameters of the author. Using the Parameters Object we can specify the following properties:

- The total number of times that the model should be executed starting from the initial marking, denoted by  $r_{max}$ .
- The maximum number of steps within a single execution, denoted by  $s_{max}$ .
- The emission rate of the author by specifying a delay in between the emission of two consecutive packets, denoted by  $e_r$ .
- The case identification technique. This property specifies which transitions will be emitted on the stream upon firing and how the corresponding events are identified. Currently we have implemented two approaches being repetition based and CPN variable based.
- Event decoration. We can choose whether we want to emit all variables associated
  to the firing of a transition within a data packet or only the core elements, being the
  trace identifier and the event name.

In the repetition based case, each repetition of an execution of the CPN Model is used as a basis for identifying a case. Thus all transitions fired in the first repetition will have *I* as a case identifier, all transitions fired in the second repartition will

http://www.cpntools.org/

<sup>&</sup>lt;sup>3</sup>The screen-cast is published at https://svn.win.tue.nl/repos/prom/Packages/EventStream/Tags/publications/screen\_captures/2015\_bpm\_demo\_streams\_cpn/2015\_demo\_cpn\_tools\_screen\_capture.tar.gz



(a) CPN Model suitable for repetition based case(b) CPN Model suitable for CPN variable based identification technique. case identification technique.

Fig. 1: Two CPN Model fragments representing different examples of case identification technique.

have 2 as a case identifier etc. In this identification technique, every transition that is fired will be emitted as an event where the transition name acts as an event name. As an example of a CPN Model suitable for a repetition based case identification technique, consider Figure 1a. Within the CPN model we have defined two variables of type INT, i.e. var i, j: INT;. A possible stream originating from the CPN Model, where  $r_{max}, s_{max} \geq 2$ , including event decoration could be  $S_r = \langle \{ \text{trace=1}, \text{concept:name=t1}, \text{i=1} \}, \{ \text{trace=2}, \text{concept:name=t2}, \text{j=1} \}, \{ \text{trace=2}, \text{concept:name=t2}, \text{j=1} \}, \dots \rangle$ . Note that within the repetition based case, first all events related to trace I are emitted before events related to trace I are emitted, i.e. cases do not run concurrently.

In the CPN variable based approach, the user specifies a specific variable present within the CPN model to act as a case identifier. In this case, only those transitions that fire and that have the specified variable associated will be emitted to the event stream. Consider Figure 1b which depicts a CPN model suitable for CPN variable based case identification. Again we have defined two variables, i.e. var i, j: INT;. If we define variable i as the trace identification variable, given  $r_{max} \geq 1$ ,  $s_{max} \geq 3$ , a possible stream originating from the CPN Model could be  $S_v = \langle \text{trace=1}, \text{concept:name=t1}, \text{i=1} \rangle$ ,  $\text{trace=2}, \text{concept:name=t1}, \text{i=2} \rangle$ ,  $\text{trace=3}, \text{concept:name=t1}, \text{i=3}, \dots \rangle$ . Note that using CPN variable based case identification allows us to hide certain transitions present within the model from the stream.

All graphical components w.r.t. the author entity are inherited from the streaming framework presented in [4]. The only new graphical component of the Stream/CPN integration framework is the author configuration screen which provides of means to select  $r_{max}$ ,  $s_{max}$ ,  $e_r$ , the case identification technique and the event decoration. For an impression of the UI of the plug-in we refer to the screen-cast accompanying this paper.

### 3 Case Study

As an explanatory case we have designed a hierarchical CPN model that is used as a basis for stream generation. The model consists of one root model and two sub models. The root model is depicted in Figure 2<sup>4</sup>. The CPN model consists of two variables, i.e. var trace, ignore: INT. The initial marking of the root model is one token of colset INT, i.e. 1'1, in place source. The transition labeled start is connected to

<sup>&</sup>lt;sup>4</sup>The CPN model can be found at: https://svn.win.tue.nl/repos/prom/Packages/ EventStream/Tags/publications/data/2015\_bpm\_demo\_streams\_cpn/demo1.tar.gz

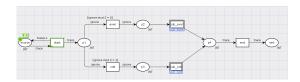
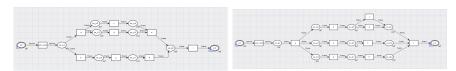


Fig. 2: Root CPN Model of the hierarchical model used within the case study



(a) CPN sub-model executed in case of a token(b) CPN sub-model executed in case of a towith an even INT value. ken with an odd INT value.

Fig. 3: Two CPN sub-models used within the case study

place source and acts as a token generator. In its binding it uses the trace variable. If transition source fires, it produces a token with the value of trace in place p1 and it produces a token with value trace + 1 in place source. All tokens with an even INT value will be routed to the sub-model named sub-even whereas all tokens with an odd INT value will be routed to the sub-model named sub-odd. In routing to the sub-models the variable ignore is used. The two sub-models are depicted in Figure 3.

After importing the hierarchical model in the ProM framework, we configure an event stream with the following parameters:  $r_{max}=1$ ,  $s_{max}=\infty$ ,  $e_r=50~ms$ ., case identification = *CPN Variable* with value trace and event decoration is true. After the event stream object is created we connect a stream based implementation of the Inductive Miner [9]. After receiving a total of 1905 events, the stream based Inductive Miner returns the Petri net depicted in Figure 4.

Although the stream based miner is not able to discovery hierarchy the resulting model aligns rather acceptable with the input model, i.e., from a control-flow perspective it exactly describes all possible traces that are emitted onto the event stream.

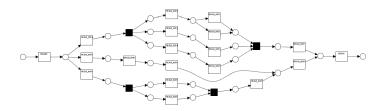


Fig. 4: Result of applying a stream based implementation of the Inductive Miner to the event stream generated by the hierarchical CPN model.

#### 4 Conclusion

The newly presented CPN extension of the stream framework within ProM enhances researchers, business users and developers to experiment with the concept of streaming data within a process mining context. The extension allows the user to import a CPN model, using any concept present within CPN tools, i.e. time, hierarchy etc., within ProM. The user is able to specify several parameters of the accompanying stream such as, emission rates, event decoration and trace identification technique.

An interesting direction for future work concerns support for the use of multiple case identification variables. This allows to discover multiple perspectives of the model under study. Another interesting direction is the development of a stream evaluation framework which allows us to manipulate certain elements of the stream, e.g. case arrival rates, throughput time, etc., in order to investigate the impact of these parameters w.r.t. the stream-based algorithm under study.

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