Analysis and Improvement of Business Process Models Using Spreadsheets

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Abstract

Software in general is thoroughly analyzed before it is released to its users. Business processes often are not – at least not as thoroughly as it could be – before they are released to their users, e.g., employees or software agents. This article ascribes this practice to the lack of suitable instruments for business process analysts, who design the processes, and aims to provide them with the necessary instruments to allow them to also analyze their processes. We use the spreadsheet paradigm to represent business process analysis tasks, such as writing metrics and assertions, running performance analysis and verification tasks, and reporting on the outcomes, and implement a spreadsheet-based tool for business process analysis. The results of two independent user studies demonstrate the viability of the approach.

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1. Introduction

The analysis of a piece of software, e.g., an algorithm or a mobile app, is a highly technical and daunting task typically performed by developers or testers who have the necessary technical background to know what to analyze and how. What is important is that the piece of software is analyzed by someone with the right skills, tools and methodologies.

Interestingly, when it comes to business processes (BPs) this is not common practice. In fact, the BP analysts, who design the processes to be executed, often do not have the necessary instruments to analyze their artifacts, i.e., the business process models.

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In the context of Business Process Management Systems (BPMSs), the tasks in the process models are typically implemented using web services (Weske, 2007). The web services can be either fully automated or it can provide a web application that allows human operators to perform the tasks through suitable user interfaces. For this type of business processes, which implementation requires involving developers, the analysis is, therefore, done again by the developers, if at all. This in turn means that the concerns of the actual owners of the artifacts, the BP analysts, may not be properly taken into account before implementing and running the production processes. Identifying issues at this late stage of the process lifecycle can be time-consuming and costly.

Let us consider, for example, the travel expense reimbursement process in Figure 1(a). Furthermore, let us assume that the process is currently in use in a service-based BPM system and that some problems has been identified by the BP analyst of the company. More concretely, he has noticed that with the current resources assigned to operate this process, only 70% of all the reimbursement requests are processed on time. The BP analyst would like to change the process in order to improve its performance without the need of having to increase the amount of resources assigned to the process. In addition to this, he has also noticed that the amount of many reimbursement requests are far below the operational costs of having to run the BP to process the request and that, in such cases, it may just be better to immediately reimburse the employee without having to run the whole process and incur in costs that are not justified by the requested amount.

Before investing the necessary effort for implementing and deploying changes in the process, the BP analyst needs to find answers to key questions, such as how many reimbursement requests, per quarter of the year, fall within the 30% of requests that are not processed on time, what should the value be for the amount requested under which the request is immediately reimbursed, and whether all these requests can be reimbursed without exceeding the maximum amount of 15K euros imposed by the accounting department. These are business questions that require the possibility to try different process execution scenarios that reproduce different execution outcomes. The BP analyst needs to be able to specify the typical behavior per quarter of the year, check whether the new fast track reimbursement will comply with the constraints above, analyze and visualize the results to propose a fine tuning of the process, and communicate with the software developer working on the implementation of the process.

These tasks can be done manually if the BP is simple and the number of issues to be analyzed are small. Otherwise, analyzing a BP can turn into a daunting task that requires automation, programming and IT skills. BP analysts usually do not have these skills, and, in practice, BPs are therefore mostly analyzed by software developers that, by nature, focus more on implementation than on business aspects.

If the BP analyst nonetheless wants to analyze a given process, he needs to communicate his analysis goals, requirements and configurations to a software developer, who is able to implement and run the analysis on behalf of the analyst. Once analysis results are ready, the developer needs to communicate
(a) The travel expense reimbursement process

(b) Process execution evidences stored in a log

Figure 1: BPMN model of a travel expense reimbursement process and a possible execution log.
them back to the BP analyst, who in turn may ask for a re-run of the analysis under new settings, and so on. Understanding well a process may thus require several iterations between the analyst and the developer. This is not optimal and suffers from the same difficulties already extensively reported in the literature on software engineering in general and requirements analysis in particular, such as ineffective communication channels, inexpressive notations, and its reliant nature (Bhat et al., 2006; Sutcliffe, 2012; Wiegers and Beatty, 2013). We propose therefore an approach that enables the BP analyst to analyze BP models on his own, with less reliance on and intervention of software developers.

We rely on spreadsheets to accomplish this. Created in 1979, spreadsheets are nowadays a common business tool and the most widely used end-user programming environment (Burnett et al., 2004). Scaffi and colleagues had estimated that only in the United States, by 2012, more than 55 million people would have been using spreadsheets at the workplace, mainly for business purposes (Scaffi et al., 2005). Considering the ubiquity of electronic spreadsheets in today’s business landscape, these tools represent an ideal environment to build powerful user-centric solutions, such as BP analysis instruments that target BP analysts.

This paper describes a spreadsheet-based approach for business process model analysis that:

- maps the problem of business process performance analysis and verification to the problem of configuring and analyzing data in common spreadsheets;
- enables the generation of analysis spreadsheets from an extended business process model editor for BPMN process models (Object Management Group (OMG), 2011);
- enables the BP analyst to define own metrics, assertions and analysis reports;
- automates the simulation of BP executions and generates process execution logs.

Two independent user studies demonstrate the viability of the approach, which was implemented in a prototype tool for spreadsheet-based BP model analysis, and a detailed qualitative analysis of the state of the art in BP model analysis highlights the benefits of the tool.

Before outlining the details of the approach (Section 3), next we formalize the context and problem statement of the work. In Sections 4–6, we then explain the design, execution and analysis of BP models. In Section 7, we report on how we implemented our prototype tool, which we assess in Section 8. We conclude the article with a discussion of related works and our final considerations on the results achieved.
2. Preliminaries and background

2.1. Business processes

As illustrated in Figure 1(a), in practice, BPs are typically expressed through process models. In this paper, we represent processes using BPMN (Object Management Group (OMG), 2011), a BP modeling notation widely used both in industry and academy. The core elements of BPMN related to control-flow specification include events, tasks, gateways and sequence flows (Dijkman et al., 2008) (see Figure 2). Events can be used to signal the start (start event) and end (end event) of a process. Tasks represent atomic activities to be performed as part of the process. Gateways are routing constructs that determine the execution flow of the process. They can be one of AND gateway (for creating concurrent execution flows), XOR gateway (to select one of a number of mutually exclusive flows), or OR gateway (to select any number of flows from the set of all outgoing flows). Sequence flows are used to represent the ordering relationship between any two elements (events, tasks and gateways) presented before. BPMN includes a richer set of elements, but in this paper we focus only on the ones presented here.

For notational convenience, we define a business process model as a tuple $BP = \langle pid, start, end, N, E, P \rangle$, where $pid$ is a unique identifier, $start$ and $end$ are the events that represent the start and end of a process, $N = T \cup G$ is the set of nodes of the process, with $T$ being the set of tasks and $G$ being the set of gateways of the process, $E \subseteq N \times N$ is the set of edges that connect pairs of nodes, and $P$ is the set of properties that store business data produced and consumed during the execution of $BP$. A a task $t \in T$, $t = \langle tid, tname \rangle$ is an activity of the process, where $tid$ is a unique identifier and $tname$ is the name of the task. A gateway $g \in G$, $g = \langle gid, C, gtype \rangle$ is a control flow node, where $gid$ is a unique identifier, $C$ is the set of conditions that controls the gateway, and $gtype \in \{AND, XOR, OR\}$ is the type of the gateway that
derives from $C$. Each condition $c \in C$ is a tuple $c = \langle cid, e, expr \rangle$, with $cid$ being a unique identifier, $e \in E$ being an outgoing edge of the gateway, and $expr$ being a Boolean expression over process properties in $P$ specifying the condition to follow the outgoing edge $e$. Process properties are of type $p = \langle nid, pname, pdesc, datatype, pvalue \rangle$, with $nid$ being the identifier of the node producing a value for $p$, $pname$ being the name, $pdesc$ being the description, $datatype$ being the data type, and $pvalue$ being the value of the property ($pvalue$ may be empty at the design time).

The BP model we introduce here is mapped to BPMN as follows. $start$ and $end$ in our model corresponds to the $start$ event and $end$ event in BPMN. Tasks $T$ and gateways $G$ map to the gateways and tasks in BPMN, respectively. Furthermore, the element $gtype \in \{AND, XOR, OR\}$ in a gateway $g \in G$ determines whether $g$ refers to an AND, XOR or OR gateway in BPMN. Edges $E$ in our model corresponds to the sequence flows in BPMN. Finally, the conditions $C$ and properties $P$ in our model are represented in BPMN through attributes associated to gateways and the process itself, respectively. It is worth noticing that the formalization and mapping introduced here are simple and straightforward, and that they are tailored to the notation needs of this paper. The interested reader can find a more detailed treatment of the formal semantics and analysis of BPMN process models (e.g., using Petri Nets) in (Dijkman et al., 2008, 2007).

The execution of a $BP$ is a business process instance (or process instance for short). A business process instance $bpi$ is a concrete case of operation of $BP$ and can be represented as a tuple $\langle pid, piid, startTs, endTs, PI, TD \rangle$, where $pid$ identifies the process model $BP$, $piid$ identifies the process instance, $startTs$ and $endTs$ are the start and end timestamps of the execution, $PI$ is the set of process properties produced by the execution of $BP$, and $TD$ is a set of task durations $td = \langle tid, d \rangle$, with $tid$ being the task identifier and $d$ being the execution time of the task.

Process instances are typically tracked in the form of a process execution log (van der Aalst, 2011) for later inspection and analysis. We define a process execution log as a set of process instances $L = \{bpi\}$. For example, Figure 1(b) shows a possible log of the travel expense reimbursement process with two process instances. This representation differs from the more common, event-based representation of process logs, in that it proposes an already aggregated view on execution events. As we will see later, this choice helps us to simplify the presentation of process execution data to the BP analyst.

2.2. Business process analysis

The term business process analysis has a broad meaning and includes many different types of analyses such as simulation, diagnosis, verification and performance analysis (Van Der Aalst et al., 2003). In this paper, we focus our analysis on the combination of the last two. More specifically, we take the dynamic perspective of verification and performance analysis, i.e., we run our analysis based on the execution of the business process models.
From this dynamic perspective, verifying a business process model means analyzing whether or not the behavior of its instances matches a given expected behavior. For example, in the scenario described in the introduction, the BP analyst may want to verify whether, under different execution conditions, the sum of the amounts for the reimbursement processed using the fast track option is kept under 15K euros in each quarter of the year. In order to perform this verification, the BP analyst needs to be able to (i) specify the expected behavior of the business process, (ii) provide the inputs for the process, run it, and track its observed behavior, and (iii) analyze the expected and observed behavior in order to verify if they match. We call the joint realization of these tasks business process verification and performance analysis. For conciseness, in the rest of the paper we refer to this simply as BP analysis and explicitly refer to verification and performance analysis when needed.

The expected behavior of the process is partly specified by the process model BP. However, the process model provides only static, structural information about the process; if instead the object of the verification are the dynamics and data produced by the execution of the process as we discuss here, additional constructs are necessary. This is where the performance analysis part comes into play. More concretely, we use metrics, i.e., measures that capture the performance of a process starting from process execution evidences. Formally, we can define a metric as a function \( m(L) = \text{val} \), where \( L \) is the process execution log and \( \text{val} \in \mathbb{R} \) is the metric’s value. Although this definition allows for the computation of cross-process metrics, i.e., of metrics computed over execution evidence from different process models, for simplicity in this article we limit our attention to single-process metrics only.

The availability of metrics further allows one to express expected behavior in terms of conditions over metric values. Such conditions can be expressed as predicates, which are Boolean statements over a metric \( m \). Formally, we can represent a predicate as a function \( \text{pred}(L, m) = \text{bool} \), where \( L \) is a process execution log, \( m \) is a metric and \( \text{bool} \in \{\text{true}, \text{false}\} \) holds the evaluation of the predicate. Predicates can be combined using standard logical operators, such as AND, OR and NOT, to build assertions. An assertion can be defined as a function \( a(L, P\text{red}) = \text{bool} \), where \( L \) is a process execution log, \( P\text{red} \) is a set of predicates and \( \text{bool} \) is as defined before. Assertions allow one to write arbitrarily complex combination of predicates to specify and check the expected behavior of a process.

In order to assess the behavior of a process, we need to run the process and record its observed behavior. For already implemented processes or services, this behavior can be extracted from the log \( L \) of the process. For processes or services that have not yet been implemented, we need to find the way of generating \( L \) by exercising the process model BP. We will get back to this issue in Section 2.3.

The analysis of BP now requires evaluating the assertions and metrics over the collected execution evidence \( L \) and visualizing the respective outcomes. Doing so requires setting up a suitable analysis report. We define a analysis report as a function \( r(L, M, A) = V \), where \( L \) is a process execution log, \( M \) is
a set of metrics, $A$ is a set of assertions and $V$ is a set of tables and charts that summarize the analysis outcomes. For example, $v \in V$ can be a pie chart that shows the percentages of true and false evaluations of an assertion $a \in A$ over the set of process instances in $L$. Such reports serve not only as a means to convey the outcomes to the analyst but also as a communication tool between the analyst and the software developer implementing the process.

2.3. Business process simulation

One way to obtain the event log $L$ when it is not possible or convenient to run the real process to generate it is to use business process simulation (van der Aalst et al., 2008a), which mimics the execution of process instances, given a business process model $BP$ and a suitable configuration. We propose to use this approach to obtain the process execution log $L$.

We define a BP simulator as a function $BPsim(BP, SS) = L$, where $BP$ is a process model, $SS$ represents the settings used to simulate the BP, and $L$ is the process execution log generated by the simulation. BP simulation thus enables the BP analyst to mimic different process execution scenarios and obtain corresponding execution evidences.

2.4. Problem statement

The problem we aim to solve in this article is devising an approach that enables the BP analyst to verify and analyze the performance of business processes without the need for software development skills. The first goal is to enable the BP analyst to write own metrics $M$ and assertions $A$, to obtain a process execution log $L$, and to design own analysis reports $r$, so as to be able to autonomously analyze the behavior of a business process $BP$. The second goal is to do so in a fashion that allows the BP analyst to easily discuss his findings with the software developer in charge of implementing processes. Our hypothesis is that mapping the BP analysis problem as defined in this article to the design of a spreadsheet calculation allows us to achieve both goals at the same time, in particular, given that spreadsheets are omnipresent in business and well-known by average BP analysts (Deloitte, 2009).

3. Spreadsheet-based business process analysis

We specifically consider the case of service-based BPs, where activities are executed by web services; human actors are hidden behind web service interfaces. Obtaining a log file for this type of BPs requires either invoking real web services (if such are available and do not have any persistent side effects) or simulating web service invocations (if the web services do have side effects or are not available at all). We assume that the BP analyst is capable of designing coarse BP models using the Business Process Modeling Notation (BPMN) and that he is familiar with spreadsheet tools like Microsoft Excel or Google Spreadsheets. We also assume that there is a software developer implementing the process and its web services, starting from the coarse BP models.
3.1. Requirements

Given these assumptions and the above problem statement, we identify a set of functional requirements. We group them into categories that correspond to the BP analysis phases (Section 2) they are related to:

**Specification of expected behavior:**

- **R1:** The solution shall enable the design of the BP model $BP$, along with its process properties $P$.
- **R2:** The solution shall enable the writing of metrics $M$ and assertions $A$ over the process execution log $L$.
- **R3:** The solution shall enable the storage of metrics and assertions for later reuse.

**Obtainment of observed behavior:**

- **R4:** The solution shall enable the configuration and simulation of $BP$ to obtain a corresponding log $L$.
- **R5:** The solution shall enable the use of existing implementations of web services used by $BP$ that do not produce unwanted side effects.
- **R6:** The solution shall enable the configuration and simulation of web services used by $BP$ that do have unwanted side effects or that do not exist yet.
- **R7:** The solution shall enable the storage of the generated log $L$.

**Analysis and reporting:**

- **R8:** The solution shall enable the creation of analysis reports $r$ based on $BP$, $M$, $A$ and $L$.
- **R9:** The solution shall enable the storage of reports for future reuse, e.g., for re-running the analysis under different conditions.
- **R10:** The solution shall enable the sharing of BP analysis configurations and results with other stakeholders (e.g., with software developers).

The implicit, non-functional requirement is that the solution’s tools that target the BP analyst shall not need any software development skills.
The overall approach proposed in this paper takes into account the fact that there are tasks that require specific technical skills that BP analysts may not have and that may prevent them from being able to analyze BPs. For example, the detailed design of executable process models in BPMN and the configuration of the more technical aspects is usually out of the reach of typical BP analysts. The design of executable BP models may in fact require the developer to use a larger set of modeling constructs than introduced in Section 2.1 (e.g., events or messages); the analyst only needs to master the subset of constructs introduced in Section 2.1 to be able to run his analysis. We therefore propose to separate the tasks related to (i) the design and configuration of executable process models, and (ii) the analysis of BPs. Task (i) is assigned to software developers, while task (ii) to BP analysts (see Figure 3). This separation of duties is not only practical and realistic, but it also leverage on the skills and interests of each role.

In order to approach the requirements discussed in Section 3.1, we leverage on BPMN and the spreadsheet paradigm to provide an approach for the analysis of BPs. BPMN is used to model executable BPs. The spreadsheet is used as interface toward the BP analyst and as communication instrument between the analyst and the developer. Simulation is used to safely generate behavioral information for those web services of the service-based BP that may have persistent side effects in the system or do not yet have a readily usable implementation. Figure 3 illustrates our approach.

Starting from a draft of BP (step 1), the software developer refines and implements the process (2) using an extended BPMN editor \( R_1 \). This produces BP, including the set of process properties \( P \) that can be used for analysis.
Given these ingredients, the BPMN editor generates a so-called configuration spreadsheet (3), which contains the process properties and a set of simulation parameters. Simulation parameters are used to configure the dynamics of the simulation (R4); they include parameters such as the number of process instances to be simulated, the rate at which instances are to be generated, and the execution time of simulated web services (R6). Process properties $P$ are used to configure business data for different process execution scenarios; they are associated to the nodes of $BP$ and may refer to both real or simulated web services (R5, R6). Activities of $BP$ that refer to real services are marked as such and pre-configured by the developer in the extended BPMN editor; the BP analyst can configure the behavior only of simulated services. He does so simply by editing the spreadsheet and defining values for the simulation parameters and process properties (4). Once the simulation is configured, the BP simulator reads the configuration and $BP$ and runs the simulation, mimicking the web service behaviors defined by the BP analyst and invoking existing web services. As a result, it generates (5) a process execution log $L$ that contains the observed behavior, which is again stored as a process execution spreadsheet (R7). The actual verification and performance analysis, is again done by the BP analyst using an analysis spreadsheet (6). In this spreadsheet, the analyst defines the metrics $M$ and assertions $A$ over the generated log $L$ as standard spreadsheet functions (R2). The spreadsheet automatically performs the necessary calculations, and allows the BP analyst to define charts or tables for the visualization of results (V), turning the analysis spreadsheet into a report $r$ that can easily be saved (R3, R8, R9) and shared with the developer (R10) (7).

In the following, we detail how processes are modeled, how the simulation is performed, and how predicates, assertions and analysis reports are calculated.

4. Business process modeling and simulation configuration

Setting up a BP analysis requires a suitable design of $BP$ and the configuration of the simulation to be performed. The definition of the process properties $P$ by the developer and their configuration by the BP analyst play an important role in setting up the BP analysis. The relevance of process properties its twofold: First, gateway conditions that determine the control flow of a process are defined over process properties. That is, they determine the runtime scenarios of process instances. Second, they are the starting point for the design of metrics and, hence, for the actual design of the verification and performance analysis. The simulation parameters allow the BP analyst to define the time behavior of simulated tasks and the number of task instances to be generated.

The developer models BPs using BPMN (Object Management Group (OMG), 2011), which is a standard process modeling notation targeting both BP analysts and software developers. The BPMN constructs presented in Section 2.1 allow him to associate process properties to tasks. For example, Figure 4(a) shows the definition of three process properties for the task Fill travel expense reimbursement form, namely, ExpenseAmount, Duration and EmployID. Each property requires a name, description and datatype, separated by commas,
matching the model $p = (\text{nid}, \text{pname}, \text{pdesc}, \text{datatype}, \text{pvalue})$ introduced before. The value for \text{nid} is automatically derived by selecting a task in the process editor; \text{pvalues} are defined by the BP analyst using a configuration spreadsheet \text{CS}. In the same vein, by using standard BPMN constructs the developer can define conditional rules to control the execution of the process. Conditions are defined over process properties. Figure 5(a) shows the conditions that regulate the execution flow over the outgoing arcs of the highlighted decision point. Each condition is set by the developer, who defines a Boolean expression, e.g., ExpenseAmount $\leq$ ExpenseThreshold, over each gateway’s outgoing arc.

A spreadsheet is a bi-dimensional array $s$ where each element $s(i,j)$, with $i$ representing the column index and $j$ the row index, represents a cell. A cell $s(i,j)$ can contain one of (i) a value that consists of an alfa-numeric datum, such as in $s(i,j) \leftarrow \text{“AJ487”}$, (ii) a reference to a different cell, such as in $s(i,j) \leftarrow s(p,k)$, or (iii) a formula that combines functions, values and references to other cells such as in $s(i,j) \leftarrow \text{sum}(13,s(p,k))$.

The configuration spreadsheet \text{CS} imports the process properties $P$ of BP for their configuration. To do so, we can follow the simple rule of mapping each property $p_j \in P$ to one spreadsheet row, like in $\text{CS}(1,j) \leftarrow p_j$.\text{pname}, $\text{CS}(2,j) \leftarrow p_j$.\text{pdesc} and $\text{CS}(3,j) \leftarrow p_j$.\text{datatype}. Figure 4(b) shows how the properties are represented in the spreadsheet. The spreadsheet’s cells are indexed using letters for columns and numbers for rows. For example, the process property ExpenseAmount associated to the BP model is mapped to row 8 in the spreadsheet using the mapping \text{CS}(A,8) \leftarrow “ExpenseAmount”,
CS(B,8) ← "Amount spent in the trip", and CS(C,8) ← "Number". The last cell (CS(D,8)) is used to set the values of the property. The rest of the properties are mapped following the same mapping logic.

To define values for properties, the BP analyst can use a constant value, as in the cell CS(D,10), or he can choose to write a formula that generates values for the cell. For example, the cell CS(D,8) uses the function RANDBETWEEN(800, 1400) to compute random values between 800 to 1400. By assigning non-constant values to the properties involved in the conditional expressions of gateways, the BP analyst can model the execution of alternative paths. As Figure 5(b) illustrates, the random function RANDBETWEEN, used to define the values of the property ExpenseAmount, is what models the conditional execution of either the Yes or No-labeled outgoing arcs of the gateway highlighted in Figure 5(a). The values obtained from these formulas are computed for each process instance at BP simulation time.

Simulation parameters are configured similarly to process properties. The parameters we consider are three: (i) the number of process instances to be simulated, (ii) the arrival rate for process instances, and (iii) the task duration for each task \( t \in T \) in the process model \( BP \). These properties do not need to be explicitly defined by the developer or BP analyst. They are added automatically to the spreadsheet at its generation time. Figure 5(b) shows that rows 6 and 7 are filled with the parameters NbrProcessInstances and ArrivalRate, respectively. From row 12 on, rows are filled with parameters that define the duration of the tasks. Values for these parameters are defined like values for
process properties.

Basic nondeterministic human-behaviors, such as task completion time, can be modeled directly by the BP analyst on the spreadsheet. He can set, for example, a fixed or a normally distributed task completion time. Modern spreadsheet software offers a vast collection of built-in mathematical and statistical functions that enable BP analysts to model the dynamics of human-based tasks by approximating it via mathematical or statistical formulas. Predefined function can be employed also to define process properties, conditional statements and simulation parameters of almost any complexity. If more complex human interventions are required the software developer needs to implement additional pieces of software, e.g., a text recognition algorithm.

The results of the business process modeling and analysis design are a business process model \(BP\) and a configuration spreadsheet \(CS\) that are consumed by the BP simulator.

Our approach supports only the BPMN constructs introduced in Section 2.1. Tasks of type service, exclusive, parallel and inclusive gateways and, sequence flows are supported, while events of types different from start and end are not yet considered in this work.

5. Business process simulation

The BP simulator is in charge of simulating the execution of \(BP\) based on the configurations provided in \(CS\), thereby producing a process execution log. Refining our definition of Section 2, the BP simulator can be seen as \(BP_{sim}(BP, CS) = ES\), where \(BP\) is the process model, \(CS\) is the configuration spreadsheet and \(ES\) is the resulting process execution spreadsheet (the log in spreadsheet format).

The process execution spreadsheet \(ES\) is a spreadsheet that holds process execution data that results from the simulation of \(BP\). Each tuple in \(ES\) represents a business process instance \(bpi\), and each cell within the tuple stores the runtime values of the elements of \(bpi\) as defined in Section 2.1. The idea of using this representation, as opposed to an event-based representation, is to keep the querying of business process instances simple and intuitive for the BP analyst and to avoid the need for writing complex event aggregation logics to reconstruct process instances. Thus, to store a business process instance \(bpi_{j}\) in \(ES\), where the associated \(BP\) has a number of \(k\) properties and \(l\) tasks, we map the elements of \(bpi_{j}\) to \(ES\) as:

\[
ES(1,j) \leftarrow bpi_{j}.piid, \\
ES(2,j) \leftarrow bpi_{j}.P[1].value, ..., ES(k + 1,j) \leftarrow bpi_{j}.P[k].value, ES(k + 2,j) \leftarrow bpi_{j}.TD[1].dur, ..., ES(k + l + 2,j) \leftarrow bpi_{j}.TD[l].dur.
\]

In other terms, we store the \(piid\) in the first column, followed by all the process properties of \(BP\) and by the durations of the tasks participating in the BP.

Figure 6 shows an example of how \(ES\) looks like for our travel expense reimbursement process. Using letters to index columns, we have that row 8 holds the process instance \(bpi_{439}\). The mapping is done as follows: \(piid\) is mapped to the first column \(ES(A, 8) \leftarrow 439\), then, we have the mapping for the process
Properties $\text{ExpenseAmount}$, $\text{Duration}$, $\text{EmployID}$ and $\text{ExpenseThreshold}$ as $\text{ES}(B, 8) \leftarrow 1.319$, $\text{ES}(C, 8) \leftarrow 6$, $\text{ES}(D, 8) \leftarrow \text{"AJ938"}$ and $\text{ES}(E, 8) \leftarrow 600$, respectively. Finally, we have the task durations, of which we show in Figure 6 only the one corresponding to the task $\text{Fill travel expense reimbursement form}$ as $\text{ES}(F, 8) \leftarrow 550$.

6. Analysis and Visualization of Results

Recall the definition of metrics as a function $m(L) = \text{val}$, where $L$ is the process execution log and $\text{val} \in \mathbb{R}$ is the metric’s value. Within a spreadsheet, $m$ corresponds to a formula that can be specified using the standard spreadsheet functions provided by the adopted spreadsheet tool, $L$ corresponds to the process execution log $\text{ES}$, and $\text{val}$ corresponds to the output produced by the spreadsheet formula. Figure 7(a) shows the metrics spreadsheet $\text{MS}$ we use for computing metrics.

In this example, if we consider Google Spreadsheets\(^1\) as our spreadsheet tool, the BP analyst computes the metric $\text{Sum of expense amounts (first 30\% of instances)}$ using a combination of the spreadsheet functions $\text{SORT(range, sortColumn, isAscending, sortColumn2, isAscending2)}$ and $\text{DSUM(database, field, criteria)}$. The function $\text{SORT(...)}$ sorts the process instances in ascending order based on the expense amount. The

\(^1\)http://docs.google.com/spreadsheet/
The function `DSUM(...)` sums the amount requested for first 30% number of instances that appear in the sorted list. Due to the lack of space to fully explain the use of the aforementioned spreadsheet functions, we put in Figure 7(a) only the reference to the formula used to compute the sum.

An assertion was defined as a function $a(L, P_{red}) = bool$, where $L$ is the process execution log, $P_{red}$ is a set of predicates and $bool \in \{true, false\}$. In turn, a predicate is a function $pred(L, m) = bool$, where $L$, $m$ and $bool$ are as defined before. In order to define assertions in a spreadsheet, we use the standard logical operators provided by spreadsheet tools. Figure 7(b) shows an example of an assertions spreadsheet $AS$ that can be used to compute assertions. For instance, the spreadsheet formula `'Metrics MS'!B2 > 15.000` checks whether or not the sum of expenses for the first 30% of the instances exceeds the maximum amount allocated for the fast track reimbursement. This assertion is composed of a single predicate that compares the metric `Sum of expense amounts (first 30% of instances)` (cell $MS(B, 2)$) with the maximum amount allowed (i.e., 15K euros). While this is a simple assertion, the BP analyst can construct fairly complex ones by combining logical operators (such as `AND`, `OR` and `NOT`) and predicates.

Also the definition of analysis reports relies on the built-in data visualization tools (charts and tables) provided by the spreadsheet. Recall that reports are of the form $r(L, M, A) = V$, with $L$ being the process execution log and $M$ and $A$ being the sets of metrics and assertions, respectively. $V$ are the vi-
visualization widgets (e.g., charts or tables) used to construct the report. Charts and tables conveniently summarize the results obtained from the computation of metrics and assertions. For example, Figure 7(a) uses a simple bar chart that plots an histogram of the expense amounts requested. This enables the visual analysis of its distribution. The dataset for this chart was prepared using metrics computed with standard spreadsheet formulas (the details are skipped in this article). The exact design of the report is up to the BP analyst, who knows best how to design the report so as to most effectively communicate his findings to the developer.

Thus, using metrics, assertions and charts, operationalized with the help of the built-in functions of the spreadsheet tool, the BP can analyze the outcomes of the simulations generated by different configuration settings that reproduce the various execution scenarios in study. Back to our reimbursement process, we can see in Figure 7 that under the configuration settings of the simulation parameters, the BP analyst can learn that reimbursing all 30% of the of lowest request amounts is not possible (the total sum of amounts exceeds the budget), and that, for example, either the target percentage should be lowered, the budget for the fast track reimbursement option should be incremented, or the company should still tolerate a delay in a (fewer) number of reimbursement requests.

7. Implementation

Figure 8 illustrates the functional architecture of the prototype of our solution. On the client side, we have the process model editor used to design BPs and the spreadsheet tool used to work with the spreadsheets CS, ES, MS and AS. On the server side, we have all the components that are in charge of configuring the environment for the simulation and analysis of BPs. When a process
model \(BP\) is ready for simulation, it is sent to the *simulation server*, which is in charge of managing the requests for BP simulations. This request is forwarded to the *simulation configurator*, which takes the model \(BP\) and performs three tasks: First, it creates the mock services that mimic the tasks in \(BP\) at simulation time and saves them in the *service repository*. Then, it stores \(BP\) into the *BP model repository* for future use. Finally, it requests the *spreadsheet manager* to create the templates for the spreadsheets \(CS, ES, MS\) and \(AS\), which are stored into the *spreadsheet repository* for reuse in the following phases.

Back to the client side, the BP analyst can use the *spreadsheet tool* to configure \(CS\) and send it to the *simulation server* for the simulation of \(BP\). The *simulation server*, in turn, forwards \(BP\) and \(CS\) to the *simulation manager*, which is in charge of managing the deployment of \(BP\) (using the configurations in \(CS\)) into the *BP simulator*. The latter queries \(ES, MS\) and \(BP\) from the *spreadsheet, service and BP model repositories*, respectively, simulates the \(BP\), and stores the obtained process execution data into \(ES\). The resulting \(ES\) is stored back into the *spreadsheet repository* and made available, together with \(MS\) and \(AS\), to the *spreadsheet tool* for analysis.

The current implementation of our analysis suite uses Signavio\(^2\) as *process model editor*, Google Spreadsheets\(^3\) as *spreadsheet tool*, and Activiti\(^4\) as internal engine of the *BP simulator*. Signavio has been extended to enable the generation of the configuration spreadsheet. Google Spreadsheets has been extended to interface with the simulation back-end, acting as user interface of the BP analyst for simulation and BP analysis. Both extensions are implemented via JavaScript; the back-end components are implemented as standard web applications using Java. The screenshots in Figures 4–7 show the look and feel of the prototype at work. At http://goo.gl/v4k2Yj we show a video of the tool in action. The source code of the tool can be downloaded from https://sites.google.com/site/ssbptester/.

8. User studies

We ran two user studies to validate the viability of the proposed approach. First, we assessed the suitability of spreadsheets with real BP analysts, then the whole approach with master students. We summarize both studies next; details of the study (scenarios, questionnaires, raw data) can be found at http://sites.google.com/site/ssbptester.

8.1. Business process analysis with spreadsheets

The objective of this study was to understand whether our approach achieves the first goal of our problem statement, i.e., enabling BP analysts to analyze business processes. This study specifically focused on the configuration of the

\(^2\)http://code.google.com/p/signavio-core-components
\(^3\)https://docs.google.com/spreadsheet
\(^4\)http://www.activiti.org
BP simulation and the analysis of process execution data. The participants to the study were three employees of the Paraguayan subsidiary of DHL, who operate as BP analysts at an everyday basis. Each BP analyst participated separately in a one hour session, conducted within the premises of the company. All participants were familiar with spreadsheets; only one of them knew Google Spreadsheets. None of them had a background in computer science. The user study was structured as a usability test followed by a retrospective probing in the form of a semi-structured interview (Lazar et al., 2010).

For the usability test, participants were introduced to our tool and watched a video exemplifying the features of the tool. Then, they were presented with an analysis scenario based on a simplified version of the BP model shown in Figure 1(a), provided with a suitable configuration spreadsheet $CS$, and asked to analyze the BP according to the scenario. For the retrospective probing, we asked participants questions about their experience, thoughts and actions after the usability test.

All participants agreed that the BP analysis tasks were easy to understand, but in some cases difficult to perform. The main reason for this was that, while participants did not have problems in writing simple spreadsheet formulas like sums, averages and standard deviations, they faced difficulties in defining complex formulas that involved conditional and statistical distribution functions. This problem was exacerbated by the fact that the nomenclature of the func-
tions in Google Spreadsheets differs from the one found in Microsoft Excel, which they were more acquainted with. The language of the tool also contributed to the issue: the spreadsheets for the analysis of the BP were all in English, and so were the names of the functions. Although all participants declared good knowledge of English, they were used to work with spreadsheets in Spanish, which made it sometimes difficult to find even functions they knew. These problems were however easily overcome with a small help from the person running the study.

As for the general feeling and mood after the study, two of the participants said they felt comfortable, while the third one said the experiment was long and stressful. All three agreed that they had to learn new concepts and terminology they were not familiar with regarding both BP simulation (e.g., arrival rates) and spreadsheet functions (e.g., conditional and statistical distribution functions). All participants agreed that using spreadsheets for analysing BPs is useful and close to their working experience for two main reasons: (i) they are familiar with spreadsheets, and (ii) spreadsheets are suitable to analyze data in a tabular format, helped by the pre-built spreadsheet formulas, filters and charts. When participants were asked if they would use the tool, all of them responded positively and stated that the approach would indeed be effective in helping them to autonomously analyze BPs, provided they have a good working knowledge of the spreadsheet’s predefined functions.

Although this study was conducted with only three BP analysts and, hence, does not have statistical relevance, we nevertheless consider the study a good indicator for the suitability of using our spreadsheet-based approach for analyzing BPs. Participants intuitively understood their tasks and unanimously agreed on the viability of the approach.

8.2. Modeling, analyzing and reporting

The second study aimed to provide end-to-end coverage of our approach and statistical tests. The study therefore also aimed to understand whether the approach facilitates the discussion of findings between the BP analyst and the software developer. To do so, we involved a total of 22 MSc students taking part of a BPM course at the University of Trento, Italy, all of them with a background in computer science, BP modeling (BPMN) and spreadsheets. The study lasted around 1 hour and 15 minutes and it was carried out in the laboratory of the university.

This study was also structured as a usability test with retrospective probing. We again offered a training session in the form of a live demo to introduce the tool to the students. The retrospective probing took the form of an online survey. Students were paired up, one playing the role of the BP analyst and one the role of the software developer. Each role was assigned a number of tasks related to the role, specified in two scenarios based on a simplified version of the BP model in Figure 1(a). The first scenario consisted in performing one analysis cycle as presented in Figure 3. The second scenario asked the BP analyst to communicate a change in the BP model to the software developer and to analyze the modified BP again.
To collect feedback, we prepared two surveys with 26 and 19 questions for the BP analyst and software developer, respectively. The questions were answered using a Likert scale of 1 (strongly agree) to 5 (strongly disagree) (Lazar et al., 2010). We consider answers 1 and 2 as positive answers, 3 as neutral answer, and 4 and 5 as negative answers. The survey included also open questions that allowed participants to provide free feedback. The results of the survey are reported in the form of descriptive statistics using the mean and median of the sample. We also use a two-sided hypothesis test to test the significance of the answers for each single question. The test makes use of a Wilcoxon, signed rank test with a significance level of \( p = 0.05 \) and a null hypothesis \( H_0: \eta = 3 \) (where \( \eta \) represents the median) and an alternative hypothesis \( H_A: \eta \neq 3 \). In other words, the null hypothesis is that participants of the study have a neutral answer for each question, against the alternative hypothesis that they are lean towards positive or negative answers.

**Software developers:** Figure 9 shows the feedback from software developers regarding their overall experience. Most questions were answered positively (which is confirmed by our tests: \( p - values \) are lower than 0.05), with the exceptions of the questions regarding the time taken by the experiment and the overall satisfaction with the experience, in which we obtained rather neutral answers (for these two questions, the \( p - values \) where slightly higher than 0.05).

**BP analysts:** Figure 11 summarizes the overall experience by BP analysts. All questions were answered positively and confirmed by our statistical tests. The exceptions are the questions related to the easiness for setting up process properties, the similarity of the BP editor’s UI w.r.t. the original version of Signavio and the preference of our tool over the others, for which we obtained rather neutral answers (we cannot reject \( H_0 \)). One participant pointed out positively that the joint use of the BP model editor with Google Spreadsheets “permits teams to work together in real-time,” while another participant recommended to improve the debugging functionalities of the editor “in order to be able to find errors in the BP model before running the BP simulation.”
<table>
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<th>Negative</th>
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<td>0</td>
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</table>

Figure 12: Survey results regarding the use of spreadsheets for BP analysis.

The results of both user studies provide good evidence of the potential of our approach. Our observations testify that the interaction between the BP analysts and the developers were well-disposed and facilitated by our approach, confirming the suitability of the approach for collaborative BP analysis. The company’s BP analysts were more inclined towards the use of mainstream spreadsheet tools, such as Microsoft Excel, given their familiarity with such tools. BPM students, instead, appreciated more the use of Google Spreadsheets, due to its suitability for real-time collaboration. Given their higher familiarity with conditional expressions and complex statistical functions compared to the BP analysts, the BPM students felt much more comfortable in using spreadsheets to analyze the

8.3. Discussion of results

The results of both user studies provide good evidence of the potential of our approach. Our observations testify that the interaction between the BP analysts and the developers were well-disposed and facilitated by our approach, confirming the suitability of the approach for collaborative BP analysis. The company’s BP analysts were more inclined towards the use of mainstream spreadsheet tools, such as Microsoft Excel, given their familiarity with such tools. BPM students, instead, appreciated more the use of Google Spreadsheets, due to its suitability for real-time collaboration. Given their higher familiarity with conditional expressions and complex statistical functions compared to the BP analysts, the BPM students felt much more comfortable in using spreadsheets to analyze the
process execution log. This suggests that, in order to bring our tool to a real setting, it is necessary to make sure BP analysts have the necessary training in using spreadsheets. However, it is important to note that all participants in both user studies easily understand the mapping of the BP analysis problem to the problem of analyzing data in spreadsheets. Once participants figured out the right spreadsheet functions to use, they were able to easily define metrics and assertions over the process execution log organized into process instances. In conclusion, we accept our hypothesis that mapping the BP analysis problem to the design of a spreadsheet calculation enables the BP analyst to analyze BPs autonomously.

9. Qualitative Analysis

We complement the above user studies with a qualitative analysis of our tool (Spreadsheet-based BP Analyzer). The analysis consists in a comparison of our tool against state of the art solutions used for BP analysis and simulation. The analysis includes today’s most representative commercial tools in the realm of BP analysis as well as academic and open source solutions. In particular, we considered Websphere Business Modeler (v. 6.2) (IBM, 2009), the licensed solution of IBM for simulation and analysis of BPs, TIBCO Business Studio (v. 3.9) (TIBCO, 2014), the proposal of TIBCO Software Inc. for modeling, analysis, and simulation BPs. Free alternative solutions are also included in our analysis. In concrete, we consider BizAgi Modeler (BizAgi, 2015), the free solution offered by BizAgi to document and analyze BPs, Bonita Open Solution (v. 5.6) (Bonitasoft, 2011), the open-source proposal of BonitaSoft for modeling and simulating BPs represented in BPMN, and Adonis Community Edition (BOCGroup, 2015), the free modeling and simulation BPM tool offered by BOC Group. Two academic tools are also present in the comparative analysis. The first one, Signavio Process Editor (v. 9.0) (Signavio, 2015), started as an academic project and recently turned into a commercial solution. The proposal targets business practitioners in the context of modeling and simulation of BPs. The second one, SimQuick (Hartvigsen, 2004), is an entirely academic BP simulation tool that uses spreadsheets and Microsoft Excel macros to enable the design and simulation of BPs.

9.1. Comparison Framework

The comparison framework used to conduct the analysis is based on five categories of functionalities, namely, BP modeling, simulation configuration, simulation, analysis, and reporting. These five categories represent the typical phases of the BP design and analysis process proposed by the analyzed tools. The framework in particular aims to highlight those concerns that are related to the BP analyst inside this BP design and analysis process. Although BP modeling is not the focus of our work, it was included in the analysis given its crucial role in the proposed process lifecycle. The BP modeling capabilities of the tools are analyzed under three dimensions: notation,
i.e., which BP modeling notation is supported by the tool, model verification, i.e., what types of modeling errors can be detected, and debugging, i.e., what types of features are available to find and fix the modeling errors.

The analysis of functionalities offered by the tools for the BP simulation configuration is mainly focused on the statistical facilities offered by the tools to configure process tasks, resources, and domain-specific parameters (e.g., expenses, interest rate). We also analyze the variety of domain-independent variables (e.g., number of simulation instances, instances arrival rate, loading period) that are possible to setup with the tools.

We capture the simulation capabilities of the tools through two dimensions, namely, runtime monitoring and task behavior. The first one is related to the features offered by the tools to monitor the execution of the simulation, e.g., animation and runtime statistics, while the second one is associated to the simulation capacities of the tools w.r.t. the simulation of task behavior.

Regarding the analysis capabilities of the tools, we focus on understanding the flexibility of the tools in giving BP analysts the freedom to write their own analysis instruments (metrics and assertions) and the power of the tools in assisting analysts in obtaining information about the behavior of the process.

The reporting dimension includes the flexibility of the tools to create custom analysis reports, the features offered to foster collaborative analysis and reporting, and whether the simulation output is accessible and in which format.

Because not all the tools could be installed for their study, we exclusively base our analysis on the official documentation provided for each solution. We therefore highlight the situations in which the documentation provides insufficient information to draw a conclusion about a particular dimension.

Figure 13 presents the analysis in a table where columns represent the description of each tool. The gray-shaded column contains the descriptions of our Spreadsheet-based BP Analyzer.

9.2. Analysis

BP modeling. All the tools offer similar modeling functionalities. Almost all of them have BPMN editors equipped with features that perform automatic model checking (syntax validation and deadlocks verification). They also provide model debugging functions through warning and errors messages and by coloring the conflicting elements. In addition, TIBCO provides features to semi-automatically fix simple syntax errors. SimQuick is an exception within this category because it has its own modeling notation implemented through Microsoft Excel macros. In our case, the full modeling functionalities is based on Signavio Core Components, the free and open-source version of Signavio Process Editor and thus our modeling features are equivalent to that of Signavio.

BP simulation configuration. No big differences are appreciated when comparing the features for configuring the simulation. All the tools enable the use of statistical distributions to configure the duration of the tasks. Spreadsheet-based BP Analyzer and SimQuick allow for the use of distributions to set up the costs associated to the execution of the process tasks while the rest of the
<table>
<thead>
<tr>
<th>BP Modeling</th>
<th>IBM WebSphere Business Modeler</th>
<th>TIBCO Business Studio</th>
<th>Bizagi Modeler</th>
<th>Bonita-Open Solution</th>
<th>Business Process Editor</th>
<th>Spreadsheet-based BP Analyzer</th>
<th>SimQuick</th>
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<tr>
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</table>

Figure 13: Qualitative analysis of our tool (Spreadsheet-based BP Analyzer) and state-of-art business process analysis and simulation solutions.
tools allow only for the use of constant values to define task execution costs. TIBCO provides features to set up tasks duration using historical data.

Half of the tools, namely TIBCO, Bonita, Adonis, and Spreadsheet-based BP Analyzer, offer the possibility to model, through constants, expressions, and even by statistical distributions, the value of domain-specific parameters, such as interest rate, expenses, return of investment. All the tools enable the definition of domain-independent variables. In SimQuick and Spreadsheet-based BP Analyzer a couple of variables can be configured, i.e., number of simulation instances and arrival rate, while in the rest of the tools a large variety of parameters can be defined, such as timespan in which the instances are created and workload calendar.

Almost all the tools, excluding ours, enable the definition of quantity, work schedule and cost per units of the resources associated to the process execution. Being a tool that is constructed on top of a spreadsheet editor (Microsoft Excel), SimQuick provides the chance to employ statistical distributions to model resources while in the other tools only constant values can be used.

**BP Simulation.** Only TIBCO, WebSphere and BizAgi support runtime-monitoring functions. The first two provide interactive animation features that allow users to follow the execution of the simulation step-by-step. In addition, all of them display descriptive statistics at runtime, e.g., average tasks duration, most expensive simulation instance, number of current idle resources and average waiting time. The main advantage of Spreadsheet-based BP Analyzer in relation to the others is the possibility to go from 0% to 100% on the simulation of the behavior of the process tasks. In other words, Spreadsheet-based BP Analyzer provides the flexibility to choose whether a task in the BP should be simulated or carried out by a real web service. This feature, which is not available in any other state-of-art tools that offer the chance of simulating the execution time, cost, domain-specific process parameters and resource utilization of the tasks, provides the possibility to verify whether the real services operate as expected.

**BP Analysis.** Spreadsheet-based BP Analyzer and SimQuick are the only ones that fully empower BP analysts to define their own metrics and assertions. In order to provide such flexibility, they leverage on the power of the spreadsheet’s built-in functions. Similarly, WebSphere provides user-defined metrics but from the documentation is not clear whether this functionality is available to measuring simulation data or real data.

The rest of the tools offer standard, predefined metrics, e.g., duration, cost, throughput and resource utilization, through which the analyst can get information about process behavior. Also, only SimQuick and Spreadsheet-based BP Analyzer are the only ones that give BP analysts the chance of writing assertions on top of the metrics computed over the process execution data.

A useful and important feature is the possibility to assist analysts in the evaluation of their business processes. Only TIBCO, IBM WebSphere and BizAgi offer additional analysis functions. WebSphere and BizAgi provide features to conduct what-if-analysis while TIBCO enables the comparison of resource utilization between simulation instances. In addition, IBM WebSphere provides
features that ease the comparison between the ideal BP model (to-be) and the current version of the model (as-is).

**BP Analysis Report.** Most of the tools provide predefined reports. However, SimQuick and Spreadsheet-based BP Analyzer offer the users the possibility to create their own reports. By exploiting the graphical and analytical features of spreadsheets editor programs (Excel, Google Spreadsheet) SimQuick and Spreadsheet-based BP Analyzer allow BP analysts to draw custom reports of any complexity. Following a more restricted approach, WebSphere also offers user-defined report features. By employing a drag-and-drop designer, BP analysts are able to build custom reports which are based on a limited set of predefined elements, such as text-fields, tables, and summary statistics (counts, sums, and averages).

The spreadsheet editors used by SimQuick and Spreadsheet-based BP Analyzer offer, in addition, built-in functionalities that enable the possibility to add comments and notes on the reporting documents, which can facilitate the communication between BP analysts and developers. In the case of Spreadsheet-based BP Analyzer this communication may even happen in real-time thanks to the chat functionalities of Google Spreadsheet.

Neither IBM Websphere nor TIBCO provide the raw output of the simulations, which limits the possibility of running further BP analyses. The rest of the tools do allow users to access the simulation output promoting the use of alternative tools to further analyze BPs. In this sense, BizAgi provides the output in a tabular, exportable and event-based format; Signavio in an event-based, downloadable MXML format (van der Aalst, 2011), while Adonis, Spreadsheet-based BP Analyzer, and SimQuick offer the raw simulation output in a tabular, trace-based and exportable scheme. Bonita’s documentation states that simulation outputs can be downloaded but their content and format are not specified.

**9.3. Discussion**

The above analysis helps to understand better the distinctive and innovative aspects of the Spreadsheet-based BP Analyzer. The combination of BP simulation and a spreadsheet-based UI (Google Spreadsheet) accompanied with a standard BPMN process model representation equips also BP analysts with limited technical skills with a single tool that enables them to simulate, verify and analyze the performance of BPs through personalized instruments (metrics, assertions and custom reports). The use of a full-fledged BP engine to run the simulation enables the flexible invocation of both simulation web services that mimic the behavior of other web services as well as real, production web services. This turns the Spreadsheet-based BP Analyzer into an instrument that can be used both by the BP analyst for his BP verification and simulation and by the developer for the testing of how real web services perform inside a simulated process. As an add-on, the live collaboration support by Google Spreadsheet offers a new and dynamic communication alternative that can ease the interaction between BP analysts and developers.

Two limitations of the Spreadsheet-based BP Analyzer w.r.t. to some of the tools presented in this comparative analysis are the lack of support for
the simulation of human resource utilizations (given our focus on service-based processes) and the lack of additional analysis aids such as what-if-analyses. We plan to extend the Spreadsheet-based BP Analyzer with these functionalities in future work.

10. Related work

The analysis of business process has triggered many research efforts, yielding a variety of different approaches. In the following, we discuss those related works that fall into the context of this paper, which includes service-based BP testing, BP simulation, compliance checking, process mining and modeling and testing spreadsheet.

The problem of service-based BP testing took significant relevance with the SOA and its use to support the operation of BPs. In particular, many approaches have been proposed to address this problem in the context of BPs represented with BPEL (Andrews et al., 2003). For example, some of the works in this context are dedicated to perform unit tests of web service compositions. Unit tests in this context means testing each web service and their corresponding interfaces, i.e., each operation offered and invoked by the service (Mayer and Lübbe, 2006; Li et al., 2008a). A side problem associated to the testing of BPEL processes is the generation of test cases. Works like (García-Fanjul et al., 2006; Yuan et al., 2006; Yan et al., 2006) propose approaches for the generation of test cases using techniques from model checking, graph search and concurrent path analysis. Other types of tests performed on service-based BPs include regression testing (Li et al., 2008b) and integration testing (Bucchiarone et al., 2007). All these approaches require special software testing and development skills.

BP simulation has been employed for the purpose of testing BPs. For example, Aguilar et al. (1999) propose a BP simulation methodology to analyze the performance of financial BPs in unforeseen, potential situations. In the context of service-based BPs, Narayanan and McIlraith (2003) propose the use of simulation to test the preservation of properties (e.g., safety conditions) associated to the services responsible for the BP execution by combining Petri-Nets and DAML-S. Chandrasekaran et al. (2003) use simulation to monitor and analyze the performance of individual web services involved in a BP. Tan and Takakuwa (2007) and Wynn et al. (2008) use simulation to evaluate the impact of BP re-engineering tasks on process performance.

In the context of process mining (van der Aalst, 2011), techniques such as conformance checking and process discovery have been employed to check whether or not a BP behaves as expected. Conformance checking (Rozinat and van der Aalst, 2008; van der Aalst et al., 2008b) verifies whether the traces of execution of a BP conform with a given BP model. In order to do so, the approach proposes to replay the real process execution data on the BP model to detect if there are mismatches between the two. Conformance checking therefore checks if the event log structurally matches the process model, or, in other words, it checks if the control flow that underlies the event log matches that of the process model. It therefore only considers the structure of the BP model as
the specification of the expected behavior and does not focus on process-specific metrics. Moreover, while the main use case of conformance checking consists in using a real event log to check against a predefined process model \textit{a posteriori} (i.e., after a real execution of the process), in our case we use a simulated log to check \textit{a priori} (i.e., before the real execution of the process) if a BP behaves as expected.

\textbf{Process discovery} is the task of inferring a BP model from process execution data (van der Aalst et al., 2004; Motahari-Nezhad et al., 2011). Testing a BP with process discovery can be done by first inferring the BP model from the process execution data and then comparing if the inferred model corresponds to the expected model. This comparison can be done either manually or using automatic techniques such as those based on BP similarity (Dijkman et al., 2009). There are a set of commercial (e.g., ARIS PPM, HP BPI, and ILOG JViews) and academic (e.g., EMiT, Little Thumb, InWoLvE, Process Miner, and MinSoN) process mining tools. The main goal of the academic tools is to extract knowledge from event logs for discovering processes. The commercial tools are more oriented to the design, analysis and optimization of BPs using for example, charts and dashboards. In addition, HP BPI can discover a BP from event logs. Our approach differs from these process mining techniques in that they are meant to be used \textit{after} the real process has been executed. This contradicts our purpose of using BP testing as an instrument to prevent unwanted behaviors. Moreover, these two approaches focus only on the structure of the BP, while our approach tests the dynamics and data produced by the execution of the BP through the use of user-defined metrics and assertions.

\textbf{Compliance checking} is the problem of verifying whether a BP model or its execution adheres to a set of compliance rules (i.e., the \textit{expected behavior}) that typically emerge from laws, regulations and standards. This problem has been addressed both statically and dynamically. In \textit{static compliance checking}, only the model of the BP is checked against the compliance rules. For example, Liu et al. (2007) address the problem by expressing the BP model in Pi-Calculus and the corresponding compliance rule in Linear Temporal Logic. Using this representation, model checking techniques are used to check whether the BP model complies with the compliance rules. Governatori et al. (2006) proposes a logic-based formalism to represent both the semantics of contracts and compliance checking procedures. The formalism used is Format Contract Language, which is based on Deontic Logic and helps in representing and checking contrary-to-duty obligations in contracts. In \textit{dynamic compliance checking}, BP execution evidences are used to check for compliance. Works by Rodríguez et al. (2013) and Silveira et al. (2010) propose the use of so-called Key Compliance Indicators (KCI)s to measure the compliance level of service-based BPs from process execution data, e.g., to measure the fulfilment of Service-Level Agreements (SLAs). In a similar approach, Casati et al. (2007) and Sayal et al. (2002) propose to warehouse process execution data to enable the monitoring, analysis and reporting on the performance of BPs, e.g., to check the duration of process execution instances when they are constrained in time. The approach we present in this paper is similar to dynamic compliance checking, with the difference that we
enable the use of simulated data, next to real data, to check different execution scenarios. Our approach can thus be used for simulation-based compliance checking if process properties provide access to the data necessary to express compliance concerns (assertions).

A topic that is also related to our work is that of spreadsheet modeling and testing. Here, the focus is put on modeling and testing the spreadsheet content itself. In particular, spreadsheet testing and debugging is important because it positively influences spreadsheet accuracy (Kruck, 2006). Burnet et al., who coined the term “end-user software engineering”, proposed an approach to support assertions in spreadsheets (Burnett et al., 2003). The assertions are built on top of cells to check the execution of formulas contained in cells. The approach provides the possibility to create assertions by the end-user, following an abstract syntax that is implemented through both graphical and textual concrete syntaxes. Hermans proposes Expector, an Excel-based tool for helping users in improving their testing practices, e.g., by helping them in achieving better testing coverage, more meaningful names for outcomes of the testing, among other features (Hermans, 2013). In the same paper, the author presents an interesting study on the use of testing within spreadsheets. They found out that 8.8% of the spreadsheets from the EUSES corpus (Fisher and Rothermel, 2005) contain testing formulas that use only the spreadsheet’s built-in functions. Rothermel et al. present a methodology for the test adequacy criterion in form-based visual programs (the authors place spreadsheets under this category) (Rothermel et al., 1998). In their methodology, the authors propose to check for the definition-use adequacy of a test suite based on the all-uses data flow adequacy criterion. The prototype, implemented in the research language Forms/3 (Burnett and Ambler, 1994), provides visual feedback to the users about the “testedness” of their spreadsheet. The research works presented above focus more on modeling and testing the spreadsheets itself. Our approach, instead, focuses on testing an external artefact with the help of spreadsheets. Yet, the contributions made in these works can complement our solution because they can help to improve the accuracy of the spreadsheets we used for BP analysis.

11. Conclusion

In this article we approached a relevant and timely issue in today’s business process management practice, i.e., that of analysing processes. We specifically emphasized the role of the BP analyst, not only in providing input for the design of processes but also in analyzing them. In order to enable BP analysts to perform own analyses without the need for programming skills, we conceived a technique that is specifically tailored to the average skills of BP analysts. The intuition we followed for the implementation of the technique is to adopt common spreadsheets as abstraction and user interface for both setting up analyses and computing analysis reports.

As confirmed by our user studies, the spreadsheet abstraction has indeed the potential to enable BP analysts to perform fairly complex analyses autonomously and to effectively discuss findings with software developers, so as
to iteratively improve their models. The qualitative analysis of the approach complements the user studies with a discussion of its strengths and weaknesses compared to the state of the art in BP model analysis.

The positive results we obtained encourage further extensions of the approach. Specifically, we would like to allow BP analysts to also obtain a concrete feeling of how their processes behave if deployed in a real BP system by emulating web services and visualizing process progress in a process monitoring dashboard. Comparing log data produced during the analysis of a process with real log data obtained after the deployment of the process would further enable the fine-tuning of the simulation/emulation. This, in turn, would improve analysis precision and turn the simulator into a viable tool, for example, to produce synthetic data for the testing of process mining algorithms.

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**References**


