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Line 21	“is” has been changed to “are”.
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Line 60	“their effectiveness captured” has been changed to “their effectiveness has captured”.
Line 62	“bot” has been changed to “both”.
Line 111	“is” has been changed to “are”.
Line 126, 639, 680	“A first” has been changed to “The first”.
Line 171	“define” has been changed to “defines”.
Line 187	“,” has been changed to “;”.
Line 255	“an” has been deleted.
Line 255, 256	“3” has been changed to “three”.
Line 317	“repeated” has been changed to “was repeated”.
Line 344	“provide” has been changed to “provides”.
Line 405	“already commented” has been changed to “has already commented”.
Line 414, 527	“requirement” has been changed to “requirements”.
Line 424	“Table” has been changed to “Tables”.
Line 567	“payed” has been changed to “paid”.
Line 617	“A further development” has been changed to “Further development”.
Line 642	“outcome” has been changed to “outcomes”.



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# 1 Effectiveness of different requirements checklists for 2 novice designers

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7 **Abstract** Working under constrained conditions can boost or kill creativity, depending on the nature of the constraints  
8 (organizational, personal or task-related). However, a design process without clearly identified constraints, which set  
9 the project objectives, could lead to inefficiencies and unfruitful iterations. Some of the most acknowledged  
10 procedures to support requirement definition are focused on the use of specific checklists. However, notwithstanding  
11 the importance of the task, little attention was dedicated to the verification of the effectiveness of these tools. In such  
12 a context, the paper presents an investigation aimed at assessing the performance of three checklists that exploit  
13 different strategies to elicit requirements. To that purpose, a sample of fifty engineering students was asked to use the  
14 checklists to define the requirements for a specific design case. The outcomes of the experiment were assessed  
15 according to well-acknowledged effectiveness metrics, i.e. quantity, operationality, validity, non-redundancy, and  
16 completeness. The result of the assessment highlights that checklists based on more general questions or abstract  
17 stimuli can better support novice designers in making explicit internally felt design constraints that can potentially  
18 lead to more innovative design.

19 **Keywords:** Conceptual design; design tools; product development; requirements elicitation

## 20 1. Introduction

21 ~~Translating customer requirements into technical requirements are~~ typically addressed by the Quality  
22 Function Deployment (QFD) method (Akao, 1990), which is well known to support designers to transform  
23 customer requirements into technical requirements. It maps Customer Attributes (or requirements) and  
24 Engineering Characteristics (technical requirements) for the product. Moreover, QFD allows to rank  
25 customer requirements according to the perceived level of importance, and to consider more sophisticated  
26 requirements classifications (Kano et al., 1984; Matzler & Hintertuber, 1998). QFD, however, leaves the  
27 definition of technical requirements to the designers' talent as it just maps the relationships between  
28 customer attributes and technical requirements. In other words, QFD supports the designers in translating  
29 external constraints (the brief and the set of customer attributes) into technical requirements, but it does not  
30 provide any help in defining internal constraints, which are claimed to push for novelty.

31 An effective and efficient design process needs a set of properly identified and formalized requirements,  
32 as this influences the creativity of the related outcomes (Arrighi, Le Masson, & Weil, 2015; Johnson-Laird,  
33 1988; Finke, 1990; Stokes, 2001), both in terms of novelty and variety of the generated ideas (Worinkeng,

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Joshi, & Summers, 2015; Herrmann, Goldschmidt and Miron-Spektor, 2018). Accordingly, the most acknowledged engineering design handbooks (e.g. Eder & Hosnedl, 2008; Cross 2000; Pahl et al., 2007; Ullman, 2010; Pugh, 1991; Ulrich & Eppinger, 2012) consider requirements as the technical description of objectives that characterize the design process since the very beginning. They help manage the problem(s) and set the goals for the design, potentially reducing the complexity of choices (due to a limited number of available alternatives to consider for the achievement of a goal). Moreover, requirements constrain the boundaries of the design space that the designer explores for suitable solutions. As said above, an excessive use of constraints can also hinder creativity (Roskes, 2015; Caniëls & Rietzschel, 2015; Peterson et al., 2013) as they might trigger a cognitive overload. In fact, constraints play an active role across the cognition-demanding design activities of analysis, synthesis and evaluation (Cascini, Fantoni, & Montagna, 2013; Evbuomwan, Sivaloganathan, & Jebb, 1996; Fiorineschi, Rotini, & Rissone, 2016; Nikulin, Cascini, Viveros, R., & Barbera, 2014). For what concerns the analysis step, their formalization represents a structured list of objectives to achieve. Besides, requirements also support the synthesis of solutions, since turning them into a design proposal deals with the generative process of ideation (Boden, 2009). Additionally, the generation of new solutions can be fostered when in presence of conflicting requirements (Baldussu, Becattini, & Cascini, 2011; C. M. Eckert, Stacey, Wyatt, & Garthwaite, 2011). Concerning the Evaluation step, requirements provide the fundamentals of the evaluation parameters needed to perform the comparisons among candidate solutions, for evaluating them and selecting the most promising.

However, the design specification is constantly updated and drives the design process across all the stages. On the one hand, this progressive refinement is due to the increasing detail level of the solution as the design process proceeds across its stages (e.g. conceptual, embodiment and detail design). On the other hand, this might also depend on a poor planning of the Fuzzy front-end (Bacciotti, Borgianni, Cascini, & Rotini, 2016). While the former case is predictable and expected, the need to adjust the specification “on the fly” due to poor planning can trigger unexpected consequences (e.g. significant additional expenses).

Checklists for the design specification support designers in the hard task of defining requirements with a list of items to be considered potentially relevant as design objectives, but their effectiveness has captured little attention so far. Unfortunately, a comparison between checklists to provide evidences about their pros and cons is currently lacking. The literature claims that design constraints affect (both positively and negatively) the creative process (Bonnardel & Bouchard, 2017) and that their formalization can help the designers in problem decomposition and management (Caniëls & Rietzschel, 2015). Thus, the number of requirements, their distribution across the different phases of the product life-cycle as well as their completeness are therefore crucial for an effective design process. This leads to the formulation of the following research question:

*“Does any difference emerge in the outcomes of a requirements definition task with the use of alternative and/or competing requirements checklists?”*

as a properly defined requirements checklist allows the designer to externalize its internally perceived design constraints, have a clear list of goals and objectives to attain and directions/licit moves for the generation of ideas.

To answer this question, the authors investigated three checklists, which appear to be suitable for a benchmarking study, since they have by different structures and principles to elicit requirements from designers and stakeholders.

According to this purpose, section 2 introduces the Requirement Checklists considered for the study and specifies the motivations behind this work. Section 3 clarifies the research method, together with the protocol for the execution of the experiment and the related acquisition of data. Section 4 presents the results of the experimental investigation with reference to acknowledged characteristics for requirements and specifications (Roozenburg & Eekels, 1991). Before the conclusion, the results get discussed with reference to the impact the new findings might have.

## 82 2. Background

83 The Cambridge Dictionary defines the noun “checklist” as “a list of things that you must think about, or  
84 that you must remember to do”. From this perspective, “requirements checklists” are not exceptions. They  
85 aim at supporting designers to leverage their own knowledge about the design task or project they have to  
86 address, and defining the conditions that the solution should meet to satisfy needs and goals. However, a  
87 checklist is not sufficient to formulate requirements since the designer, typically, must interpret the contents  
88 of the checklists and adapt them to the situation at hand. Indeed, not all the items in the checklist should be  
89 considered relevant, while some others will probably have to be adjusted to suit the specificity of the needs  
90 and the goals behind the design. To this purpose, several requirement checklists are available in the  
91 literature.

92 The following subsections review the current state of the art and describe the checklists considered for  
93 this work.

### 94 2.1. Current lacks and motivations behind this work

95 The management of requirements is a topic per se, as there are handbooks specifically tailored to support  
96 scholars and practitioners in carrying out activities as requirements elicitation, refinement, analysis, etc.  
97 (Dick et al., 2017). So far, the literature presents most of the contributions about requirements, their  
98 elicitation and management from the perspective of the Information Technology domain. Despite the  
99 importance of this topic in engineering design, its related literature focused most on the efforts for providing  
100 new approaches (Jones & Kyoung-Yun, 2015; Mokhov et al., 2016) or improving the existing ones (Shu et  
101 al., 2017; Sumesh et al., 2020) for the management or the elicitation of requirements (Brace & Cheutet,  
102 2012). More in general, the requirements engineering literature often fails to describe how the requirements  
103 have been identified and formalized or it simply reports the list of requirements used in the study, without  
104 providing too many details about the technique used to generate them (e.g. through experts’ opinion, as  
105 witnessed in Tompkins et al. 2018). Nevertheless, the selection of the elicitation technique/approach is also  
106 gathering more and more attention in recent years (e.g. Wellsandt et al., 2014). A recent paper by Horkoff  
107 et al. (2018), still from the perspective of software development, considers the issue of selecting the “right”  
108 elicitation technique according to four dimensions: the acceptability for the user and the subject (i.e. effort  
109 required for elicitation), the perceived satisfaction and its usefulness. However, Carrizo, Ortiz, and Aguirre  
110 (2016), in their survey about requirements elicitation techniques claimed that the metrics to choose the  
111 elicitation technique are not unique and there is no common way to compare their performance and run a  
112 meaningful comparison among them.

113 Beyond the review of well acknowledged checklists presented by Brace & Cheutet (2012), recently  
114 other checklists appeared in literature with the purpose of targeting specific domains, such as the  
115 environment-related requirements checklist by Michelin et al. (2015) as well as the more “need-oriented”  
116 checklists presented in Becattini & Cascini (2014) or in Brglez and Dolšak (2016). Nevertheless, few  
117 studies have previously checked the effectiveness of checklists (e.g. Becattini, Cascini, & Rotini, 2015),  
118 but unfortunately, it is not possible to find a comprehensive benchmark for checklists in literature.

119 Moreover, the checklist can present different items and/or use different modalities to stimulate the  
120 designer towards the identification of requirements. Accordingly, the checklists considered in this paper  
121 (described in the following paragraphs) present such intrinsic differences, thus allowing for purposeful  
122 comparisons and providing evidence to answer the research question.

### 123 2.2. Checklists selected for this work

#### 124 2.2.1. The Pahl and Beitz’s checklist (PBCL)

125 Pahl et al. (2007) suggest two different requirements checklists, which can be used in different phases  
126 of the design process. The first checklist concerns the elicitation of the information to support the activities  
127 involved in the conceptual design phase, while the other checklist mainly focuses on the elicitation of

128 specific criteria to support assessment and selection of solutions during the embodiment and detailed design  
 129 phases. For the purposes of this work, the authors considered the checklist that Pahl and Beitz proposed for  
 130 conceptual design (hereinafter called PBCL). The reason behind such a choice is that the role of the  
 131 requirements is relevant especially in the early phases of the design process, as a poor definition of the  
 132 specification might trigger several costly design iterations. This led the authors to focus on the conceptual  
 133 design stage of the process. PBCL guides the exploration of requirements through the administration of a  
 134 set of stimuli to the user. The stimuli cover different categories of product features, which can be briefly  
 135 summarised as follows:

- 136 ○ functional performance: flows of force/energy, material and signal/information
- 137 ○ life cycle issues: assembly, transportation, operation maintenance and end-life
- 138 ○ human factors: safety and ergonomics
- 139 ○ specific features of the system: geometry and kinematics
- 140 ○ quality: regulations, standards and testing
- 141 ○ costs and schedules: investments, costs, planning and controls of the development process, time  
 142 for the development.

143 With the aim of providing an idea about the formulation of stimuli belonging to PBCL, two examples  
 144 are presented in the following for the category “human factors” (Pahl et al., 2007):

145 *“Safety: Direct safety systems, operational and environmental safety.”*

146 *“Ergonomics: Man-machine relationship, type of operation, operating height, clarity of layout, sitting  
 147 comfort, lighting, shape compatibility.”*

148 As shown, the stimuli belonging to each category are lists of examples related to aspects, performance,  
 149 features, and parameters of the system, presented in a general form, and textually described, which might  
 150 result relevant under particular conditions or situations.

### 151 **2.2.2. The Pugh’s checklist (PCL)**

152 The checklist proposed by Pugh (Pugh, 1991), hereinafter called PCL, still considers the main categories  
 153 of requirements introduced in PBCL, although it is more detailed, especially concerning life cycle issues.  
 154 PCL relies on questions as triggers to elicit requirements. To provide an example, the stimuli for safety and  
 155 ergonomics suggested by PCL are presented as follows:

156 *“Safety: Should any special facilities be provided for the safety of users and non-users?”*

157 *“Ergonomics: Which requirements, with regard to perceiving, understanding, using, handling, etc. does  
 158 the product have to meet?”*

159 Therefore, PCL works with a different logic if compared to PBCL since it proposes a set of specialised  
 160 questions to guide the user towards the definition of the relevant requirements the system should satisfy.

### 161 **2.2.3. A third checklist for the comparison (BCL)**

162 Eventually, the third instrument considered for the comparison is the checklist suggested by Becattini  
 163 & Cascini (2013), hereinafter called BCL. It is based on textual stimuli that are organised according to the  
 164 terms of Ideality in TRIZ (Altshuller, 1984; Gadd, 2011; Salamatov, 1999). More in particular, according  
 165 to the “law of Ideality increase” suggested by TRIZ, which states that systems evolve by increasing the  
 166 delivered useful functions and by reducing generated harmful effects and consumption of resources.  
 167 Therefore, the exploration of requirements according to the perspective suggested by Ideality allows the  
 168 user to take into consideration future desired features, potentially relevant for system and stakeholders. The  
 169 above-introduced three categories contain sub-classes that refer to specific aspects of the system at different  
 170 levels, in different phases of the life cycle and for different stakeholders. The stimulus provided to the user  
 171 is the textual description that defines the sub-class it belongs to. For instance, still considering issues related  
 172 to safety and ergonomics, these stimuli belong to an abstract class of side effects directly due to the technical  
 173 system itself:

- 174 ○ Side effects due to the action of the technical system:
  - 175 ○ Ex1: Production scraps (e.g. process waste, amount of materials to be reprocessed, etc.)

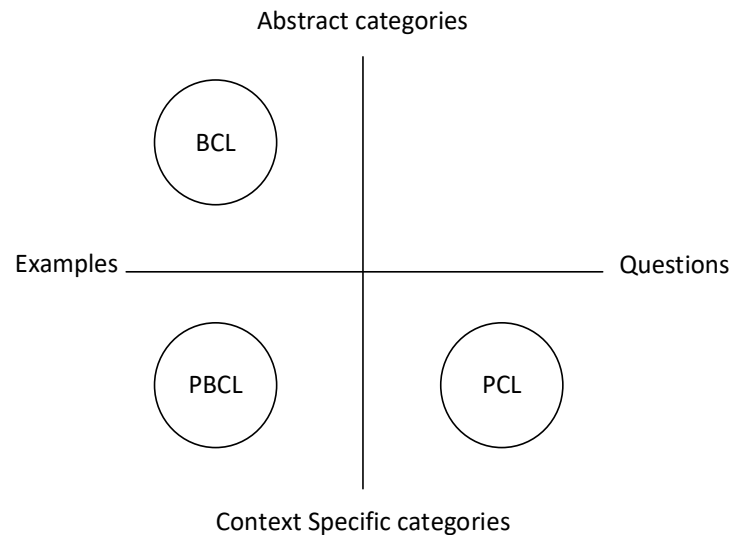


- 176           ○ Ex2: Environment pollution (e.g. heat dissipation, noise level, effects of chemicals on the  
177           eco-system, etc.)  
178           ○ Ex3: Comfort, ergonomics, safety (e.g. standing/seated operator, grabbing force, # of  
179           required movements to carry out an operation, etc.)  
180           ○ Ex4: Reliability (e.g. expected mean time between failures, failure rate, etc.)  
181           ○ Ex5: ... (any other issue related to side effects generated by the system)

182           It is quite evident that the BCL's classification of requirements strongly differs from those adopted by  
183           the other checklists, since it classifies the requirements according to a perspective explicitly based on the  
184           evolution of needs, related system features, and functional role played by the latter in satisfying the  
185           stakeholders.

#### 186   2.2.4. *Characterization of the main differences among the selected checklists*

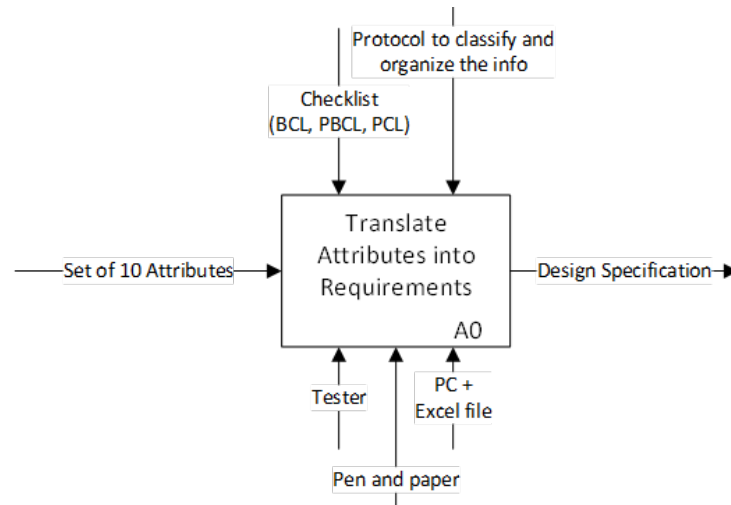
187           The three checklists presented in this section differ by their stimulation strategies; their opposed features  
188           are depicted in representative quadrants (Figure 1). In particular, the considered metrics can be discerned  
189           in terms of the specific way to provide the stimuli (Examples vs Questions in Figure 1), and in terms of  
190           abstraction of the provided stimuli (Abstract vs Context specific in Figure 1). The comparison of the  
191           considered metrics is expected to clarify what kind of strategy better stimulates the identification and  
192           formalization of requirements.



193           **Figure 1. Checklists Classified by Characteristic of the Provided Stimuli to Generate Requirements.**  
194

### 195   3. Research Method

196           The experimental set-up can be schematically represented as shown in Figure 2, where the task requires  
197           novice designers to translate a set of ten product attributes into a more comprehensive set of engineering  
198           requirements.



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**Figure 2. IDEF0 Diagram of the Experimental Set-Up (left arrow: input; right arrow: output; downwards arrows: process controlling elements; upwards arrows: mechanisms for process execution)**

203 The detailed description of the experimental set up and the analysis process is reported in the following  
204 subsections.

### 205 3.1. Sample of participants involved in the experiment

206 Fifty students of the Master of Science (MS) degree in Mechanical Engineering (University of Florence,  
207 Italy) composed the sample considered for the experiment (Testers in Figure 2). They were all from Italy  
208 and only three of them were female. They attended the first part of a course that provides a framework and  
209 a set of creative tools for the analysis of stakeholders' needs, and the definition of new benefits/advantages  
210 the product to innovate should provide.

211 A specific definition of function was provided to students, i.e. that based on the Energy-Material-Signal  
212 (EMS) flows (Pahl et al., 2007), since it is one of the most largely taught and diffused. It is worth noticing  
213 that the concept of function can be interpreted in many ways (Eckert et al., 2011; Eckert, 2013; Vermaas &  
214 Eckert, 2013) and therefore a common framework to define functions is beneficial, at least to support the  
215 identification of functional requirements (target shared by all the three checklists, without exceptions).

216 Eventually, a short briefing (thirty minutes overall) was performed with students before the experiment  
217 to explain the checklists' logic and their use. Then, the sample was subdivided into three groups (one for  
218 each checklist – according to the left downward arrow of Figure 2): 18 students worked with the PCL, 18  
219 with the PBCL and 14 with the BCL (one female for each group). The numerical differences between  
220 groups are due to the room setup, as the administration of checklists was done to avoid cross-contamination  
221 among participants. More specifically, students were asked to work individually to avoid the mixing of  
222 individual thinking and to extract a greater amount of data for the subsequent analysis process. The test  
223 requires no control group as the aim is to compare different checklists. Previous studies have already  
224 verified the benefits of using a checklist against no support (N. Becattini & Cascini, 2014).

### 225 3.2. Design task

226 The experiment was structured by relying on a particular academic case study: “a device for teeth and  
227 mouth hygiene (e.g. an innovative toothbrush)” which also holds the attributes of Table 1, i.e., a set of given  
228 design constraints that are represented as input arrow in Figure 2.



229

**Table 1. Initial Set of Desired Attributes**

Attribute	Description of the objectives
Hygienic aspects	Performances about this attribute should be comparable to those of existing products of the same type. Nevertheless, the system to be designed should comply at least with standard safety requirements, in order to avoid problems in the oral cavity.
Comfort	No particular performances are expected in terms of comfort.
Aesthetic pleasantness	The ideal solution should be pleasant and perfectly integrated in the environment where it is normally hosted.
Versatility of use	It is expected the possibility to perform multiple cleaning operations within the oral cavity. Besides the teeth cleaning, tongue, palate, and gingival interstices should be considered.
Cleaning effectiveness	The teeth cleaning effectiveness must be maximised.
Ease of use	The system should be as easy as possible.
Multiple functions	Besides the cleaning functionalities, the system should provide other functionality types. In particular, the system should allow to listen to music and/or daily news.
Customisation	The system should allow to be configured according to the user preferences.
Size	The system can be bigger than existing products with similar functionalities.
Energy saving	No particular restrictions are provided in terms of energy consumption.

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Students were asked to use the checklists as a guide to extract and/or find the design information needed for the engineering development of the product. Since no additional data were provided, students could extrapolate whatever they wanted in terms of additional data to formulate engineering requirements (represented in Figure 2 by the right arrow). For example, for the attribute “Size” (Table 1), the objective reports that the system can be bigger than the existing ones, but without explicit limits for the maximum allowable size. It is a choice of the student (conditioned by the specific checklist) to establish and indicate missing information (i.e., to externalize additional constraints not otherwise made explicit).

**238 3.3. Testing protocol**

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The material administered for the experiment consisted in a paper sheet containing the list of attributes reported in Table 1, another paper sheet with a short description of the parameters composing the checklists, and a spreadsheet file containing a structured matrix (Figure 3). Accordingly, students were asked to use their own laptop to list their requirements on the spreadsheet (central and right upwards arrow in Figure 2).

	Main function 1			...			Main function M		
Checklist parameter 1	R 1.1.1	..	R 1.1.j	...	...	...	R M.1.1	..	R M.1.j
Checklist parameter 2	R 1.2.1	..	R 1.2.j	...	...	...	R M.2.1	..	R M.2.j
...	...								
Checklist parameter N	R 1.N.1	..	R 1.N.j	...	...	...	R M.N.1	..	R M.N.j

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**Figure 3. Generalized Version of the Matrix (N checklist parameters and M functions) Administered to Students for the Experiment (Fiorineschi, Becattini, Borgianni, & Rotini, 2020). The Number of Column Under Each Function Is Not Fixed and Can Vary By Subject. Empty Boxes Are Allowed When It Is Not Possible to Extract/Infer Relevant Requirements.**

248 The structured matrix was provided to support students (novices in requirements definition) in  
249 performing the required task.

250 The time allotted for the test was 60 minutes and students were asked to save the spreadsheet file and  
251 send it to a specific email address for data collection.

### 252 3.4. Data analysis

#### 253 3.4.1. Metrics for requirements evaluation

254 As the goal is to compare the results of the application of different requirements checklists, the results  
255 require appropriate evaluation metrics. Roozenburg and Eekels (1991) defined the three ideal characteristics  
256 of a design specification, together with the three main elements a requirement should comply with, to be  
257 included into the requirement list. They are presented here in the following, in the same order proposed by  
258 Roozenburg and Eekels (1991):

- 259 ○ Validity - as the capability of the requirement to discriminate the extent of the achievement of a  
260 certain objective.
- 261 ○ Completeness - as the capability of the whole specification to cover all the objectives in the  
262 different domains where stakeholders are involved.
- 263 ○ Operationality - as the capability of the requirement to make the objective measurable, to avoid  
264 subjective evaluation of (partial) solutions.
- 265 ○ Non-redundancy - as the capability of the specification to be free from duplicates.
- 266 ○ Conciseness - as the capability of the specification to contain just the meaningful requirements,  
267 without neglecting important facets to be taken into account (not too many, not too few).
- 268 ○ Practicability - as the capability of the requirements to be tested, e.g. with simulations or by  
269 exploiting available information.

270 To assess Validity and Operationality the authors relied on a coding scheme based on the ENV model  
271 (Cavallucci & Khomenko, 2005). The ENV model describes parameters of entities by clarifying the  
272 Element they belong to, their Name and the Value it takes. For instance, the sentence “A tomato is a round  
273 and red vegetable” describes two parameters: “Tomato” is the element (E), whose considered parameters  
274 are (N) colour and shape. They assume, respectively, the values (V): red and round. As Validity describes  
275 the requirement to satisfy/target to achieve, a requirement is valid just if it defines both the name of the  
276 (measurable) parameter to achieve (N) and the element (E) it belongs to. Operationality requires verifying  
277 the achievement of a target value (V) for the above parameter. A requirement is operational just if it clarifies  
278 what value to measure otherwise it is not.

279 Non-redundancy concerns with the exclusion of duplicates from the specification: redundant  
280 requirements (in each specific set) are those targeting the same E-N-V triad.

281 For the Completeness metric, the stakeholders involved in the life-cycle of the specific product have  
282 been identified with an “a-posteriori approach” on the entire set of requirements generated by all the  
283 subjects (see Table 2). Accordingly, a (more) complete specification involves a larger set of considered  
284 stakeholders.

285 The evaluation of conciseness becomes impossible to define with an a-priori logic, being it dependent  
286 on the direction of development a requirement holds in itself, as a piece of designer’s externalized  
287 knowledge. On the other hand, the evaluation of practicability depends on the reference simulation or  
288 testing system considered for such a purpose. The arbitrariness of these two characteristics also makes them  
289 potential biasing factors. For this reason, Conciseness and Practicability have been neglected from the final  
290 metrics.

291 Subjects were not aware of the metrics to apply on the results to avoid biased outcomes.

292

293

**Table 2. The Stakeholders Identified from the Entire Set of Requirements Generated by Students**

Stakeholder	Short description	Phase
Seller	The person who handle the product until it is out of the store	Sale
Buyer	The person who is interested in buying the product and that operates the selection	
User	The person handling the product from the first moment after the purchase, up to its retirement (except for maintenance intervals)	Use/Benefit
Beneficiary	The person receiving the benefits provided by the product (non-necessarily the same person of the user).	
Dentist	The person that is indirectly affected by the benefits provided by the product.	Other
Transporter	The person that transport the product	Transport
Maintainer	The person handling the product during the maintenance operations	Maintenance
Disposal guy	The person handling the product during the disposal operations	Disposal

294 **3.4.2. Data collection and management**

295 In order to manage data, spreadsheets were collected by group, so that the results can be classified by  
 296 the subject participating in the data collection process (Figure 2 – right downwards arrow). In each students’  
 297 spreadsheet, each requirement has been analysed to verify the presence of the three parameters of the ENV  
 298 triad. Moreover, for each requirement, the affected stakeholders (Table 2) were identified by means of  
 299 additional columns in the same worksheet (Table 3).

300 Concerning the Non-redundancy metric, each worksheet (one for each student) enabled this assessment  
 301 with a specific matrix (see Table 4).

302 **Table 3. Table Used to Assess the Requirement Sets Produced by Each Student. “1” Or “0” Are**  
 303 **Attributed to Each of the ENV if Respectively the Parameters Are Present or Are Missing. “1” Is**  
 304 **Assigned to Each Stakeholder Actually Affected by the Requirement.**

	Affected stakeholders											
	E	N	V	Transporter	Seller	Buyer	User	Beneficiary	Dentist	Maintainer	Disposal guy	
Requirement 1	1/0	1/0	1/0	1/0	1/0	1/0	1/0	1/0	1/0	1/0	1/0	
Requirement 2	1/0	1/0	1/0	1/0	1/0	1/0	1/0	1/0	1/0	1/0	1/0	
...												
Requirement n	1/0	1/0	1/0	1/0	1/0	1/0	1/0	1/0	1/0	1/0	1/0	

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**Table 4. Non-Redundancy Assessment Matrix. The Value “1” Is Introduced in Those Boxes (In the Lower Triangle of The Matrix) Where It Has Been Identified as Redundancy Between the Requirement in the Row with the Requirement in the Column.**

	Requirement 1	Requirement 2	Requirement 3	Requirement 4	...	Requirement n
Requirement 1						
Requirement 2						
Requirement 3						
Requirement 4						
...						
Requirement n						

### 309 3.4.3. Experimental data processing

310 According to the considered metrics, three evaluators coded the specification produced by students.  
311 More precisely, each evaluator coded the results of two out of the three considered checklists, so that two  
312 different coders assess the results of each of the three treatments. Each evaluator worked with all the  
313 considered metrics. Krippendorff’s Alpha test (Hayes & Krippendorff, 2007) returns Inter-Rater Reliability  
314 score. When Alpha scores below 0,66 for a specific metric, the coding results were shared and discussed,  
315 and the specific coding activity was repeated. This iterative process was then repeated until the Alpha scores  
316 overcome the threshold value 0,66.

317 The adopted metrics and the related coding activity enabled the distinction of suitable requirements from  
318 those that are not (univocally) interpretable or not measurable. This skimming process started from the  
319 whole quantity of items and progressively considered the criteria of validity and operationality (which  
320 applies on every requirement). Then, the specifications got also reduced to remove duplicates (non-  
321 redundancy). The remaining requirements, per responding subject involved in the experiment, constituted  
322 the individually generated design specification. The degree of completeness of each specification followed  
323 to the criterion described at the beginning of section 2.4.

324 All the individually generated tentative specifications, as well as their progressive refinements towards  
325 the final set of selected design requirements, constituted the data points of distributions by checklists. These  
326 distributions got analysed in terms of descriptive statistic estimators (average and standard deviation) to  
327 highlight the performance of each checklist and compare them against each other.

## 328 4. Results

329 The statistical analyses reported in the following paragraphs mainly concern the descriptive statistics  
330 and the analysis of variance test (ANOVA). The latter was performed to assess the difference in the number  
331 of requirements produced by students using the BCL, PBCL and PCL checklist. SPSS version 18.0 is used  
332 to calculate all statistical comparisons and the level of significance was set to p-value < 0.05.

### 333 4.1. Overall productivity by subject and checklist

334 Table 5 shows, per each of the subject involved in the investigation, the number of requirements that  
335 populate every tentative specification considered for the application of metrics. These values, then, reflect  
336 the overall productivity of each subject.

337

**Table 5 Number of Requirements Generated by Each Subject**

Checklist																			Tot.	AVG.	St. dev.
	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18			
BCL	31	17	28	43	15	8	13	10	12	38	11	11	14	18					269	19.2	11.2
PBCL	13	15	7	6	17	17	20	9	28	10	21	15	17	16	11	11	5	9	247	13.7	5.9
PCL	21	13	29	10	12	15	45	23	10	17	13	11	9	4	11	10	9	17	279	15.5	9.4

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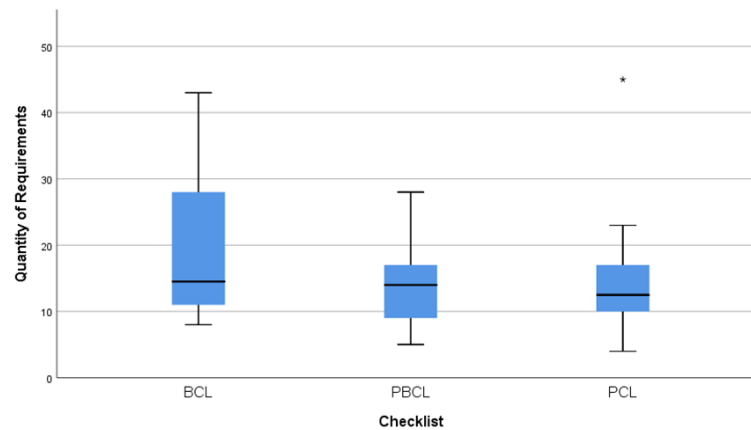
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Descriptive statistics, reported in Table 5, show that on average BCL provides more populated tentative checklists when requirements are formalised starting from the same set of solution attributes as input for the synthesis of solutions. PCL, then, appears to be more productive than PBCL. Both BCL and PCL also have large variability among the recorded performance, while PBCL provides less performing but more stable results. The boxplots depicted in Figure 4 graphically summarize the distribution of recorded outcomes.



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**Figure 4. Boxplot of the Number of Requirements Produced by the Different Groups of Students (each of them using a different checklist).**

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Figure 4 also highlights that there is a generally wider distribution of results in the third and the fourth quartile for BCL, while PBCL and PCL have more narrow distributions and similar interquartile ranges, which should be statistically investigated. Moreover, the largest value for PCL application displayed in the boxplot in Figure 4, is separated from the bulk of the data, and can, therefore, be tested as an outlier. Looking at the interquartile range (IQR) computed from Tukey’s hinges (Schwertman, Owens, & Adnan, 2004); the selected value exceeds three times the index, and can be labelled as an outlier, and hence dropped from the analysis. Therefore, the final distribution of the PCL requirements presents a more stable distribution, with a lower standard deviation as reported in Table 6.

356

357

On the new set of data, the analysis of variance test (ANOVA) estimates the difference among the BCL, PBCL and PCL checklist. However, no significant difference among the three groups is found.

358

**Table 6 Number of Requirements for the PCL Checklist After the Eliminated Outlier**

Subject																			Tot.	AVG.	St. dev.
	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18			
n° req.	21	13	29	10	12	15	-	23	10	17	13	11	9	4	11	10	9	17	234	13.76	6.10

359

360

**4.2. Outcomes of the investigation by metrics for requirements and specifications**

361

362

363

This subsection provides details about the validity and the operability of requirements and the completeness and non-redundancy for the whole specification, consistently with the metrics adapted from Roozenburg and Eekels (1991).

364 **4.2.1. Validity**

365 Table 7 presents results about the validity of the requirements, based on the same set of data reported in  
 366 Table 5 and Table 6. In particular, Table 7 shows, for each checklist, the total number of requirements  
 367 generated by each student (rows "a" of the table) and those remaining in the specification after the check  
 368 for validity, both in absolute (rows "b" of the table) and percentage terms (rows "c" of the table).

369 Data in Table 7 mainly confirm the results already commented in subsection 3.1. In fact, both PCL and  
 370 BCL show higher values, on average, of valid requirements generated by students. Still, the BCL  
 371 distribution results in the one with the largest value of standard deviation.

372 Figure 5 presents the normalised results according to the number of subjects exposed to each testing  
 373 condition with checklists. The results are organised in clusters of requirements spanning four different  
 374 contiguous design specification sizes (e.g. from 4 to 7 requirements in the specification, from 8 to 11, etc.).  
 375 To make results homogeneous and fully comparable, they are normalised and scaled to 100% to make a  
 376 direct estimation of the probability of finding more or less populated specifications within the samples.

377 There is approximately 30% of chance to get a design specification with 20 requirements or more with  
 378 BCL, while this probability drops to 5% for both PCL and PBCL. To better check for differences among  
 379 the three groups, an ANOVA test is performed on the percentage of valid requirements generated by each  
 380 student. The Tukey post hoc test is also used to detect significant differences in the pairwise comparisons  
 381 (i.e. BCL vs PBCL, BCL vs PCL and PBCL vs PCL). The level of significance is again set at  $p\text{-value} < 0.05$ .  
 382 The results are in Tables 8 and 9.

383 **Table 7. The Total Number of Requirements for Each Student (rows "a" of the table) and the**  
 384 **Number of Valid Ones in Absolute (rows "b") and Percentage Terms (rows "c").**

Checklist	Number of requirements for each subject																		Tot.	AVG.	St. dev.	
	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18				
<b>BCL</b>																						
a)	31	17	28	43	15	8	13	10	12	38	11	11	14	18						269	19.2	11.2
b)	25	13	21	34	15	6	9	8	10	29	10	9	8	14						211	15.1	8.8
c)	.8	.8	.8	.8	1	.8	.7	.8	.8	.8	.9	.8	.6	.8						-	0.8	0.1
<b>PBCL</b>																						
a)	13	15	7	6	17	17	20	9	28	10	21	15	17	16	11	11	5	9	247	13.7	5.9	
b)	12	8	6	4	11	17	15	8	25	9	14	11	13	8	8	7	4	7	187	10.4	5.2	
c)	.9	.5	.9	.7	.6	1	.8	.9	.9	.9	.7	.7	.8	.5	.7	.6	.8	.8	-	0.8	0.2	
<b>PCL</b>																						
a)	21	13	29	10	12	15	-	23	10	17	13	11	9	4	11	10	9	17	234	13.8	6.1	
b)	19	12	23	9	10	11	-	19	9	15	12	9	8	4	11	6	9	12	198	11.7	4.9	
c)	.9	.9	.8	.9	.8	.7	-	.8	.9	.9	.9	.8	.9	1	1	.6	1	.7	-	0.9	0.1	

385 **Table 8 Summary of the ANOVA Test Using the Percentage of Valid Requirements (df – degrees of**  
 386 **freedom; F – Fisher Snedecor statistics; Sig – Significance)**

	Sum of Squares	df	Mean Square	F	Sig.
<b>Between Groups</b>	0.093	2	0.046	3.412	0.042
<b>Within Groups</b>	0.626	46	0.014		
<b>Total</b>	0.719	48			

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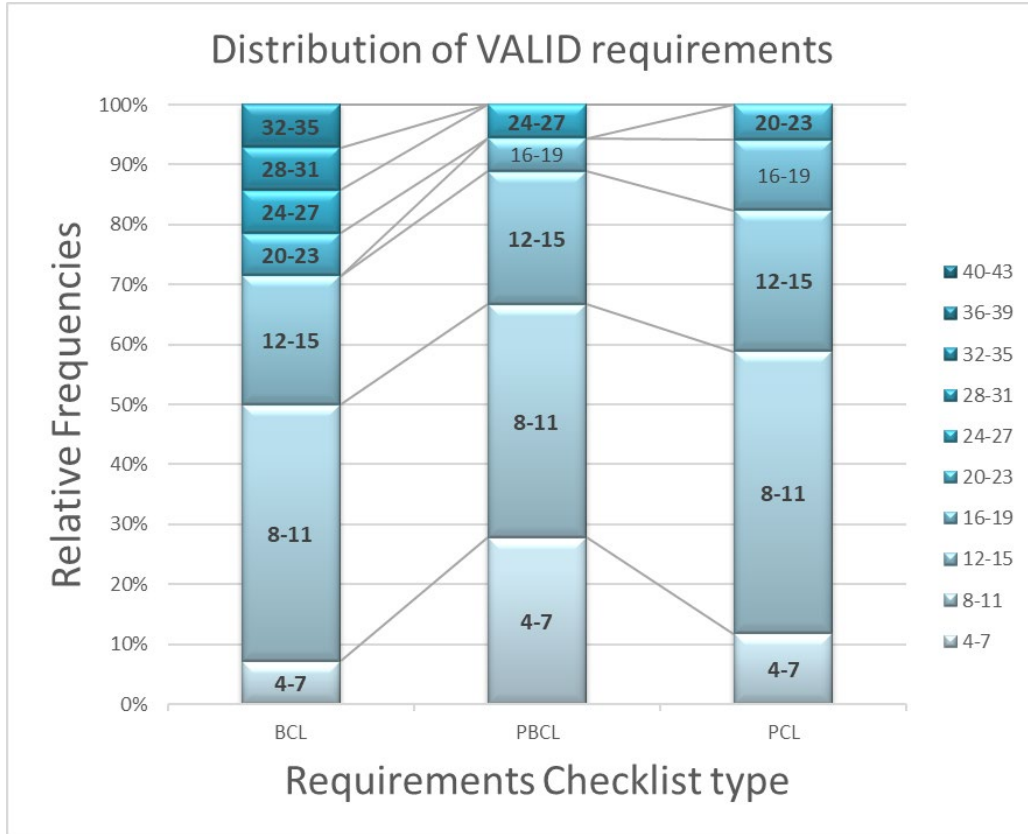


Figure 5. Normalized Results According to the Amount of Subjects Exposed to Each Testing Condition with Checklists

Table 9 Summary of the Pairwise Comparisons Between the Three Checklists Using the Percentage of Valid Requirements (I – rows; J – Columns)

		J				
		BCL	PBCL	PCL		
I	BCL	-	Mean diff. (I-J)	0.028	Mean diff. (I-J)	-0.073
			Std. Error	0.041	Std. Error	0.041
			Sig.	0.781	Sig.	0.069
	PBCL	-	-	Mean diff. (I-J)	-0.10	
				Std. Error	0.041	
				Sig.	<b>0.039</b>	
PCL	-	-	-	-		

As the results of Table 8 show, the difference in the production of valid requirements concerning the three checklists is significant ( $F=3.412^*$ ,  $p\text{-value}<0.05$ ). According to the results of the Tukey post-hoc test, significant differences exist amongst the group of students using the PCL and the PBCL checklists ( $p\text{-value}=0.039$ ), as reported in Table 9. In particular, it is possible to observe that PCL, on average, produces 10% more valid requirements than PBCL.

The Shapiro-Wilk test verifies the assumption that the data follows a normal distribution (null hypothesis), which is confirmed ( $p\text{-value} 0.197$ ).

401 **4.2.2. Operationality**

402 Table 10 collects the amount of valid and operational requirements, by subject and by checklist, as for  
 403 Table 7.

404 **Table 10. Amount of Total and Valid and Operational Requirements, by Subject and by Checklist**

Checklist	Number of requirements for each subject																		Tot.	AVG.	St. dev.	
	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18				
<b>BCL</b>																						
a)	31	17	28	43	15	8	13	10	12	38	11	11	14	18						269	19.2	11.2
b)	25	13	21	34	6	6	9	8	10	29	10	9	8	14						202	14.4	9.1
c)	.8	.8	.8	.8	.4	.8	.7	.8	.8	.8	.9	.8	.6	.8						-	0.7	0.1
<b>PBCL</b>																						
a)	13	15	7	6	17	17	20	9	28	10	21	15	17	16	11	11	5	9	247	13.7	5.9	
b)	7	8	4	1	11	10	13	8	25	7	14	8	8	8	7	4	7	158	8.8	5.1		
c)	.5	.5	.6	.2	.6	.6	.7	.9	.9	.7	.7	.5	.5	.5	.7	.6	.8	.8	-	0.7	0.2	
<b>PCL</b>																						
a)	21	13	29	10	12	15	-	23	10	17	13	11	9	4	11	10	9	17	234	13.8	6.1	
b)	17	10	23	8	10	11	-	19	8	15	12	9	8	4	10	6	9	11	190	11.2	4.8	
c)	.8	.8	.8	.8	.8	.7	-	.8	.8	.9	.9	.8	.9	1	.9	.6	1	.6	-	0.8	0.1	

405 In general, Table 10 confirms what has already commented for the set of valid requirements. Both PCL  
 406 and the BCL averagely generated a higher percentage of requirements (rows “c” of Table 10), both valid  
 407 and operational. Comparing Table 10 with Table 7, it is interesting to note that BCL produces almost all  
 408 valid and operational requirements, having only nine requirements dropped from the analysis (from 211  
 409 valid requirements to 202 valid and operational ones). The other two checklists, PBCL and PCL, have very  
 410 different behaviours: once again PCL performs better than PBCL as the number of dropped requirements  
 411 is smaller (9 against 29).  
 412

413 Figure 6, consistently with Figure 5, shows normalised data for clusters there represented to clarify the  
 414 probability mass functions, one per each of the experimental treatments.

415 In such a case, Figure 6 has clusters with a shift towards lower values, so as to keep track of the whole  
 416 set of experimental data. Almost 30% of specifications formalised with BCL holds more than 20  
 417 requirements, while it is approximately 5% for PCL and PBCL. In this case, as well, ANOVA and a  
 418 following Tukey post-hoc test investigate the differences among the three groups. Tables 11 and 12  
 419 summarize the results.

420 **Table 11 Summary of the ANOVA Test Using the Percentage of Valid and Operational**  
 421 **Requirements (df – degrees of freedom; F – Fisher Snedecor statistics; Sig – Significance)**

	Sum of Squares	df	Mean Square	F	Sig.
<b>Between Groups</b>	0.345	2	0.173	9.068	.000
<b>Within Groups</b>	0.876	46	0.019		
<b>Total</b>	1.221	48			

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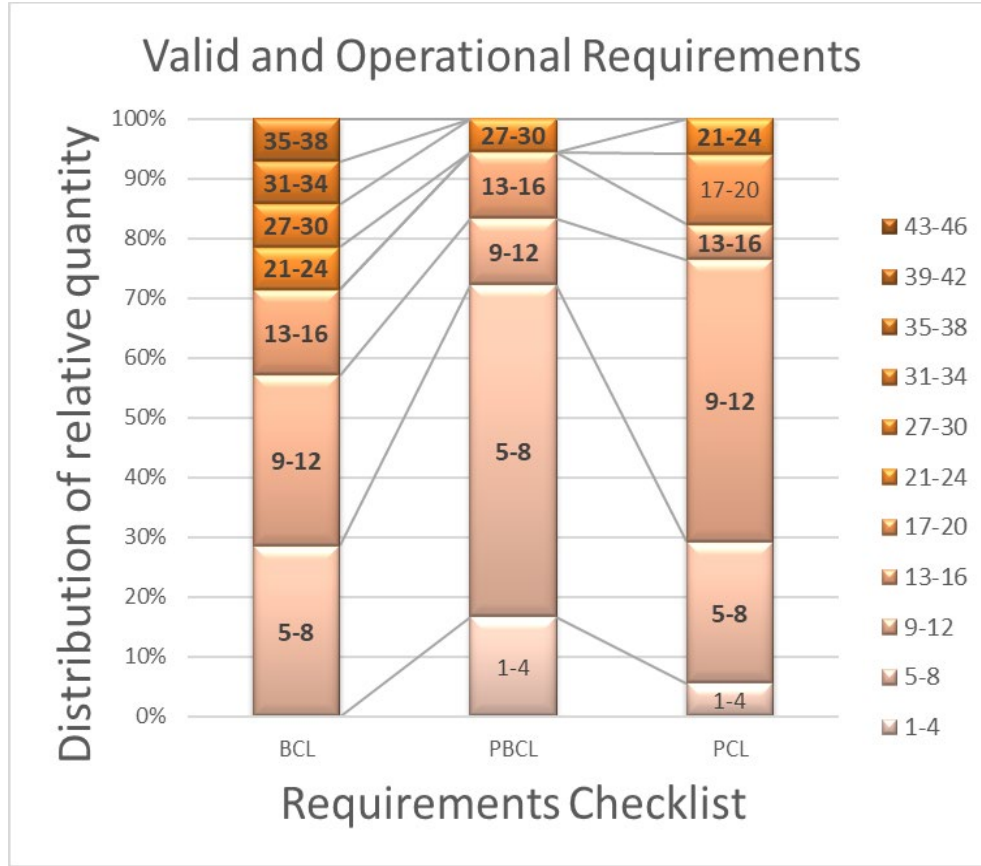


Figure 6. Normalised Data for Clusters There Represented to Clarify the Probability Mass Functions, One Per Each of the Experimental Treatments.

Table 12 Summary of the Pairwise Comparisons Between the Three Checklists Using the Percentage of Valid and Operational Requirements (I – rows; J – columns)

		J				
		BCL	PBCL		PCL	
I	BCL	-	Mean diff. (I-J)	0.17	Mean diff. (I-J)	-0.08
			Std. Error	0.05	Std. Error	0.05
			Sig.	0.06	Sig.	0.189
	PBCL	-	-	Mean diff. (I-J)	-0.20	
				Std. Error	0.04	
				Sig.	<b>0.00</b>	
PCL	-	-	-	-		

The difference between the PCL, PBCL and BCL groups of generated requirements is significant (F=9.068\*\*\*, p-value=0.000), as reported in Table 11. Looking at Table 12, it is possible to observe that the statistically significant difference is again between the PCL and the PBCL (p-value= 0.000), where PCL perform better than PBCL in generating both valid and operational requirement of about 20%. As for the previous check, also this case confirms the assumption of the normality in data distribution (p-value>0.05 for the Shapiro-Wilk test).

446 **4.2.3. Non-redundancy**

447 Table 13 shows the results for non-redundancy by subject and requirements checklist considered for this  
 448 benchmarking study.

449 **Table 13. Amount of Total and Valid, Operational and Non-Redundant Requirements, by Subject**  
 450 **and By Checklist.**

Checklist	Number of requirements for each subject																		Tot.	AVG.	St. dev.	
	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18				
<b>BCL</b>																						
a)	31	17	28	43	15	8	13	10	12	38	11	11	14	18						269	19.2	11.2
b)	16	7	19	7	4	5	8	6	8	28	8	7	7	8						138	9.9	6.6
c)	.5	.4	.7	.2	.3	.6	.6	.6	.7	.7	.7	.6	.5	.4						-	0.5	0.2
<b>PBCL</b>																						
a)	13	15	7	6	17	17	20	9	28	10	21	15	17	16	11	11	5	9	247	13.7	5.9	
b)	5	5	4	1	6	4	12	5	6	4	5	5	5	7	5	5	4	3	90	5.0	2.0	
c)	.4	.3	.6	.2	.4	.2	.6	.6	.2	.4	.2	.3	.3	.4	.5	.5	.8	.3	-	0.4	0.2	
<b>PCL</b>																						
a)	21	13	29	10	12	15	-	23	10	17	13	11	9	4	11	10	9	17	234	13.8	6.1	
b)	9	8	21	7	7	8	-	17	8	7	12	8	6	4	10	6	6	6	150	8.8	4.3	
c)	.4	.6	.7	.7	.6	.5	-	.7	.8	.4	.9	.7	.7	1	.9	.6	.7	.4	-	0.7	0.2	

451  
 452 Based on the descriptive analyses shown in Table 13, BCL and PCL are the checklists that have similar  
 453 performance and, in percentage, generate more valid, operational and non-redundant requirements (rows  
 454 "c" of Table 13). The sample of subjects exposed to PBCL provided on average the lowest percentage of  
 455 valid, operative and non-redundant requirements compared to the total generated by each student. The  
 456 average, set to specifications having a size of 5 requirements, is less than the number of attributes used as  
 457 input for requirements identification and formalisation.

458 Figure 7 presents the distribution of specification sizes, with the same logic used for Figure 6 and Figure  
 459 5.

460 The distributions of probabilities are now substantially different from the ones considered in the above  
 461 diagrams. The highest threshold in terms of the final size of the specification is achieved using BCL, but  
 462 more than the 60% of the collected specifications did not reach a size larger than eight requirements.

463 The ANOVA test, reported in Table 14, still shows a statistically significant difference among the three  
 464 groups ( $F=11.608^*$ ,  $p\text{-value}=0.000$ ). Looking at the results in Table 15, the difference is significant in the  
 465 pairwise comparisons between PBCL vs BCL ( $p\text{-value}=0.044$ ) and PBCL vs PCL ( $p\text{-value}=0.000$ ). The  
 466 results are in line with the descriptive analysis reported in Table 13: there are no significant differences  
 467 between BCL and PCL ( $p\text{-value}=0.110$ ), as they generate on average the same percentage of non-  
 468 redundancy. However, BCL produces on average 15% more valid, operational and non-redundant  
 469 requirements than PBCL. PBCL, on the other hand, continues to perform worse than PCL, producing on  
 470 average 28% less non-redundant requirements. The Shapiro-Wilk test confirms the null hypothesis of  
 471 normality ( $p\text{-value}>0.05$ ).

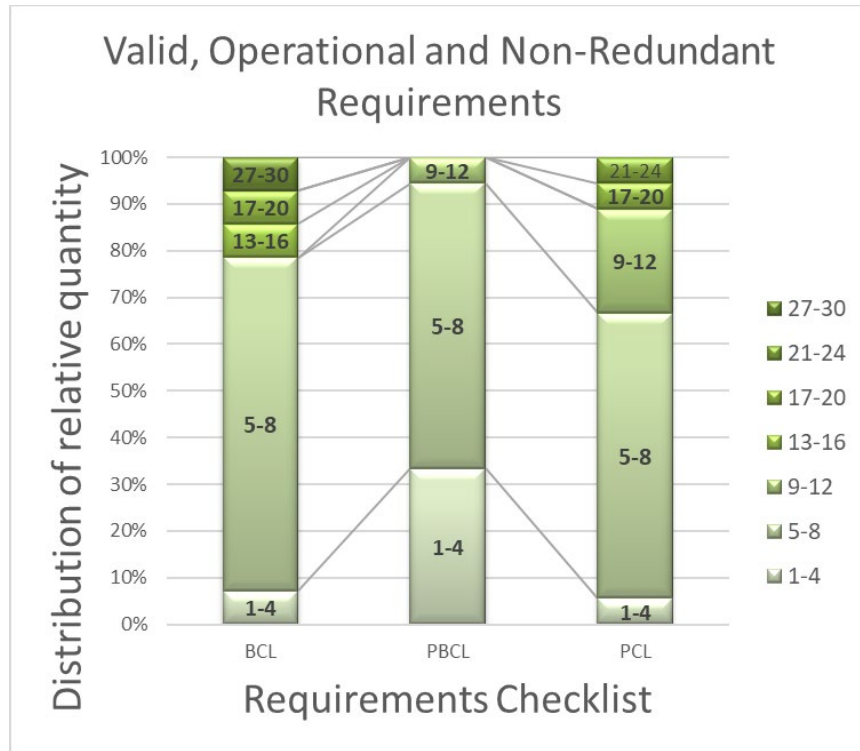


Figure 7. Distribution of Specification Sizes in Clusters of 4 Requirements, with Probabilities Which Are Proportional to Bar Heights and Colours Whose Intensity Grows with The Size of the Design Specification

Table 14 Summary of the ANOVA Test Using the Percentage of Valid, Operational and Non-Redundant Requirements (df – degrees of freedom; F – Fisher Snedecor statistics; Sig – Significance)

	Sum of Squares	df	Mean Square	F	Sig.
Between Groups	0.666	2	0.333	11.608	0.000
Within Groups	1.320	46	0.029		
Total	1.986	48			

Table 15 Summary of the Pairwise Comparisons Between the Three Checklists Using the Percentage of Valid, Operational And Non-Redundant Requirements (I – rows; J – columns)

		J				
		BCL	PBCL	PCL		
I	BCL	-	Mean diff. (I-J)	0.15*	Mean diff. (I-J)	-0.13
			Std. Error	0.06	Std. Error	0.06
			Sig.	0.044	Sig.	0.110
	PBCL	-	-	Mean diff. (I-J)	-0.28***	
				Std. Error	0.06	
				Sig.	0.00	
PCL	-	-	-	-		

The mean difference is significant at the \* p <0.05, \*\* p <0.01, \*\*\* p <0.001

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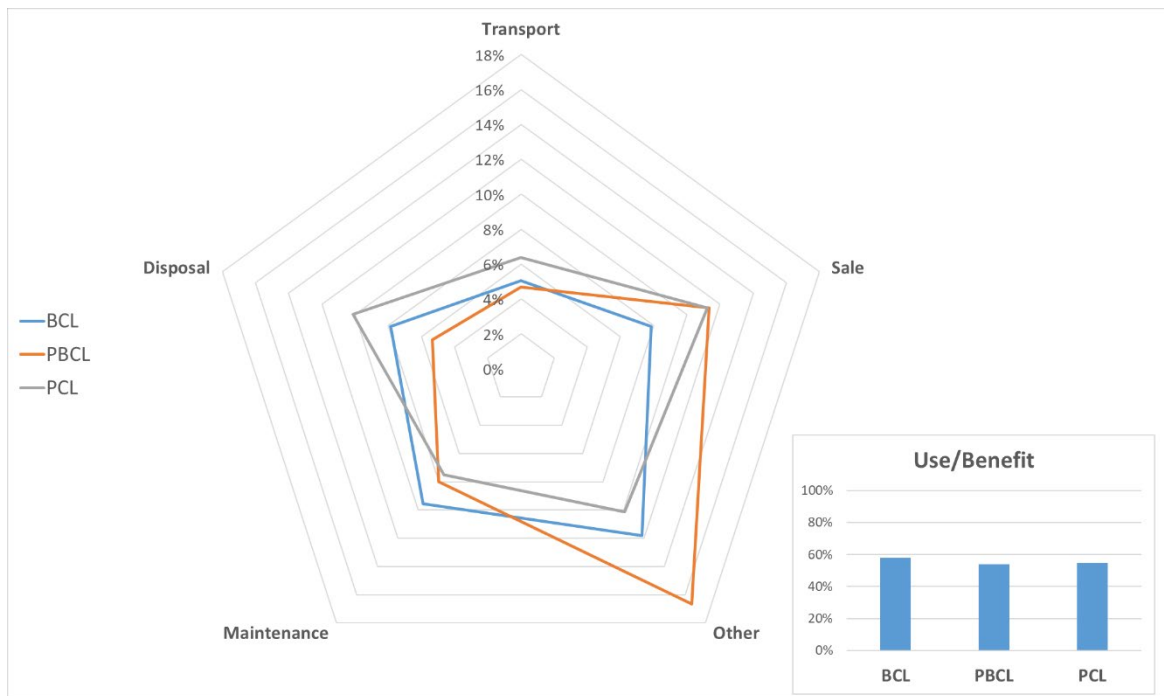
#### 482 4.2.4. Completeness

483 The results about the completeness metric follow a requirement-centred perspective, in place of subject-  
 484 centred. It means that results are still organised by checklist, but they refer to requirements as the collection  
 485 of all the requirements produced by the whole set of subjects participating in the study.

486 Figure 8 shows a radar plot of the distribution of requirements, consistently with the class of stakeholders  
 487 /solution's lifecycle stage they refer to, except for use stage. In fact, the requirements generated with all the  
 488 three checklists appear to focus particularly on the use stage, i.e. when the solution works to fulfil the need  
 489 of a beneficiary (which is not necessarily the user of the solution itself). For this reason, the values have  
 490 been separated in a histogram chart. Despite small differences, all of them trigger the identification of these  
 491 requirements in 50-60% of cases.

492 One of the most evident results is that all the three checklists do not trigger any particular stimulus  
 493 towards the definition of requirements concerning the transportation phase, despite there is an explicit  
 494 reference to this phase both in PBCL and in PCL. On the other hand, it is also worth noticing that PBCL  
 495 allows for a wider exploration of requirements classified under the label "other", at least with reference to  
 496 the whole set of requirements, when compared to the results obtained with BCL and PCL.

497 The results, then, show that the three checklists almost equally span the remaining classes considered  
 498 here, with no notable differences except for BCL in the "sale" category.  
 499



501 **Figure 8. Distribution of Requirements in Terms of Spanned Life Cycle Phases**

## 502 5. Discussion

### 503 5.1. Obtained results

504 Concerning productivity, the three checklists present some statistically significant differences,  
 505 highlighted in the summary of the results provided in Table 16 below.



506  
507

**Table 16 Summary of the Pairwise Comparisons**

Metrics	Results of the pairwise comparisons
Overall productivity	BCL>PBCL; BCL>PCL; PBCL<PCL
Validity	BCL>PBCL; BCL<PCL; PBCL<PCL*
Validity and Operationality	BCL>PBCL; BCL<PCL; PBCL<PCL***
Validity, Operationality and Non-Redundancy	BCL>PBCL*; BCL<PCL; PBCL<PCL***

The mean difference is significant at the \* p <0.05, \*\* p <0.01, \*\*\* p <0.001

508 Looking at the total amount of generated requirements, the difference between BCL, PBCL and PCL is  
509 not significant, according to the high standard deviation of the three distributions. It means that although  
510 some of the checklists provide different types of stimuli (see Figure 1), there is a high level of subjectivity  
511 in their interpretation and exploitation. Nevertheless, as shown in Figure 4, BCL presents higher  
512 potentialities for generating more populated checklists, suggesting that it has a (slightly) stronger  
513 effectiveness in stimulating designer’s formalization of design constraints.

514 When applying the filter of “validity”, both PCL and BCL show a higher percentage of valid  
515 requirements generated by students. According to Figure 5, the three checklists have (approximately) the  
516 same probabilities to generate requirements in the clusters 8-11 and 12-15, but PBCL is characterised by a  
517 higher probability to obtain the cluster 4-7, together with the absence of any subject capable to obtain more  
518 than 27 valid requirements. It implies that PBCL appears to be the less performant in terms of validity, also  
519 confirmed by the ANOVA test (see Tables 8 and 9). However, the significant difference is verified only  
520 between the PBCL and PCL. Very similar considerations can be made when applying the further filter of  
521 “operationality”, where PBCL still appears as the less performant (p-value<0.05) compared to PCL. These  
522 results can be explained by the extreme conciseness and simplifications characterising PBCL. If this can  
523 speed up the process; conversely, it can lead to misleading outcomes, especially if used by inexperienced  
524 users (as for the students participating at the experiment). In fact, in the context of this study, the guidance  
525 provided by PCL to inexperienced users resulted in a generation of a higher amount of valid and operational  
526 requirements, if compared to PBCL. When considering the additional non-redundancy filter, the worst  
527 performances of PBCL are even more extreme. The difference of performance resulted statistically  
528 significant also in comparison with BCL (see Table 15). Therefore, when dealing with a set of product  
529 attributes, PBCL demonstrated to be not particularly effective in the generation of a comprehensive design  
530 specification. This is consistent with previous results, which showed that the general productivity of PBCL  
531 is not particularly effective when used in conceptual design (Becattini et al., 2015). In general, the three  
532 checklists show a sensible and comparable reduction rate. PBCL, then, demonstrated to be the least capable  
533 of supporting the designer in the formalization of design constraints. The limited number of requirements  
534 in the related design specifications suggests that PBCL is also not particularly effective in supporting the  
535 externalization of internally perceived design constraints, despite a specific investigation of differences  
536 between internal and external constraints is left for future investigations.

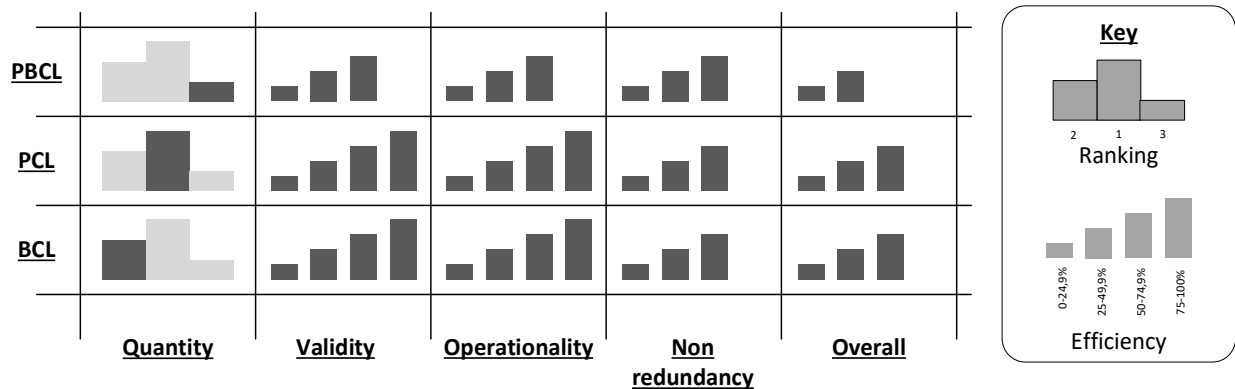
537 However, BCL and PCL are the most affected by the non-redundancy filter. Indeed, they have similar  
538 performances also confirmed by the absence of statistically significant differences in their comparison (see  
539 Table 15). This appears to be controversial, as these checklists have opposite characteristics (Figure 1). The  
540 results potentially show that there is an undisclosed variable (beyond the human factor) which is affecting  
541 the phenomenon and that still needs to be explored.

542 Resuming, PBCL triggered the definition of 247 items to populate the design specification. Once  
543 skimmed, just 90 of these items could be considered requirements by validity, operationality and non-  
544 redundancy (thus design task-related constraints). The ratio between the final and the initial set of  
545 requirements provides a preliminary, but quantitative, measure of the effectiveness of the checklist. The  
546 average efficiency of PBCL, therefore, is approximately 36% (90/247). BCL and PCL, instead, produce a  
547 much more comparable result: the efficiency of BCL is approximately 50% (138/269), while for PCL is  
548 64% (150/234). From this perspective, the best performance is achieved by PCL. As most of these

549 efficiency drops depend on the non-redundancy filter, it appears that in different ways these checklists can  
 550 trigger some fixation in the designer. This kind of fixation could be probably reflected on the solutions  
 551 generated to address the requirement list; nevertheless, these experimental data just show that the subjects  
 552 were focusing their attention on the same few problems.

553 For what concerns the completeness metric, Figure 8 shows that all the three checklists led the students  
 554 to generate requirements that focus on the same life cycle phase, i.e. the use of the product. This is a further  
 555 confirmation that the novice designers participating in the study paid particular attention to specific  
 556 objectives, potentially neglecting some others that are relevant and that facilitate the variety and the novelty  
 557 of the generated ideas.

558 The results of this experiments show that unexperienced designers should preferably use PCL, since it  
 559 has, overall, the best performance score. However, BCL is a potential alternative based on a completely  
 560 different strategy (see Figure 1), which also shows a good performance. Nonetheless, it is strongly based  
 561 on a specific definition of a function, which needs to be introduced to users before BCL use. In general,  
 562 under the limitations summarised in the following section, the investigation has shown that, at least for  
 563 novice designers, a checklist whose stimulation strategy uses examples and is too much dependent on the  
 564 context, seems to perform worse than other strategies that exploit sets of more general questions or abstract  
 565 stimuli. Notwithstanding, the investigated checklists organise the stimuli into categories that take into  
 566 consideration the several conditions and scenarios under which the system could operate, the lack of  
 567 completeness appears to be the negative aspect shared by all the considered approaches. Accordingly,  
 568 Figure 9 summarises the findings above discussed.



569

570 **Figure 9. Summary of the Performance for The Checklists Considered in The Study, According to**  
 571 **The Metrics Defined for The Assessment. The Checklists Are Distinguished by Their Capability to**  
 572 **Elicit Requirements as A Podium Ranking (Column: Quantity). An Efficiency Diagram, Described**  
 573 **As 4 Bars of Increasing Height for Increasing Performance Displays the Results for Validity,**  
 574 **Operationality and Non-Redundancy. The Last Column Shows the General Performance of The**  
 575 **Checklists as An Overall Indicator of Efficiency (ratio of “survived” requirements in the**  
 576 **specification after skimming).**

## 577 5.2. Limitations and future developments

578 A non-negligible limitation of the work is the limited number of checklists that have been experimented,  
 579 which inevitably reduces the general validity of the considerations presented in the previous section.  
 580 Widening the sample of tested checklists could strengthen the validity of the results, especially with  
 581 reference to the categories of strategies shown in Figure 1. Therefore, a first research work for the future  
 582 should repeat the same study with an enlarged sample of checklists. This can be done by taking into account  
 583 also contributions coming from other fields like software engineering, which is a sector very sensitive to  
 584 engineering requirements. This would also complement the viewpoint on tools to support designers in  
 585 externalizing internally perceived design constraints, which is what checklists aim at doing. The cross-

586 domain comparisons of checklists would also help understanding the differences between different levels  
587 of abstraction for design constraints, beyond the subtle differences emerged through the comparison of a  
588 checklist based on abstract stimulation (BCL) and one based on concrete examples (PBL).

589 Another limitation of the work comes from the predefined input delivered to students that introduced a  
590 common set of external constraints for all the participants (that could probably be the reasons of the shared  
591 focus observed in Figure 8). This need arose to ensure a common vision of the task, but it inevitably led to  
592 a condition that can sensibly differ from a real case. In addition, the considered sample of participants  
593 (engineering students from the same institution) implies non-negligible limitations. Indeed, besides the  
594 effects of their limited expertise (which led to results that could not be applicable to more experienced  
595 practitioners), it is also not clear to what extent the results are affected by ethnological aspects.  
596 Consequently, it is necessary to perform additional experiments extended to students from different  
597 institutions, disciplines and/or countries. This could be useful to obtain statistically significant evaluations,  
598 and, then, more robust indications for selecting the most suited checklist, according to the user needs.

599 Furthermore, besides the generic indications provided in this paper, it is still unclear whether more  
600 detailed and statistically significant correlations can be obtained between each metric presented in Section  
601 3 and parameters like the complexity of the task, the expertise of the designer and/or the design team and  
602 the type of product. Moreover, it is also unclear whether a specific checklist can be more suited for analysis,  
603 synthesis or evaluation purposes. This kind of information could pave the way for the development of  
604 criteria to identify the most effective requirement checklist for the design context that supports the related  
605 creative design activity.

606 Further development of the work could be the evaluation of the “quality” of the obtained specifications  
607 in terms of possible help or hindrance in exploring the design space, which is crucial to complement the  
608 missing step of this research: how a more or less complete design specification stimulates the generation of  
609 more creative ideas. Indeed, if on the one hand abstract and generic requirements may be useful in order to  
610 avoid undesired fixations (Vasconcelos et al., 2018; Jansson & Smith 1991) on specific designs; on the  
611 other hand, they can lead to several design iterations. However, excessively detailed specifications can lead  
612 to the opposite effects. The reduction of the efficiency for the whole design process due to fixation or  
613 iterations would require attention on the side of organizational (time-related) constraints. Therefore, future  
614 studies should perform additional investigations on design outcomes coming from design tasks where the  
615 requirements emerge from the use of specific checklists. To this purpose, well known creativity or idea  
616 generation effectiveness metrics can be successfully adopted (e.g. Shah et al., 2003; Sarkar & Chakrabarti  
617 2011; Nelson et al., 2009).

618 Eventually, another future development concerns the usability and the perceived onerousness of the  
619 checklists. To this purpose, future experiments could consider the use of simple surveys to be administered  
620 to participants, in order to extract the required information. Predefined frameworks like the NASA TLX  
621 (Hart, 2006; Sandra & Staveland, 1988) can address this objective.

### 622 5.3. Expected impact

623 The comparison performed in this work led to important indications about what checklist novice  
624 designers or students should use to start their design process, so that they have a clear guidance for the  
625 analysis, synthesis and evaluation of problems and solutions, despite this can be potentially boosting or  
626 undermining their creativity. Additionally, the experimental approach used for the work presented in this  
627 paper can be repeated with any design checklist. Moreover, the obtained numerical results might represent  
628 the first reference to run comparisons with results coming from further experiments using the same  
629 checklists.

630 Therefore, this work paves the way for future studies on requirement checklists, where one of the  
631 potential outcomes is the definition of a framework to start linking the design phases of product planning  
632 and conceptual design. To that purpose, the matrix shown in Figure 3 constitutes a preliminary tool proposal  
633 to help novices in better exploiting the potentialities of checklists.

634 Concluding, the experimental approach, the obtained results and the research hints provided in this  
635 section are expected to promote new research about how to support designers in defining comprehensive  
636 sets of requirements.

## 637 **6. Conclusion**

638 The paper aims at unveiling potential differences between requirements checklists, as they help the  
639 designer to set the initial constraints to steer the whole design process across its cycles of analysis, synthesis  
640 and evaluation. The identification and formalization of these design constraints help the designer to  
641 decompose the problem and reduce its complexity with a more efficient allocation of cognitive resources.  
642 The comparison among checklists for their applicability and related outcomes is here measured in terms of  
643 the metrics proposed by Roozenburg and Eekels (1991). More specifically, the metrics adopted in this paper  
644 were Validity, Operationality, Non-redundancy and Completeness. The group of 50 MS Mechanical  
645 Engineering students that took part in this study worked individually to generate a meaningful tentative  
646 design checklist. The outcomes of the experiment showed that there are differences among the three  
647 checklists here considered, i.e. Pugh's checklists (PCL), Pahl and Beitz's checklist for conceptual design  
648 (PBCL), and a recent design specification checklist as proposed in Becattini & Cascini (2013) (BCL). The  
649 benchmark between the checklists is consistent, as all the subjects participating in the study address the  
650 same design brief with the same set of proposed product attributes and translate it into a set of engineering  
651 requirements.

652 The obtained numerical results are, to the knowledge of the authors, one of the few quantitative examples  
653 of checklist application, despite the paper does not focus on how these checklists can stimulate a creative  
654 design process and/or outcomes. Nevertheless, the results showed that PBCL is not particularly effective in  
655 the generation of a sufficient comprehensive design specification as they collect a limited set of elements,  
656 that are poorly representative of the external and the designer's constraints. In other words, PBCL does not  
657 effectively support the formalization of design constraints, thus it probably poorly supports creative  
658 behaviour and outcomes for novice designers.

659 PCL and BCL, on the contrary, showed very similar performance overall, despite they have very  
660 different characteristics. PCL presents a set of questions that span the various life cycle stages of the solution  
661 to be designed (thus context specific categories). BCL, conversely, proposes a set of more abstract concepts,  
662 displayed as examples, which the checklist user has to contextualize to properly define requirements. BCL  
663 allowed the generation of a higher number of requirements per person than PCL, but the number of  
664 redundant items in the specification align the performance of the two checklists, whose results are,  
665 eventually, comparable. From the perspective of completeness, all the checklists appear to be mostly  
666 focused on the use stage of the solution lifecycle, while there are opportunities for the combined use of PCL  
667 and BCL in order to cover different moments of the solution life cycle and overcome potential lacks that  
668 might emerge in case just one of these checklists is used.

669 The results presented in the paper might represent the first benchmark to run comparisons with results  
670 coming from further experiments using the same checklists. Moreover, they could be useful for future  
671 activities focused on the development of guidelines for the selection of the most suitable design checklist.

672 Notwithstanding the above-mentioned results, some limitations might be ascribed to this work.  
673 Accordingly, the experimental subjects are students, thus not representative of the actual industrial  
674 perspective. Moreover, the initial design brief is quite simple, for a product with a relatively low degree of  
675 complexity (toothbrush). Therefore, the applicability of these results to the industrial context is not  
676 immediate. This requires additional investigations: with new experiments could be tailored to the specific  
677 engineering domain at hand as well as involve subjects with a higher level of expertise. Beyond what  
678 suggested here, the authors expect that peers interested in performing additional analysis and/or at  
679 validating the results presented in this paper, might reuse the presented research approach and the  
680 experimental protocol. Additionally, several hints for further research activities are provided in this paper,  
681 among which, that of investigating around the actual support that the obtained requirement lists can provide  
682 in terms of design space exploration.

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