



An Empirical Investigation of the Intuitiveness of Process Landscape Designs

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Abstract. Process landscapes define the scope and relationships between an organization’s business processes and are therefore essential for their management. However, in contrast to business process diagrams, where nowadays BPMN prevails, process landscape diagrams lack standardization, which results in numerous process landscape designs. Accordingly, our goal was to investigate how intuitive are current landscape designs to users with low expertise, as well as users having expertise in BPMN and landscape modeling. A total of 302 subjects participated in the research showing that previous expertise impacts the interpretation of landscape elements and designs whereas, in the case of having contextual information, subjects responded more consistently. The results also show that the basic relationships between processes are intuitive to users, also in the case when only proximity between shapes is facilitated. Our findings may imply future designs of languages for process landscapes. They also may be useful for those who actually model process landscape diagrams and search for suitable notations.

Keywords: Process landscape · Diagram · Semantic transparency · BPMN

1 Introduction

A process landscape diagram represents the top-level part of a process architecture, which stands for “*an organized overview of business processes that specifies their relations, which can be accompanied by guidelines that determine how these processes must be organized*” [1]. The specification of a process landscape model is the most important challenge for the establishment of a process architecture [2]. For example, a high amount of organizations’ services requires an efficient representation of the entire

process landscape in order to measure the overall organization's potential [3]. A process landscape represents processes as 'black-boxes' and so focuses on interrelationships between processes and external participants. In this manner, a process landscape enables an organization to maintain an overview of processes, which simplifies process-related communication and may represent a starting point for process discovery.

Accordingly, a process landscape model has to be comprehensible by all major stakeholders of an organization [2, 3]. This implies the usage of a common, compact, and intuitive language for the creation of process landscape diagrams. However, no standardized languages for creating process landscapes exist [4], whereas BPMN 2.0 does not cover the wide landscapes and complexities that exist in the process-modeling domain [5, 6]. Consequently, organizations, as well as process modeling tool vendors (e.g., ARIS Express, Visual Paradigm, Vizi Modeler and Signavio), define their own 'overviews of processes' most commonly by imitating 'value chain' diagrams.

As a result, landscape diagrams differ from each other, and while there is no common landscape modeling language, an inexperienced user could infer a different meaning from the appearance of a language element, which could negatively impact the comprehension of a diagram and the corresponding decisions made. And while the graphical representation significantly impacts the cognitive effectiveness of a diagram [7–9], it is important to specify a common palette of comprehensible symbols fitting with the process landscapes domain.

According to these challenges, the main goal of our work was to investigate the intuitiveness of the representations of process landscape designs as found in academia and industry, i.e., to test if representations of landscape concepts are intuitive (i.e. semantically transparent, clear) to people with 'near-to-zero' knowledge of a process landscape design. In this light, we defined the following research questions which could be tested empirically:

- *RQ1: Are common landscape designs semantically transparent to 'novice users'?*
- *RQ2: How does the previous knowledge impacts the comprehension of process landscape designs?*

Accordingly, the paper is organized as follows. The introduction chapter already identified the problem and motivation for the research, whereas the second chapter contains background on process landscapes, semiotics, and semantic transparency. The third chapter introduces empirical research, which was applied to provide answers to the stated research questions. The fourth chapter presents and discusses the results of our research. The last chapter brings conclusions and limitations of our approach as well as implications and future research directions.

2 Research Background

2.1 Process Landscapes

A high-level model of an organization that represents an overall structure of business processes and their relationships, emerged as a tool to aid process-oriented companies in managing large business process collections [10]. With roots in the early 1980s,

when Porter [11] introduced the value chain model, the concept is commonly specified as a ‘process landscape’ and represents a set of interconnected processes within an organizational system. Alternative terms in use are ‘process overview’ [12], and ‘process map’. However, according to the findings of Poels et al. [13], the term ‘process map’ may either represent a model of a business process architecture or an entry-level model of a business process model architecture.

A process landscape model (Fig. 1) shows the structure, grouping, modularity, functionality, and technology of chain processes, business processes, and working processes. In contrast to business process models, processes on the landscape level are modeled as ‘black-boxes’ whose internal complexity is hidden for the sake of simplicity and clarity.

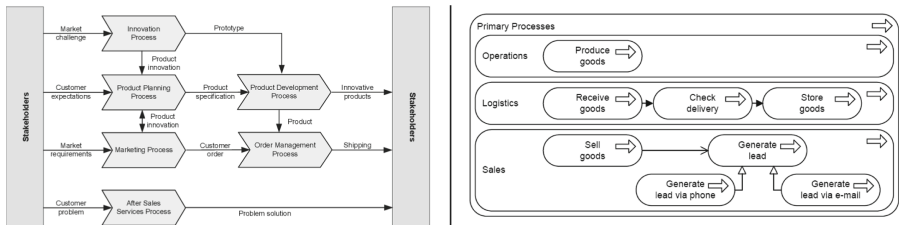


Fig. 1. Examples of process landscape diagrams [15, 16]

Process landscape diagrams may be used in numerous ways, addressing the concerns of business-oriented users as well as technically-oriented ones [12]. While being specified on the macro level, they provide a comprehensive understanding and highlight different types of relationships or dependencies with other processes and artifacts [14]. Process landscape diagrams help process owners, quality managers, and other process-related stakeholders to ease the maintenance of their processes by offering a quick overview of processes. Afterward, in detailed process diagrams, individual business processes may be decomposed into finer levels of detail (i.e., sub-processes and tasks). In summary, like modeling individual processes is a starting point for any process improvement effort, modeling the architecture of an organization’s collection of business processes is required for any analysis, design or improvement effort that transcends the level of individual processes [13].

Figure 1 represents two common process landscape diagrams, with processes depicted as chevron arrows (left) or rectangles (right), whereas arrows represent between-process relationships. The left diagram in Fig. 1 additionally connects organizational processes with the environment by specifying connections to external participants (i.e., stakeholders).

2.2 Modeling Process Landscapes with BPMN 2.0

The evidence from academia and practice show that BPMN is used for modeling of process landscapes in an informal way [17, 18]. In our previous work [18], three different BPMN-based approaches for modeling of process landscaped have been investigated: (1) use of black-box Pools and Message flows for modeling of BPMN Collaboration

diagrams with hidden details; (2) use of BPMN Collaboration diagrams and (3) use of Enterprise-wide BPMN Process diagrams.

An analysis performed in [18] demonstrates that none of the BPMN approaches results in diagrams with a graphical similarity to common landscape diagrams (e.g., value chains). Analytically, this was confirmed by Malinova et al. [19], who performed a semantical mapping between BPMN and ‘Process maps’. Their results show that BPMN in its current form is not appropriate for process landscape design.

2.3 Semiotics

The theoretical foundation on how a visual vocabulary of a notation is defined can be explained by semiotics [20], i.e. a study of signs, an investigation into how the meaning is created and communicated. According to semiotics, a sign consists of a signifier (i.e. any material thing that is signified, be it an object, words on a page, or an image) and signified (i.e. the concept which the signifier refers to) (Fig. 2, left).

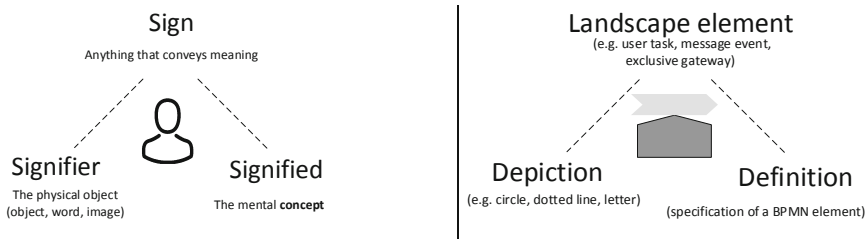


Fig. 2. Main concepts as specified in semiotics (left) and OMG’s namespace (right)

Based on the relation between the signifier and signified, semiotics defines three types of signs: (1) icon, where a signifier physically resembles the signified (i.e. person sign on Fig. 2); (2) symbol, where the signifier presents the signified with an arbitrary or conventional relation; and (3) index, where the signifier is related to the signified by an associative relation (i.e. Fig. 2, right – the darker symbol supports the lighter one (from below), which is analogous to common real-life situations). In the process languages’ space (e.g. BPMN, CMMN, DMN), a sign is commonly referred to as an element, whereas the signifier is commonly referred to as a depiction of an element [21]. The definition of a process element has an equal meaning as a ‘signified’ in semiotics, meaning the specification of a language concept. Since the focus of our investigation is on process languages, we will use the terms according to the process languages namespace, i.e. a ‘*process element consists of its definition and depiction*’.

2.4 Semantic Transparency

Caire et al. [22] stated that “*The key to designing visual notations that are understandable to naïve users is a property called semantic transparency*”, which means that the meaning (semantics) of a sign is clear (i.e. intuitive, transparent) from its appearance alone

(as in the case of onomatopoeia in spoken languages). Therefore, addressing semantic transparency is recognized as one of the most powerful approaches for improving the understandability, especially for novice users [22].

Semantic transparency of a sign is a continuous function with two endpoints. On the positive side, a sign may be semantic transparent, which means that a novice reader could accurately deduce the meaning of a sign from its appearance (e.g., a drawn tree representing a tree). Semantically transparent signs tend to be defined either by similarity or an associative relationship (index). In contrast, semantic perversity means that a novice reader would likely deduct an incorrect meaning from the sign's appearance (e.g., an arrow directed in an opposite way to an actual flow). On the midpoint, a sign may be defined as semantically opaque, which means that it is defined by a convention [22]. Semantically transparent signs reduce cognitive load because they have built-in mnemonics: as a result, their meaning can be either perceived directly or easily learned [23]. Such representations speed up recognition and improve intelligibility to naïve users [24, 25]. Indeed, one of the main challenges of modeling languages is to model the diagram in a precise and user-friendly way, where each applied graphical element should be intuitive for users [26], which positively implies the acceptance of a modeling technique [27].

3 Empirical Research

3.1 Research Model and Process

In respect to *RQ1*, we identified two independent variables, namely 'depiction of a landscape concept' and 'contextual information'. In the case of the former, we searched for the levels of the independent variable among the various sources of process landscape diagrams whereas the visual vocabularies of the corresponding notations have been identified (i.e. levels represented common landscape symbols). The 'contextual information' variable was additionally introduced, since the meaning of an individual element may be more precisely identified when putting it into the context (element level and diagram level). The dependent variable was defined as 'semantic transparency', which represents the extent to which the meaning of a symbol can be inferred from its appearance. We operationalized the 'semantic transparency' as the number of correctly identified meanings of the investigated elements, i.e. comprehension. Higher values were preferred, representing semantic immediate symbols (i.e. a novice reader can infer the meaning from its appearance alone).

In respect to *RQ2*, two types of previous knowledge (i.e. expertise) have been considered as independent variables: 'BPMN expertise' and 'Landscape modeling expertise'. 'BPMN expertise' was considered since BPMN represents ISO and the de-facto standard for process modeling, where some attempts in academia and practice actually apply BPMN for landscape modeling [28]. Accordingly, we presume that BPMN expertise may impact the subjects on how they perceive landscape elements. In a similar manner, 'Landscape modeling expertise' may impact the subjects' answers. In both cases, the 'expertise' was investigated on a 7-point Likert scale from highest to lowest degree of disagreement, with items adapted from Recker [29].

To provide answers to the stated research questions, we performed empirical research as follows. To test to what extent ‘novice users’ would be likely to infer the correct meaning for the common landscape designs, subjects were introduced with different landscape elements and (partial) diagrams, where their task was to identify the meaning for the provided depictions. The applied research protocol may correspond to pre-experimental designs, more specifically ‘one group posttest only design’ [30]. In our case, the treatments were associated with the instructions, provided to subjects, whereas the observations are associated with the subject’s responses.

3.2 Subjects and Sampling

Since we investigated intuitiveness, the ideal candidate for the research would be an individual who (1) understands the meaning of the concepts, which are used in landscape modeling, yet has (2) no experiences with the corresponding landscape modeling notations. According to this, IT and business students of the same degree were selected as suitable candidates for the research.

3.3 Research Instrument

The focal research instrument was an online questionnaire, which was categorized into the following parts. In the first part, subjects were asked to provide basic demographic information (age, gender), and their experiences in BPMN as well as in landscape modeling (both measured on a 7-point Likert scale from lowest to the highest degree of experience, and self-reported number of modeled diagrams). In the second part, subjects were introduced by alternative depictions of common landscape elements (i.e. landscape elements as used in academia and industry, including BPMN and Archimate), where they were asked to associate the most appropriate meaning to them (including the ‘undecided’ answer). In addition, partial diagrams were presented to subjects to test if they would more effectively infer the meaning if using a diagram’s contextual information. To minimize learning effects, the individual items as well the answers were randomized. In the third part of the questionnaire, a “two-treatments” alike design was applied to test the alternative notations used in landscaped modeling. Due to the paper’s length limitations, this part was excluded from this paper. The instrument was prepared in Slovenian and English version and was completely anonymized. The actual research was performed in January 2019. In total, 588 subjects were invited to participate, 347 subjects actually opened the questionnaire or partially completed it, whereas 302 subjects successfully completed the questionnaire. Out of them, 65% of the subject came from Slovenia, whereas 35% came from Ukraine.

4 Results

The results were collected and partially analyzed in 1KA (<https://www.1ka.si/d/en>), an advanced open-source application that enables services for online surveys. Afterward, the data was exported into MS Excel as well as SPSS, to perform additional analysis.

4.1 Descriptive Statistics

As previously mentioned, 302 subjects successfully completed the questionnaire, 161 of them male (53%) and 141 female (47%). In average subjects were 21.3 years old when completing the questionnaire. 197 subjects (65%) come from Slovenia, whereas 105 subjects (35%) came from Ukraine and so completed the survey in the English language. On average, it took eight minutes and four seconds to complete the questionnaire. Based on subjects' expertise with BPMN and landscapes modeling, subjects were classified into the following levels of expertise (Table 1): (1) inexperienced - the subjects who partially or fully disagreed on having any experience in BPMN or landscapes modeling (in all Likert items); (2) BPMN experts - subjects who partially to fully agreed on having expertise in BPMN (in all Likert items); (3) landscape modeling experts - subjects who partially to fully agreed on having expertise in landscape modeling (in all Likert items).

Table 1. Descriptive statistics with respect to subjects' expertise

Expertise	Male	Female	Slovenia	Ukraine	Total	T. [%]
Inexperienced	80	68	102	46	148	49%
BPMN expertise	34	18	44	8	52	17%
Landscape modeling expertise	7	9	10	6	16	5%
All	161	141	197	105	302	100%

As evident from Table 1, the sum of all levels of expertise does not match all subjects, since those subjects who specified 'undecided' on their level of expertise were not classified into any level of expertise.

4.2 Comprehension of Process Landscape Elements

In order to investigate the comprehension of the depictions of common landscape elements, subjects were asked to associate the correct meaning to the stated symbols, where they had an additional 'undecided' option to answer. In total 291 subjects completely answered this question, choosing 'undecided' at an average of 25% of the answers. Table 2 summarizes these results by showing the percentage of answers associated with individual definitions. We bolded the highest (preferred) values for individual depictions and highlighted (in the gray background) the correct definitions (i.e. as specified or used in praxis). As evident from Table 2, subjects associated the symbols to the correct definitions, consistently and at all levels of expertise in the cases of all process elements (depictions D1, D2, D3, and D5) as well in the cases of 'data elements' (depictions D8, and D9). In the remaining cases, the associations of definitions of the stated symbols depend on the level of expertise as follows. The rectangle (D4) was recognized either as a process or as a participant, which may have roots in BPMN's specification of the concept of a Pool. A BPMN Pool may either be treated as a participant or as a process in case it references one.

Table 2. Comprehension of landscape elements

D	Process landscape element depiction	Expertise	Process	Participant	Supp. pro.	Mng. Proc.	Datastore	Document	Proc. collect.	Triggering r.	Inf. flow	Cond. trigger	Gen.-spec. r.	Undecided	Valid a.
1		I	29%	3%	1%	6%	3%	0%	3%	5%	7%	4%	4%	34%	146
		B	39%	4%	2%	10%	0%	0%	6%	2%	8%	2%	0%	27%	51
		L	31%	0%	6%	19%	0%	0%	6%	6%	6%	0%	0%	25%	16
		A	32%	4%	3%	8%	2%	0%	3%	4%	8%	3%	3%	31%	291
2		I	18%	6%	8%	10%	3%	1%	4%	2%	3%	5%	5%	35%	146
		B	25%	4%	8%	8%	2%	0%	2%	0%	2%	8%	4%	37%	51
		L	25%	6%	19%	0%	0%	0%	0%	0%	0%	0%	6%	44%	16
		A	16%	5%	8%	10%	4%	1%	3%	2%	4%	8%	4%	35%	291
3		I	5%	11%	12%	10%	3%	1%	3%	1%	3%	4%	2%	44%	146
		B	4%	4%	29%	10%	0%	2%	0%	0%	2%	0%	0%	49%	51
		L	6%	6%	19%	6%	0%	0%	0%	0%	0%	6%	6%	50%	16
		A	5%	11%	17%	9%	4%	2%	2%	1%	2%	5%	2%	40%	291
4		I	9%	27%	1%	1%	12%	6%	1%	0%	1%	2%	1%	38%	146
		B	41%	20%	2%	6%	2%	2%	0%	0%	0%	0%	2%	25%	51
		L	44%	19%	0%	13%	0%	6%	0%	6%	0%	0%	6%	6%	16
		A	19%	24%	1%	3%	10%	5%	1%	0%	1%	2%	2%	32%	291
5		I	3%	8%	12%	12%	2%	3%	2%	3%	3%	5%	1%	45%	146
		B	2%	0%	18%	25%	0%	2%	0%	2%	2%	2%	2%	45%	51
		L	0%	0%	13%	25%	6%	0%	0%	6%	6%	0%	6%	38%	16
		A	3%	6%	12%	14%	3%	3%	3%	5%	3%	3%	2%	42%	291
6		I	8%	7%	18%	5%	3%	1%	5%	3%	3%	6%	5%	34%	146
		B	8%	0%	16%	12%	0%	0%	22%	0%	2%	8%	6%	27%	51
		L	0%	0%	25%	6%	6%	0%	13%	0%	6%	6%	6%	31%	16
		A	7%	4%	18%	6%	3%	1%	9%	3%	3%	7%	6%	32%	291
7		I	3%	5%	1%	1%	59%	1%	8%	3%	1%	2%	1%	16%	146
		B	2%	2%	0%	0%	88%	0%	2%	0%	0%	0%	0%	6%	51
		L	0%	6%	0%	0%	69%	0%	19%	0%	0%	0%	0%	6%	16
		A	2%	7%	1%	1%	58%	1%	9%	2%	1%	2%	1%	16%	291
8		I	1%	1%	1%	1%	2%	85%	1%	0%	1%	1%	0%	6%	146
		B	0%	0%	0%	2%	6%	92%	0%	0%	0%	0%	0%	0%	51
		L	0%	0%	0%	0%	13%	88%	0%	0%	0%	0%	0%	0%	16
		A	1%	0%	0%	1%	3%	88%	1%	0%	1%	0%	0%	5%	291
9		I	18%	0%	3%	5%	1%	1%	1%	14%	32%	3%	6%	17%	146
		B	4%	2%	0%	0%	2%	0%	2%	55%	25%	4%	2%	4%	51
		L	0%	6%	6%	0%	0%	0%	6%	31%	31%	0%	0%	19%	16
		A	15%	1%	2%	5%	1%	1%	1%	22%	30%	3%	7%	14%	291
10		I	6%	1%	3%	4%	0%	0%	4%	13%	15%	10%	16%	27%	146
		B	6%	0%	4%	2%	0%	0%	2%	10%	4%	31%	24%	18%	51
		L	6%	0%	0%	6%	0%	0%	0%	19%	25%	25%	6%	13%	16
		A	7%	1%	4%	4%	1%	0%	3%	12%	13%	14%	17%	24%	291

Expertise: I=inexperienced; B= BPMN expert; L= landscapes modeling expert; A= all subjects.

The symbol for the collapsed processes collection (D6) was correctly identified by BPMN experts, which may have roots in an analogous representation of a collapsed BPMN subprocess. Expanded collections of processes (D9 and D10) haven't been identified correctly. However, this may have roots in the research instrument since subjects focused on the relationships between processes present on the collection instead of the collection itself.

Individual landscape elements, as well as the relationships between them, were additionally investigated by considering contextual information, i.e. by putting elements into (partial diagrams). Initially, subjects were asked to identify the type of process by providing them a simple value chain-based landscape diagram (Fig. 3).

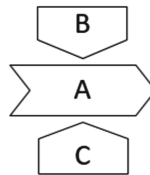


Fig. 3. Common depictions of process types in a value chain model

In this case, subjects recognized the core process correctly in 73%, whereas the supporting process was recognized correctly by 68% of subjects. This is a significant increase when compared to the individually investigated symbols (17%, Table 2, D3). In the case of the management process, the success rate was 59%, whereas in the individual investigation it was 14% (Table 2, D5).

4.3 Comprehension of Between-Processes Relationships

The explicit relationships between processes in a process landscape diagram were investigated individually and by providing subjects the following two diagrams (Fig. 4).

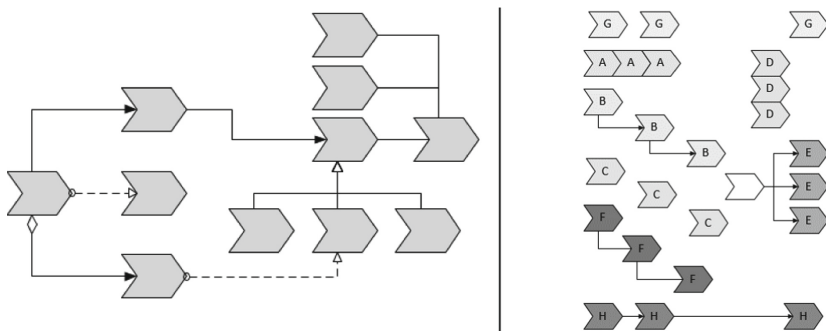

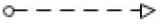

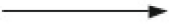



Fig. 4. Investigation of relationships between processes (part 1, left; part 2, right)

The focus of the left diagram in Fig. 4 was to investigate the relationships between processes as specified by Eid Sabbagh et al. [31], namely composition, specialization, trigger, and information flow, with the last two being specified as behavioral ones (Table 3, the highest values are bolded). While two symbols may share the same meaning in praxis which stands for ‘symbol overload’ (e.g. solid line and arrow, different types of arrows), we did specify any correct definitions in this case.

Table 3. Comprehension of explicit relationships

D	Depiction of a between-process relationship	Expertise L.	Information flow	Sequential flow (trigger)	Conditional sequential f.	Gen.- spec. process	Undecided	Valid a.
11		I	21%	23%	23%	19%	14%	145
		B	6%	12%	49%	25%	8%	51
		L	0%	25%	50%	19%	6%	16
		A	17%	20%	29%	21%	13%	289
12		I	14%	17%	38%	20%	10%	145
		B	51%	8%	35%	0%	6%	51
		L	50%	0%	31%	0%	19%	16
		A	24%	16%	35%	13%	11%	289
13		I	17%	34%	21%	17%	11%	145
		B	14%	20%	18%	43%	6%	51
		L	6%	31%	13%	38%	13%	16
		A	19%	28%	21%	21%	11%	289
14		I	36%	33%	6%	10%	15%	145
		B	27%	57%	2%	14%	0%	51
		L	31%	31%	13%	25%	0%	16
		A	34%	36%	7%	13%	11%	289
15		I	26%	8%	6%	31%	29%	145
		B	27%	18%	6%	18%	31%	51
		L	25%	38%	0%	19%	19%	16
		A	24%	10%	6%	29%	31%	289

Expertise: I=inexperienced; B=BPMN expert; L=landscapes modeling expert; A=all subjects.

Table 3 reveals that subjects responded the most consistently (in respect to different levels of expertise) in the case of a conditional trigger relationship (D11), which may be associated with the intuitiveness of a diamond shape symbol, commonly representing a decision-point. Information flow (D12) was correctly recognized by experts, which may be related to the fact that the depiction is equal to BPMN’s message flow. In a similar manner, the ‘generic-specific’ relationship (D13) was correctly identified by the experts, who may have the knowledge either of UML class diagrams or Archimate. The sequential relationship (D14) was not correctly recognized by inexperienced users, whereas all other

expertise levels including all answers inferred the correct meaning. The answers were also inconsistent in the case of a solid line, whereas the majority of subjects reported as being a ‘generic-specific’ relationship (i.e. as common in organizational charts).

The focus of the right diagram in Fig. 4 was to investigate the implicit relationships between processes, which commonly occur on a landscape diagram, especially value-chain based. In this manner, subjects were asked to specify the relationships between the processes sharing the same letter and color (Table 4).




Table 4. Comprehension of explicit and implicit relationships

Element set	Sequentially	Parallel	Independent	Undecided	Valid	Element set	Sequentially	Parallel	Independent	Undecided	Valid		
A	I	72%	17%	1%	9%	144	E	9%	61%	16%	14%	144	
	B	82%	16%	2%	0%	51		8%	75%	14%	4%	51	
	L	69%	31%	0%	0%	16		6%	63%	19%	13%	16	
B	A	72%	20%	2%	6%	288	F	10%	63%	17%	11%	288	
	I	78%	6%	6%	10%	144		65%	12%	10%	13%	144	
	B	82%	10%	6%	2%	51		82%	8%	4%	6%	51	
C	L	75%	13%	6%	6%	16	G	88%	6%	0%	6%	16	
	A	75%	11%	6%	9%	288		66%	13%	9%	11%	288	
	I	5%	4%	80%	11%	144		11%	4%	69%	16%	144	
D	B	2%	4%	92%	2%	51	H	14%	2%	80%	4%	51	
	L	6%	6%	88%	0%	16		19%	6%	75%	0%	16	
	A	5%	5%	81%	9%	288		11%	5%	72%	12%	288	
E	I	8%	67%	13%	11%	144	I	77%	3%	2%	17%	144	
	B	6%	76%	16%	2%	51		94%	0%	4%	2%	51	
	L	0%	94%	6%	0%	16		81%	0%	13%	6%	16	
F	A	7%	73%	12%	8%	288	J	79%	4%	4%	13%	288	
	I	8%	67%	13%	11%	144		K	77%	3%	2%	17%	144
	B	6%	76%	16%	2%	51			94%	0%	4%	2%	51
L	0%	94%	6%	0%	16	81%	0%		13%	6%	16		
G	A	7%	73%	12%	8%	288	L	79%	4%	4%	13%	288	
	I	8%	67%	13%	11%	144		M	77%	3%	2%	17%	144
	B	6%	76%	16%	2%	51			94%	0%	4%	2%	51
L	0%	94%	6%	0%	16	81%	0%		13%	6%	16		
H	A	7%	73%	12%	8%	288	N	79%	4%	4%	13%	288	
	I	8%	67%	13%	11%	144		O	77%	3%	2%	17%	144
	B	6%	76%	16%	2%	51			94%	0%	4%	2%	51
L	0%	94%	6%	0%	16	81%	0%		13%	6%	16		
I	A	7%	73%	12%	8%	288	P	79%	4%	4%	13%	288	
	I	8%	67%	13%	11%	144		Q	77%	3%	2%	17%	144
	B	6%	76%	16%	2%	51			94%	0%	4%	2%	51
L	0%	94%	6%	0%	16	81%	0%		13%	6%	16		
J	A	7%	73%	12%	8%	288	R	79%	4%	4%	13%	288	
	I	8%	67%	13%	11%	144		S	77%	3%	2%	17%	144
	B	6%	76%	16%	2%	51			94%	0%	4%	2%	51
L	0%	94%	6%	0%	16	81%	0%		13%	6%	16		
K	A	7%	73%	12%	8%	288	T	79%	4%	4%	13%	288	
	I	8%	67%	13%	11%	144		U	77%	3%	2%	17%	144
	B	6%	76%	16%	2%	51			94%	0%	4%	2%	51
L	0%	94%	6%	0%	16	81%	0%		13%	6%	16		
L	A	7%	73%	12%	8%	288	V	79%	4%	4%	13%	288	
	I	8%	67%	13%	11%	144		W	77%	3%	2%	17%	144
	B	6%	76%	16%	2%	51			94%	0%	4%	2%	51
L	0%	94%	6%	0%	16	81%	0%		13%	6%	16		
M	A	7%	73%	12%	8%	288	X	79%	4%	4%	13%	288	
	I	8%	67%	13%	11%	144		Y	77%	3%	2%	17%	144
	B	6%	76%	16%	2%	51			94%	0%	4%	2%	51
L	0%	94%	6%	0%	16	81%	0%		13%	6%	16		
N	A	7%	73%	12%	8%	288	Z	79%	4%	4%	13%	288	
	I	8%	67%	13%	11%	144		AA	77%	3%	2%	17%	144
	B	6%	76%	16%	2%	51			94%	0%	4%	2%	51
L	0%	94%	6%	0%	16	81%	0%		13%	6%	16		
O	A	7%	73%	12%	8%	288	AB	79%	4%	4%	13%	288	
	I	8%	67%	13%	11%	144		AC	77%	3%	2%	17%	144
	B	6%	76%	16%	2%	51			94%	0%	4%	2%	51
L	0%	94%	6%	0%	16	81%	0%		13%	6%	16		
P	A	7%	73%	12%	8%	288	AD	79%	4%	4%	13%	288	
	I	8%	67%	13%	11%	144		AE	77%	3%	2%	17%	144
	B	6%	76%	16%	2%	51			94%	0%	4%	2%	51
L	0%	94%	6%	0%	16	81%	0%		13%	6%	16		
Q	A	7%	73%	12%	8%	288	AF	79%	4%	4%	13%	288	
	I	8%	67%	13%	11%	144		AG	77%	3%	2%	17%	144
	B	6%	76%	16%	2%	51			94%	0%	4%	2%	51
L	0%	94%	6%	0%	16	81%	0%		13%	6%	16		
R	A	7%	73%	12%	8%	288	AH	79%	4%	4%	13%	288	
	I	8%	67%	13%	11%	144		AI	77%	3%	2%	17%	144
	B	6%	76%	16%	2%	51			94%	0%	4%	2%	51
L	0%	94%	6%	0%	16	81%	0%		13%	6%	16		
S	A	7%	73%	12%	8%	288	AJ	79%	4%	4%	13%	288	
	I	8%	67%	13%	11%	144		AK	77%	3%	2%	17%	144
	B	6%	76%	16%	2%	51			94%	0%	4%	2%	51
L	0%	94%	6%	0%	16	81%	0%		13%	6%	16		
T	A	7%	73%	12%	8%	288	AL	79%	4%	4%	13%	288	
	I	8%	67%	13%	11%	144		AM	77%	3%	2%	17%	144
	B	6%	76%	16%	2%	51			94%	0%	4%	2%	51
L	0%	94%	6%	0%	16	81%	0%		13%	6%	16		
U	A	7%	73%	12%	8%	288	AN	79%	4%	4%	13%	288	
	I	8%	67%	13%	11%	144		AO	77%	3%	2%	17%	144
	B	6%	76%	16%	2%	51			94%	0%	4%	2%	51
L	0%	94%	6%	0%	16	81%	0%		13%	6%	16		
V	A	7%	73%	12%	8%	288	AP	79%	4%	4%	13%	288	
	I	8%	67%	13%	11%	144		AQ	77%	3%	2%	17%	144
	B	6%	76%	16%	2%	51			94%	0%	4%	2%	51
L	0%	94%	6%	0%	16	81%	0%		13%	6%	16		
W	A	7%	73%	12%	8%	288	AR	79%	4%	4%	13%	288	
	I	8%	67%	13%	11%	144		AS	77%	3%	2%	17%	144
	B	6%	76%	16%	2%	51			94%	0%	4%	2%	51
L	0%	94%	6%	0%	16	81%	0%		13%	6%	16		
X	A	7%	73%	12%	8%	288	AT	79%	4%	4%	13%	288	
	I	8%	67%	13%	11%	144		AU	77%	3%	2%	17%	144
	B	6%	76%	16%	2%	51			94%	0%	4%	2%	51
L	0%	94%	6%	0%	16	81%	0%		13%	6%	16		
Y	A	7%	73%	12%	8%	288	AV	79%	4%	4%	13%	288	
	I	8%	67%	13%	11%	144		AW	77%	3%	2%	17%	144
	B	6%	76%	16%	2%	51			94%	0%	4%	2%	51
L	0%	94%	6%	0%	16	81%	0%		13%	6%	16		
Z	A	7%	73%	12%	8%	288	AX	79%	4%	4%	13%	288	
	I	8%	67%	13%	11%	144		AY	77%	3%	2%	17%	144
	B	6%	76%	16%	2%	51			94%	0%	4%	2%	51
L	0%	94%	6%	0%	16	81%	0%		13%	6%	16		
AA	A	7%	73%	12%	8%	288	AZ	79%	4%	4%	13%	288	
	I	8%	67%	13%	11%	144		BA	77%	3%	2%	17%	144
	B	6%	76%	16%	2%	51			94%	0%	4%	2%	51
L	0%	94%	6%	0%	16	81%	0%		13%	6%	16		
AB	A	7%	73%	12%	8%	288	BB	79%	4%	4%	13%	288	
	I	8%	67%	13%	11%	144		BC	77%	3%	2%	17%	144
	B	6%	76%	16%	2%	51			94%	0%	4%	2%	51
L	0%	94%	6%	0%	16	81%	0%		13%	6%	16		
AC	A	7%	73%	12%	8%	288	BD	79%	4%	4%	13%	288	
	I	8%	67%	13%	11%	144		BE	77%	3%	2%	17%	144
	B	6%	76%	16%	2%	51			94%	0%	4%	2%	51
L	0%	94%	6%	0%	16	81%	0%		13%	6%	16		
AD	A	7%	73%	12%	8%	288	BF	79%	4%	4%	13%	288	
	I	8%	67%	13%	11%	144		BG	77%	3%	2%	17%	144
	B	6%	76%	16%	2%	51			94%	0%	4%	2%	51
L	0%	94%	6%	0%	16	81%	0%		13%	6%	16		
AE	A	7%	73%	12%	8%	288	BH	79%	4%	4%	13%	288	
	I	8%	67%	13%	11%	144		BI	77%	3%	2%	17%	144
	B	6%	76%	16%	2%	51			94%	0%	4%	2%	51
L	0%	94%	6%	0%	16	81%	0%		13%	6%	16		
AF	A	7%	73%	12%	8%	288	BJ	79%	4%	4%	13%	288	
	I	8%	67%	13%	11%	144		BK	77%	3%	2%	17%	144
	B	6%	76%	16%	2%	51			94%	0%	4%	2%	51
L	0%	94%	6%	0%	16	81%	0%		13%	6%	16		

RQ1: Are common landscape designs semantically transparent to ‘novice users’?

When observing the subgroup of subjects, who did not report any level of expertise in BPMN and landscapes modeling (Table 1, ‘inexperienced’), we may conclude that they were able to successfully associate the stated depictions of process landscape elements to the meanings they share in the diagrams (Table 2), especially in the cases, where symbols share the same meaning over different notations (e.g. D7 and D8 in Table 2). They also successfully recognized all process symbols, especially when considering them in the context, as summarized in Table 5.

Table 5. Process depictions as recognized by inexperienced users

Symbol			
Individual element level	12%	29%	12%
Diagram level (context)	58%	72%	67%

When comparing the results of alternative process representations (e.g. D1 and D2 in Table 2), subjects reported better in the case of the chevron symbol (D1) despite the fact D2 is used in a formal specification (ArchiMate). In respect to between-process connections, the ‘inexperienced’ subgroup of subjects failed to associate the stated depictions to the proper meanings as used in landscape diagrams (Table 3), despite the fact that the investigated depictions were put into the context (Fig. 4, left). However, the same subjects reported consistently with experienced users when identifying sequentially, parallel or independent processes either connected explicitly with arrows or implicitly by facilitating the mechanism of proximity between the elements.

RQ2: How does the previous knowledge impacts the comprehension of process landscape designs?

Two types of previous knowledge were investigated in our research: experiences with BPMN and experience with landscape modeling (Table 1). The results of our investigation show that by considering previous knowledge subjects responded differently in several cases when compared to ‘inexperienced’ subgroup of subjects. These were most evident in the cases of a rectangle (Table 2, D4) and chevron arrow with a plus sign (Table 2, D6). Table 6 summarizes these by comparing the comprehension of process landscape elements by considering their definitions as specified or used in praxis.

Besides, the connection elements, which depictions are mainly specified in a conventional way (they do not have any ‘built-in mnemonics’) have reported different comprehension levels, when considering different levels of expertise. E.g. in the case of BPMN experts a dotted arrow was successfully associated with an information flow (in BPMN it represents a Message flow), whereas they successfully associated solid arrow with a triggering relationship (in BPMN it represents a Message flow).

Table 6. Comprehension of landscape elements in respect to levels of expertise

Expertise	Process (chevron)	Process (ArchiMate)	Support process	Participant	Mngm. process	Process collection	Database	Document	Collection (chevron)	Collection (ArchiMate)	Average "undecided"
I	29%	18%	12%	27%	12%	5%	59%	85%	1%	4%	27%
L	39%	25%	29%	20%	25%	22%	88%	92%	2%	2%	21%
B	31%	25%	19%	19%	25%	13%	69%	88%	6%	0%	21%

Expertise: I=inexperienced; B= BPMN expert; L= landscapes modeling expert;

5.1 Implications

We foresee several implications of our investigation. First, experts involved in developing and evolving process landscape languages may consider our research results to select and/or specify the depictions of elements which are intuitive to modelers. Secondly, the selection of visual elements for process landscape design should consider related notations (e.g. BPMN), where elements depictions may already have standardized meanings and therefore should not be overridden. Our research may also be of use for researchers who investigate and propose simplifications of complex process languages, as well as for the ones who extend existing visual languages.

5.2 Research Limitations and Future Work

The results of this research should be considered with the following internal and external limitations in mind. With respect to the external validity, there is a certain degree of risk of generalizing results above the research sample. While students reported as not being skilled in BPMN and landscape modeling languages, another group of subjects could provide different results (e.g. subjects from another environment could be impacted by other signs in their everyday life). Besides, the sample of subjects experienced in landscape design was rather small (16 subjects). Secondly, there is also a certain degree of risk associated with the instrument, where the subject may not be able to correctly interpret the depictions as well the semantics of the symbols out of the instructions (e.g. as in the case of expanded process collections).

Our future work will be focused on specifying a modified landscape modeling notation, based on these results and test if the resulting diagrams are more cognitively effective when compared to existing ones. Besides, we may extend the research to other regions to test on how cultural differences may impact the intuitiveness of symbols.

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