A Static Change Impact Analysis Approach based on Metrics and Visualizations to Support the Evolution of Workflow Repositories

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ABSTRACT:

In ever-changing business environments, organizations continuously refine their processes to benefit from and meet the constraints of new technology, new business rules, and new market requirements. Workflow management systems (WFMSs) support organizations in evolving their processes by providing them with technological mechanisms to design, enact, and monitor workflows. However, workflows repositories often grow and start to encompass a variety of interdependent workflows. Without appropriate tool support, keeping track of such interdependencies and staying aware of the impact of a change in a workflow schema becomes hard. Workflow designers are often blindsided by changes that end up inducing side- and ripple-effects. This poses threats to the reliability of the workflows and ultimately hampers the evolvability of the workflow repository as a whole. In this paper, we introduce a change impact analysis approach based on metrics and visualizations to support the evolution of workflow repositories. We implemented the approach and later integrated it as a module in the HP Operations Orchestration (HP OO) WFMS. We conducted an exploratory study in which we thoroughly analyzed the workflow repositories of 8 HP OO customers. We characterized the customer repositories from a change impact perspective and compared them against each other. We were able to spot the workflows with high change impact among thousands of workflows in each repository. We also found that while the out-of-the-box repository included in HP OO had 10 workflows with high change impact, customer repositories included 11 (+10%) to 35 (+250%) workflows with this same characteristic. This result indicates the extent to which customers should put additional effort in evolving their repositories. Our approach contributes to the body of knowledge on static workflow evolution and complements existing dynamic workflow evolution approaches. Our techniques also aim to help organizations build more flexible and reliable workflow repositories.

KEY WORDS:
workflow evolution; change impact analysis; metrics; visualizations; dependency management;
INTRODUCTION

Large-scale workflow repositories, which may encompass thousands of workflows in real world settings, are intrinsically complex. Workflows in these repositories frequently link to each other, forming a complex network of dependencies. As workflows evolve, their number of elements and interconnections tend to increase. Furthermore, organizations often heavily rely on some of the out-of-the-box (OOTB) workflows provided by vendors. This means that modifying or replacing these core workflows can affect the large amount of other workflows that depend on them. Therefore, evolving workflow repositories poses a challenging task.

In this context, two problems may occur. First, workflow designers may become reluctant to apply changes to workflows. In this case, the repository becomes less flexible, since it neither leverages opportunities nor deals with the constraints of new technology, new market requirements, and new legislation (Casati, Ceri, Pernici, & Pozzi, 1998). Second, workflow designers may end up performing changes to workflows without knowing the associated impact, because it is too difficult to be aware of all interdependencies and evaluate how critical they are. In this case, the repository becomes less reliable, since inappropriate changes may induce side- and/or ripple-effects (Arnold, 1996). A side-effect is an error or other undesirable behavior that occurs as a result of a modification (Freedman & Weinberg, 1982). In turn, a ripple-effect occurs when a small change to a system affects many other parts of this same system (Stevens, Myers, & Constantine, 1974). In fact, previous research already showed that making software changes without visibility into their effects can lead to poor effort estimates, delays in release schedules, degraded software design, unreliable software products, and premature retirement of software systems (Mens & Demeyer, 2008; Souza & Redmiles, 2008; Swanson & Beath, 1989). In summary, by being less flexible and less reliable, the workflow repository also becomes less evolvable.

This paper reports the results of joint efforts from researchers and engineers from the University of São Paulo, HP Labs, and HP software in seeking innovative workflow evolution solutions to be integrated into the HP Operations Orchestration (HP OO) product. HP OO is a professional industry Workflow Management System (WFMS) that provides an OOTB workflow repository targeted to help organizations automate common IT operations. Customers can also leverage this repository to build their own custom workflows. Table I depicts HP OO common usage scenarios.
Driven by customers’ feedback, we decided to focus on enhancing HP OO’s change impact analysis features. Software change impact analysis concerns “identifying the potential consequences of a change, or estimating what needs to be modified to accomplish a change” (Arnold, 1996). The analysis aims to make the existing relationships among artifacts more explicit to humans, so that they can maintain and evolve software systems more easily. Change impact analysis information can then support planning changes, approving changes, accommodating certain types of changes, and tracing through the effects of changes (Arnold, 1996). Naturally, mitigating side- and ripple-effects have also been two common goals of change impact analysis (Kagdi & Maletic, 2006).

Despite its benefits, change impact analysis has long been one of the most tedious and difficult parts of the software evolution process. According to Arnold (1996), tools frequently either provide limited analyses scopes or are too complex so that only specialists are able to deal with it. Moreover, manually inspecting artifacts to determine change impact is often labor intensive, ad-hoc, and definitely does not scale for large systems. Building on our previous work on dependency management (Gustavo Ansaldi Oliva & Gerosa, 2012), we conceived a static interdependency-based change impact analysis approach to support workflow designers in evolving their workflow repositories. It is static because the analyses rely on the workflow schema (structure), which is the definition of the sequence in which activities are executed (Casati et al., 1998). In other words, we are tackling the problem at the workflow type level (and not at the instance level) (Dadam & Rinderle, 2009). It is interdependency-based because we determine change impact by detecting and analyzing the interdependencies (call relationships) among the workflows of a repository. We decided to focus on inter-workflow analysis, since most industrial tools already support intra-workflow change impact analysis. Therefore, our approach is applicable to any kind of WFMS containing workflows that call each other. This also means that the way workflow activities are actually implemented (e.g. Java applications, web services, or human intervention) is irrelevant to our approach.

Our approach relies on two metrics (change scattering and impact) and two visualizations (call-graphs and treemaps) to enable both low-level and high-level analyses. While the former focuses on the relationships of a certain workflow, the latter enables analyzing the repository as a whole.

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In particular, the visualizations we implemented make relationships among workflows explicit and more easily understandable to humans, thus increasing their awareness about change impact levels. We also tried to make the visualizations as intuitive as possible, so that workflow designers would not have difficulties to interpret them. Driven by HP OO customers’ needs, the primary goals of the approach we conceived were to (i) identify workflows possibly impacted when a certain workflow is changed, (ii) determine the likelihood of impact for each of these workflows, and (iii) offer mechanisms to enable the analysis of the change impact levels of the repository as a whole.

We implemented our approach and later integrated it into the HP Operations Orchestration (HP OO) product. In this paper, we describe the approach and an exploratory study we conducted. In such study, we characterized and analyzed 8 workflow repositories, each belonging to a different HP OO customer. The metrics and visualizations triggered a series of insights about each repository. For instance, we found that one customer developed most part of his workflows with high change impact. We also found that while the out-of-the-box repository had 10 workflows with high change impact, customer repositories included 11 (+10%) to 35 (+250%) workflows with this same characteristic.

This paper is structured as follows. In the next section, we introduce and discuss related studies in the field of workflow evolution. After that, we describe our approach and its main features. Afterwards, we present the setup of the exploratory study. Next, we show and discuss the results and the limitations of such study. Finally, in the last section, we state our conclusions and plans for future work. This paper extends our previous work published in the proceedings of the IEEE 20th International Conference of Web Services (ICWS 2013) (Gustavo A. Oliva, Gerosa, Milojicic, & Smith, 2013).

RELATED WORK

Business processes, which are often called workflows when implemented and automated within a WFMS, live in an environment that is typically highly dynamic (Dadam & Rinderle, 2009). As a consequence, workflows have to evolve in order to keep up with such an environment and remain useful. The challenges to evolve workflows have been investigated from different perspectives and several solutions have been proposed so far.

Casati et al. (1998) focused on the problem of running workflow instances when their respective schema is modified, i.e. changing existing workflows while they are operational. They introduced formal criteria to determine which running instances can be transparently migrated to the new version. In fact, dealing with running instances when updating workflow schemas is a classic problem of workflow evolution (Dadam & Rinderle, 2009). Our proposed approach has a different focus. Instead of dealing with the runtime effects of changes, we take a step back and offer an approach to support workflow designers in both planning and evaluating the impact of changes in a static fashion during design time. In a certain sense, we want to increase the awareness of workflow designers regarding the levels of change impact for the whole repository. Therefore, these approaches can be seen as complimentary, as one supports the other. Indeed, the interplay between concurrently applied workflow schema and instance changes (e.g., discovering the degree with which they overlap) is a fruitful research topic (Dadam & Rinderle, 2009).

Wang and Capretz (2011) conceived a change impact analysis approach targeted to Service-Oriented Systems. Similarly to our proposal, their approach is also based on dependency analysis. However, they define the dependencies in terms of messages exchanged among services. Fig. 1 depicts message exchanges (M1, M2, ..., M7) among the services of a hypothetical order process.
Data relative to messages (D0, D1, …, D7), which they refer to as *model elements*, are also taken into account. The goal of their work is to estimate the impact of changing the dependencies network (e.g. by adding a new service that receives and sends new messages) and data elements correspondence (e.g. by removing a certain data element from a message) on services and on the entire system. The authors’ implied notion of service collaboration has been called by other researchers as a *service choreography* (Barker, Walton, & Robertson, 2009; Ben Hamida et al., 2012; Issarny et al., 2011; Leite et al., 2013).

The core of their approach is formed by metrics they conceived based on the concepts of information entropy (Shannon & Weaver, 1963) and link analysis (e.g., HITS algorithm (Kleinberg, 1999)). These metrics include: (model) element entropy, dependency entropy, service entropy, and system entropy. Based on such metrics, they defined change impact metrics, which include: service impact, system impact, and symmetrical effect. While their approach seems very promising, their evaluation was constrained to the calculation of such metrics to the example depicted in Fig. 1 and fictitious change tasks. Here we leverage the degree of realism of our evaluation, which was conducted with real customer repositories.

In a previous study (S. Wang & Capretz, 2009), the same authors developed a change impact analysis model for web services evolution that relies on the extraction and analysis of service dependencies. Since they are dealing with lower level entities (web services), the way they capture dependencies is fundamentally different from ours. In general terms, the authors link web services according to the dependencies that exist among their respective elements (e.g., the output elements of a web service x are the input elements of a web service y). Furthermore, the authors also capture the existing relations among the inner elements of a web service (intra-dependency). Relying on these two kinds of dependencies, the authors provide (i) a metric to identify services that are difficult to modify and (ii) another metric to calculate the impact of changing a specific element of a web service. We also highlight the methodology they developed for automating changes to web services. A supporting tool was developed as part of Wang’s PhD thesis (S. Wang, 2010). In summary, our goals are quite similar to theirs, although we tackle the problem at a higher level of abstraction. Since our analysis relies only on call relationships among workflows, its implementation is simpler (especially with relation to the extraction of dependencies). Besides providing metrics to calculate change impact, we also leverage two visualizations that help workflow designers cope with the complexity of analyzing their workflow repositories as whole.
Wang et al. (Y. Wang, Yang, Zhao, & Su, 2012) conceived a comprehensive change impact analysis approach for service-based business process. While we treat the building blocks of workflows as black boxes and do not distinguish between the various kinds of workflow schema changes, their approach focuses on how service changes affect process and how process changes affect services. The authors define two layers: the process layer, which contains the internal processes of an organization, and the service layer, which consists of services that are each an external view of the internal process from the viewpoint of a specific business partner. In other words, they consider a model in which services expose observable behaviors (a.k.a. behavioral interfaces) in the form of a set of operations and invocation relations between these operations. In fact, previous studies have already discussed this modeling perspective (Zaha, Barros, Dumas, & Hofstede, 2006), and languages for describing it have been conceived (e.g., WSCI²). Wang and colleagues also present a taxonomy for service changes (Fig. 2) and processes changes (Fig. 3), as well as a derived set of change impact patterns (Fig. 4). In addition, they report a prototype tool that implements their approach.

![Fig. 2. Taxonomy for service changes (Y. Wang et al., 2012)](http://www.w3.org/TR/wsci)
Lins et al. (2008) analyzed workflow provenance (a.k.a. audit trail, lineage, pedigree) in order to extract information about workflow evolution. The authors conducted an initial case study and showed, for instance, that analyzing how much time is spent in workflow design can help in the understanding of how users interact with workflow systems. It also helps to discover the amount of effort spent to accomplish tasks, such as creating new workflows or modifying existing ones. This study thus exemplifies the potential of mining workflow evolution history. Other studies discuss the application of workflow evolution to specific areas. For instance, Chinthaka et al. (2011) state that scientists working on eScience environments frequently use workflows to carry out their experiments. Since workflows evolve as the research itself evolves, the authors analyze workflow evolution to track the evolution of the research itself. Regarding industry tools, we highlight that no other orchestration products (Microsoft Opalis, BMC Atrium, Cisco Tidal, etc.) provide the level of analysis and visualization offered by our approach. A summary of the related work is presented in Table II.

### Table II. Summary of Related Work

<table>
<thead>
<tr>
<th>Title</th>
<th>Target</th>
<th>Focus</th>
<th>Contribution</th>
<th>Evaluation</th>
<th>Tool</th>
<th>Ref.</th>
</tr>
</thead>
<tbody>
<tr>
<td>Workflow Evolution</td>
<td>Workflows</td>
<td>Schema changes</td>
<td>Dynamic schema evolution approach</td>
<td>Example</td>
<td>No</td>
<td>(Casati et al., 1998)</td>
</tr>
<tr>
<td>Dependency and Entropy Based Impact Analysis for</td>
<td>Service-</td>
<td>Service dependencies extracted from message exchanges</td>
<td>Change impact analysis metrics</td>
<td>Example</td>
<td>No</td>
<td>(S. Wang &amp; Capretz, 2011)</td>
</tr>
<tr>
<td>Service-Oriented System Evolution</td>
<td>Oriented Systems</td>
<td></td>
<td></td>
<td></td>
<td></td>
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</tr>
<tr>
<td>A Dependency</td>
<td>Service-</td>
<td>Service dependencies</td>
<td>Change Impact Analysis</td>
<td>Example</td>
<td>Yes</td>
<td>(S.</td>
</tr>
</tbody>
</table>
Impact Analysis Model for Web Services Evolution

- Extracted from WSDL:
  - Output elements of x are the input elements of y
  - Semantic mapping or correspondence built between elements of x and y
  - Manually/automatically designed relations for elements of x and y

Metrics and Methodology

- Taxonomies for service and process changes
- Change impact patterns
- Change impact analysis algorithms

Example

Yes

Wang & Capretz, 2009

Change impact analysis in service-based business processes

- Workflows and Implementing Services
  - The way service changes affect business processes and vice-versa

Examining statistics of workflow evolution provenance: A first study

- Workflow
  - Workflow evolution provenance (history)

Analysis of workflow evolution provenance generated by 30 subjects who worked on 6 distinct exploratory tasks (e.g., creating a visualization, mining a data set) over 4 months

Preliminary Case Study

Yes

(Y. Wang et al., 2012)

Other studies discuss change impact analysis in broader terms. Arnold (1996) extensively covered the foundations of change impact analysis in his classic book. He presents basic concepts, terminology, difficulties in applying change impact analysis in practice, different natures of change, etc. Lehnert (2011a) argues that although several impact analysis approaches have been developed over the years, there is no solid framework for classifying and comparing them. The author thus proposes a taxonomy for classifying change impact analysis approaches, taking into account aspects such as scope of analysis, used techniques, style of analysis, granularity of target entities, existence of tool support, supported languages, and asymptotical complexity of both time and space. The same author also produced a technical report with an extensive review of change impact analysis techniques (Lehnert, 2011b). Finally, for more information on the definition, historical background, foundations, and future directions of workflow evolution, we refer the reader to a book chapter written by Dadam and Rinderle (2009).

THE WORKFLOW CHANGE IMPACT ANALYSIS APPROACH

In this section, we introduce the change impact analysis approach we conceived to support the maintenance and evolution of workflow evolution issues. We focused our efforts on the following three main questions that arose from needs of HP OO customers:

(RQ1) Which workflows are possibly impacted when a certain workflow is changed? By change to workflows, we mean any kind of change applied to their schema (structure).

(RQ2) Given the list of workflows obtained from RQ1, then how different is the likelihood of impact for each of these workflows? Obtaining the list of possibly impacted workflows is necessary, but not sufficient. Workflow designers should know where to focus their maintenance efforts. Therefore, we also investigate the likelihood of impact for each of the possibly impacted workflows.
(RQ3) How can one evaluate the repository as a whole? Since workflow repositories are usually large and complex, analyzing the change impact caused by each individual workflow becomes infeasible. Therefore, we also support repository-wide analyses by means of visualization techniques.

By answering those questions, we intend to provide a way to identify the potential effects of a change. We focus on inter-workflow change impact analysis, since intra-workflow change impact analysis is simpler and already covered by a variety of tools. Therefore, typical use cases would include using our approach to support workflow schema modification, workflow version upgrades, and the identification of core workflows (i.e., those that potentially affect a large portion of the repository). As key benefits, we highlight that our approach increases the awareness of workflow maintainers, thus fostering more confident and responsible changes (as opposed to ad-hoc changes). In the end, this should mitigate side- and ripple-effects of changes. Furthermore, since our approach is capable of quantifying the change impact of workflows, it helps organizations to estimate change effort. As a desirable consequence, it should reduce the occurrence of statements like “it was more complicated than I first thought,” which are often heard during software maintenance tasks. Moreover, our approach helps organizations target their testing routines, which should ultimately lead to more reliable and less buggy workflow repositories. Regarding the audience, our solutions is meant to be used primarily by workflow designers in their own environment, so that they can analyze and report on their workflow repositories. Finally, it should also help managers quickly track the overall change impact levels of workflows and compare repositories against each other.

The remainder of this section is organized as follows. We first present the vocabulary we used and the assumptions we made for this work. Then, we present the internal analytical model we rely on. Next, we present the proposed metrics and visualizations. Finally, we provide some implementation details.

Vocabulary and Assumptions
We organized the vocabulary of our approach as a domain model (a.k.a. conceptual model) (Larman, 2004), which is depicted in Fig. 5. Domain models describe the main entities of a domain, as well as how these entities relate to each other. We employ the domain model to establish the assumptions we make regarding the kinds of workflow constructs we support.

We assume the existence of a Repository, which contains a series of sections. Sections are pretty much like the folders of a file system, and workflow designers use them to organize workflows and operations according to some criteria. Workflows (or simply, flows) contain interconnected steps, each representing a certain activity. Subflow steps are those that invoke another workflow. Operation steps are those that invoke a standalone operation (e.g., function, script, or even a packaged application). Fork steps are those that split into two or more Lanes, which are executed in parallel. The Join step merges all lanes upon their ending. Elementary steps include the start step and the final steps. In particular, we assume that workflows have a single start and one or more final steps (just like State Machines). The set of concepts in our domain model covers all workflow modeling constructs available in the HP OO product. In particular, HP OO employs a proprietary process modeling language inspired by BPMN2.3

3 http://www.omg.org/spec/BPMN/2.0
Internal Analytical Model

Our solution heavily relies on static call-graphs. A static call-graph is a directed graph that represents calling relationships between subroutines in a computer program. In our context, we build flow static call-graphs to support change impact analysis. In our flow call-graph, each node represents a flow, and each directed edge \((F_i, F_j)\) indicates that the flow \(F_i\) calls flow \(F_j\) (i.e., \(F_i\) has a subflow step that invokes \(F_j\)). We also say that \(F_i\) is a client of \(F_j\), and that \(F_j\) is a subflow of \(F_i\).

Since calculating a single call-graph for the whole workflow repository would likely result in a large and complicated structure, we calculate one call-graph per flow. This results in a much simpler and smaller structure to analyze. We start with the chosen flow and then discover its clients (i.e., all the other flows that can possibly call the chosen flow). We do it recursively until no more client flows are found. An example is shown in Fig. 6, which depicts the call-graph of a
hypothetical flow $F_{12}$. In our implementation, we obtain this information by manipulating HP OO XML files that describe the schema of each workflow in the repository. These XML files can be seen as a complete serialization of the repository.

**Elementary Metrics: Change Scattering and Impact**

We assess change impact according to two main metrics: change scattering and impact. The former addresses RQ1 and the later addresses RQ2. In the following, we define such metrics and introduce the algorithms we use to compute them.

**Change Scattering.** We define $\text{Scattering}(F_i)$ as the quantity of flows that are possibly impacted when $F_i$ is changed. We directly employ the analytical model to calculate this metric. Consider the example shown in Fig. 6, which depicts the call-graph of a hypothetical flow. In such case, the change scattering of $F_{12}$ is equal to 8. We also say that these 8 flows are clients of $F_{12}$. Finally, having identified the clients of $F_{12}$, it becomes straightforward to determine which and how many sections are also possibly impacted.

**Impact.** We define $\text{Impact}(F_i, p)$ of a flow $F_i$ as the quantity of flows that have a high chance of being impacted when $F_i$ is changed, where “high chance” means any probability higher than or equal to $p$. Therefore, $\text{Impact}(F_i, p) \leq \text{Scattering}(F_i)$. The pseudo code for calculating impact is as follows.
To illustrate the rationale behind the metric, consider again the call-graph depicted in Fig. 6. If $F_{12}$ is called in every possible execution path of $F_3$, then the likelihood of $F_3$ being impacted by a change in $F_{12}$ becomes high. However, if $F_{12}$ is called in only one among many possible execution paths inside $F_3$, then the likelihood of $F_3$ being impacted becomes much lower. The chances of $F_6$ being impacted are then determined based on the results for $F_3$ and so on. In summary, we
analyze the execution paths of all flows included in the call-graph of F_{12} to determine their likelihood of being impacted by a change in F_{12}. For the sake of simplicity, hereafter we refer to such flows as client flows.

The first step in algorithm I concerns creating an empty <Key,Value> map, where key is a flow and value is the respective chances of it being impact by a change in F_{i}. We initialize the map inserting the entry <F_{i}, 1.0>. Now we need to determine the order in which we will process each client flow of F_{i}. For instance, in order to calculate the chances of F_{3} being impacted, we need to first calculate the chances of F_{2} being impacted, because F_{3} calls F_{2}. Therefore, the call-graph of F_{i} constraints the order in which the chances of impact need to be calculated. We solve this problem with a calculation of the topological order of the call-graph of F_{i}. One possible topological order for our example is: F_{12}, F_{5}, F_{6}, F_{5}, F_{8}, F_{1}, F_{4}, F_{7} (F_{i} is always the first vertex in the topological order). In lines 04 and 05, we calculate the topological order and remove the first item (F_{i}) respectively. In lines 06-10, we calculate the chances of every client flow being impacted by a change in F_{i} (in topological order). We invoke Algorithm II in line 08, which is responsible for determining the chances of a client F_{j} being impacted by a change in F_{i}. In line 12, we determine the number of map entries that have an impact likelihood larger than or equal to the constant p. In our previous study (Gustavo A. Oliva et al., 2013), we considered parallel lanes as anonymous inner workflows and included them in the map as well. In this new version, the chances of parallel lanes being impacted are attributed to the hosting workflow. In other words, only workflows from the call-graph of F_{i} are now included in the map. As a consequence, the calculation of Impact(F_{i},p) is more precise and realistic.

The first step in algorithm II concerns determining all possible execution paths of F_{j}. More precisely, if F_{j} has n steps, then one valid execution path Q of F_{j} is an ordered list of steps where Q[0] is a start step, Q[n-1] is an end step, and Q[i] is connected to Q[i+1] for 0 ≤ i < n-1. Obtaining all execution paths can be quite complicated in the cases where the flow’s schema includes cycles and parallel lanes (forks/joins). To deal with cycles, we build execution paths such that a certain cycle is not included twice in the same path. As for parallel lanes, we treat each as a separate workflow and only consider the one that has the highest chance of being impacted. After that, we determine the probability with which each execution path will result in a call (either directly or through other client flows) to F_{i} (line 02). We call this measure path impact. We then take the average path impact as a measure to represent the chances of F_{i} being impacted (line 08).

In algorithm III, we show how we calculate path impact. We look at every step included in the path and discover whether it refers to a flow call. If the flow being called is included in the chancesOfImpact map (line 05), it means that such flow is either F_{i} or a client of F_{i}. In this case, the step impact measure is calculated by multiplying the value from the map by a coefficient. This coefficient is determined according to the position of the step in the execution path (line 06). Steps that occur early in the path receive a higher coefficient, while steps that occur late in the path receive a lower coefficient. We took this approach since we believe that the chances of a flow F_{j} being impacted by a flow F_{i} are greater when F_{j} calls F_{i} right in the beginning of its execution. For instance, if F_{i} happens to have a bug and return an incorrect value to F_{j}, then all subsequent steps of F_{j} will be susceptible to wrong behavior. In the extreme case, the first step in F_{j} would be invoking F_{i}. In this case, the position coefficient would be equal to 1. The algorithm then returns the maximum step impact found (line 14).

**Derived Metrics**

To support the analysis of large repositories, we use color schemes to classify flows and sections. The color scheme for flows is as follows. We say that a flow is red when both change scattering
and impact are high. We say that a flow is yellow when either value is high. Finally, we say that a flow is green when both values are low. We define “high” in a relative manner by doing a quartile analysis of the values and picking the extreme outliers. The extreme outliers in a quartile analysis are those higher than \( Q_3 + 3 \times IQR \), where \( Q_3 \) stands for the third quartile and \( IQR \) stands for the interquartile range. Hence, the color of a flow can only be determined by analyzing the whole repository (i.e., both the change scattering and impact distributions are needed). If a certain value in a distribution is not high, then we just consider it low.

In turn, we color sections according to the flows that they contain. If a section contains at least one red flow, it is colored red. Otherwise, if a section contains at least one yellow flow, then it is colored yellow. If a section has only green flows, then it is colored green. If a section has no flows (i.e., it has only subsections), then we color it gray. We also employ color shading to enable visual comparison of sections of the same color. For instance, a red section with 5 red flows will be darker than one with 2 red flows. The same applies to yellow sections. Table III summarizes the color schemes for flows and sections.

<table>
<thead>
<tr>
<th>Color</th>
<th>Flow</th>
<th>Section</th>
</tr>
</thead>
<tbody>
<tr>
<td>Red</td>
<td>High Scattering AND High Impact</td>
<td>Contains at least one red flow</td>
</tr>
<tr>
<td>Yellow</td>
<td>High Scattering XOR High Impact</td>
<td>Contains at least one yellow flow (and no red flows)</td>
</tr>
<tr>
<td>Green</td>
<td>Low Scattering AND Low Impact</td>
<td>Contains only green flows</td>
</tr>
<tr>
<td>Gray</td>
<td>[Not applicable]</td>
<td>Contains no flows (empty section)</td>
</tr>
</tbody>
</table>

**Dispersion.** When most part of yellow and red flows are concentrated in a single repository section, it implies that potentially problematic flows are collocated. This way, it becomes easier to spot which part of the repository should receive more attention. For instance, when red and yellow are dispersed, one needs to say that flow \( F_i \) from section \( S_a \), flow \( F_j \) from section \( S_b \), and flow \( F_k \) from section \( S_c \) need to undergo rigorous testing. On the other hand, when red and yellow flows are collocated, one simply can state that section \( S_i \) needs more testing. Furthermore, different repository sections could be maintained by different teams. In this case, identifying how dispersed red and yellow flows are may reveal how many different teams should be involved in refactoring or testing activities.

We measured the dispersion of red flows by calculating the ratio \( \frac{\text{number of red sections}}{\text{number red flows}} \). If the number of red sections and red flows are the same, it means that each red flow lies in a different section. Hence, we say that the dispersion is 100% in this case. The other extreme is when all red flows lie in the same section. The dispersion of yellow flows is calculated analogously.

**Visualizations**

Our approach relies on two specific visualization techniques, namely call-graphs (Fig. 7) and treemaps (Fig. 8). While call-graphs help address research questions RQ1 and RQ2, treemaps help address research question RQ3. In both visualizations, we apply the color scheme presented in the previous section. In the following, we describe such visualization techniques.

**Graph visualization.** In our approach, we use call-graphs to depict the change scattering of a specific flow. In other words, this visualization shows all the flows in the repository that call a specific one, either directly or indirectly (as exemplified in Fig. 6). This way, before changing a
specific flow, one can first check its change scattering and impact metric values and then investigate which specific flows depend on it.

Fig. 7. Call-graph visualization of flow “[ApplicationService]”

In our implementation, we made the visualization interactive, so that a user can move nodes around the screen, zoom in, zoom out, etc.

**TreeMap.** Treemap is an efficient and compact visualization method that uses nested rectangles to display information with hierarchical characteristics (Shneiderman, 1992). We use treemaps as a means to visualize the color of each repository section. This enables workflow designers to quickly spot repository sections that require more attention.
Given that workflow repositories can be quite large, we decided to use the *squarified layout algorithm* introduced by Bruls et al. (2000). This layout subdivides rectangular areas in a way such that the resulting subrectangles have a lower aspect ratio when compared to the results produced by the original treemap layout algorithm. Consequently, the squarified layout uses space more efficiently and produces rectangles that are easier both to point at in interactive environments and to estimate with respect to size. Finally, in our implementation, we made the treemap visualization interactive, so that one can discover which flows exist within a particular repository section.

**Implementation**

We implemented the approach as a Java 2 SE library and integrated it in HP OO, thus enhancing the tool’s change impact mechanisms. Our library relies on two important frameworks:

**Jung.** The Java Universal Network/Graph Framework (JUNG) is a library that provides a common and extensible language for modeling, analyzing, and visualizing any kind of data that can be represented as a graph or network. We rely on Jung classes and interfaces to implement the graph data structure itself. Hence, the core domain entities of our implementation are built and manipulated using Jung types and algorithms. Furthermore, we relied on Jung’s visualization framework to implement the call-graph visualization. More information about Jung can be found at its website.⁴

**Prefuse.** Prefuse is a Java-based toolkit for building interactive information visualization applications. Prefuse relies on the Java 2D graphics library and supports a rich set of features for data modeling, visualization, and interaction. We used Prefuse to build the interactive treemaps. More information about Prefuse can be found at its website.⁵

**EXPLORATORY STUDY**

We conducted an exploratory study to assess our proposed dependency-centric change impact analysis approach. In summary, we implemented the approach in Java and incorporated it in the HP Operations Orchestration tool, which is an industry tool that supports the authoring, execution, and management of workflows from the IT operations domain. Afterwards, driven by the research questions, we thoroughly analyzed eight workflow repositories, each belonging to an HP OO customer. We also highlighted insights and trends we identified while analyzing the results.

In the following subsections, we present the setup of this study. In particular, we describe the HP Operations Orchestration tool, the way we implemented our approach, and the steps we followed to conduct the analysis of the customer repositories.

**HP Operations Orchestration**

HP Operations Orchestration is a professional industry tool for authoring, executing, and managing IT operations workflows. HP OO also provides a workflow repository out-of-the-box (OOTB) with standard flows and operations to automate common IT processes. HP OO has a broad range of international customers, including Turkcell⁶ and Evergreen⁷ companies. Fig. 9 is a

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⁴ [http://jung.sourceforge.net](http://jung.sourceforge.net)
⁵ [http://prefuse.org](http://prefuse.org)
screenshot of HP OO Studio, which is the module used to author workflows. All out-of-the-box workflows are included below the “Accelerator Packs” folder in the left-hand side of the figure. Workflow categories include: database, network, virtualization, etc. On the right-hand side, the “Power on Virtual Machine” workflow is displayed. More information is available at the product’s website.8

Fig. 9. HP Operation Orchestration: Power on Virtual Machine Workflow

The Study
We applied our approach to eight HP OO customer repositories, which were selected and provided by HP Software. We first characterized each repository by calculating change scattering and impact (p = 0.75) metrics for every flow and then by analyzing the distributions of these metrics using descriptive statistics. Afterwards, we calculated the number and percentage of red, yellow, and green flows of each repository. Analogously, we also calculated the number and percentage of red and yellow sections. Based on the results and insights we obtained, we explored specific repository sections in more detail to uncover which flows should deserve more attention because of their change impact. In the following, we present the results we obtained.

Characterizing the Repositories
To provide an overview of the customer flow repositories, we obtained their number of flows and calculated descriptive statistics for change scattering and impact metrics. We included the HP OO out-of-the-box workflow repository (OOTB) in our analysis, since it serves as a baseline to compare results with. We also highlight that every customer repository includes the out-of-the-box content in its own repository. The results are shown in Table IV.

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7 http://www.evergreensys.com/hp-operations-orchestration-case-study/
Table IV. Customer Repository Overview: Descriptive Statistics for Scattering and Impact

<table>
<thead>
<tr>
<th></th>
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<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>OOTB</td>
<td>1687</td>
<td>434</td>
<td>25.7%</td>
<td>397</td>
<td>5.33</td>
<td>28.35</td>
<td>12.36</td>
<td>159.43</td>
<td>434</td>
<td>25.7%</td>
<td>0</td>
<td>124</td>
<td>1.31</td>
<td>6.81</td>
<td>0</td>
<td>14.88</td>
</tr>
<tr>
<td>C1</td>
<td>1695</td>
<td>441</td>
<td>26.0%</td>
<td>397</td>
<td>5.27</td>
<td>28.13</td>
<td>12.46</td>
<td>162.03</td>
<td>441</td>
<td>26.0%</td>
<td>0</td>
<td>124</td>
<td>1.30</td>
<td>6.76</td>
<td>0</td>
<td>15.00</td>
</tr>
<tr>
<td>C2</td>
<td>1712</td>
<td>449</td>
<td>26.2%</td>
<td>397</td>
<td>5.28</td>
<td>27.88</td>
<td>12.57</td>
<td>164.89</td>
<td>449</td>
<td>26.2%</td>
<td>0</td>
<td>124</td>
<td>1.29</td>
<td>6.70</td>
<td>0</td>
<td>15.12</td>
</tr>
<tr>
<td>C3</td>
<td>1726</td>
<td>471</td>
<td>27.3%</td>
<td>397</td>
<td>5.15</td>
<td>27.22</td>
<td>12.88</td>
<td>173.21</td>
<td>471</td>
<td>27.3%</td>
<td>0</td>
<td>124</td>
<td>1.24</td>
<td>6.55</td>
<td>0</td>
<td>15.47</td>
</tr>
<tr>
<td>C4</td>
<td>1780</td>
<td>471</td>
<td>26.5%</td>
<td>397</td>
<td>5.33</td>
<td>27.30</td>
<td>12.75</td>
<td>170.77</td>
<td>471</td>
<td>26.5%</td>
<td>0</td>
<td>124</td>
<td>1.33</td>
<td>6.67</td>
<td>0</td>
<td>14.71</td>
</tr>
<tr>
<td>C5</td>
<td>1968</td>
<td>624</td>
<td>31.7%</td>
<td>397</td>
<td>7.23</td>
<td>24.92</td>
<td>12.57</td>
<td>181.73</td>
<td>624</td>
<td>31.7%</td>
<td>0</td>
<td>124</td>
<td>1.75</td>
<td>7.43</td>
<td>1</td>
<td>10.97</td>
</tr>
<tr>
<td>C6</td>
<td>2016</td>
<td>497</td>
<td>24.7%</td>
<td>397</td>
<td>5.19</td>
<td>26.68</td>
<td>13.02</td>
<td>178.31</td>
<td>497</td>
<td>24.7%</td>
<td>0</td>
<td>124</td>
<td>1.20</td>
<td>6.47</td>
<td>0</td>
<td>15.45</td>
</tr>
<tr>
<td>C7</td>
<td>2913</td>
<td>1171</td>
<td>40.2%</td>
<td>361</td>
<td>4.09</td>
<td>16.85</td>
<td>17.62</td>
<td>345.20</td>
<td>1171</td>
<td>40.2%</td>
<td>0</td>
<td>124</td>
<td>1.16</td>
<td>4.93</td>
<td>0</td>
<td>16.91</td>
</tr>
<tr>
<td>C8</td>
<td>3769</td>
<td>994</td>
<td>26.4%</td>
<td>428</td>
<td>4.75</td>
<td>21.67</td>
<td>15.31</td>
<td>269.53</td>
<td>994</td>
<td>26.4%</td>
<td>0</td>
<td>130</td>
<td>1.07</td>
<td>5.38</td>
<td>0</td>
<td>16.85</td>
</tr>
</tbody>
</table>

Repository size, in terms of number of flows, ranged from 1687 (OOTB) to 3769 (C8). Hence, we notice that the C8 repository is more than twice as large as the OOTB repository. By looking at the N(%) column of either the change scattering or the impact portions of the table, we observe that C5 and C7 repositories have a distinct high percentage of flows that have at least one client. In other words, flows in these repositories are more interconnected. In turn, the largest change scattering is found in C8’s repository. Moreover, C8 also has the maximum impact value. In other words, C8 has at least one flow that is likely to affect 130 other flows when it is changed.

Regarding change scattering, we notice that C5 repository has a distinct high mean value. Furthermore, its mean impact value is also the highest among all repositories. In fact, it is the only repository whose median value for impact is above zero. At the same time, standard deviation for impact in C5 is also the highest. This indicates that some specific flows might be responsible for the high average impact value. On the other hand, we see that the mean change scattering and impact values for C8 are lower when compared to others. This suggests that despite the high number of flows it has, change impact levels are somewhat controlled in such repository. The lower standard deviation values in this repository also support this conclusion. C7 repository is in an interesting position. Although its mean change scattering is the lowest one, its mean impact is just a little lower than most of others. The standard deviation for impact is also the lowest among all repositories. This suggests that the impact statistical distribution is more uniform in this repository.

Finally, we computed skewness and kurtosis to better understand the shape of the distributions. Qualitatively, a positive skew indicates that the tail on the right side is longer than the left side, the bulk of the values (possibly including the median) lie to the left of the mean, and there are relatively few high values. Change scattering and impact skewness are positive for every customer repository, being particularly high for C7 and C8. Interestingly, impact skewness is much lower for C5, thus providing some evidence that this repository has a larger amount of high values for impact when compared to other repositories. Qualitatively, positive kurtosis indicates that the distribution has a more acute peak around the mean and fatter tails. Change scattering and impact kurtosis are positive for every customer repository, being particularly high for C7 and C8 again. In addition, impact kurtosis is much lower for C5, thus providing more evidence that its impact distribution is different from the others. In summary, by inspecting the values in Table IV, we notice that OOTB, C1, C2, C3, C4, and C6 share similar distributions for both change scattering and impact. Analogously, C7 and C8 are similar to each other. Finally, C5 has
particular distributions for the metrics, showing symptoms that workflow coupling is just starting to become out of control.

**Workflows and their Colors**

In order to further investigate the repositories, we calculated their percentage of yellow and red flows. Differently from the previous characterization, the analysis of workflow colors puts the metrics together and thus provides a more general view of the repository. The results are depicted in Fig. 10.

![Fig. 10. Percentage of red and yellow flows for each repository](image)

Using \(p=0.75\) for the impact metric calculation, we notice than no more than 4% of all flows were classified as either yellow or red in each customer repository. As we suspected, C7 and C5 have the larger ratios of red and yellow flows. Hence, these two repositories are in a more worrying situation when compared to the others.

We also calculated the absolute number of red and yellow flows in each customer repository (Fig. 11). Such number indicates the amount of effort required to maintain and evolve the repositories.
In absolute measures, C7 has the larger amount of red and yellow flows, followed by C8 and C5. More precisely, C7 has 35 flows that have a high change impact (3.5x more than OOTB). Therefore, the team responsible for evolving the C7 repository should devote special attention to a larger number of flows. Interestingly, although C5 repository has almost half of the size of C8 repository, its numbers of yellow and red flows are similar to those of C8. The remaining repositories have similar amounts of yellow and red flows. In particular, the number of yellow and red flows in C2, C1, and C3 are almost equal to that of OOTB. This shows that these particular customer repositories diverge very little in terms of change impact when compared to the baseline represented by OOTB. This is also due to their size, which is very similar to that of OOTB.

**Sections and their Colors**
In addition to analyzing the color of flows, we also quantitatively analyzed the color of sections. The goal is to understand how dispersed yellow and red flows are. Analogously to the previous analysis, we started by calculating the percentage of red and yellow sections for each customer repository. The results are given in Fig. 12.
C7 have distinct large ratios of yellow and red sections. C5, in turn, has a high ratio of yellow sections. The other customer repositories have similar ratios of yellow and red sections. We also calculated the absolute number of red and yellow sections for each customer repository. This analysis indicates how many different sections in the repository deserve more attention in terms of change impact.
Interestingly enough, C7 not only has the highest ratios of yellow and red sections, but also has the largest absolute numbers of yellow and red sections. Following C7, we have C8 and C5. In particular, while C8 has more red sections than C5, it has less yellow sections.

Analyzing the Dispersion of Flows among Sections

Given the results of the previous subsection, we decided to further analyze how dispersed yellow and red flows are. The results for the flow dispersion metric are given in Table V.

<table>
<thead>
<tr>
<th>Client</th>
<th>#Red sections</th>
<th>#Red flows</th>
<th>Red flows dispersion</th>
<th>#Yellows sections</th>
<th>#Yellow flows</th>
<th>Yellow flows dispersion</th>
</tr>
</thead>
<tbody>
<tr>
<td>C5</td>
<td>14</td>
<td>22</td>
<td>63.6%</td>
<td>24</td>
<td>52</td>
<td>46.2%</td>
</tr>
<tr>
<td>C8</td>
<td>19</td>
<td>26</td>
<td>73.1%</td>
<td>22</td>
<td>55</td>
<td>40.0%</td>
</tr>
<tr>
<td>C7</td>
<td>30</td>
<td>35</td>
<td>85.7%</td>
<td>48</td>
<td>79</td>
<td>60.8%</td>
</tr>
<tr>
<td>C6</td>
<td>10</td>
<td>11</td>
<td>90.9%</td>
<td>18</td>
<td>31</td>
<td>58.1%</td>
</tr>
<tr>
<td>C4</td>
<td>12</td>
<td>12</td>
<td>100.0%</td>
<td>14</td>
<td>30</td>
<td>46.7%</td>
</tr>
<tr>
<td>C2</td>
<td>10</td>
<td>10</td>
<td>100.0%</td>
<td>14</td>
<td>26</td>
<td>53.8%</td>
</tr>
<tr>
<td>C1</td>
<td>10</td>
<td>10</td>
<td>100.0%</td>
<td>13</td>
<td>24</td>
<td>54.2%</td>
</tr>
<tr>
<td>C3</td>
<td>10</td>
<td>10</td>
<td>100.0%</td>
<td>13</td>
<td>24</td>
<td>54.2%</td>
</tr>
<tr>
<td>OOTB</td>
<td>10</td>
<td>10</td>
<td>100.0%</td>
<td>13</td>
<td>24</td>
<td>54.2%</td>
</tr>
</tbody>
</table>

While C5 has a large ratio of red and yellow flows (Fig. 10), the results indicate that the dispersion is low for both yellow and red flows. This corroborates our findings from the analysis of Table IV. At the same time, while C7 also has a large ratio of red and yellow flows (Fig. 10), the results indicate the dispersion is much higher than that of C5. Such findings become even more evident when comparing the treemaps of C7 and C5 (Fig. 14). Clearly, yellow and red flows are less dispersed in the C5 customer repository. In the following subsection, we further investigate this repository.
Analyzing the Workflow Repository of C5

According to our previous findings, C5 has a large number of red flows and they are quite concentrated into few repository sections. Taking a closer look at the C5 treemap (Fig. 14), we notice that most part of the red and yellow sections are included in an upper section in the hierarchy called CSA. This implies that most part of yellow and red flows were actually developed by the customer itself. In Fig. 15, we depict the treemap for the CSA section only.

The CSA section treemap reveals a particularly dark red subsection, denoting that such subsection hosts a large number of red flows. By means of the interactive mechanisms we implemented in
the treemap, we discovered that such subsection hosts 9 flows, 7 of which are red. We selected one of these red flows and analyzed it using the call-graph visualization (Fig. 16).

![Call-graph of visualization of the “[ServiceComposite]” workflow](image)

**Fig. 16. Call-graph of visualization of the “[ServiceComposite]” workflow**

The panel in the left-hand side shows interesting information. We can see that the flow’s change scattering and impact have the same value (36). This means that this particular flow is core to the system, as it has a high chance of impacting every flow included in its call-graph. Furthermore, 36 is a very high value for the metrics of change scattering and impact, since the thresholds for being red in this repository are 17 and 4 respectively.

**Threats to Validity and Limitations**

Some factors may have influenced the validity of our study. In the following, we present the threats to the validity of this study, as well as general limitations:

**Assumption regarding workflow schema.** Although we believe our domain model should be complete enough to represent and calculate change impact for most workflows, we acknowledge the missing support for some constructs, including BPMN’s Inclusive Gateways, Complex Gateways, and Events. However, we highlight that our approach does not depend on how workflows activities are implemented (e.g. Web Services, Java standalone applications, etc.), since it relies exclusively on the concepts depicted in the domain model we conceived.

**Analysis scope.** The simplicity, straightforwardness, and flexibility of our approach comes at a cost. We do not take into account data dependencies that might exist in the contexts of intra- and inter-workflow analysis. We neither consider the case in which workflows compete for shared resources.

**Triangulation of results.** We did not conduct a qualitative study with customers to collect their opinion and feedback about the results we obtained. This remains as a future work.
CONCLUSIONS AND FUTURE WORK

Although workflow management systems have emerged as a technical solution that supports the development and control of complex workflows, several challenges still exist. In this paper, we discuss the problem of change impact in the context of workflow evolution. We introduced a static dependency-centric change impact analysis approach that provides metrics and visualizations to assist workflow developers. Furthermore, instead of creating something from scratch, we focused on porting tried-and-true impact analysis techniques from the Software Engineering domain to the area of workflow management. We conducted an exploratory study in which we applied our approach to eight different industrial workflow repositories. We followed a top-down strategy, starting from a repository-wide analysis to a client individual section. The mechanisms offered by our approach triggered a series of insights about the change impact health of each repository and allowed us to compare repositories with each other. We noticed that repositories substantially vary in size (from 1687 to 3769 workflows) and both in the number and percentage of flows with relevant change impact levels (from 34 to 114 workflows and from 2.8% to 6.3% respectively). Repositories also considerably vary in terms of the dispersion of red flows among repository sections (from 63.6% to 100%). We also discovered that most of the yellow and red workflows from the HP OO customer C5 repository were developed by the customer itself. Its repository also had distinguishing high means for the metrics of change scattering (7.23) and impact (1.75), showing symptoms that workflow coupling is starting to get high.

The results we obtained provided some evidence that our approach is both feasible and effective. Indeed, we achieved a level of workflow repository analysis and visualization that is not available in other industry products. At the same time, we acknowledge that a deeper validation of the approach should be conducted by collecting and reasoning about the feedback of the workflow repository owners. In summary, the approach itself and the results of the exploratory study should support researchers seeking lightweight ways to effectively manage large and complex workflow repositories. In practical terms, we think the use of our approach fosters planned changes (as opposed to ad-hoc changes) and ultimately improves the flexibility and reliability of workflow repositories. Finally, we believe our approach contributes to the body of knowledge on static workflow evolution.

Other issues addressed by our implementation and that are not in the scope of this paper include identifying flows that share common steps. By common steps, we mean those that invoke the same flow or operation. Identifying these common patterns throughout the repository leverages opportunities for refactoring and encapsulation, thus increasing the maintainability of the workflow repository. To implement this feature, we relied on the SimPack package developed by Bernstein and Kiefer from the University of Zurich (http://www.ifi.uzh.ch/ddis/simpack.html). As future work, it should be possible to enhance our approach by discovering “zones” in the workflows that might be safe to change, even if it is a red flow. Other improvements could be accomplished by uncovering data dependencies (Kopp, Khalaf, & Leymann, 2008), as well as analyzing data produced during runtime. For instance, workflow execution logs could be mined to discover the number of times each execution path is run for each flow, which could then be used to calibrate the calculation of the impact metric. Finally, we think that combining our approach with existing mechanisms that transparently apply workflow schema changes during runtime would be a major step towards safer and more efficient workflow evolution.
ACKNOWLEDGMENTS

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REFERENCES


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