

Equality in Cumulative Voting: A Systematic Review with an Improvement Proposal

K. Rinkevičs^a, R. Torkar^{a,b}

^a*Blekinge Institute of Technology, Sweden*

^b*Chalmers University of Technology and University of Gothenburg, Sweden*

Abstract

Context. Prioritization is an essential part of requirements engineering, software release planning and many other software engineering disciplines. Cumulative Voting (CV) is known as a relatively simple method for prioritizing requirements on a ratio scale. Historically, CV has been applied in decision-making in government elections, corporate governance, and forestry. However, CV prioritization results are of a special type of data—compositional data.

Objectives. The purpose of this study is to aid decision-making by collecting knowledge on the empirical use of CV and develop a method for detecting prioritization items with equal priority.

Methods. We present a systematic literature review of CV and CV analysis methods. The review is based on searching electronic databases and snowball sampling of the found primary studies. Relevant studies are selected based on titles, abstracts, and full text inspection. Additionally, we propose Equality of Cumulative Votes (ECV)—a CV result analysis method that identifies prioritization items with equal priority.

Results. CV has been used in not only requirements prioritization and release planning but also in e.g. software process improvement, change impact analysis and model driven software development. The review presents a collection of state of the practice studies and CV result analysis methods. In the end, ECV was applied to 27 prioritization cases from 14 studies and identified nine groups of equal items in three studies.

Conclusions. We believe that the analysis of the collected studies and the CV result analysis methods can help in the adoption of CV prioritization method. The evaluation of ECV indicates that it is able to detect prioritization items with equal priority and thus provide the practitioner with a more

fine-grained analysis.

Keywords: Cumulative voting, prioritization, requirements engineering

1. Introduction

Software products are becoming larger and more complex. Each product is usually affected by a large number of factors such as functional requirements, quality attributes, or software process improvement issues. Since time, funding, and resources are limited, it is seldom possible or even desirable to fully address all the factors. Therefore, the level of attention to a particular factor should be decided according to its importance (e.g. business value), cost, risk, volatility, dependencies between the factors and other such criteria. These type of decisions are made by product stakeholders: users, clients, managers, sponsors, developers, and other persons associated with the product. In order to make decisions regarding a large number of factors it is highly advisable to prioritize the factors in a systematic way [1].

Prioritization is commonly used in requirements selection and release planning. First, project stakeholders prioritize software requirements. Priority values then can be used to determine the order in which the requirements are going to be implemented. Requirements with higher priority could be implemented early while requirements with lower priority may be postponed for later releases or left out.

Another example could be prioritization of potential security threats. It is done by security professionals, software developers and system administrators to assess the level of risk and to select risk mitigation activities.

One of the prioritization methods used in software engineering is Cumulative Voting (CV) [2]. The main advantage of CV is that it is relatively simple and fast, yet produces priorities in ratio scale [1, 3]. This allows us to not only determine what prioritization items are more important but also how much more important they are. (Ratio scale prioritization is particularly important in software release planning and cost-value analysis [4, 5].)

Prioritization is usually performed by multiple stakeholders where individual priorities are combined into a single priority list. Each stakeholder's preferences may have different weight in the final priority. Such prioritization provides more information than just the priorities of factors. In the end, it may be useful to analyze the results of the prioritization to assess disagreement between stakeholders, measure stakeholder satisfaction with the results

or find distinct groups of stakeholders.

The purpose of this study is to help industry practitioners and academia researchers in adopting, using and developing CV, while the importance of prioritization in software engineering and the prospectiveness of CV constitutes a need to do further research in this area.

This study presents a systematic literature review on the empirical use of CV and CV result analysis methods. CV results correspond to special type of data—compositional data. Principles of compositional data analysis are described in this paper. A new method for CV result analysis, called Equality of Cumulative Votes (ECV), is proposed. The method identifies prioritization items with *equal* priority. ECV is evaluated using a considerable amount of data, which was obtained from the primary studies identified by the systematic review (through the kindness of the authors of said studies).

The remainder of this paper is structured as follows. We introduce definitions and place this study in a context in the next section. In Section 3 we give a short presentation of related studies. In Section 4 research questions and the methods used in this study are presented. In Section 5 the execution of the systematic literature review (SLR) is presented; however, we wait with presenting the results of the SLR. In Section 6 the design of our method of analysis, Equality of Cumulative Votes (ECV), is given, while the results of the SLR and the corresponding evaluation of ECV are presented in Section 7. Section 8 provides discussions, presents threats to validity and concludes the paper.

2. Background

This section presents definitions and places this study in a context. In the coming sections we will cover: a description of software requirements prioritization methods; examples of CV result analysis methods; and a description of compositional data analysis and CV.

2.1. Prioritization Methods

Some of the most popular prioritization methods are the analytical hierarchy process (AHP), cumulative voting (CV), ranking, numerical assignment, top-ten, the planning game, minimal spanning tree, bubble sort and binary search tree [1, 6]. Ranking and numerical assignment methods perform prioritization on an ordinal scale. AHP and CV are, on the one hand, considered

to be harder to use and also more time consuming compared to other methods but, on the other hand, produce priorities in ratio scale.

Ratio scale priorities have several advantages over ordinal scale priorities. Ratio scale shows not just the order of items but also relative distance between them. This enables the priority of a group of items to be calculated by summing up the priorities of individual items [4]. It is possible to say that one item or set of items has higher priority than another set of items. Supposing stakeholders have to choose between several low priority items and one item with higher priority; with ordinal scale, the item with highest priority will always be selected first. However, if priorities are given on a ratio scale, it is possible that lower priority items will be selected if their cumulative priority is higher.

Finally, the ratio scale allows the combining of multiple priority factors by calculating ratios between them. One example of this is the cost-value ratio that shows which requirements give more value for less money [5].

2.2. Prioritization Result Analysis

Disagreement between stakeholders happens when two or more stakeholders have assigned a different priority to one prioritization item. If the level of disagreement is high it may indicate potential conflicts between stakeholders. Such conflicts may be of technical character, as well as social or cultural.

The satisfaction a stakeholder has with the final prioritization results is determined by the difference between the results and the individual priorities of the stakeholder. A smaller level of difference leads to higher satisfaction. In the end, stakeholder satisfaction is important because it is necessary to achieve stakeholder commitment.

In some cases a part of stakeholders may form a group of some kind and, therefore, prioritize requirements similarly. It may be useful to detect whether a group of stakeholders has different preferences compared to other stakeholders. As an example, in [7], domain experts, technical experts, managers, project managers, testers, and developers use CV to prioritize software process improvement issues and the CV results are analysed using disagreement charts and satisfaction charts. Finally, principal component analysis (PCA) is used to identify distinct groups of stakeholders.

The same items can be prioritized by the same stakeholders multiple times from different perspectives. In this case it is useful to determine correlation between the priorities in different perspectives to assess the differences between the perspectives. As an example, in [8], CV is used by developers,

105 testers and managers to prioritize quality attributes. The same quality at-
106 tributes are prioritized from two perspectives: the perceived situation today
107 and the perceived ideal situation. Correlation between the two perspectives
108 is evaluated using the Spearman rank correlation matrix. This allows an
109 analysis of how well the company balances the priorities of software quality
110 attributes.

111 In [9] change impact issues are prioritized by developers, testers, man-
112 agers, and system architects. The prioritization is done with respect to three
113 perspectives: strategic, tactical, and operative. In order to determine corre-
114 lation between the perspectives, CV results are analyzed using the Kruskal-
115 Wallis test. In [10] the results of [9] are further analyzed using PCA, bi-plot,
116 and ternary plot. In this case, PCA is used to find correlated issues, bi-
117 plot shows variance, correlation, difference between the priorities of issues,
118 and the viewpoints of stakeholders, while ternary plots are used to show the
119 relative number of issues that received high, medium, and low priority.

120 As can be seen above, from the examples above, prioritization has been
121 performed with various stakeholders, using different perspectives and, in the
122 end, also analyzed using various techniques. We will next describe in more
123 detail one of the more common methods to manage prioritization issues—
124 cumulative voting—which has been used in software engineering for some
125 time. (CV has its roots in corporate governance and biology.)

126 2.3. Cumulative Voting

127 CV is a prioritization method for prioritizing a list of items [2] and has
128 been studied and applied in various fields.

129 In forestry it is used to take into account opinions of different parts of
130 society while planning forest harvesting [23]. CV has also been used as a
131 voting mechanism in government elections [24] and to aid decision making
132 in corporate governance [12]. In computer science we have seen CV being
133 part of various software algorithms, e.g. in [25] it is used as part of pattern
134 detection algorithm that is used to locate the optic nerve in a retinal image.

135 In software engineering CV has been applied not only in requirements
136 engineering and software release planning [26] but also in software security
137 [27], software quality [8], software metrics [28], software process improvement
138 [7], and software verification and validation [29].

139 Studies have also used CV as part of a research method itself. For in-
140 stance, in [30] software impact analysis issues are elicited in structured in-
141 terviews and afterwards the importance of each issue is determined with the

142 help of CV. Whether CV has been used in a particular domain or as part of
143 a methodology is in itself quite irrelevant as long as one takes into account
144 the type of data CV results consist of.

145 CV has many synonyms in literature: hundred (100) dollar (\$) method/
146 test and hundred (100) point method.

147 In CV a stakeholder is given 100 points, imaginary dollars or units of
148 percentages that can be spent on the prioritization items. In the simplest
149 case, the stakeholder can spend any amount of points on any number of items
150 as long as the total amount adds up to 100. The more points assigned to an
151 item, the higher the priority of the item (and implicitly, the lower priority
152 to the other items). The stakeholder may spend all points on just one item
153 or distribute them among all or some of the items. Once again, this is the
154 simplest case; other variants exist, which we will see next.

155 Often prioritization is done by more than one stakeholder. The final prior-
156 ity of an item can be calculated by adding up the points each stakeholder has
157 spent on it. Sometimes the vote of some stakeholders may be more important
158 than the votes of others. For example, a manager may be more influential or
159 shareholders may have different amount of shares. In such a case the prior-
160 ities of each stakeholder may be multiplied by an individual coefficient or a
161 stakeholder may be given a more points to perform the prioritization.

162 Worth mentioning in this context is that it is advisable to randomize the
163 order of items in a prioritization list. This is necessary in order to minimize
164 the effect of order on the prioritization results, which has shown to have an
165 effect [16].

166 2.3.1. *Benefits and Drawbacks of Cumulative Voting*

167 Compared to analytical hierarchy process (AHP), CV is faster and easier
168 to learn and use [1, 3]. AHP benefits from consistency check, but CV does
169 not require this because all prioritization items are evaluated simultaneously
170 [3].

171 There are, however, a few problems with CV. First of all, it cannot be
172 repeated for the same stakeholders and prioritization items due to stakeholder
173 bias [2] (c.f. Section 2.3.4). Secondly, CV becomes more difficult to use when
174 the number of prioritization items increases [17].

175 2.3.2. *Example of Cumulative Voting with Several Stakeholders*

176 Let us next give an example of CV with several stakeholders. Suppose
177 Robin, Alice, and John are three friends who want to buy some beverages in

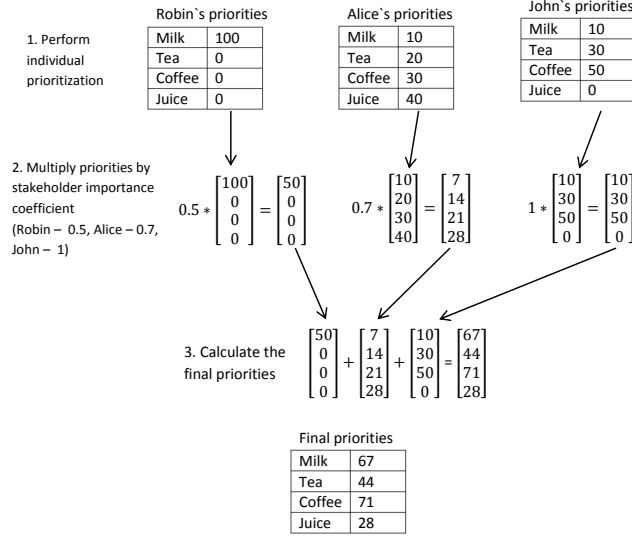


Figure 1: Example of CV with several stakeholders.

178 a store. They have different preferences but do not want to buy too many
 179 drinks. Therefore, they decide to use CV to decide what to buy. Each of
 180 the friends distributes 100 points between four items: milk, tea, coffee, and
 181 juice (Step 1 in Figure 1). In this case each of them will spend a different
 182 amount of money on the purchase, hence, their priorities are multiplied by
 183 different coefficients (Step 2 and the stakeholder importance coefficient in
 184 Figure 1). The final beverage priorities are calculated by summing up the
 185 weighted priorities of stakeholders (Step 3 in Figure 1).

186 2.3.3. Stakeholder Bias

187 Prioritization using CV may be biased if a stakeholder knows the pref-
 188 erences of other stakeholders. She may manipulate the results by spending
 189 more points on items that are important to her but not to the other stake-
 190 holders. On the one hand, stakeholder bias makes it unreasonable to repeat
 191 CV with the same prioritization items and stakeholders. On the other hand,
 192 this property of CV may be useful in giving more power to important mi-
 193 nority stakeholders, such as security experts or software testers. Suppose the
 194 same software requirements are prioritized for a second time using CV. A
 195 developer might know that all vital functionality is selected by other stake-
 196 holders, but his toy feature is left out. In effect, the developer could spend

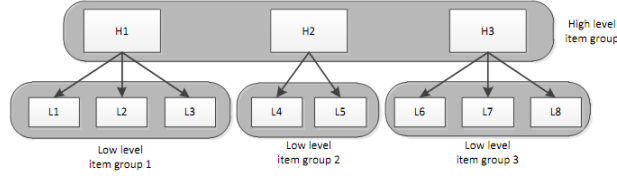


Figure 2: Example of prioritization item hierarchy.

all his points on this feature to put it in the next release.

Stakeholder bias may be mitigated by setting a maximum priority that can be assigned to an item. This way each stakeholder is forced to distribute the money between several prioritization items [4].

Another bias is that people in general tend to assign round priority values. This is likely caused by lack of objective judgement criteria. Either way it seems to be a problem not acknowledged by many since all prioritization is largely based on expert opinion.

2.3.4. Scalability of Cumulative Voting—Hierarchical Cumulative Voting

The standard CV approach has a low scalability. If the number of prioritization items is high, stakeholders may lose sight of the bigger picture and assign priorities to a limited number of items. One, unsophisticated, solution to the problem is to provide more points for prioritization (1,000 or 10,000 instead of 100); however, one could take another approach.

When the number of prioritization items is high they can usually be grouped hierarchically by forming a tree structure (Figure 2) and, thus, parent-child dependencies will exist between many items.

In [4] the authors propose a method for prioritizing hierarchically structured items called Hierarchical Cumulative Voting (HCV). It may be seen as combination of the hierarchical part of the Analytical Hierarchy Process (AHP) [1, 18] and the CV prioritization method. Since items are prioritized in smaller sets, stakeholders do not lose sight of the bigger picture during prioritization, and the prioritization of a large number of requirements is considered easier.

2.3.5. Compensation Factors

HCV deals with the problem of prioritization scalability but it comes at a cost. Low level item groups may consist of different numbers of items, but the number of points spent on each group is the same, i.e. in a small-sized

group, the same amount of points is distributed among fewer items. Hence, items in smaller groups are statistically more likely to have a higher priority, on average, compared to items in larger groups. To balance this difference each low level prioritization item can be multiplied by a compensation factor [4].

As an example, suppose an item (A) in a group of 10 items is assigned 60 points. Hence, A will receive 600 compensated points. In this case it is impossible for any item in a group smaller than 6 items to compete with A . Even if item (B) in a group of 5 is assigned the maximum number of points (100), the maximum compensated priority value B can receive is 500.

In [17] the authors suggest that compensated prioritization is more favorable compared to uncompensated. But neither compensated nor uncompensated prioritization is perfect and, as a general rule, it is better to keep the size of prioritization item groups similar.

2.3.6. HCV Execution

According to [4], HCV is conducted with the following steps (Steps 4–5 are optional):

1. Construct hierarchy. Prioritization items need to be divided into one high and several low level item groups. Each low level item group is child to exactly one high level item. And each high level item has one low level item group. One low level item may belong to several item groups. Even if parts of the items are not logically connected they can be grouped separately and assigned a fake parent item, e.g. ‘misc. items’. HCV does not, as far as we know, provide any instructions for creating a requirements hierarchy.
2. Each high and low level item group is prioritized separately using CV. The stakeholder may prioritize all item groups at once or one by one. But it should be possible to prioritize groups in any order and repeatedly, because the stakeholder might learn more about the items while performing the prioritization.

In particular the stakeholder is likely to learn more about a high level item when prioritizing its low level item group [17]. Some stakeholders may prioritize only part of the groups and each group may be prioritized by different stakeholders.

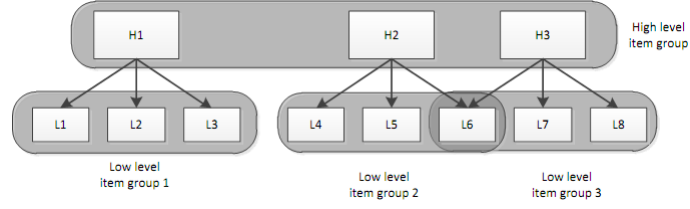


Figure 3: Overlapping prioritization item hierarchy example.

- 259 3. The priority of each low level item is normalized by dividing it with
260 the sum of all low level priorities of each item in all groups.
- 261 4. The final priority of each low level item is calculated by multiplying it
262 with the priority of its parent high level item.
- 263 5. Then one applies the compensation factor to all low level requirements
264 as described in Section 2.3.5.
- 265 6. Finally, when multiple stakeholders have performed the prioritization,
266 priorities of low level items are combined as in standard CV.

267 It is possible that one low level item is child of more than one high level
268 requirement and, thus, belongs to two or more low level requirement groups
269 (see Figure 3). Such requirements participate in the standard HCV prioritization
270 process and are prioritized two or more times with each group they
271 belong to. At the end of the prioritization they receive several priority values.
272 These values can be summed together to form the final priority of the item.
273 (This is done because the item adds value to both parts of the hierarchy.)

274 2.3.7. Example of Hierarchical Cumulative Voting

275 Suppose six requirements for a mobile phone operating system need to be
276 prioritized: ‘reminder alarm’, ‘specify repeated event’, ‘hide contact’, ‘add
277 picture to phonebook’, ‘search contact’, ‘make video call’. Three high level
278 requirements can be identified: ‘Calendar’, ‘Phonebook’, ‘Call’. The low level
279 requirements are then grouped as sub-requirements of high level requirements
280 as shown in Figure 4. The ‘Search contact’ requirement is a sub-requirement
281 and has two parent requirements: ‘Phonebook’ and ‘Call’. The computation
282 of the final priorities of requirements is shown in Table 1.

283 After requirements are grouped, and a hierarchy is defined, each group of
284 requirements are then prioritized using CV. The final priority of a low level

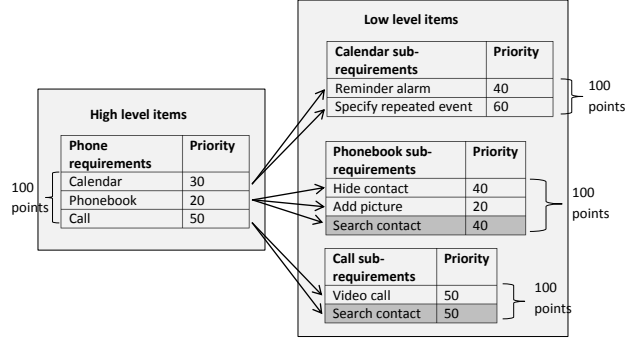


Figure 4: Example of hierarchical cumulative voting with requirement hierarchy.

Table 1: Example of hierarchical cumulative voting.

Phone requirements	Compensation factor	Sub-requirements	Priority calculation	Final priority
Calendar	2	Reminder alarm	$40 \times 30 \times 2$	2400
Calendar	2	Specify repeated event	$60 \times 30 \times 2$	3600
Phonebook	3	Hide contact	$40 \times 20 \times 3$	1600
Phonebook	3	Add picture	$20 \times 20 \times 3$	800
Phonebook & Call	3 & 2	Search contact	$40 \times 20 \times 3 + 50 \times 50 \times 2$	7400
Call	2	Video call	$50 \times 50 \times 2$	2500

285 requirement is computed by multiplying the priority of the requirement with
 286 the priority of its parent high level requirement and the compensation factor.
 287 The compensation factor in this particular case is the number of elements in
 288 a group, two for the ‘calendar’ and ‘call’ sub-requirements and three for the
 289 ‘phonebook’ sub-requirement.

290 2.4. Compositional Data Analysis

291 CV results can be seen as a special type of data, i.e. compositional data.
 292 Compositional data does not contain absolute values. It shows only the
 293 relative weight of a component compared to the whole. In [10] the authors
 294 propose the use of compositional data analysis for the statistical analysis of
 295 CV.

296 A compositional data item is a vector (x) of positive components with a
 297 constant sum k :

$$x = (X_1; X_2; \dots; X_n) \text{ where } x_i \geq 0 \text{ and } \sum_{j=1}^n x_j = k. \quad (1)$$

298 The property of the sum of the items being restricted is called the constant
 299 sum constraint. In CV, priorities assigned by a stakeholder to the items of
 300 a prioritization set is a compositional data vector with a constant sum of
 301 100. The value of k (i.e. 100 in this case) is arbitrary and does not affect
 302 the analysis of the data because the information is contained in the ratios
 303 between the components of the vector. The vector can sum up to any number
 304 but still hold the same data, i.e. vectors (1, 2, 7) and (10, 20, 70) are in this
 305 case considered equivalent. This principle is called *scale invariance*.

306 Another property of compositional data items is *subcompositional coher-*
 307 *ence*. Consider that two compositions are analysed. One composition is a
 308 subcomposition of the other. *Subcompositional coherence* means that the re-
 309 sults of the analysis are the same for the common parts of the compositions
 310 [19]. This property is important for the analysis of HCV results. Statements
 311 that are made regarding each smaller group of prioritization items are also
 312 true for all items prioritized with HCV.

313 The priority of an item is relative to the priority of the other items in
 314 the set. Hence, the priority of an individual item is meaningless without
 315 context, i.e. the complete set of items. The same item may receive different
 316 priority when put in two different prioritization sets. If the item is put in a
 317 set of items with high priority it will receive a lower relative priority. This
 318 also holds true the other way around i.e. if the item is put in a set with low
 319 priority items its priority will be higher.

320 When doing analysis of compositional data one must take into account
 321 that compositional data special type of data and should be analysed differ-
 322 ently than ordinary data.

323 When doing analysis of compositional data one must take into account
 324 that compositional is a special type of data and should be analysed differently
 325 than other data types. Ordinary unconstrained variables are free to take any
 326 positive or negative values, whereas, compositional data values can only be
 327 positive and have a constrained maximum value. Moreover, components of
 328 compositional data vectors are not independent from each other. The fact
 329 that an item is assigned 70 priority points means that the next item can take
 330 only values between 0 and 30. Hence, there is a negative correlation between
 331 the items.

Standard parametric statistical tests require that data vectors have multivariate normal distribution. Vector $X = (X_1, X_2, \dots, X_n)$ is considered to have multivariate normal distribution if any linear combination of its parts is normally distributed, and linear combination is defined by:

$$Y = a_1X_1 + a_2X_2 + \dots + a_nX_n, \quad (2)$$

where Y is the product of lineal combination and a_i is any real number. Now, since the sum of priorities assigned in CV must add up to 100, or any other constant number, at least one linear combination of X is not normally distributed because it always adds up to 100:

$$Y = 1 \cdot X_1 + 1 \cdot X_2 + \dots + 1 \cdot X_n = 100. \quad (3)$$

In our opinion, the above indicates, quite strongly, that CV results do not follow a multivariate normal distribution and, hence, it follows that they should be analyzed using non-parametric statistical tests [20].

2.4.1. Problem of Zeroes

Compositional data analysis requires that log-ratios between any components in a vector can be computed. But computing a log-ratio with a zero value is, in this case, meaningless. This is a problem since CV allows stakeholders to assign zero priorities to some prioritization items (we would even strongly argue that this is very common).

In compositional data there are two types of zeroes: essential and rounded. Essential zeroes mean that a data component is not present. Rounded zeroes mean that the component is present but its value is very low. We, as others have before us, conjecture that zeroes in CV results are rounded because the priority of an item is a completely abstract notion and the instrument for measuring priority is human judgement [10].

Before compositional data analysis can be applied to CV results, we should first remove zeroes in the data. One approach can be to forbid stakeholders to assign zero priorities. This approach is used in e.g. [7]. But this can add some unnecessary complexity to the prioritization process and, explicitly, delimits an expert's freedom. In [10] the authors propose the use of a multiplicative replacement strategy (as defined in [21]) for CV result analysis.

This method replaces rounded zeroes with small values using the expression

$$r_j = \begin{cases} \delta_j, & \text{if } x_j = 0, \\ (1 - \frac{\sum_{k|x_k=0} \delta_k}{c})x_j, & \text{if } x_j > 0, \end{cases} \quad (4)$$

where δ_j is the imputed value and c is the constant sum constraint. In order for the total sum of components to stay constant, the equation subtracts some value from the items with a priority higher than zero. More is subtracted from components with higher values than from components with lower values (and the value of the imputed δ_j is arbitrary).

2.4.2. Isometric log-ratio transformation

In order to apply standard statistical methods to compositional data it should be transformed to remove the inherent correlation of the values. Compositional data analysis proposes special transformations that change the compositional data values to unconstrained real values. One such transformation is isometric log-ratio (*ilr*) transformation (as proposed by [20, 22]).

After compositional data vectors are transformed using zero replacement and *ilr*, any standard statistical tests can be applied.

3. Related Work

In the previous sections we introduced requirements prioritization methods, some examples of CV result analysis methods and a more detailed description of compositional data analysis and CV.

In this section we only present systematic literature reviews performed in this field and how they relate to our study.

A systematic review of requirements prioritization methods is presented in [33]. The study focuses on prioritization method comparison and selects eight relevant studies. Two of the studies use CV. These two studies are also included in the systematic literature review conducted as part of this study. In [33] the author concludes that there is little research on requirements prioritization and studies usually deal with a small number of requirements.

In the next section we will cover the methodology of this study. As will be presented later, the systematic literature review had two purposes: to assemble data that have been used in CV and to investigate if there existed a method of analysis that would identify prioritization items with equal priority.

392 4. Methodology

393 This section covers the research questions of this study and the methods
394 used to answer them.

395 4.1. Selection of Research Methods

396 The main purpose of this study is to collect knowledge on the use of CV
397 in order to help software engineers and researchers in adopting it.

398 One way of collecting this knowledge is to conduct an empirical study. A
399 survey in a large number of software companies can be used to quantify the
400 level of adoption of CV in industry (similarly to the study by [34]), while a
401 case study can be used to receive qualitative feedback on the use of CV [35].

402 Knowledge on the empirical use of CV can also be obtained from existing
403 studies. This may be done by means of a systematic literature review. Several
404 studies have used CV in industry as well as in academic settings. Neverthe-
405 less, there are no studies that provide an overview of the current state of the
406 practice in this field (as reported by research studies). Therefore, before con-
407 tinuing with the refinement of CV and conducting new empirical studies (i.e.
408 case study or experiment), a systematic literature review would be required.

409 This paper proposes a new method for CV result analysis, called Equality
410 of Cumulative Votes (ECV). (ECV groups prioritization items into groups of
411 items with similar priority.) As will be presented later, the systematic review
412 did not reveal any methods that solve this problem; however, ECV needs to
413 be evaluated and, hence, applied to CV results.

414 There are two options to obtain CV results in order to test ECV. One is
415 to conduct a new empirical study. The second option is to collect CV results
416 from existing studies. The latter approach also has the added benefit of
417 trying to replicate the results from previous studies and, if data from several
418 other studies are used, a larger amount of data can be obtained. Moreover,
419 the generalizability of the evaluation increases when prioritization results
420 from different sources and domains are used. On the other hand, the main
421 benefit of conducting a separate empirical study is the possibility to control
422 the conditions of CV.

423 In our study we evaluated ECV by obtaining data from previously con-
424 ducted studies as found by the systematic literature review. In order to
425 obtain the data, authors of relevant primary studies were contacted.

426 In short, this study consists of two parts: a systematic literature review
427 (SLR) of CV and an evaluation of ECV based on the data from the primary

428 studies found in the SLR.

429 *4.2. Research Questions*

430 The systematic review should focus on catching studies that empirically
431 use CV. Information about place, time, scale, and domain of the studies
432 should be collected and the results of the review will hopefully aid academic
433 researchers by identifying paths for further investigation of CV. Hence, the
434 first research question is:

435 **RQ 1.** What is the state of practice in empirical studies that use CV?

436 The level of trust in research results considering CV is determined by the
437 quality of the studies that use CV, hence this study includes an evaluation
438 of the quality of primary studies identified by the systematic review.

439 Next, a valuable aspect of decision-making is the analysis of prioritization
440 results. Thus, the second research question is:

441 **RQ 2.** What CV result analysis methods have been presented in papers as
442 identified by RQ 1?

443 Finally, the evaluation of ECV answers the third research question:

444 **RQ 3.** Is ECV capable of identifying prioritization items with equal priority?

445 **5. Systematic Literature Review**

446 This section presents the design of the systematic literature review. For
447 the results of the execution please see Section 7.1 and 7.2.

448 Table 2 presents an overview of activities performed during the systematic
449 literature review. The review protocol was developed by one researcher and
450 evaluated by another researcher. Studies were searched for in two iterations.
451 The first search was performed using databases. The second search was
452 performed using snowball sampling [36] (snowball sampling examines the
453 references of primary studies revealed by the first search). References that
454 are relevant to the review, i.e. they pass the selection criteria, are then added
455 to the set of primary studies.

456 The search for papers was performed by a single researcher. Study se-
457 lection, on the other hand, was performed by two researchers. First, one
458 researcher examined all found studies. Next, another researcher re-examined

Table 2: Review activities.

Review phase		Researchers involved
Trial search in databases		A
Develop review protocol		A
Evaluate review protocol		B
Paper search and selection from databases	Search in databases	A
	Search string validation	A
	Selection based on metadata	A and B
	Selection based on full text	A and B
Pilot data extraction (3 papers)		A
Paper selection from the reference lists	Selection based on metadata	A and B
	Selection based on full text	A and B
Data extraction		A and B
Data synthesis		A

all studies classified as primary studies in addition to 20 randomly selected excluded studies to ensure the quality of the selection.

To ensure the quality of the review, the quality evaluation and data extraction was performed independently by two researchers. Inter-rater analysis was performed using Krippendorff’s Alpha statistics [37, 38].

5.1. Data Sources and Search Strategy

The SLR was designed based on the guidelines by Kitchenham [39]. First a trial search in electronic databases was conducted. In order to scale the review to a manageable, yet sufficient size, databases were searched with different search strings. Relevant papers that were found during the trial search were used to extract additional search strings. The trial search revealed that the number of studies that use CV is not very large. Therefore, we decided to include not only software engineering studies but also studies in other research areas, such as forestry or corporate governance, since one key aspect we intended to investigate was analysis methods for CV.

Since CV is frequently used in studies without mentioning this in the

abstract, full text search in databases is preferable. Unfortunately not all databases support full text search. Full text search was performed in the IEEE Xplore and Springer Link databases. In ACM Digital Library, Inspec/Compendex, ISI Web of Knowledge, and SCOPUS only metadata was searched. The search strings used, consisting of a Boolean expression (A or B or C or D or E or F or G), where:

- | | |
|-----------------------|---------------------------|
| (A) Cumulative voting | (E) hundred dollar method |
| (B) 100 dollar method | (F) hundred dollar test |
| (C) 100 dollar test | (G) hundred point method |
| (D) 100 point method | |

Search strings contained only synonyms of CV and they did not limit the research area to software engineering. The search was performed independently using each of the search strings in each database. All search results were combined and documented using reference management software. The quality of the search strings and the selection of electronic databases were validated against a previously known core set of papers—[3, 10, 28, 40]—checking that all papers from the core set were found by the search.

5.2. Study Selection

To select relevant papers a set of criteria were designed. The criteria for paper selection are presented in Tables 3 and 4.

Papers were selected in two phases: based on metadata and based on full text.

Obviously, the main criterion for inclusion of a paper is that it must present empirical use of CV or present an analysis of the results of using CV. However, there are papers that pass this criterion but are not relevant for this review. CV is frequently used in computer algorithms. There is a significant difference between the way humans and computers make decisions. Since this review is concerned with human decisions we excluded papers that present CV that is not performed by humans. In addition, only papers that were written in English were selected and duplicate studies were automatically excluded by the citation management software used in this review. We searched for papers between 2001–2011. By then performing a snowball sampling of these papers we are convinced that we have a representative sample

Table 3: Paper search and selection in the databases.

Selection phase	Inclusion criteria	Number of papers selected
Search in databases	published 2001–2011 (databases last accessed Feb. 20, 2011)	256
	contains search strings	
Selection based on metadata	exclude duplicates and tables of contents	177
	written in English	
Selection based on full text	full text is available	127
	study involves empirical use of CV or presents analysis of empirical use of CV	58
	CV is done by humans and not software	25

Table 4: Paper selection from the reference lists of the selected papers.

Selection phase	Inclusion criteria	Number of papers selected
Selection from references	papers included in the reference lists of relevant papers found in databases	467
Selection based on metadata	written in English	462
	reference is already revealed by search in databases	450
Selection based on full text	full text is available	329
	study involves empirical use of CV or presents analysis of empirical use of CV	15
	CV is done by humans and not software	

511 and, furthermore, that the bulk of the studies are relevant from a software
512 engineering perspective.

513 5.3. *Quality Evaluation*

514 The goal of quality evaluation is to determine the best primary studies
515 according to some measure of quality. Since the number of studies that use
516 CV is not large, quality evaluation was not used as an exclusion criterion.

517 The quality of a study obviously depends on the correctness of the study
518 process including planning, operation, analysis and interpretation of the re-
519 sults (is the study right?) The correctness of the process can be measured
520 by evaluating the description of the study or replicating the study. Thus,
521 to gain the trust of industry practitioners and other researchers, the process
522 of the study should be rigorously described. In short, the description has to
523 facilitate the replication of the study as well as the presentation of limitations
524 and validity threats.

525 Even the most correct and rigorously described study is useless if it does
526 not contribute to the industry or research community (is it the right study?)
527 The topic of the research ought to address important goals and issues. The
528 findings of the study should also be significant, i.e. there is a high probability
529 of the results of the study are true. The significance of the findings depends
530 on how realistic the study is, the correctness of the process and the results
531 of the study, as well as the statistical significance of the findings.

532 **Realism** of a study depends on the context, scale, and subjects of the
533 study. The study should be conducted in a **setting** that is similar or equal
534 to the setting in which the findings of the study are intended to be used.
535 Hence, studies that are conducted in an industrial setting are in many cases
536 valuable. The **subjects** of a study should be similar to the people who are
537 supposed to use the findings of the study. The subjects ought to have appro-
538 priate work experience, role in the organization, skills, cultural background,
539 motivation, and so forth. The **scale** of a study refers to the size of the study
540 objects. In the case of this systematic review the scale of a study is mea-
541 sured as the number of prioritization items. Study in academia may have a
542 large number of prioritization items. At the same time, an industrial study,
543 with professionals as subjects, may involve a smaller number of prioritization
544 items.

545 Each study may have a different level of realism. Some studies involve
546 industry practitioners in an academic setting to simulate real word practice in
547 a laboratory environment. Other studies may involve academic researchers
548 that execute a project. For example, researchers may be developing open
549 source software. On the reality scale these studies are somewhere in between
550 the purely academic and industrial studies.

551 The **type** of the research study can be considered as a criterion for the
552 evaluation of study realism. Reference [41] suggest that study designs that
553 are more rigorous (e.g. experiments) are more realistic than observational
554 studies (e.g. case study) due to a higher level of control. On the other hand
555 [42] rate study designs based on other criteria, i.e. how frequently each type
556 of study design is used in an industrial or academic setting. If a study design
557 is used more in an industrial setting, then it is considered more realistic.
558 For instance, in software engineering, case studies are frequently used in
559 industrial settings, whereas, experiments are usually performed in academia
560 using students as subjects. Therefore, [42] argue that case studies are more
561 realistic than formal experiments. Obviously the effect of study design on
562 the study realism may be interpreted in different ways. Therefore, we will

563 not use this parameter in our quality evaluation.

564 The statistical significance of the results of a study can be used to evaluate
565 the significance of the study findings. This measure will not be used, because
566 the studies that are evaluated belong to very different research areas, i.e. the
567 significance levels of the findings of the studies are not directly comparable
568 for meta-analysis. Additionally, sometimes no result is more interesting than
569 a significant result, i.e. it may reveal important gaps in existing knowledge.

570 The ultimate goal of research, at least in software engineering, is in many
571 cases industry impact. However, most of the time ideas need to be devel-
572 oped and validated in academia before industry professionals will risk to
573 adopt them. Therefore, academic impact is important as well. Academic
574 impact is usually measured by the number of citations. Academic impact is
575 also measured for particular researchers, using the number of papers she has
576 published and the number of times her papers have been cited. This measure
577 will not be used in our quality evaluation because it is somewhat biased. The
578 number of citations is likely to be lower for newer papers and the number
579 of papers that a researcher has published gives little information about the
580 actual quality or impact of her research.

581 5.3.1. *Rating of the Studies*

582 The quality evaluation in our review is based on the evaluation of: (i)
583 Study realism. (ii) Study scale. (iii) Availability of raw results of CV. (iv)
584 Quality of the research methodology.

585 Realism of the studies is rated in three aspects: subjects, setting, and
586 scale. The subjects and setting is rated according to Table 5. The total
587 rating of study realism is determined by summing up the ratings of the two
588 aspects. For instance, if a study is conducted with industry professionals
589 as subjects in an academic context the study will receive rating 1 (out of 2
590 maximal points).

591 In order to rate the scale of a study the number of prioritization items was
592 counted. If a paper presents several prioritization cases only the prioritization
593 with the largest number of the prioritization items is considered. If HCV is
594 used all of the prioritization items on different levels are counted together.
595 However, if an item is present in several groups in the hierarchy it is counted
596 only once.

597 The availability of raw results from the application of CV is rated sepa-
598 rately because it is especially important for our purposes (and for most other
599 researchers in order to replicate a study). The data availability rating criteria

Table 5: Rating of study reality level.

Aspect	Contribute to relevance (rating 1)	Do not contribute to relevance (rating 0)
Subjects	Industry professionals	Academia students or teachers, or other
Context	Industrial	Academia

Table 6: Research data availability rating.

Rating	Study rating criteria
0	CV results was not provided in the paper and we was unable to obtain the results from the authors.
1	CV results are not provided in the paper but the data was obtained from the authors. Part of the data is lost or corrupted.
2	CV results are not provided in the paper but all the data was obtained from the authors.
3	All CV results are included in the paper or reference is given to online source where all the data can be accessed.

is given in Table 6. If the data of a study is not available it is not possible to validate the results of the study and, hence, the credibility of the findings is lower. Ideally the data collected in the study should be presented directly in the paper. An alternative may be to make the data freely available online and reference the online source.

The quality of the research methodology of a paper is rated according to a checklist presented in Appendix C. The checklist is based on guidelines for presenting research studies (as presented in [43, 44]) and the guidelines for quality evaluation of research studies as presented in [39, 42]. Evaluation is done with regard to the rigor of the description and correctness of the research process and reasoning. Checklist items represent issues that research studies should implement and present in a research paper. The checklist also contains item descriptions or questions that are used to evaluate the quality. Each item in the checklist is rated according to criteria presented in Table 7. The final rating of correctness of the research process of a study is computed by summing up the ratings assigned to all items in the checklist.

Study rating criteria was validated during a trial data extraction. Two researchers each rated three randomly selected papers. Afterwards, differences in ratings were discussed and study rating criteria were updated to avoid differences in interpretation.

As a result of the rating each study was assigned four rating values on an ordinal scale. In order to perform a more advanced analysis of the quality evaluation results these ratings were then converted into ratio scale ranks.

Table 7: Rating of correctness of research process.

Rating	Study rating criteria
0	No description provided.
1	Only basic information is provided about the checklist item. Or significant validity threats exist with regard to this item.
2	Description is sufficient. Some minor questions are left unanswered. Validity threats may exist but they are not likely to affect the results of the study.
3	Description is rigorous and clear. Questions presented in quality evaluation checklist in Appendix C are answered. Decisions of the study are well justified, alternatives are discussed. No unhandled validity threats can be identified.

Table 8: Example of rating values.

Study	Realism	Research data availability	Correctness of research process	Number of prioritization items
ST1	2	0	15	6
ST2	1	3	20	69
ST3	0	3	10	6

For each study, the number of studies that had received lower ratings were counted. The resulting number is the rank of the study; thereby, the quality of a study is expressed as four rank values.

An example of rating values is shown in Table 8. Table 9 shows ranking values computed for the studies in Table 8. We can observe that study realism level rating for ST3 is 0. There are no studies that have a lower study realism. Therefore, realism ranking for ST3 is 0. ST1 on the other hand has the highest realism rating. Since ST1 has higher reality level than both ST2 and ST3 it is assigned reality level rank 2.

5.4. Data Extraction

The goal of data extraction is to understand how and why CV is used and how CV results are analysed in research studies. Ultimately, this will allow us to answer the first and second research questions in our study.

Table 9: Example of ranking values.

Study	Reality level	Research data availability	Correctness of research process	Number of prioritization items
ST1	2	0	1	0
ST2	1	1	2	2
ST3	0	1	0	0

636 Data extraction was documented with the help of spreadsheet software.
637 Extracted data items are available from [45].

638 6. Equality of Cumulative Votes

639 In the previous section we described the execution of the systematic lit-
640 erature review. In order to perform a more thorough analysis later we here
641 present the design of ECV before presenting the results of the systematic
642 literature review. For the results of the evaluation of ECV please see Sec-
643 tion 7.3 (ECV is implemented in the *R* programming language [46] and the
644 code can be found at [47].)

645 In CV stakeholders may assign similar or equal values to several prior-
646 itization items. As a result the difference between the items is small. The
647 variation in priorities is caused not only by the difference between prioritiz-
648 ation items but also by human error and lack of information. For instance,
649 people tend to simplify the task of prioritization by assigning rounded values
650 to items or giving equal values to several items [48].

651 During prioritization it may be beneficial to know which items are equal.
652 A common example is software release planning where requirements are dis-
653 tributed among several product releases. If two or more requirements are
654 considered equal they can be interchanged between the releases regardless of
655 their priority. That allows other criteria, such as cost or effort, to be used as
656 sole indicators for planning that particular release.

657 6.1. Testing Equality of Two Items

658 There are two ways to determine which prioritization items have similar
659 priority. One approach is to find items that are different and consider other
660 items as equal. Another approach is to find items that are equal.

661 The first approach uses statistical tests to evaluate differences between
662 e.g. two sample means, in order to determine that two items are different.
663 Samples in this case consist of priorities assigned by all stakeholders to a
664 particular prioritization item. The number of stakeholders that perform the
665 prioritization is frequently small. Hence, the size of the sample is very often
666 too small for statistical tests to detect a significant difference in the tests,
667 thus, identify too many equal items to make any useful conclusions.

668 ECV, in contrast, uses the second approach. It finds items that are
669 similar and the rest of the items are considered different. This method tests
670 the probability of the difference between the means of two items being smaller

671 than the given value. In short, ECV tests the probability of the means of two
672 prioritization items differing by less than 25%. If the probability is higher
673 than 70% the items are considered equal.

674 The input to ECV is an $n \times p$ matrix A that contains the raw results of
675 the prioritization. The columns of the matrix represent prioritization items
676 while rows represent stakeholders. ECV performs the following operations
677 for the priorities of each of the two prioritization items:

- 678 1. Replace zeroes in CV results.
- 679 2. Transform the data using *ilr* transformation.
- 680 3. Determine distribution function using kernel density estimation.
- 681 4. Use the distribution function to find the probability that the difference
682 between two prioritization items is smaller than 25%.
- 683 5. Form groups of equal prioritization items.

684 Since CV results are compositional data, zeroes in A are replaced with
685 other values. This is done using the multiplicative replacement strategy
686 which is described in Section 2.4.1. Next, two columns are extracted from
687 matrix A to create the new matrix B :

$$B = [a_{*,k} a_{*,l}], \quad (5)$$

688 where a is an element of matrix A , and k and l are the columns that repre-
689 sent items that are tested for equality, "*" denotes all rows of corresponding
690 column.

691 The *ilr* transformation is then applied to each row of the matrix B and
692 the new vector C is obtained. The equation for calculating elements of C
693 using *ilr* transformation is:

$$c_i = ilr(b_{i1}, b_{i2}) = \sqrt{0.5} \log(b_{i1}/b_{i2}), \quad (6)$$

694 where c_i is the i^{th} element of C and b_{i1} and b_{i2} are the first and second
695 elements in the i^{th} row of B . Each value c_i represents a log-ratio between
696 values of columns k and l . The mean of the values of C can be interpreted as
697 an average log-ratio between the items that expresses the difference between
698 the items.

699 After the data is transformed into log-ratios statistical test can be applied.
700 The purpose of the test is to determine what the probability is of the relative
701 difference between two prioritization items k and l being less than 25%. Or
702 in terms of log-ratios it means determining the probability of c_i (obtained
703 from priorities assigned to k and l) as being in the range of $\frac{3}{4}$ to $\frac{4}{3}$. Hence,
704 the objective of the test is to determine the probability of the sample mean
705 (i.e. mean value of the items of C) laying between the two values.

706 The probability that the mean takes a particular value can be expressed
707 in the form of a cumulative distribution function. The probability of the
708 mean being between two values a and b (where a is smaller than b) can be
709 determined by subtracting the probability of the mean being smaller than a
710 from probability of the mean being smaller than b .

711 However, CV result data may or may not have multivariate normal dis-
712 tribution. If the data is normally distributed a Student's t -test can be used;
713 otherwise, a non-parametric estimation of the distribution function is needed.

714 Otherwise a non-parametric estimation of the distribution function could
715 be performed. In our case, the CV result data obtained from the primary
716 studies identified by the systematic review, were tested for normality using
717 the Anderson-Darling test. Before applying the test the data was transformed
718 using methods of compositional data analysis. To compute the test we used
719 method *adtestWrapper* from R language library *robCompositions*.

720 The tests we performed indicated, quite strongly, that in most of the
721 prioritization cases the data is not normally distributed. Hence, our rec-
722 ommendation is that, in general, a non-parametric approach should be used
723 to determine the probability density function, and one such, common, ap-
724 proach would be to use the kernel density estimation. (In our implementation
725 of ECV in the R programming language, kernel density estimation is per-
726 formed using the package *ks*.)

727 To determine the probability of \bar{x} being between a and b the following
728 equation is used:

$$p = P(b) - P(a), \quad (7)$$

729 where P is the cumulative distribution function obtained by applying
730 kernel density estimation on *ilr*-transformed priority values denoted by vector
731 C . Variable a is equal to *ilr*(3, 4) and b is equal to *ilr*(4, 3). (A graphical
732 interpretation of Equation (7) is presented in Figure 5.) The area that is
733 denoted by letter p represents the probability computed by the equation.

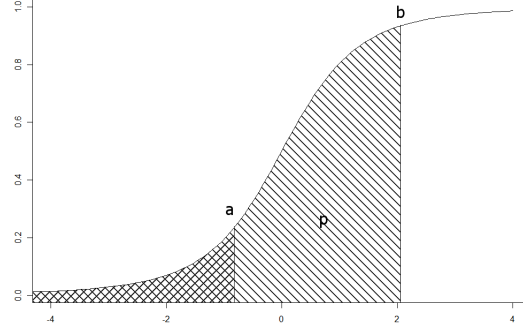


Figure 5: Cumulative distribution function of the log-ratio c_i between the items k and l (area p denotes probability that c_i is between $\frac{3}{4}$ and $\frac{4}{3}$.)

Table 10: Example of an equality table.

prioritization items	i1	i2	i3	i4
i1	equal	equal	-	equal
i2	equal	equal	-	-
i3	-	-	equal	-
i4	equal	-	-	equal

734 After both prioritization items are tested for equality it may be convenient
735 to display the equality of different items in the form of a table. Please see
736 Table 10 for an example.

737 6.2. Grouping Prioritization Items

738 When equal items are determined they can be divided into groups of equal
739 items. Division is performed in such a way that each two items in a group
740 are equal. The test for equality of the items described in Section 6.1 is not
741 transitive. Hence, if prioritization item A is equal to B and B is equal to C
742 then it does not automatically imply that A is equal to C . Therefore, there
743 may be several ways to group the equal items. The two possible division
744 criteria that we have considered in this study are:

- 745 1. Maximize the number of items that have a group.
- 746 2. Maximize the number of items in each group.

747 Current implementation of ECV (available from [47]) does not include
748 the division of items into groups. In this study the division is done manually,
749 so that each two items in a group are equal.

7. Results

This section presents the results of this study including the systematic literature review and the application of ECV on industry and academic data collected from the primary studies. Data extracted from primary studies and the results of the quality evaluation are available in [45].

7.1. State of Practice in Empirical Studies that use CV or Analyze the Results of CV (RQ 1)

The study search resulted in 634 unique studies. The search in databases revealed 180 papers, while an additional 454 papers were discovered using snowball sampling. The study selection resulted in 40 primary studies. Hence, 94% of the studies were excluded by the selection criteria. Snowball sampling revealed 15 (36%) out of all primary studies. The study selection criteria and the number of papers excluded by each criterion are shown in Tables 3 and 4. In total 163 of 634 studies were excluded because full text was not available.

All results of the study selection are available online and can be obtained by contacting the authors of this paper. For each study we specify keywords and databases that were used to find the study. If a study has been excluded, the exclusion criteria are provided.

The number of papers revealed by each search string and database is presented in Table 11. It should be noted that several papers were found by more than one search string or in more than one database. Table 11 shows that the search string ‘cumulative voting’ was the most frequently used in the research community to denote CV. Therefore, researchers should use or reference this term when discussing CV.

To perform snowball sampling we examined the references of primary studies that were found during the database search. References were used to search for the papers in the Google and Google Scholar search engines. Studies that were found in the search and passed the study selection criteria were added to the set of primary studies.

After the primary studies were selected, data extraction and quality evaluation was performed by two researchers. One researcher examined all studies while the second researcher did quality evaluation and data extraction for 10% of the studies. The studies were randomly selected. Inter-rater agreement were calculated by means of Krippendorff’s alpha coefficient. Agreement for data extraction results was 0.86 and agreement for the quality evalu-

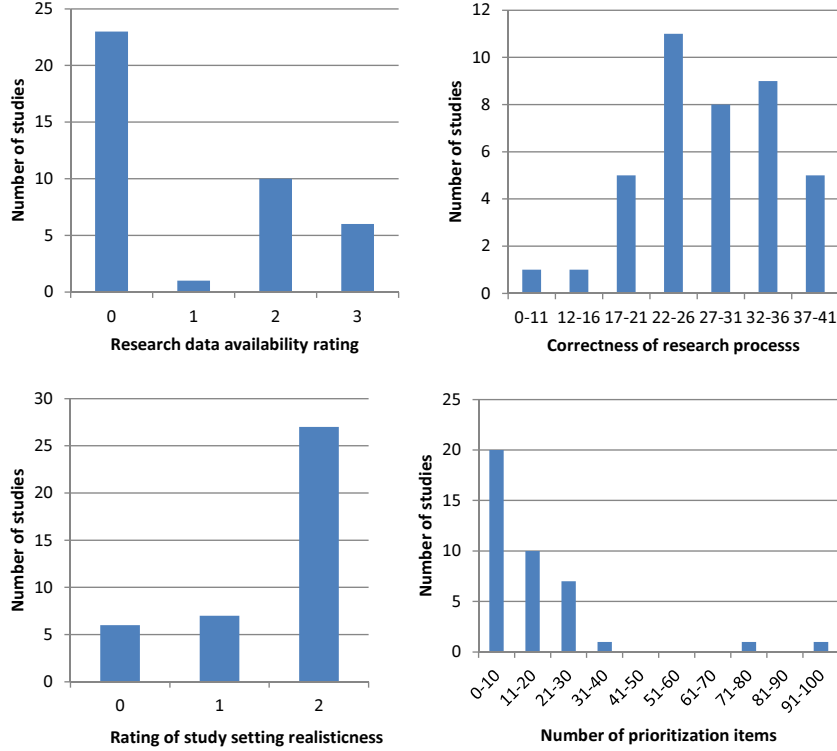
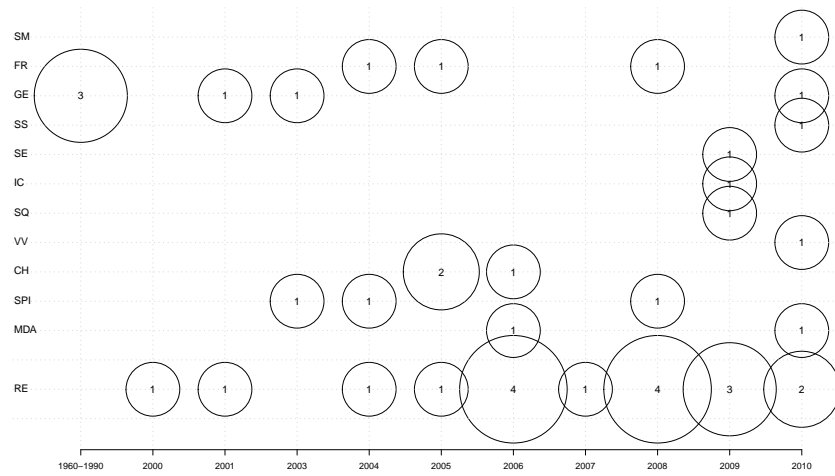


Figure 6: Study quality ratings.

786 ation was 0.73. According to [38] it is common to require agreement above 0.8
787 and the lowest acceptable agreement is 0.667. Therefore, we conclude that
788 the agreement calculated for this study is sufficient. Ratings of the study
789 setting, correctness, research data availability, and number of prioritization
790 items are presented in Figure 6.

791 Table 12 shows the studies with the highest quality according to our cri-
792 teria. These studies show a high level of rigor in a realistic setting. Moreover,
793 authors of the studies manifest confidence by providing raw data for further
794 use and evaluation.

795 Figure 7 shows a bubble chart of the distribution of studies over research
796 areas and time. The figure shows that CV was, as far as we know, first ap-
797 plied some time ago in research of government elections. Nowadays, though,
798 CV has been adopted in a wide range of software engineering areas, most
799 frequently in requirements engineering and software release planning. Eight



MDA - model driven software development	FR - forestry
CH - change impact analysis in software engineering	GE - government elections
RE - requirements engineering and software release planning	SS - software security
IC - intellectual capital in software company	SQ - software quality
SPI - software process improvement	SM - software metrics
V&V - software verification and validation	SE - software engineering in general

Figure 7: Distribution of studies over time.

Table 11: Number of papers found in the databases.

database	search strings							unique papers found	primary studies selected
	"100 point method"	"100 dollar method"	"100 dollar test"	"hundred point method"	"hundred dollar method"	"hundred dollar test"	"cumulative voting"		
ACM	2	0	0	1	2	3	31	34	7
IEEE	3	2	0	1	2	6	38	46	11
Inspec/Compendex	1	0	0	1	1	1	22	14	7
ISI web of science	0	0	0	0	1	1	15	16	6
SCOPUS	2	0	0	0	1	2	24	25	9
Springer	2	0	2	0	2	2	89	95	6
unique papers found	6	2	2	1	4	11	165	180	
primary studies selected	1	2	1	1	2	4	18		25

Table 12: Top ranked studies.

	Correctness of research process	Research data availability	Study setting	Number of prioritization items
Barney 2009 [49]	36	2	2	17
Berander 2009 [17]	41	2	0	29
Barney 2009 [50]	40	2	2	5
Barney 2009 [8]	31	2	2	27
Barney 2008 [51]	34	2	2	14
Laukkanen 2005 [52]	22	3	2	30
Hu 2006 [53]	34	2	1	14
Feldt 2010 [29]	24	3	2	8
Regnell 2001 [40]	21	3	2	91
Svahnberg 2008 [30]	34	1	1	7

800 studies use CV in academia while the remaining 32 studies report on using
801 CV in industry.

802 7.2. CV Result Analysis Methods Identified by RQ 1 (RQ 2)

803 The papers identified in the review use various CV result analysis meth-
804 ods. The main goals for CV result analysis are presented in Table 13 and
805 a summary of methods used in the primary studies can be found in Section
806 Appendix B.

807 In order to present prioritization results many studies use charts or tables.
808 These charts and tables show the average priority of each prioritization item
809 that is computed from priorities assigned by all stakeholders. In [54] a table

810 of five items with highest total priority is presented. [55] shows tables with
811 min , max , \tilde{x} , \bar{x} and σ of priorities assigned by different stakeholders to a
812 particular prioritization item. Finally, in [55, 56] error bars are added to the
813 chart of final priorities (denoting σ of priorities).

814 In a few cases final priorities are presented in the form of ranks and
815 CV results are degraded from ratio to ordinal scale. This is done when the
816 interest lies only in the order of final priorities.

817 Several papers are interested in the difference between priorities from dif-
818 ferent prioritization perspectives (e.g. current and ideal situation) or stake-
819 holder groups (e.g. software developers and management). Pearson or Spear-
820 man correlation coefficients are commonly used to determine what the level of
821 similarity is between all priorities from two perspectives. Whereas, Wilcoxon,
822 Kruskal-Wallis, Nemenyi-Damico-Wolfe-Dunn tests and the χ^2 statistic are
823 used to detect if there is a significant difference in the value of one prioritiza-
824 tion item from two or more perspectives. In addition, PCA is used to detect
825 if there are distinct groups of stakeholders with common priorities [7, 10, 57].

826 In some cases, a stakeholder may assign equal priority to several prioritiza-
827 tion items or leave several items unrated, e.g. the stakeholder may not have
828 carefully considered all prioritization items. Hence, the difference between
829 the items may have been unnoticed.

830 In [4] the scalability of prioritization is measured using two charts. The
831 first chart shows the average percentages of items given a non-zero value.
832 The second chart shows average percentages of divergence of values. If a
833 stakeholder assigns equal priorities to many prioritization items the diver-
834 gence of values is low. Unfortunately it is unclear from [4] how the average
835 percentage of divergence is calculated.

836 In [58] distribution, disagreement, and satisfaction charts are presented.
837 The distribution chart shows how the final value of a prioritization item
838 is constructed from priorities assigned by different stakeholders. This chart
839 shows how much each stakeholder has contributed to the final value of a prior-
840 itization item. The disagreement chart shows the level of agreement between
841 different stakeholders on the value of a particular prioritization item. The
842 satisfaction chart shows stakeholder satisfaction with prioritization results
843 by calculating the correlation between final priorities and priorities assigned
844 by a stakeholder.

845 The use of bi-plots and ternary plots are proposed in [10]. A bi-plot shows
846 final priorities and stakeholder viewpoints in a two dimensional plane while a
847 ternary plot shows prioritization items inside a triangle. Ternary plots show

Table 13: Goals for CV result analysis.

Purpose of the method	Name
Show the final priority of each prioritization item. Stakeholder priorities are combined into one value.	Chart or table of final priorities
Difference between priorities assigned by different perspectives (status quo, ideal situation) or different stakeholder groups (developers, management) [10]	Bi-plot
detect stakeholder groups with similar priorities [10]	Bi-plot
show the relative number of issues that have received high, medium, or low priority [10]	Ternary plot
detect stakeholder groups with common priorities [10]	PCA
how the final value of prioritization item is constructed from priorities assigned by different stakeholder. This chart shows how much each stakeholder has contributed to the final value of prioritization item [58]	Distribution chart
the level of agreement between different stakeholders on value of particular prioritization item [58]	Disagreement chart
satisfaction of a stakeholder with the prioritization results by the calculating correlation between the final priorities and priorities assigned by a stakeholder [58]	Satisfaction chart
percentage of the divergence of the priorities assigned by a stakeholder [4]	average percentage of divergence
average percentage of items given a non-zero value [4]	
detect equal prioritization items (presented in this paper)	ECV

848 how many low, medium or high priorities are assigned to a prioritization
849 item. The corners of the triangle represent high, medium, and low priority,
850 e.g. if a prioritization item has received mostly high priority values then it is
851 shown closer to the high priority corner.

852 7.2.1. Problems with Data Analysis in Primary Studies

853 A few primary studies, as revealed by the systematic review, have prob-
854 lems with the data analysis. These studies disregard the compositional nature
855 of CV results.

856 In [7, 57] standard PCA is performed without applying log-ratio trans-
857 formations to compositional data. According to [59], this is likely to be
858 inadequate and in [60], a more appropriate method for performing PCA on
859 compositional data is presented.

860 The normality of compositional data is defined in [61]. It is stated that
861 it is convenient to transform compositional data using isometric log-ratio
862 transformation before the tests for normality can be applied. [54] violates
863 this requirement by applying the Shapiro-Wilk test for normality to untrans-
864 formed compositional data.

865 The Kruskal-Wallis test is used in [54] to analyze compositional data.
866 The test is used to evaluate the difference between three organization levels.

Table 14: Identified groups of equal items.

Paper identifier & Description	Type of CV	Pairs of equal items	Groups of equal items
Barney 2009 [50] Perceived priorities of software product investments in an ideal situation	comp. HCV	(A2, B4) (B4, B5) (B4, C1) (B5, B15) (B6, B7) (B7, B8) (B14, B15) (B14, B18) (B17, B18)	(A2, B4) (B4, C1) (B5, B15) (B6, B7) (B14, B15) (B17, B18)
	uncomp. HCV	(B4, B5) (B4, B8) (B5, B15) (B6, B7) (B7, B12) (B14, B15) (B14, B18) (B16, B17) (B12, B13)	(B4, B5) (B5, B15) (B6, B7) (B14, B15) (B16, B17) (B12, B13)
Berander 2009 [17] Software requirements for course management system	uncomp. & comp. HCV	(3:2, 3:3)	(3:2, 3:3)
Svahnberg 2008 [30] The view of academia researchers on the requirements understandability criteria	CV	(Development, Verification & Validation) (Development, Product Planning 1)	(Development, Product Planning 1)

867 The Kruskal-Wallis test assumes that variables within each sample are in-
868 dependent [62]. However, values within compositional data vectors are not
869 independent (as described in Section 2.4). Hence, we claim the Kruskal-
870 Wallis test to be somewhat misused in [54].

871 7.3. Identifying Prioritization Items with Equal Priority Using ECV (RQ 3)

872

873 This section presents the results of applying ECV to the industrial and
874 academic CV data as found through the systematic literature review. Six
875 primary studies included the raw prioritization results in the paper itself or
876 referenced online sources where the data was available. To collect the data
877 from the remaining 34 papers, the authors of all papers were contacted.

878 First, the email addresses provided in the papers were used. If no answer
879 was received authors were searched for using Google, Facebook and LinkedIn.
880 Authors from 11 papers provided us with data to be used in the evaluation
881 of ECV. However, due to confidentiality reasons we can not publish this data
882 directly.

883 In short, ECV was applied to 27 CV prioritization cases from 14 studies.
884 In the cases of HCV, ECV was applied two times to the same data to test both
885 compensated and uncompensated priorities. Equal items were detected in
886 three prioritization cases. A summary of the results is presented in Table 14
887 and below follows a summary of each relevant study.

888 In [30] a prioritization of requirement understandability criteria is pre-
889 sented. One of the main findings of the paper is that two criteria - "De-
890 velopment" and "Verification & Validation" - are most important from an
891 academic viewpoint. ECV adds new knowledge to these results. It shows
892 that "Development" and "Verification & Validation" are equally important,
893 i.e. it is not true that either one of the criteria is more important.

894 A prioritization of software requirements for an academic course man-
895 agement system is presented in [17]. ECV detected that two features—
896 Assignment Submission and Assignment Feedback—have the same priority.
897 If the system is developed in several releases Assignment Submission and As-
898 signment Feedback features can be freely interchanged between the releases
899 and, hence, in this way ECV simplifies release planning.

900 In [50] software product investments are prioritized with HCV. The re-
901 sults of ECV was different for uncompensated and compensated HCV. When
902 compensated HCV was used ECV detected equal items that belonged to dif-
903 ferent high level prioritization groups (*A*, *B* and *C*) indicating that ECV
904 provided a more fine-grained view. In the case of uncompensated HCV, on
905 the other hand, all equal items belonged to one high level prioritization group
906 (group *B*).

907 8. Discussion and Conclusions

908 This section discusses the results of the systematic review and evaluation
909 of ECV conducted as part of this study.

910 CV has been applied in various areas, but most frequently in requirements
911 prioritization and release planning, and quite often also as part of research
912 methodologies. A large part of the studies have been conducted in Sweden,
913 at Ericsson AB. One can see a slight increase in the interest in CV. During
914 the last five years there have been more studies that use CV than between,
915 say, 2000–2005.

916 Overall, studies that use CV or analyze the results of CV have a high
917 quality in terms of correctness of research process and study realism. How-
918 ever, very few studies present prioritization of more than 30 items and the

919 availability of research data is somewhat limited. In our particular case we
920 were able to obtain data from 43% of the primary studies.

921 *8.1. Implications for Practitioners*

922 The results of this study provide decision support for industry practition-
923 ers. We believe that a collection of state of the practice studies help the
924 adoption of CV prioritization method. (The top studies are summarized in
925 Table 12.) In addition, a set of CV analysis methods enables comprehen-
926 sive understanding of the prioritization results. (The analysis methods are
927 presented in Table 13.) One of the most common goals of CV analysis is to
928 display the prioritization results and, thus, to show the difference between
929 several prioritization perspectives.

930 Additionally, we present ECV—a novel method for CV analysis. Priori-
931 tization often results in the assignment of similar priorities to several prior-
932 itization items. CV results contain both ‘real priorities’ and random errors.
933 Due to random errors, equal prioritization items may receive different pri-
934 orities. ECV identifies such items. It allows stakeholders to disregard the
935 random part of the CV results. Thus, ECV simplifies the understanding of
936 the prioritization results.

937 ECV identifies prioritization items with similar priority and tests whether
938 these items can be considered equal. In this case, ECV can be used in
939 software release planning. For example, let us suppose that a set of software
940 requirements are prioritized with regard to the implementation costs. First of
941 all, ECV can then detect items with equal cost. Second, the equal items can
942 be freely interchanged between the releases. Finally, the decision to allocate
943 a requirement to a particular release can be made based on another criteria,
944 such as risk or business value.

945 ECV has been successfully applied on a considerable amount of CV data
946 and, additionally, has also detected equal items in different groups of HCV
947 hierarchies.

948 *8.2. Implications for Academia*

949 In the systematic review 36% of papers were revealed by the snowball
950 sampling. That is a considerable amount. Several studies do not mention
951 the name of the prioritization method (i.e. cumulative voting or hundred
952 dollar test). Others are not available through selected databases because
953 they are conference publications or theses. It shows, in our opinion, that
954 snowball sampling ought to be used in all systematic literature reviews.

CV results are a special type of data—compositional data. Standard statistical analysis methods that assume the independence of the samples cannot be applied to CV results. In [63] methods for the analysis of compositional data have been presented. The systematic review conducted as a part of this study revealed that 22 studies analyze CV results; yet, only one study uses compositional data analysis methods, i.e. [10]. None of the studies, including [10], present methods for detecting items with equal priority in CV results. Hence, ECV is, in this respect, a unique method.

The small use of compositional data analysis is really not surprising, since literature describing CV does not state that the results are compositional data. Standard statistical analysis methods may produce useful results for compositional data. However, there are cases when they are misleading or even faulty. Section 7.2.1 contains evidence of inappropriate use of statistical methods by several papers.

This study has collected a set of compositional data analysis methods for CV analysis (see Table 13). We believe that this could help researchers to improve the analysis of CV results with appropriate methods.

Since CV is associated with compositional data, it might be tempting to choose another requirements prioritization method. However, it would not solve the problem *per se*, because any ratio scale prioritization, for instance AHP, contains compositional data.

The principal implications for the academia are mainly the following:

1. All systematic literature reviews should include snowball sampling.
2. Researchers can improve their statistical analysis of CV results using compositional data analysis methods collected and developed by this study.
3. When CV or any other ratio scale prioritization method is taught, compositional data analysis should also be presented as part of the solution.

8.3. *Validity Threats*

The validity of the systematic review is mainly limited by the chosen databases, the design of the review, and human judgement in study selection and data extraction.

988 To mitigate the threats we use the most popular databases in the field
989 of software engineering. In the beginning of the systematic review a re-
990 view protocol was developed, peer-reviewed, and revised. Search strategy
991 was validated against a set of previously known papers obtained from other
992 researchers.

993 One of many terms used to name cumulative voting is ‘\$100 method’. We
994 were not able to search for this term because non of the chosen databases sup-
995 port search for special characters like ‘\$’ and the search string ‘100 method’
996 yields too many hits. To increase the likelihood of discovering relevant studies
997 snowball sampling was extensively used.

998 To increase the validity of study selection, all included studies and 20
999 randomly selected excluded studies were examined by two researchers. There
1000 were no disagreement on the inclusion/exclusion of the studies.

1001 The large number of studies identified by snowball sampling (15 out of
1002 40 studies) may be caused by faulty design or by faulty execution of the
1003 search in the databases. There are several reasons why the studies revealed
1004 by snowball sampling are not revealed by the search in databases. (Reason
1005 for each study is given in Table Appendix A.2.) Based on these reasons we
1006 argue that snowball sampling does not indicate any problems with the design
1007 of the search in the databases.

1008 Four studies were not found because they were not available through
1009 databases used in this systematic review. Out of them one is a master thesis,
1010 two are conference publications and one is a publication in the area of forestry.
1011 Seven studies do not mention the name of the prioritization method (i.e.
1012 hundred dollar method or cumulative voting). Only phrases like “distribution
1013 of a predefined amount of fictitious money (\$100,000) over the items to be
1014 prioritized” or “1,000 points” allowed us to identify that CV was indeed
1015 used. One paper used a previously unknown name for CV, i.e. the 100-point
1016 technique.

1017 The quality of the data extraction and quality evaluation was validated
1018 using inter-rater agreement analysis. In our case, 10% of the studies were
1019 rated by two researchers and Krippendorff’s alpha was calculated. The agree-
1020 ment for the data extraction results was 0.86 and the agreement for the
1021 quality evaluation was 0.73 (indicating a credible level of quality).

1022 There are two main validity threats with ECV itself. First, ECV may not
1023 detect prioritization items with equal priority. Second, ECV may produce a
1024 false positive result, i.e. there may be a real difference between items that
1025 ECV claims as being equal.

1026 To mitigate the first threat ECV was applied on artificially created test
1027 data with and without items with similar priority. ECV worked correctly in
1028 both cases.

1029 To mitigate the second threat we visually inspected the results of the
1030 application of ECV on the real world data from the primary studies. We
1031 concluded that items identified by ECV can be considered equal.

1032 CV results used in the evaluation of ECV were tested for normality. The
1033 tests indicated that CV results do not have multivariate normal distribution.
1034 Therefore, the design of ECV was based on a non-parametric statistical test.

1035 8.4. Future Work

1036 With respect to future work one can distinguish two interesting paths:
1037 Scalability and improvements to ECV.

1038 First, there are very few studies that apply CV on prioritization sets of
1039 more than 30 items. However, in requirements engineering, industry prac-
1040 titioners need to prioritize much larger numbers of software requirements.
1041 Therefore, the state of art could benefit from the application of CV and
1042 HCV to large prioritization sets.

1043 The proposed method, ECV, has now been evaluated on existing research
1044 data. To further evaluate ECV, it would be appropriate to apply it in direct
1045 industry practice and in prioritization cases with a larger number of priori-
1046 tization items (>30). Additionally, compositional data analysis methods, as
1047 the ones identified by this paper, should be tried with other prioritization
1048 methods that produce ratio scale results.

1049 Second, ECV may be improved to find groups of equal items not just
1050 pairs. Equality of a pair (or a group) of items to another item can be tested
1051 with the help of compositional balances.

1052 The CV process itself can also be improved with the help of compositional
1053 data analysis. Weighting of stakeholder priorities could be done using com-
1054 positional powering, which could be presumed as better compared to using
1055 a multiplication that is removed in a log-ratio transformation.

1056 Additionally, compensation of priority values in HCV is not *subcomposi-*
1057 *tionally coherent*; thus, sequential binary partition could quite possibly be
1058 used to improve the compensation.

1059 8.5. Conclusions

1060 CV prioritization results are special type of data – compositional data.
1061 Any analysis of CV results must take into account the compositional nature

of the CV results.

This study presents a systematic literature review of the empirical use of CV. CV has been applied in various areas, but most frequently in requirements prioritization and release planning. The review has resulted in a collection of state of the practice studies and CV result analysis methods. We believe that it can help the adoption of CV prioritization method.

In our case, snowball sampling was performed as a part of the review. Since it revealed 36% out of all primary studies, we believe that in future snowball sampling should be used in all systematic reviews.

Additionally, we present ECV—a novel method for CV analysis. As suggested by our evaluation, ECV is able to detect prioritization items with equal priority (i.e. items that have insignificant difference in priority). The evaluation of ECV was based on the data obtained from the authors of the primary studies.

References

- [1] P. Berander, A. Andrews, Requirements prioritization, in: A. Aurum, C. Wohlin (Eds.), *Engineering and managing software requirements*, Springer-Verlag, Berlin/Heidelberg, 2005, pp. 69–94.
- [2] D. Leffingwell, D. Widrig, *Managing software requirements: A unified approach*, Addison-Wesley Professional, 1999.
- [3] V. Ahl, An experimental comparison of five prioritization methods, Master's Thesis, School of Engineering, Blekinge Institute of Technology, Sweden.
- [4] P. Berander, P. Jönsson, Hierarchical cumulative voting (HCV) - Prioritization of requirements in hierarchies, *International Journal of Software Engineering and Knowledge Engineering* 16 (6) (2006) 819–850.
- [5] J. Karlsson, K. Ryan, A cost-value approach for prioritizing requirements, *IEEE Software* 14 (5) (1997) 67–74.
- [6] J. Karlsson, An evaluation of methods for prioritizing software requirements, *Information and Software Technology* 39 (14-15) (1998) 939–947.
- [7] F. Pettersson, M. Ivarsson, T. Gorschek, P. Öhman, A practitioner's guide to light weight software process assessment and improvement planning, *Journal of Systems and Software* 81 (2008) 972–995.

- 1095 [8] S. Barney, C. Wohlin, Software product quality: Ensuring a common
1096 goal, in: Q. Wang, V. Garousi, R. Madachy, D. Pfahl (Eds.), Trust-
1097 worthy Software Development Processes, Vol. 5543 of Lecture Notes in
1098 Computer Science, Springer-Verlag, Berlin, Heidelberg, 2009, pp. 256–
1099 267.
- 1100 [9] P. Jönsson, C. Wohlin, A study on prioritisation of impact analysis
1101 issues: A comparison between perspectives, Software Engineering Re-
1102 search and Practice in Sweden.
- 1103 [10] P. Chatzipetrou, L. Angelis, P. Rovegard, C. Wohlin, Prioritization of
1104 issues and requirements by cumulative voting: A compositional data
1105 analysis framework, in: Proceedings of the 2010 36th EUROMICRO
1106 Conference on Software Engineering and Advanced Applications, IEEE
1107 Computer Society, Washington, DC, USA, 2010, pp. 361–370.
- 1108 [11] R. L. Engstrom, D. A. Taebel, R. L. Cole, Cumulative voting as a rem-
1109 edy for minority vote dilution: The case of Alamogordo, New Mexico,
1110 Journal of Law & Politics 5 (1988) 469–497.
- 1111 [12] S. Bhagat, J. A. Brickley, Cumulative voting: The value of minority
1112 shareholder voting rights, Journal of Law and Economics 27 (2) (1984)
1113 339–365.
- 1114 [13] V. Hiltunen, J. Kangas, J. Pykalainen, Voting methods in strategic forest
1115 planning - Experiences from Metsähallitus, Forest Policy and Economics
1116 10 (3) (2008) 117–127.
- 1117 [14] P. Boldi, F. Bonchi, C. Castillo, S. Vigna, Voting in social networks, in:
1118 Proceedings of the 18th ACM conference on Information and knowledge
1119 management, ACM, New York, NY, USA, 2009, pp. 777–786.
- 1120 [15] H. Ayad, M. Kamel, Cumulative voting consensus method for partitions
1121 with variable number of clusters, IEEE Transactions on Pattern Analysis
1122 and Machine Intelligence 30 (1) (2008) 160–173.
- 1123 [16] M. Svahnberg, A. Karasira, A study on the importance of order in re-
1124 quirements prioritisation, in: Proceedings of the Third International
1125 Workshop on Software Product Management, IEEE Computer Society,
1126 Washington, DC, USA, 2009, pp. 35–41.

- 1127 [17] P. Berander, M. Svahnberg, Evaluating two ways of calculating priorities
1128 in requirements hierarchies - An experiment on hierarchical cumulative
1129 voting, *Journal of Systems and Software* 82 (2009) 836–850.
- 1130 [18] T. L. Saaty, *The analytic hierarchy process*, McGraw-Hill, New York,
1131 1980.
- 1132 [19] J. Aitchison, J. J. Egozcue, Compositional Data Analysis: Where Are
1133 We and Where Should We Be Heading?, *Mathematical Geology* 37 (7)
1134 (2005) 829–850.
- 1135 [20] V. Pawlowsky-Glahn, J. J. Egozcue, Compositional data and their anal-
1136 ysis: An introduction, Geological Society, London, Special Publications
1137 264 (1) (2006) 1–10.
- 1138 [21] J. A. Martín-Fernández, C. Barceló-Vidal, V. Pawlowsky-Glahn, Deal-
1139 ing with zeros and missing values in compositional data sets using non-
1140 parametric imputation, *Mathematical Geology* 35 (3) (2003) 253–278.
- 1141 [22] P. Filzmoser, K. Hron, Outlier detection for compositional data using
1142 robust methods, *Mathematical Geosciences* 40 (2008) 233–248.
- 1143 [23] S. Laukkanen, T. Palander, J. Kangas, Applying voting theory in par-
1144 ticipatory decision support for sustainable timber harvesting, *Canadian*
1145 *Journal of Forest Research* 34 (7) (2004) 1511–1524.
- 1146 [24] D. Cooper, A. Zillante, A comparison of cumulative voting and gener-
1147 alized plurality voting, *Public Choice* 150 (1-2) (2010) 363–383.
- 1148 [25] A. Hoover, M. Goldbaum, Locating the optic nerve in a retinal image
1149 using the fuzzy convergence of the blood vessels, *Medical Imaging, IEEE*
1150 *Transactions on* 22 (8) (2003) 951–958.
- 1151 [26] V. Heikkilä, A. Jadallah, K. Rautiainen, G. Ruhe, Rigorous support for
1152 flexible planning of product releases - A stakeholder-centric approach
1153 and its initial evaluation, in: *2010 43rd Hawaii International Conference*
1154 *on System Sciences*, IEEE Computer Society, 2010, pp. 1–10.
- 1155 [27] D. Baca, K. Petersen, Prioritizing countermeasures through the coun-
1156 termeasure method for software security (CM-Sec), in: M. Ali Babar,

- 1157 M. Vierimaa, M. Oivo (Eds.), Product-Focused Software Process Im-
 1158 provement, Vol. 6156 of Lecture Notes in Computer Science, Springer-
 1159 Verlag, Berlin, Heidelberg, 2010, pp. 176–190.
- 1160 [28] P. Berander, P. Jönsson, A goal question metric based approach for ef-
 1161 ficient measurement framework definition, in: Proceedings of the 2006
 1162 ACM/IEEE international symposium on Empirical software engineer-
 1163 ing, ACM, New York, NY, USA, 2006, pp. 316–325.
- 1164 [29] R. Feldt, R. Torkar, E. Ahmad, B. Raza, Challenges with software ver-
 1165 ification and validation activities in the space industry, in: Proceedings
 1166 of the 2010 Third International Conference on Software Testing, Verifi-
 1167 cation and Validation, IEEE Computer Society, Washington, DC, USA,
 1168 2010, pp. 225–234.
- 1169 [30] M. Svahnberg, T. Gorschek, M. Eriksson, A. Borg, K. Sandahl,
 1170 J. Börster, A. Loconsole, Perspectives on requirements understandabil-
 1171 ity – For whom does the teacher’s bell toll?, in: Proceedings of the 2008
 1172 Requirements Engineering Education and Training, IEEE Computer So-
 1173 ciety, Washington, DC, USA, 2008, pp. 22–29.
- 1174 [31] J. Aitchison, The statistical analysis of compositional data, Journal of
 1175 the Royal Statistical Society. Series B (Methodological) 44 (2) (1982)
 1176 139–177.
- 1177 [32] J. Egozcue, V. Pawlowsky-Glahn, G. Mateu-Figueras, C. Barceló-
 1178 Vidal, Isometric logratio transformations for compositional data analy-
 1179 sis, Mathematical Geology 35 (3) (2003) 279–300.
- 1180 [33] K. Khan, A systematic review of software requirements prioritization,
 1181 Master’s thesis, Blekinge Institute of Technology, Ronneby, Sweden.
- 1182 [34] F. Zahedi, The analytic hierarchy process: A survey of the method and
 1183 its applications, Interfaces 16 (4) (1986) 96–108.
- 1184 [35] P. Runeson, M. Höst, Guidelines for conducting and reporting case study
 1185 research in software engineering, Empirical Software Engineering 14 (2)
 1186 (2008) 131–164.
- 1187 [36] L. Goodman, Snowball sampling, The Annals of Mathematical Statistics
 1188 32 (1) (1961) 148–170.

- 1189 [37] K. Krippendorff, Bivariate agreement coefficients for reliability of data,
1190 Sociological Methodology 2 (1970) 139–150.
- 1191 [38] K. Krippendorff, Content analysis: An introduction to its methodology,
1192 2nd Edition, Sage Publications, 2003.
- 1193 [39] B. Kitchenham, S. Charters, Guidelines for performing systematic liter-
1194 ature reviews in software engineering, Tech. Rep. EBSE 2007-001, Keele
1195 University (2007).
- 1196 [40] B. Regnell, M. Höst, J. Natt och Dag, P. Beremark, T. Hjelm, An
1197 industrial case study on distributed prioritisation in market-driven re-
1198 quirements engineering for packaged software, Requirements Engineer-
1199 ing 6 (1) (2001) 51–62.
- 1200 [41] B. Kitchenham, Procedures for performing systematic reviews, Tech.
1201 Rep. TR/SE-0401, Keele University (2004).
- 1202 [42] M. Ivarsson, T. Gorschek, A method for evaluating rigor and industrial
1203 relevance of technology evaluations, Empirical Software Engineering 16
1204 (2011) 365–395.
- 1205 [43] C. Wohlin, P. Runeson, M. Höst, Experimentation in software engineer-
1206 ing: An introduction, Springer Netherlands, 2000.
- 1207 [44] A. Jedlitschka, D. Pfahl, Reporting guidelines for controlled experiments
1208 in software engineering, in: Proceedings of the 2005 International Sym-
1209 posium on Empirical Software Engineering, IEEE Computer Society,
1210 2005, pp. 95–104.
- 1211 [45] K. Rinkevics, R. Torkar, Data extraction and quality evaluation results
1212 (2011).
1213 URL [http://rinkevic.wordpress.com/2011/11/26/
1214 data-extraction-and-quality-evaluation-results/](http://rinkevic.wordpress.com/2011/11/26/data-extraction-and-quality-evaluation-results/)
- 1215 [46] R. Ihaka, R. Gentleman, R: A language for data analysis and graphics,
1216 Journal of computational and graphical statistics 5 (3) (1996) 299–314.
- 1217 [47] K. Rinkevics, R. Torkar, ECV implementation source code in R (2011).
1218 URL [http://rinkevic.wordpress.com/2011/08/14/
1219 ecv-implementation-in-r/](http://rinkevic.wordpress.com/2011/08/14/ecv-implementation-in-r/)

- 1220 [48] R. M. Groves, F. J. Fowler, M. P. Couper, J. M. Lepkowski, E. Singer,
1221 Survey methodology, John Wiley and Sons, 2009.
- 1222 [49] S. Barney, A. Aurum, C. Wohlin, The relative importance of aspects
1223 of intellectual capital for software companies, in: Proceedings of the
1224 2009 35th Euromicro Conference on Software Engineering and Advanced
1225 Applications, IEEE Computer Society, 2009, pp. 313–320.
- 1226 [50] S. Barney, C. Wohlin, Software product quality: Ensuring a common
1227 goal, in: Proceedings of the International Conference on Software Pro-
1228 cess: Trustworthy Software Development Processes, Springer-Verlag,
1229 Berlin, Heidelberg, 2009, pp. 256–267.
- 1230 [51] S. Barney, A. Aurum, C. Wohlin, A product management challenge:
1231 Creating software product value through requirements selection, Journal
1232 of Systems Architecture 54 (6) (2008) 576–593.
- 1233 [52] S. Laukkanen, T. Palander, J. Kangas, A. Kangas, Evaluation of the
1234 multicriteria approval method for timber-harvesting group decision sup-
1235 port, Silva Fennica 39 (2) (2005) 249–264.
- 1236 [53] G. Hu, A. Aurum, C. Wohlin, Adding value to software requirements:
1237 An empirical study in the Chinese software industry, in: Proceedings of
1238 the Seventeenth Australasian Conference on Information Systems, 2006.
- 1239 [54] P. Jönsson, C. Wohlin, Understanding impact analysis: An empiri-
1240 cal study to capture knowledge on different organisational levels, in:
1241 Proceedings of International Conference on Software Engineering and
1242 Knowledge Engineering, IEEE Computer Society, 2005, pp. 707–712.
- 1243 [55] L. Kuzniarz, L. Angelis, Empirical extension of a classification frame-
1244 work for addressing consistency in model based development, Informa-
1245 tion and Software Technology 53 (2011) 214–229.
- 1246 [56] P. Rovegard, L. Angelis, C. Wohlin, An empirical study on views of im-
1247 portance of change impact analysis issues, IEEE Transactions on Soft-
1248 ware Engineering 34 (4) (2008) 516–530.
- 1249 [57] C. Wohlin, A. Aurum, Criteria for selecting software requirements to cre-
1250 ate product value: An industrial empirical study, in: S. Biffl, A. Aurum,

- 1251 B. Boehm, H. Erdogan, P. Grünbacher (Eds.), Value-based software en-
1252 gineering, Springer-Verlag, Berlin, Heidelberg, 2006, pp. 179–200.
- 1253 [58] B. Regnell, M. Höst, J. Natt och Dag, Visualization of agreement and
1254 satisfaction in distributed prioritization of market requirements, in: Pro-
1255 ceedings of REFSQ2000, 6th Int. Workshop on Requirements Engineering:
1256 ing: Foundation for Software Quality, 2000, pp. 1–12.
- 1257 [59] J. Aitchison, Principal component analysis of compositional data,
1258 Biometrika 70 (1) (1983) 57–65.
- 1259 [60] P. Filzmoser, K. Hron, C. Reimann, Principal component analysis for
1260 compositional data with outliers, Environmetrics 20 (6) (2009) 621–632.
- 1261 [61] V. Pawlowsky Glahn, J. Egozcue, R. Tolosana Delgado, Lecture notes
1262 on compositional data analysis, Tech. rep., Universitat de Girona, Spain
1263 (July 2007).
- 1264 [62] W. H. Kruskal, W. A. Wallis, Use of ranks in one-criterion variance
1265 analysis, Journal of the American Statistical Association 47 (260) (1952)
1266 583–621.
- 1267 [63] J. Aitchison, The statistical analysis of compositional data, Chapman
1268 & Hall, London, 1986.
- 1269 [64] S. Bowler, D. Brockington, T. Donovan, Election systems and voter
1270 turnout: Experiments in the United States, The Journal of Politics
1271 63 (3) (2001) 902–915.
- 1272 [65] D. Brockington, A low information theory of ballot position effect, Po-
1273 litical Behavior 25 (1) (2003) 1–27.
- 1274 [66] N. Dzamashvili Fogelström, M. Svahnberg, T. Gorschek, Investigating
1275 impact of business risk on requirements selection decisions, in: Proceed-
1276 ings of the 2009 35th Euromicro Conference on Software Engineering
1277 and Advanced Applications, IEEE, 2009, pp. 217–223.
- 1278 [67] S. Hatton, Choosing the right prioritisation method, in: Proceedings of
1279 the 19th Australian Conference on Software Engineering, IEEE Com-
1280 puter Society, Washington, 2008, pp. 517–526.

- 1281 [68] S. Hatton, Early prioritisation of goals, in: Proceedings of the 2007
1282 Conference on Advances in Conceptual Modeling: Foundations and Ap-
1283 plications, Springer-Verlag, Berlin, Heidelberg, 2007, pp. 235–244.
- 1284 [69] M. Staron, C. Wohlin, An industrial case study on the choice between
1285 language customization mechanisms, in: J. Münch, M. Vierimaa (Eds.),
1286 Product-Focused Software Process Improvement, Vol. 4034 of Lecture
1287 Notes in Computer Science, Springer-Verlag, Berlin, Heidelberg, 2006,
1288 pp. 177–191.
- 1289 [70] T. Touseef, C. Gencel, A structured goal based measurement framework
1290 enabling traceability and prioritization, in: Proceedings of the 2010 6th
1291 International Conference on Emerging Technologies, 2010, pp. 282–286.
- 1292 [71] P. Berander, C. Wohlin, Differences in views between development roles
1293 in software process improvement - A quantitative comparison, in: Pro-
1294 ceedings of the 8th International Conference on Empirical Assessment in
1295 Software Engineering, 2004.
- 1296 [72] P. Berander, Using students as subjects in requirements prioritization,
1297 in: Proceedings of the 2004 International Symposium on Empirical Soft-
1298 ware Engineering, IEEE Computer Society, 2004, pp. 167–176.
- 1299 [73] P. Berander, C. Wohlin, Identification of key factors in software process
1300 management - A case study, in: Proceedings of the 2003 International
1301 Symposium on Empirical Software Engineering, IEEE Computer Soci-
1302 ety, Washington, DC, USA, 2003, pp. 316–325.
- 1303 [74] R. L. Cole, D. A. Taebel, R. L. Engstrom, Cumulative voting in a mu-
1304 nicipal election: A note on voter reactions and electoral consequences,
1305 The Western Political Quarterly 43 (1) (1990) 191–199.
- 1306 [75] J. Kuklinski, Cumulative and plurality voting: An analysis of Illinois’
1307 unique electoral system, The Western Political Quarterly 26 (4) (1973)
1308 726–746.
- 1309 [76] J. Sawyer, D. MacRae, Game theory and cumulative voting in Illinois:
1310 1902-1954, The American Political Science Review 56 (4) (1962) 936–
1311 946.

1312 Appendix A. Primary Studies

1313 *Appendix A.1. Primary studies found in databases.*

Title	Reference
Prioritizing countermeasures through the countermeasure method for software security (CM-Sec)	Baca 2010 [27]
The relative importance of aspects of intellectual capital for software companies	Barney 2009 [49]
Software product quality: Ensuring a common goal	Barney 2009 [8]
Balancing software product investments	Barney 2009 [50]
Hierarchical cumulative voting (HCV) prioritization of requirements in hierarchies	Berander 2006 [4]
A goal question metric based approach for efficient measurement framework definition	Berander 2006 [28]
Evaluating two ways of calculating priorities in requirements hierarchies: An experiment on hierarchical cumulative voting	Berander 2009 [17]
Election systems and voter turnout: Experiments in the United States	Bowler 2001 [64]
A low information theory of ballot position effect	Brockington 2003 [65]
Prioritization of issues and requirements by cumulative Voting: A compositional data analysis framework	Chatzipetrou 2010 [10]
A comparison of cumulative voting and generalized plurality voting	Cooper 2010 [24]
Challenges with software verification and validation activities in the space industry	Feldt 2010 [29]
Investigating impact of business risk on requirements selection decisions	Fogelstrom 2009 [66]
Choosing the right prioritization method	Hatton 2008 [67]
Early prioritization of goals	Hatton 2007 [68]
Rigorous support for flexible planning of product releases: A stakeholder-centric approach and its initial evaluation	Heikkilä 2010 [26]
Voting methods in strategic forest planning: Experiences from Metsähallitus	Hiltunen 2008 [13]
Empirical extension of a classification framework for addressing consistency in model based development	Kuzniarz 2010 [55]
Evaluation of the multi-criteria approval method for timber-harvesting group decision support	Laukkanen 2005 [52]
A practitioner's guide to light weight software process assessment and improvement planning	Pettersson 2008 [7]
An empirical study on views of importance of change impact analysis issues	Rovegard 2008 [56]
An industrial case study on the choice between language customization mechanisms	Staron 2006 [69]
Perspectives on requirements understandability—For whom does the teacher's bell toll?	Svahnberg 2008 [30]
A study on the importance of order in requirements prioritization	Svahnberg 2009 [16]
A structured goal based measurement framework enabling traceability and prioritization	Touseef 2010 [70]

1315 *Appendix A.2. Primary studies revealed by snowball sampling.*

Reference	Title	Reason why the paper is not revealed by the search in databases
Ahl 2005 [3]	An experimental comparison of five prioritization methods	Selected databases does not contain the paper, master thesis at BTH
Barney 2008 [51]	A product management challenge: Creating software product value through requirements selection	Prioritization method name not mentioned, phrase "1,000 points" used instead.
Berander 2004 [71]	Differences in views between development roles in software process improvement—A quantitative comparison	Prioritization method name not mentioned, phrase "100 points" used instead.
Berander 2004 [72]	Using students as subjects in requirements prioritization	Unknown CV name: 100-point technique
Berander 2003 [73]	Identification of key factors in software process management: A case study	Prioritization method name not mentioned, phrase "100 points" used instead.
Cole 1990 [74]	Cumulative voting in a municipal election: A note on voter reactions and electoral consequences	Study published before year 2001.
Hu 2006 [53]	Adding value to software requirements: An empirical study in the chinese software industry	Prioritization method name not mentioned, phrase "1,000 points" used instead.
Jonsson 2005 [9]	A study on prioritization of impact analysis issues: A comparison between perspectives	Selected databases does not contain the paper.
Jonsson 2005 [54]	Understanding impact analysis: An empirical study to capture knowledge on different organizational levels	Selected databases does not contain the paper.
Kuklinski 1973 [75]	Cumulative and plurality voting: An analysis of Illinois' unique electoral system	Study published before year 2001.
Laukkanen 2004 [23]	Applying voting theory in participatory decision support for sustainable timber harvesting	Selected databases does not contain the paper.
Regnell 2001 [40]	An industrial case study on distributed prioritization in market-driven requirements engineering for packaged software	Prioritization method name not mentioned: "distribution of a predefined amount of fictitious money (\$100,000) over the items to be prioritized."
Regnell 2000 [58]	Visualization of agreement and satisfaction in distributed prioritization of market requirements	Prioritization method name not mentioned: "distribution of a predefined amount of fictitious money (\$100,000) over the items to be prioritized."
Wohlin 2006 [76]	Game theory and cumulative voting in Illinois: 1902–1954	Study published before year 2001.
Wohlin 2006 [57]	Criteria for selecting software requirements to create product value: An industrial empirical study	Prioritization method name not mentioned: "The subjects had 1,000 points to spend among the 13 criteria."

1317

Appendix B. CV Result Analysis Methods

1318

	Paper															
	Svohberg2008	Svohberg2009	Storou2006	Pettersen2008	Wohlin2006	Lanklaen2005a	Hu2006	Jonsson2005a	Kuzniarz2010	Rovgaard2008	Berander2006a	Berander2004a	Berander2006	Feldt2010	Barney2006b	Barney2008
Analysis method																
Table that shows final priorities	x			x											x	
Chart that shows final priorities	x			x	x	x	x									
Table of top-5 prioritization items								x								
min, max, \bar{x} , $\bar{\sigma}$ and σ of priorities assigned by different stakeholders									x	x						
Bar chart of prioritization results showing \bar{x} priority and σ of priorities									x	x						
Pearson correlation coefficient		x									x					
Nemenyi Damico Wolfe Dunn													x			
Spearman's r														x		
Kruskal-Wallis								x							x	
Wilcoxon							x									
Correlation matrix	x													x	x	
Chart for comparing priorities from two perspectives, priorities are points in two dimensional plane, x - and y -axis represent two different perspectives										x						
Difference between priorities assigned by each two stakeholders using χ^2 -statistic										x						
Median ranks		x														
CV results converted to priority ranks	x												x			
PCA				x	x											
Percentage of divergence of priorities assigned by a stakeholder																x
Average percentage of items given non-zero value										x						
Distribution chart										x						
Disagreement chart				x												x
Satisfaction chart				x												x
Bi-plot																x
Ternary plot																x

1319

Appendix C. Quality Evaluation Checklist

1320

	Item	Question or Description of the Item	Rating
1.	Background, introduction	Introduce research area	
2.	Problem statement, purpose	What is the problem [44]? Where does it occur [44]? Who has observed it [44]? Why is it important to be solved [44]?	
3.	Context, independent variables (aka. environment, setting)	Study location, time constraints, application domain, organization, tools, market, process (e.g. software development methodology), size of project, product that is being developed	
4.	Related work	Other existing work, alternative technologies, solutions, and studies	
5.	Goals and Hypotheses	Null hypothesis and one or more alternative hypotheses for each goal	
6.	Research questions		
7.	Design, Research methods		
7.1.	Design	Description of each step of the study	
7.2.	Control group	If there is a control group, are participants similar to the treatment group participants in terms of variables that may affect study outcomes[39]?	
7.3.	Randomization	Random selection of participants and objects Random assignment of treatment and objects to participants Random order of treatments in case of paired design. If each participant is assigned two treatments A and B, then part of participants perform A first and the other part start with B	
7.4.	Blocking	Group participants of the study into homogeneous groups called blocks (e.g. students in one course, database developers in one company) and implement the study design within each block independently. The idea is that variability of independent variables (e.g. experience and knowledge of subjects) is smaller within a group. That helps measuring changes in dependent variables [41].	
7.5.	Balancing	Equal number of subjects should be assigned to each treatment [41].	
7.6.	Blinding	Automated assignment of treatments to subjects [41] Automated distribution of study materials to subjects [41] Persons who grade the task results should not know which treatment was used [41] Analyst should not know which treatment group is which [41] Automated data collection from subjects [41]	
8.	Subjects (participants)		
8.1.	Population		
8.2.	Sampling	How sampling is performed? What subjects are included and excluded? [39] What is the type of the sampling (e.g. convenience, random)? Is the sample(selected participants) representative of the population?	
8.3.	"Drop outs" and response rate	Are reasons given for refusal to participate[39]?	
8.4.	Subject motivation	E.g. material benefits, course credits for students, etc.	
9.	Objects	E.g. documents and other artifacts	
10.	Measures, Data collection procedures	Who, when, and how to measure [39]? How is the measurement supported? Is it automated [39]? Are the measures used in the study the most relevant ones for answering the research questions [39]?	
11.	Analysis procedure		
11.1.	Data description	Do the numbers add up across different tables and subgroups [39]?	
11.2.	Data types (continuous, ordinal, categorical)		
11.3.	Scoring systems		
11.4.	Data set reduction, outliers		
11.5.	Statistical methods	Are the assumptions of statistical methods met?	
11.6.	Statistical significance	If statistical tests are used to determine differences, is the actual p -value given [39]? If the study is concerned with differences among groups, are confidence limits given describing the magnitude of any observed differences [39]?	
12.	Validity threats	Threats, implications of the threats, and threat mitigation	
12.1.	Side-effects during study execution	Deviations from the plan, solutions for the deviations	
13.	Most important findings	Are all study questions answered [39]? Are negative findings presented [39]?	
14.	Industry impact, inference, generalization	What implications does the report have for practice [39]? How and where the results can be used? Limitations under which findings are relevant [44]?	
15.	Future work		