

Determining the Progress of a Business Object Based on Its Object Instances: An Empirical Study

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Abstract: A fundamental task of any business process monitoring component is to continuously determine the progress of the running processes of an enterprise. This is particularly challenging when facing dynamic processes undergoing changes during run-time, which most likely affect the progress of the respective processes as well. This paper considers object-centric business processes, which consist of business objects and their relations. During run-time, these business objects may be instantiated multiple times to form object instances. The run-time behaviour of these object instances is manifested in terms of object lifecycles that interact with each other. For monitoring a single business object five alternative methods are introduced, which allow determining the progress based on average calculations, information about the semantic object relations (hierarchical order, minimal and maximal cardinality), or event logs (if available). For all methods, the precalculated progress of individual object instances is leveraged. To evaluate the different methods, an empirical study with 65 participants was conducted. As key observation, the majority of the participants that are experienced with process modelling and monitoring tools, prefer deriving the progress of a business object from event logs. The results of this paper are fundamental for determining the progress of a holistic object-centric business process.

1 INTRODUCTION

Monitoring the progress of running processes in real-time (e.g. in terms of a monitoring dashboard) is indispensable for agile enterprises. Monitoring allows economising resources (e.g. human and material consumption), efficiently planning the running process(es), and saving costs and time. With the insights gained through process monitoring future process runs can be optimised. Process progress determination is a fundamental, but often neglected task in business process monitoring. In this paper, the progress determination of a business object in the context of an *object-centric* business process is defined. The latter consists of business objects and their relations. During run-time, multiple interacting object instances are instantiated from these business objects. Furthermore, the run-time behaviour of these object instances are defined by their object lifecycles.

Process progress determination is challenging due to the multitude of object instances that may be cre-

ated (and eventually be deleted afterwards) during the execution of an object-centric process. Consequently, for a real-world entity, hundreds or thousands of corresponding object instances and their interactions with instances of other business objects may exist. In addition to this challenge, dynamic process changes (e.g. in the context of a process model evolution) might become necessary affecting the progress of running object instances as well (Andrews et al., 2021). Furthermore, at build-time, the total number of the object instances involved in an object-centric business process is not fully known, but evolves during run-time, varying across the object instance of an object-centric process. Consequently, the behaviour of the overall business process (i.e. the object-centric process) may vary as well (Steinau et al., 2021).

To tackle the challenge of progress determination, a bottom-up approach is pursued that starts with the determination of a single object instance and leads to the determination of the overall progress of a business process. This paper deals with determining the progress of one business object (e.g. object *Application* of a recruitment process) based on the precalculated progress of its object instances (Arnold et al., 2021) (e.g. *Application1* and *Application2*) in the

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context of an *object-centric* business process (i.e. a collection of concurrently executed, but independent object instances). To address this challenge five alternative methods are developed and compared. These methods are based on the average progress of context-related object instances, information about semantic object relations (hierarchical order, minimal and maximal cardinality), or event log data (i.e. the process history). The information (e.g. log data) on which these methods are based is not always available.

Sec. 2 provides background information about object-centric processes and sets out the research context of determining the progress of an object-centric business process. Sec. 3 then introduces five alternative methods for determining the progress of one business object based on the progress of its context-related object instances. Sec. 4 describes the research design of the conducted evaluation study to assess the five methods. Sec. 5 analyses and evaluates the results of this study. Sec. 6 addresses related work and Sec. 7 concludes the paper.

2 BACKGROUNDS AND RESEARCH CONTEXT

In the object-centric process management paradigm a business process is described in terms of interacting business objects that correspond to real-world entities. An implementation of this paradigm is provided by the PHILharmonicFlows framework, which enables dynamically evolving object-centric business processes that allow for both build-in flexibility and ad-hoc process changes during run-time (Andrews et al., 2021). The relations between the business objects, including their cardinalities and hierarchical structuring, are manifested by the Relational Process Structure (RPS) (Steinau et al., 2018). The RPS corresponding to a recruitment business process is shown in Fig. 1. At run-time, Object Instances (e.g. *Job Offer1* to *Job Offer4*) of a business object (e.g. *Job Offer*) are created and organised in a relational instance structure. An example of such a relational instance structure is shown in Fig. 2.

2.1 Research Context and Research Questions

Determining the progress of an object-centric business process as depicted in Fig. 2 is a challenging task. First of all, it should be possible to determine the progress of a single object instance (e.g. *Job Offer1*) based on its lifecycle process. This task has already

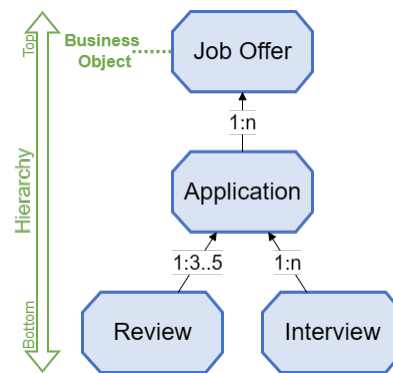


Figure 1: RPS at design-time with its *Business Objects*.

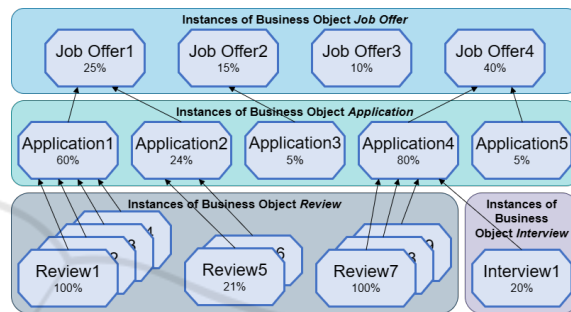


Figure 2: Relational instance structure at run-time.

been addressed by us in (Arnold et al., 2021). The method we presented in this work is based on a one-dimensional Kalman Filter which is used to determine the progress of a single object instance. Fig. 2 shows an example of an object-centric business process emphasising the progress of object instances depending on the progress of their subordinated object instances. The aim of this paper is to determine the progress of one business object (e.g. object *Review*) based on the precalculated progress (cf. (Arnold et al., 2021)) of its object instance (e.g. *Review1* to *Review9*). Note that in object-centric processes, an arbitrary number of object instances may be created or deleted during run-time, which turns the task of determining the progress of such processes into a challenging endeavour:

- For different business objects there may be a varying number of object instances and corresponding object instance interactions.
- The total number of object instances is not completely known at build-time, but dynamically evolves during run-time.
- Due to dynamic changes of an object-centric process (e.g. to add or delete object attributes or to change cardinalities of object relations) the behaviour of different object instances of an object-centric process may vary significantly.

To tackle these challenges, the following research question (RQ) is considered:

RQ: How can the progress of one business object with multiple object instances be determined in the context of an object-centric business process?

First, different methods for determining the progress of one business object based on the already precalculated progress of its object instances are presented (cf. Sub-RQ 1). Second, the different methods are investigated and compared in an empirical study (cf. Sub-RQ 2). The study focuses on the suitability of the progress determination methods and elaborates on whether they match human intuition.

Sub-RQ 1: What alternative methods exist to determine the progress of one business object based on the progress of its context-related object instances?

Sub-RQ 2: Which of these methods is suited best for users, i.e. matches human intuition best?

3 PROGRESS DETERMINATION METHODS

Determining the progress of a varying number of object instances created in the context of one business object can be accomplished in various ways. More specifically, progress calculations may be based on the given RPS (cf. Sec. 2), on average calculations, or on event log data (i.e. the process history). This section introduces five alternative methods for determining the progress of one business object ($progress_{method}$) based on the precalculated progress (cf. (Arnold et al., 2021)) of its object instances.

3.1 Method 1: Total Average

The most intuitive method to determine the progress of one business object (e.g. *Review*) is to calculate the average (AVG) progress of its object instances. Therefore, the precalculated progress $prog_i$ of each object instance i is added up and the resulting number is divided by the total number I of these object instances. This calculation is formalised by Formula (1).

$$progress_{AVG} = \frac{\sum_{i=1}^I prog_i}{I} \quad (1)$$

with $prog_i \in \{x \mid x \in \mathbb{R}, 0 \leq x \leq 100\}$
and $I \in \mathbb{N}^+$

3.2 Methods 2 and 3: Cardinalities

At design-time, each object is related to at least one other object in the RPS. Corresponding relations are equipped with a $1:n$ cardinality ($n \in \mathbb{N}$). Thus, for a particular parent object, many child objects may be created at run-time. Moreover, the respective cardinality may be restricted by a minimum and maximum. For example, in Fig. 1 the cardinality between parent object *Application* and its child object *Review* is given by $1:3..5$. This means that, only a maximum number of five *Review* instances may be created in the context of one given *Application* instance at run-time and a minimum number of three object instances are required to terminate the respective business process. Note that the $m:n$ cardinality ($n, m \in \mathbb{N}$) does not exist in our modelling tool PHILharmonicFlows.

3.2.1 Method 2: Minimum Cardinality

When using the minimal cardinality (MIN), the progress is determined most optimistically by expecting the minimum number of possible object instances to be executed, as defined by the minimum cardinality. More specifically, the precalculated progress $prog_i$ of all object instances i (with the total number of instances I) are added up and divided by the result of adding up the maximum $max()$ of the minimum cardinality c_{min} or the number of created object instances $I(p)$ for each parent object p (with the total number of parent objects P). This calculation is formalised by Formula (2).

$$progress_{MIN} = \frac{\sum_{i=1}^I prog_i}{\sum_{p=1}^P \max(c_{min}, I(p))} \quad (2)$$

with $prog_i \in \{x \mid x \in \mathbb{R}, 0 \leq x \leq 100\}$
and $I, P, c_{min}, I(p) \in \mathbb{N}^+$

3.2.2 Method 3: Maximum Cardinality

When using the maximum cardinality (MAX), the progress is determined most conservatively by expecting the maximum number of possible object instances to be executed, as defined by the maximum cardinality. More specifically, the precalculated progress $prog_i$ of all object instances i (with the total number of instances I) are added up and divided by the product of the maximum cardinality c_{max} and the total number of parents P . This calculation is formalised by Formula (3)

$$progress_{MAX} = \frac{\sum_{i=1}^I prog_i}{c_{max} * P} \quad (3)$$

with $prog_i \in \{x \mid x \in \mathbb{R}, 0 \leq x \leq 100\}$
and $I, c_{max}, P \in \mathbb{N}^+$

3.3 Method 4: Event Logs

Event logs (LOG) can be used to predict the expected number of object instances E that will be created for a business object in the context of their common parent object instance. For this purpose, all finished parent object instances (of either terminated or running processes) are continuously analysed and the expected number of created object instances is determined (e.g. based on average or machine learning). To calculate the progress based on this estimation, the precalculated progress $prog_i$ of all object instances i (with the total number of instances I) are added up and divided by the result of adding up the maximum $max()$ of the expected number of object instances E or the number of created object instances $I(p)$ for each parent object p (with the total number of parent objects P). Note that this method may only be used if an event log exists and the accuracy of the prediction depends on the quality of the log. This calculation is formalised by Formula (4).

$$progress_{LOG} = \frac{\sum_{i=1}^I prog_i}{\sum_{p=1}^P max(E, I(p))} \quad (4)$$

with $prog_i \in \{x \mid x \in \mathbb{R}, 0 \leq x \leq 100\}$
and $E \in \mathbb{R}$ and $I, P, I(p) \in \mathbb{N}^+$

3.4 Method 5: Parent Object

The Parent Object method (PO) combines two concepts:

1. Considering the object instances created for a business object in the context of each parent object instance individually (as in MIN, MAX, and LOG).
2. The average calculation (as in AVG).

For this, first the average of the progress $prog_{i \in p}$ of all object instances $I(p)$ created for a business object in the context of each parent object instance p (e.g. all reviews of the same application) is calculated individually. Second, the average of these results is calculated (with the total number of parent object instances P) to determine the overall progress of the business object (e.g. object *Review*). Note that $prog_{i \in p}$ numbers the progress of the instances i for each parent object instance individually. This calculation is formalised by Formula (5).

$$progress_{PO} = \frac{\sum_{p=1}^P \frac{\sum_{i=1}^{I(p)} prog_{i \in p}}{I(p)}}{P} \quad (5)$$

with $prog_{i \in p} \in \{x \mid x \in \mathbb{R}, 0 \leq x \leq 100\}$
and $I(p), P \in \mathbb{N}^+$

4 RESEARCH METHOD

This section summarises the research method underlying the empirical study we conducted to assess the five progress determination methods from a human perspective. In detail, this section focuses on the data collection method, the study design and structure, and the data analysis method.

4.1 Data Collection

How users perceive the progress of a business object based on the five progress determination methods is investigated in an empirical study. For conducting this study and collecting data, the web-based tool Unipark is leveraged. The study is performed based on an anonymous online questionnaire¹ and is available in both English and German language. Moreover, both language options do not differ with respect to content or structure. The questionnaire was available over a period of one month.

4.2 Study Execution

The empirical study is structured in five parts with a total of 35 questions. For this purpose, dichotomous, semi-open multiple choice questions, and open questions are used. Moreover, some of the questions are identical, but refer to different backgrounds.

Demographics and Experience. In the first part of the study, demographic data is queried from all participants. This includes information like gender, age, current profession, and professional field. Moreover, the experiences of the participants in respect to process modelling and (business process) in general are queried.

Training. The second part offers a training session for about half of the participants. The other half does not participate in any training session. For selecting the participants a random function is used. The training includes the explanation of the progress determination methods as well as the study structure.

Perception of Progress. The main part of the study aims to assess the perception of progress for three different scenarios using the described progress methods (cf. Sec. 3). For each scenario multiple questions with a varying number of object instances are given.

¹Questionnaire and responses of the 65 participants are available on Researchgate: https://www.researchgate.net/publication/378140057_Determining_the_Progress

Details about the scenarios.

- 1.) *Evaluation of research papers for a conference.*
All papers shall be evaluated by 3 to 5 reviewers to decide whether or not the paper can be accepted.
- 2.) *Grade bonus for students.* To receive the grade bonus for a particular exam, students need to achieve at least 80% of all exercise points from the 12 to 15 exercise sheets.
- 3.) *Recruitment process.* A company has published a large number of vacancies. Depending on the job offer, there are many, few, or no applicants.

With 14 semi-open multiple choice questions (i.e. seven for Scenario 1, three for Scenario 2, and four for Scenario 3), the participants have to choose their most appropriate progress calculated for each question of the given scenario. Additionally, for each progress, the calculated progress determination method is provided (i.e. "a.) 85% (AVG)"). Moreover, if none of the given methods match the participants intuition of progress, the participants may indicate their perception of progress with a short explanation.

Cognitive Strain and Behaviour during Participation. Each scenario is completed with the following three questions to investigate the behaviour of the study participants.

- 1.) Were you able to answer the previous questions clearly? If not, why?
- 2.) Did you change answers from previous questions when answering this form sheet? If yes, why and where?
- 3.) Have you chosen different methods for different scenarios? If yes, why?

Ranking the Methods: Additional to the scenarios, the participants rated the progress determination methods in a "drag-and-drop" like manner, with ranking the method they consider being most suitable at the top. As not all methods are always applicable to determine progress, the following three combinations are considered to cover the most common real-world conditions.

Available methods:

Case 1: AVG, MIN, MAX, LOG, and PO

Case 2: AVG, MIN, MAX, and PO

Case 3: AVG, MIN, and LOG

First, all methods are available for the participants to define a suitable ranking. Second, all methods except LOG is given to find a suitable ranking if no event log exists. Third, MAX and PO are not available, as the maximum cardinality is often not defined for top-level objects and the PO is not applicable to top-level

objects in general. Additionally, an optional text field is given to describe a method or procedure that does not corresponds to the five introduced progress determination methods.

Language: Finally, the participants are asked, which language version they read to verify that no differences exist in the translation and wording.

4.3 Data Analysis

The study structure and its data analysis and validation are generated on the checklist of the empirical cycle described in (Wieringa, 2014). Furthermore, all collected data of the questionnaire are analysed and evaluated based on the methodology presented in (Wieringa, 2014; Brace, 2018). The aim of the evaluation is to find a representative answer that reflects the opinion of the participants. In detail, an exploratory analysis is applied that uses *Cross Tabulations* to compare the quantitative results from different participant groups. In addition, open-ended questions are extended to investigate the cognitive strain and behaviour during participation. In the evaluation, different participant groups are defined according to their background and prior knowledge. The division into groups allows to investigate variations in the perception of progress and the associated choice of progress determination methods depending on the background knowledge of the individual participants.

5 EVALUATION

Overall, 65 participants completed the questionnaire of the empirical study. 5 of them were excluded as they had problems understanding the tasks (answering the questions as "not clear" at cognitive strain part). The following evaluation is based on the answers of the remaining 60 participants.

5.1 Demographics and Experience

In total, more male (39 | 65%) than female (19 | 31.7%) participants took part. The remaining participants define their gender as non-binary (2 | 3.3%). Altogether, the participants are between 19 and 34 and on average 25.13 years old.

Most participants have their profession field in the MINT (Mathematics - Computer Science - Natural Science - Technology) (51 | 85%). The remaining participants have a background in economics (9 | 15%). Thereby, most participants are students (43 | 71.7%) studying either in a MINT (34 | 79.1%) or an economics program (9 | 20.9%). The second major share

is given by academics (14 | 23,3%) in MINT. The remaining participants are working in industry with focus in MINT (3 | 5%). With this group of participants, confounding variables (as e.g. general school education or major age differences) could be avoided as far as possible.

Exactly one quarter (15 | 25%) of the participants have no experiences with process modelling, most of them are students (14 | 93.3%). Half of the participants (30 | 50%) are not familiar with monitoring tools, 27 of them are students (27 | 90%). More than one third (21 | 35%) of the participants have experiences with business process monitoring tools. This group is also familiar with process modelling and monitoring tools and define the expert group. In this expert group, the share of male (10 | 47,6%) and female (11 | 52.4%) is similar. On average they are 26.8 years. Most of them have their professional field in MINT (18 | 85.7%) the remaining in economics (3 | 14.3%). Most of them are academics (11 | 52.4%). The remaining participants are divided in students (9 | 42.8%) and others (1 | 4.8%). The non-experts are composed of all participants except the experts. In the following, the results of the experts (E) are compared with the ones of the non-experts (NE) to enable a profound analysis.

5.2 Perception of Progress

The perception of progress differs significantly between the experts and non-experts (cf. Tab. 1). The biggest difference can be found in the number of participants that chose the progress calculation based on LOG. In each of the 14 perception progress questions, at least 71.4% (up to 85.7%) of experts chose the progress based on LOG. In contrast, the non-expert selected this method only with at least 23.1% (up to 38.5% in the second scenario and up to 51.3% in the third scenario). In 5 of 13 cases, where AVG and LOG are used, the non-experts opted for AVG. In the other 8 cases, the majority favoured LOG. Consequently, LOG has been mostly chosen, when object instances are created from more than one parent object.

In the first scenario, the answers of the non-experts are approximately evenly distributed between the progress determination methods. The methods AVG, MIN, and PO are chosen less and less over the questionnaire (Scenarios 2 and 3). Considering the non-experts for the first scenario at most 10.3% preferred the progress based on MAX. However, in the second scenario this method was selected by up to 28.2% of the non-experts. This may be due to the fact that in the first question of Scenarios 2 only 3 of at least 12 (up to 15) exercise sheets have been com-

pleted and the progress is calculated far too high when choosing AVG. Accordingly, AVG was chosen significantly (above 10%) less. As another observation that emerged from the study responses, PO is not preferred by experts. In comparison, up to 20.5% of the non-experts chose PO. Furthermore, the number of participants choosing the same method for all questions of one scenario is increasing significantly from the first to the third scenario (to threefold) by the non-experts. In general, it can be observed that non-experts do not have a clear preferred method. In contrast, LOG was preferred by the experts in most cases (more than 80% in Scenarios 1 and 2, more than 70% in Scenario 3). The last question of scenario 3 addresses the issue of how to define progress if no object instance is created. In this case, both groups answer very similar. Above 70% voted for 0% and the remaining for 100%. Additionally, one expert remarked that no progress should be assigned in this case and the progress of the parent object should be used instead.

5.3 Ranking the Methods

In this part, the participants rank the progress determination methods considering their suitability. Therefore, the position of each method is represented by a number. For example, 1 is assigned to the top-rated method. Tab. 2 shown, the average position for each method for both groups separately.

PO is rejected by most participants in both groups. For example, non-experts rated it with 4.2 out of 5 (whereas 5 is most unsuitable) and the experts rated it with 4.6 out of 5 in the first case. The remaining ranking has shown similar results. Note that the number of available methods differs in Case 2 and 3. Furthermore, MIN is evaluated very similarly by both groups and is rejected, next to PO, as the second most unsuitable method. Compared to MIN, MAX is more suitable. In general, most participants preferred a more conservative progress calculation.

Both groups differ in their choice of the most suitable method. The non-experts selected AVG (1.6 to 2.1) as their favoured method in each case and the experts chose LOG (1.3). Note that AVG only considers existing object instances (snapshot) whereas LOG also considers object instances that are expected to be created in the future (big picture). This supports the statement that the experts have the big picture of a business process in mind and not only consider the snapshot of the current process. In contrast, the non-experts are more focused on the snapshot as on the big picture. Due to this fact, the experts evaluated MAX (1.9) as the best suitable option, that should be used when LOG is not available. The non-expert's

Table 1: Study results in percent for each question of the three scenarios. Table entries marked with “-” were not part of the response options in the study.

	Scenario 1							Scenario 2			Scenario 3		
	Q1	Q2	Q3	Q4	Q5	Q6	Q7	Q1	Q2	Q3	Q1	Q2	Q3
AVG NE	33.3	48.7	35.9	20.5	28.2	20.5	35.9	20.5	25.6	17.9	56.4	48.7	46.2
E	9.5	4.8	9.5	4.8	4.8	4.8	4.8	4.8	9.5	9.5	23.8	28.6	28.6
MIN NE	30.8	17.9	23.1	28.2	15.4	17.9	7.7	12.8	17.9	20.5	-	-	-
E	4.8	9.5	0	4.8	4.8	4.8	4.8	4.8	4.8	0			
MAX NE	7.7	10.3	10.3	10.3	7.7	7.7	7.7	28.2	25.6	20.5	-	-	-
E	4.8	4.8	9.5	0	0	0	0	9.5	4.8	4.8	-	-	-
LOG NE	28.2	23.1	30.8	25.6	33.3	33.3	33.3	38.5	30.8	33.3	41	51.3	48.7
E	81	81	81	85.7	81	85.7	85.7	81	81	85.7	71.4	71.4	71.4
PO NE	-	-	-	15.4	15.4	20.5	12.8	-	-	5.1	-	-	5.1
E	-	-	-	0	0	0	0	-	-	0	-	-	0

Table 2: Average position in the ranking for the three given cases. Results are coloured to visualise the better or worse rating comparing both groups.

	Case 1					Case 2				Case 3		
	AVG	MIN	MAX	LOG	PO	AVG	MIN	MAX	PO	AVG	MIN	LOG
NE	2.1	3.4	3.0	2.2	4.2	1.8	2.8	2.2	3.3	1.6	2.6	1.8
E	3.0	3.5	2.7	1.3	4.6	2.1	2.5	1.9	3.5	2.2	2.5	1.3

choice of the most appropriate method (AVG) is always available. Considering the distribution of the average position, the range of the experts varies largely (e.g. 1.3 to 4.6 for Case 1). Consequently, most participants rank the methods in the same way. In contrast, the range of the non-experts is smaller (e.g. 2.1 to 4.2 for Case 1). This indicates that the selected ranking of the methods differs among the non-experts, which leads to similar results for different methods. For example, in Case 1, AVG results in 2.1 and LOG in 2.2 on average. In the following, the total ranking for both experts and non-experts are given. Note that this ranking is the same for all three cases independent of non-available methods.

Non-expert: $AVG > LOG > MAX > MIN > PO$

Expert: $LOG > MAX > AVG > MIN > PO$

5.4 Training and Language

In total, 45 (69.2%) of the participants took part in a previous training. In general, no differences could be found between participants with and without training. Furthermore, only 14 (21.5%) participants read the English questionnaire whereas about one quarter are experts (4 | 28.6%). The distribution of experts and non-experts regarding the language choice is almost identical to their distribution and no differences could be found in the answers.

5.5 Limitation of the Study

In the first scenario, participants often chose different methods for the individual questions. However, as the study progressed, participants increasingly chose only one or two methods. As the order of the scenarios was the same for each participant it is unclear whether this observation is based on a learning progress of the participants or on the scenarios themselves. Furthermore, the scenarios were rather simple to allow for a better understanding. However, the transferability of the results to complex scenarios was not investigated.

6 RELATED WORK

The current research in the field of OCEL (object-centric event log) (Ghahfarokhi et al., 2021) allows an event log to be related to multiple objects that means each row is given by one object and includes, for example, its identifier and type of object. Traditionally, an event log is event-based that means each row in a table has at least an identifier, an activity, a timestamp, and related objects. The OCEL is similar to the event log generated from our PHILharmonicFlows framework. Our approach can be applied to OCEL if the progress of the individual instances is given. This precondition also exists for object-centric business pro-

cesses, but was already addressed in (Arnold et al., 2021). In (Gherissi et al., 2022), an approach for predictive process monitoring based on OCELS is discussed. In general, this approach improves the accuracy in predicting the next activity and the MAE (Mean Absolute Error) in time prediction compared to the conventional event logs by utilising the interaction between objects.

7 SUMMARY AND OUTLOOK

In this paper, two research questions were addressed to determine the progress of a collection of object instances created in the context of one business object. Regarding Sub-RQ 1, five possible progress determination methods were introduced and formally defined. To address Sub-RQ 2, an empirical study was conducted that investigates the most suitable progress determination method. In this context, progress determination was considered for three different scenarios. In addition, a method ranking was created. This ranking helps to decide which progress method will be executed first, and if this method is not an option due to non-existing conditions (e.g. the event log for LOG) which is the next favoured. For the evaluation, an expert group, which comprises the participants with experiences in process modelling and business process modelling tools, and a non-expert group, consisting of the remaining participants, are defined. In general, non-experts have no clearly preferred method as they prefer both AVG and LOG. The experts agree in all cases (with more than 70%) and consider LOG to be the most appropriate. Considering this result, LOG should be used to determine the progress method. When the information required by this method is not available progress determination should be based on MAX. Finally, if the maximum cardinality is not implemented, AVG should be used.

Future research is needed to provide fully functional progress calculation in a monitoring tool for object-centric business processes. In addition, the usability and accuracy of their resulting outcomes will be tested directly at the monitoring tool by end-users. For example, a *Delphie* study with focus on experts more complex scenarios might provide additional insight into the usability and human intuition of progress determination. The run-time behaviour (e.g. response times) of the individual methods is examined in order to check their suitability directly in our monitoring tool. The combination of methods is also being investigated in order to achieve more precise progress.

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REFERENCES

- Andrews, K., Steinau, S., and Reichert, M. (2021). Enabling runtime flexibility in data-centric and data-driven process execution engines. *Information Systems*.
- Arnold, L., Breitmayer, M., and Reichert, M. (2021). A one-dimensional kalman filter for real-time progress prediction in object lifecycle processes. In *2021 IEEE 25th International Enterprise Distributed Object Computing Workshop (EDOCW)*, pages 176–185. IEEE.
- Brace, I. (2018). *Questionnaire design: How to plan, structure and write survey material for effective market research*. Kogan Page Publishers.
- Ghahfarokhi, A. F., Park, G., Berti, A., and van der Aalst, W. M. (2021). Ocel: A standard for object-centric event logs. In *European Conference on Advances in Databases and Information Systems*, pages 169–175. Springer.
- Gherissi, W., El Haddad, J., and Grigori, D. (2022). Object-centric predictive process monitoring. In *International Conference on Service-Oriented Computing*, pages 27–39. Springer.
- Steinau, S., Andrews, K., and Reichert, M. (2018). The relational process structure. In *Int. Conf. on Advanced Information Systems Engineering*, pages 53–67. Springer.
- Steinau, S., Andrews, K., and Reichert, M. (2021). Coordinating large distributed relational process structures. *Software and Systems Modeling*, 20(5):1403–1435.
- Wieringa, R. (2014). *Design science methodology for information systems and software engineering*. Springer.