

Mapping the Knowledge Artifact Terrain

A Quantitative Resource for Qualitative Research

Federico Cabitza and Angela Locoro

*Dipartimento di Informatica, Sistemistica e Comunicazione,
Università degli Studi di Milano Bicocca, Viale Sarca, 336, 20126, Milano, Italy*

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Abstract: In this paper, we present a method by which to build a metaphorical map of a portion of the scholarly literature along conceptual dimensions that have been previously characterized in terms of positive, negative and neutral terms. The method allows to “locate” scholarly works in this space, according to multiple criteria, like the definitions that they contain; the relevant concepts that can be extracted by means of a content analysis; and relevant passages that researchers can extract in studying their content. The resulting maps are not representational, nor trying to extract any objective essence of a scientific contribution. Rather, they are resources for the qualitative research, review and *interpretation* of literature sources. As such, these maps are “knowledge artifacts” in themselves, as they visualize, so to say, the interpretation of a set of works by qualitative researchers, and allow to build a visual comprehension of topological and qualitative relationships between the considered literature contributions. We applied the method to the case of the “knowledge artifact” literature and report the main results in this paper.

1 INTRODUCTION

A Knowledge Artifact (KA) is any artifact built to support knowledge-related processes. This purposely generic definition allows to cover a broad spectrum of instances of this concept, which nevertheless present many differences and mirror different perspectives toward what knowledge is and how it can be supported.

In a recent qualitative literature survey (Cabitza and Locoro, 2014), the authors accomplished an extensive review of the heterogeneous scholarly contributions that had focused that far on the concept of Knowledge Artifact. This review resulted in an interpretative and bottom-up framework by which to characterize single instances of a KA in terms of two opposite and complementary design dimensions: *objectivity* and *situativity*. In that work objectivity was denoted as the dimension characterizing the KAs that are more oriented to a model-driven and Artificial Intelligence (AI) approach to knowledge management. Conversely, situativity was considered the dimension characterizing those KAs that adopt a more constructivist, practice- and collaboration-oriented approach to knowledge support. It is obvious that clear-cut distinctions are useful only for theoretical and analytical purposes, and that in reality both dimensions are

present at different degrees in each Knowledge IT Artifact (KITA). The above-mentioned literature review was aimed at discovering this two-dimensional mix by a qualitative analysis that was carried out by the researchers manually.

An ideal continuation of this approach may be the application of a more systematic method by which to extract a sort of polarity of the literature sources under consideration along some dimensions of interest, like the two mentioned above; and then to represent these sources in the metaphorical space defined by these conceptual dimensions, which are assumed to be orthogonal and independent.

Thus this paper can be seen as a follow-up of the contribution mentioned above in that: 1) it proposes a method by which to build a knowledge artifact supporting the study of a body of scholarly contributions; and 2) also, by applying this method in a case study focusing on the KA literature, it sheds light on the more recent and relevant contributions from this body of literature.

The method by which the conceptual space mentioned above would be populated is based on the content that the researchers extract from the literature sources during their analysis, and takes also in consideration specific lexicons (i.e., lists of words) that

Table 1: The objectivity (above) and situativity (below) lexicons.

algorithm, analytical, artifact, autonomous, autonomy, bi, body, business, capture, categorize, codification, codify, combination, communicate, communication, complete, computational, concept, conceptual, correct, crm, datawarehouse, decidable, decision, determinist, deterministic, discrete, document, dss, encode, engineer, engineering, exchangeable, expert, explicit, externalization, externalize, externalizing, factual, fix, formal, formalism, formality, functionalist, hard-coded, hard-code, independent, independently, information, intelligence, is, ka, km, knowledge, management, map, metadata, minimalistic, minimalist, model, nomothetic, objective, objectivist, objectivistic, objectivity, olap, ontology, order, outcome, passive, positivist, positivistic, predictive, prescribe, prescriptive, problem, procedural, procedure, process, processual, procedural, rational, record, regulation, repository, represent, representable, representation, representational, retrieve, semantic, semi-formal, solve, specify, static, store, structure, symbolic, top-down, transfer, valid, validity, validate.
action, actionable, activity, affordance, agency, articulate, articulation, artifact, augment, awareness, beyond, body, bottom-up, brainstorming, cad, chaotic, co-worker, co-create, co-creation, collaborate, collaboration, collectivity, collectively, communicate, communication, community, constructivism, constructivist, constructivistic, consumer, context, contextual, continuous, convey, cooperate, cooperation, cop, creative, cultural, cybernetic, decision, document, emerge, emergentist, enable, evolve, experience, externalize, embody, fit, flexibility, flexible, fluid, groupware, holistic, human-embodied, human-embody, incomplete, informal, information, innovation, input, integrate, interact, interaction, interative, intermediary, internalization, interpretation, interpretive, interpretable, ka, knowledge, learn, learner, learning, local, locality, malleability, malleable, manipulate, mediate, mediation, negotiable, nominalism, nominalist, nominalistic, organization, others, partial, perform, performance, personalization, practice, pragmatic, pragmatist, pragmatically, presence, problem, producer, product, proxy, reconcile, reconciliation, relational, result, retrieve, share, situate, situation, situational, situativity, skill, social, socialization, sociotechnical, socio-technical, solve, stakeholder, subjective, subjectivity, support, synergistic, tacit, team, think, thinking, training, transfer, undecidable, underspecify, understand, understanding, unpredictable, unstructured, unstructure, usable, use, user-driven, utilization, vehicle, voluntarist, voluntaristic, word, working.

the researchers have previously prepared for each dimension at hand (in our case, objectivity and situativity). These lexicons contain so called *polar terms* because their content is compared to the extracted content in order to measure the degree of polarity of a paper with respect to each dimension, i.e., its degree of *dimension-ness*.

As anticipated above, this space should be considered a knowledge artifact in itself for the following reasons. Points in this space would not merely “represent” literature sources, but rather help the researcher look at, in a way, her reified interpretation of those contributions, and be supported in getting a visual comprehension of the literature of her interest (in our specific case of the body of literature regarding the KA concept). The space and the objects therein located would then support reflective insight, collaborative discussion, and discovery, also, e.g., of proximities, affinities, alignments, trends that can be found among the analyzed sources (obviously still on a metaphorical level). The maps that result from the application of our method can then be seen as resources for the qualitative interpretation of a body of literature, as this latter is processed in terms of linguistic prevalence and polarity. We will see how this is accomplished in the next Section; then in Section 3, we will validate the method by applying it to the KAs research, and we will give some examples of how the considered literature can be mapped giving visual evidence of the intrinsic diversity of the scholarly contributions and their possible interpretations. At last, we

will be back to the map metaphor again in Section 4, which will close the paper.

2 METHOD

The method that we aim to propose for the visualization of scientific content along the dimensions of objectivity and situativity¹ considers three intertwined aspects that contribute in making a literature source valuable: i) the definition aspect, represented in terms of all of the sentences that in a paper give an explicit definition of a KA; ii) the design-oriented aspect, represented in terms of the sentences that in the paper describe the functionalities or the main requirements motivating the design of a KA in the paper; iii) the theoretical aspect, represented in terms of a list of nominal categories that researcher can extract by trying to understand the underlying assumptions that drove the authors of a paper in discussing the definitions as well as the design aspects of a KA. In particular, terms for describing the theoretical aspect can result from any technique of content analysis and interpretative paradigm aimed at the construction of a theory, ontology or model, through the analysis of the paper’s content, and can be seen as framework meta-keywords that can be added to the paper through an

¹The reader should mind that the method is intended to be general with respect to the dimensions of analysis, and that we applied it to these two dimensions for the sake of example only.

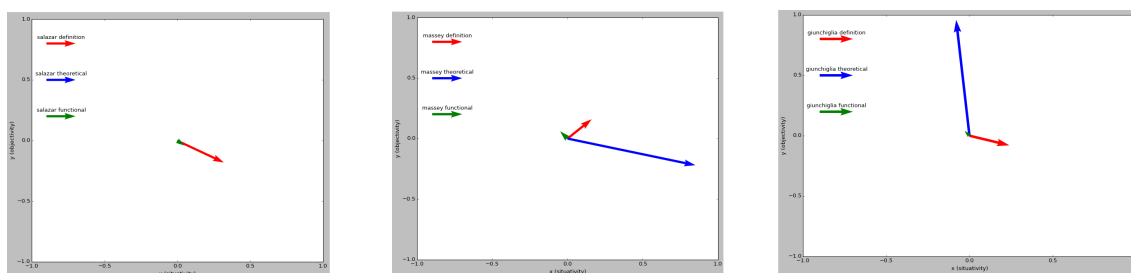


Figure 1: Visualization of the component vectors for, respectively, the definition, design, and theoretical aspects of the three papers examined.

annotation process, to enrich it or support its interpretation.

As anticipated in the Introduction, the method is proposed as an interpretative tool, which could support expert scholars in getting a multimodal (i.e., both textual and visual) comprehension of the conceptual dimensions of the works they consider, flanking traditional techniques of content analysis that require both reading and extracting meaningful material out of the research papers under consideration.

Among the most popular applications of polarity detection in written (and spoken) texts there are “opinion mining” and “sentiment analysis” (Pang and Lee, 2002). These approaches aim to “integrate emotional aspects in natural language understanding” (Cambria and Hussasin, 2015), in order to extract information on the dimension of the “sentiment” of people, that is their feeling or mood, being these voters, consumers, community-members or simply citizens. For instance, the main approaches to sentiment analysis are machine-learning and dictionary-based ones (Rice and Zorn, 2013). The latter approach exploits a list of polar terms to give a measure of the emotional valence of a text (either positive or negative), being it either a whole document, or a portion of it, like just a single sentence.

In general, common techniques used to measure a dimension of a text, like its “sentiment”, encompass the representation of texts by means of word vectors and word vector comparison by means of document classifiers (Pang et al., 2002). However, “mining opinions and sentiments from natural language [...] is an extremely difficult task as it involves a deep understanding of most of the explicit and implicit, regular and irregular, syntactical and semantic rules that are proper of a language” (Cambria and Hussasin, 2015). Moreover, for long texts and generic vocabularies “such approaches turns critically on the quality and comprehensiveness with which the dictionary reflects the sentiment in the texts to which it is applied” (Rice and Zorn, 2013). For these reasons, a polarity detection approach would more probably result

suitable whenever applied to a specific domain, with a specific domain vocabulary created by hand by domain experts, and applied to very short domain texts or group of sentences. And as a means to further comprehension and interpretation of multiple texts, rather than as an end in itself.

In this paper, we try to generalize the approach of sentiment analysis to the analysis of any dimension that can be characterized by a vocabulary of dimension-relevant (either positive or negative) terms to extract and evaluate the polarity and valence (along that dimension) of any text. In our specific case, we are interested in potential expressions of objectivity and situativity within a scientific KA-related contribution, instead of positive or negative emotions in generic content. We have applied the resulting method to the case of the body literature regarding both “representational KAs” and “socially-situated KAs” (Cabitza and Locoro, 2014).

More precisely, in our method we represent strings (i.e., sets of words, or 1-grams) in terms of *vectors* along a dimension, that is elements in a mono-dimensional vector space characterized by: a direction (i.e., a polarity, which we indicate as a coefficient α); and a magnitude, i.e., the degree of dimensionness of the string; as said above, a common example of dimension is “sentiment”, which can be either negative or positive, as well as situativity and objectivity, as in the case study reported in Section 3.

We model the vector magnitude in terms of the product of two components: β and γ . The former is the *valence* of the predominant polarity within the string with respect to the other polarity (within the same string): therefore it is a sort of *relative degree of polarization* of the string. The latter (γ) is the degree of *absolute polarity* of the string, that is how much it is polarized with respect to the total length. In other words, we take the importance of the most relevant lexicon within a text (γ), and then we weight it according to its relative importance with respect to the other lexicon within a specific string (β). Because the vector magnitude is normalized with respect to the string

length, string vectors are defined on a continuous interval $[-1, 1]$ symmetric around the origin (which represents perfect neutrality).

Obviously, strings can be represented along more than one dimension (e.g. (Mikolov et al., 2013)). In this case, they are vectors in a multidimensional vector space, that is *sum vectors*, resulting from the composition of each dimension vector defined above.

As anticipated in Section 1, to illustrate our method we consider just two particular dimensions: objectivity and situativity. For each of these two dimensions we defined one lexicon, D , divided in two partitions, D^+ and D^- : these are *unordered sets* of both neutral and positive/negative words (1-grams, excluding stop-words), respectively, with respect to the dimension at hand. Furthermore, for simplicity's sake, in this study we assume that these two dimensions are mutual opposites, that is $S_{Situativity^+} \equiv S_{Objectivity^-}$, and vice versa.

As hinted at in Section 1, our method also considers to extract three strings from each literature contribution: the set of the definitions therein given by the authors and identified as such by the researcher; the set of codes that the researcher assigns to the paper, following any hermeneutics technique; the set of the relevant passages that the researcher extracts from the paper. Each one of these strings are vectors in the bi-dimensional vector space, or objectivity - situativity plane.

Here we emphasize the main aim of the method that is represented in the three Formulas 1, 2 and 3 to calculate polarity, valence and the string vector magnitude, respectively. While a more syntactic approach would have been aimed at *representing* a research contribution within the vector space by applying the method to the whole paper content, we aim it at supporting qualitative research. Therefore, we apply the algorithm represented in the formulas mentioned above to content that is produced by the qualitative researcher during the study of each literature source. We then map the outputs of her study on the definition, design and theoretical level in terms of a cumulative vector (each vector associated with one single paper) on the conceptual plane.

$$\alpha = \begin{cases} -1 & \text{if } |S_{D^+}| < |S_{D^-}| \\ +1 & \text{if } |S_{D^+}| \geq |S_{D^-}| \end{cases} \quad (1)$$

$$|\beta_D| = \begin{cases} \frac{|S_{D^{grt}}| - |(S_{D^{lwr}} \setminus (S_{D^{grt}} \cap S_{D^{lwr}}))|}{|S_{D^+} \cup S_{D^-}|} & \text{if } |S_{D^+}| \neq |S_{D^-}| \\ \varepsilon & \text{otherwise} \end{cases} \quad (2)$$

$$\gamma_{D^i} = \beta_{D^i} \cdot \frac{|S_{D^{grt}}|}{|S|} \quad (3)$$

In Formula 1, α is the polarity coefficient; it is +1 if D^{grt} is D^+ (the positive lexicon defined for dimension D), that is when the cardinality of D^+ is greater than the cardinality of D^- (the negative lexicon defined for dimension D); α is -1, otherwise, that is if D^{grt} is D^- or the cardinality of D^- is greater than the cardinality of D^+ .

In Formula 2, S_{D^+} is the set of occurrences of the words belonging to D^+ , that has been found in the string S (also, the set of word occurrences connoted either positively or neutrally along dimension D within S). S_{D^-} is the set of occurrences of the words belonging to D^- that has been found in the string S (also, the set of word occurrences connoted either negatively or neutrally along dimension D within S). $S_{D^{grt}}$ and $S_{D^{lwr}}$ indicate which set among the two sets mentioned above has greater and lower cardinality, respectively, with respect to the other set. The ε parameter is an arbitrarily small constant (e.g., .1) to allow for the visualization of null polarity. In our method, if we detect a prevalence of either positive or negative words within the string S , we connote the neutral words, that are the words contained in $D^+ \cap D^-$, as either positive or negative, respectively. This simple context-based and majority-driven polarization of neutral words is reasonable when the difference between the cardinality of S_{D^+} and S_{D^-} is much greater than the number of neutral words defined in both lexicons, as it is often the case.

In Formula 3, γ_{D^i} is the magnitude of the string vector, that is the coordinate associated with valence β_{D^i} taken along the axis representing the i -th dimension. In calculating the cardinality of S , duplicate words are considered as many times they are replicated.

3 THE CASE STUDY

The method described in Section 2 was applied to the case of the “knowledge artifact” literature, in order to validate it and report the outcome of its application to a typical qualitative research task: literature review and study.

To this aim, we created the objectivity and situativity lexicons by hand; as anticipated above, these lexicons encompassed positively related terms, negatively-related ones in two distinct partitions, and neutral words in both of them, i.e. words that cannot be considered as either clearly positive or negative, but that nevertheless are relevant for the domain at hand (e.g., “knowledge”). We then applied our method of polarity detection to a selection of research papers identified in (Cabitza and Locoro, 2014).

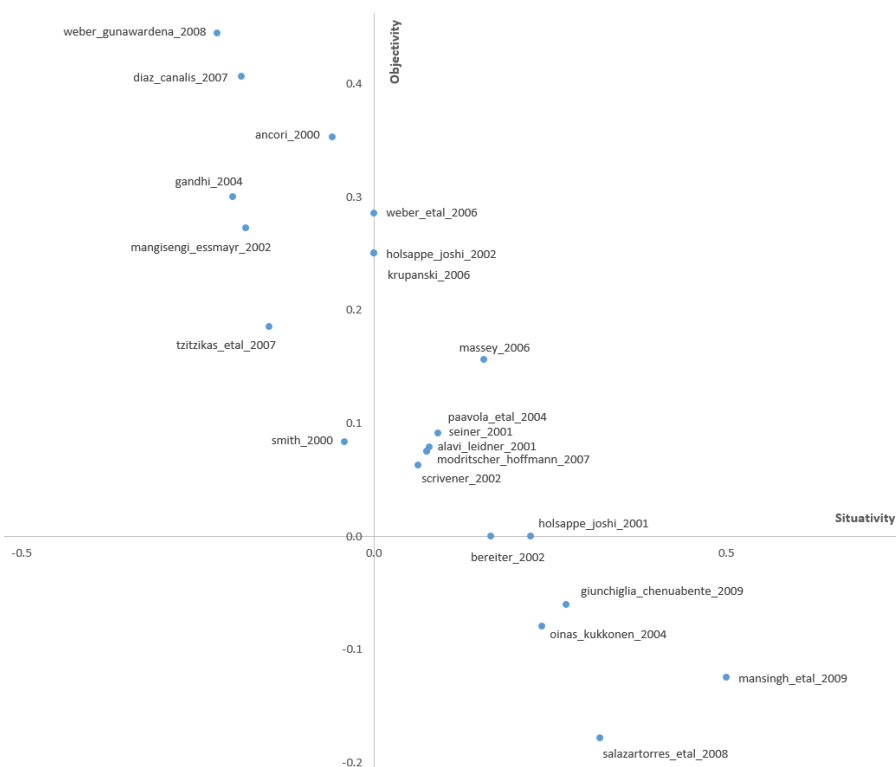


Figure 2: Mapping of the considered papers in regard to their definition aspect within the objectivity-situativity plane.

To create the lexicons we proceeded as follows. We considered 22 papers that in (Cabitza and Locoro, 2014) were found to contain one or more explicit definition statements of the KA, and 3 more research papers from the 10 papers that in that literature survey were found to be concerned with KA design. These latter works are (Massey and Montoya-Weiss, 2006; Salazar-Torres et al., 2008; Giunchiglia and Chenu-Abente, 2009). The authors extracted the terms from these primary sources independently, on the basis of their perceived relevance and capability to connote the concepts of objectivity and situativity (either positively or negatively).

The overall list of terms contained all of the terms extracted by both of us, while terms extracted by only one of the authors were reviewed jointly to decide for its inclusion on the list. Then we classified the identified terms as either positive or negative (and of course neutral), also in this case independently: inter-rater agreement was assessed in terms of Cohen’s kappa, and this process iterated each time after a short discussion on the term that had been classified differently. After a few iterations, we got sufficient agreement for both the objectivity and situativity lexicons, which are reported in Table 1. So far objectivity terms are 103, whereas the situativity lexicon contains 145 terms. We recall here that, as part of our method, the

lexicons can be continuously refined, for any further research purpose, also according to the papers considered and the raters involved.

The 22 papers mentioned above were used also to validate our procedure of polarity detection and visual mapping. In particular, we extracted the KA definitions as bag-of-words, and processed them as 1-grams by means of a Python script that executed the lemmatization of word tokens² and the removal of stopwords³. We then computed the valence mentioned in Section 2 according to Formulas 2 and 1, and computed the objectivity as well as the situativity degrees by exploiting Formula 3.

Figure 2 depicts the visual mapping of the 22 papers according to their definitions of KA in terms of situativity and objectivity coordinates, which result from the application of our method⁴.

An example of insight that this kind of visualization can suggest is how the literature has changed over time: from Figure 2, by looking at the paper position and their date of publication (marked in the point la-

²We used the Wordnet lemmatizer of the nltk package available at <http://www.nltk.org/index.html>

³The stopwords list is available at <http://algs4.cs.princeton.edu/35applications/stopwords.txt>

⁴In Figure 2 we chose to plot only the resulting points, and not the entire vector, for readability’s sake.

bel) one can observe that, in recent times, the number of situativity-oriented works is increasing over time (showing perhaps a growing interest or sensitivity towards that approach?). Another example is the observation of the formation of “clusters” of papers in particular areas of the space (e.g. the I, the II and the IV quadrants), and with particular distance proportions and shapes: e.g. the uniformity of spacing between the papers located in the II quadrant; the mutual proximity of the papers of the I quadrant; the more scattered set of papers of the IV quadrant. This can give some hints on the linguistic uniformity in the definitions of the “representational” (hence more objective) KAs and, on the contrary, of the linguistic variety in the definitions provided for the “socially-situated” KAs. Also, a further look can be given to see whether close papers are also describing similar KAs, or whether the same researcher (or group of researchers) has maintained the same theoretical “stances” over time, and so on.

In general, with this kind of qualitative visualization it is possible to inquire into the linguistic properties of the papers that the method has grouped together (spatially) to see whether they should be also grouped conceptually to some extent (e.g., by writing style, by theoretical stance, by objectivity and situativity attitude, and so on). As the researchers’ aims can vary a lot, as well as their “mental models”, different “mappings” can be open to various interpretation and insights, according to the research scope and aims that tap in this particular knowledge artifact.

For the 3 papers that focused on three specific KA applications, we proceeded as follows. We created three word sets, obtained by extracting: their definitions (as in the 22 papers mentioned above); the relevant passages describing the concrete application of the KA at hand; the keywords that the authors agreed to apply to each of them. For this latter task, we followed a hybrid approach (Fereday and Muir-Cochrane, 2008), trying to apply the lexicon terms first, but then also any other code we agreed upon, after a short cycle of iterative revisions.

Applying the script mentioned above to these three word sets, we obtained the component vectors related, respectively, to the definition-, the theory- and design-related aspects of each paper (see Figure 1 for a representation of each dimension as a vector for each paper). This representation allows to compare each of the relevant dimensions of a paper, and to compare them with the same dimensions of other papers, discovering for example that the paper (Salazar-Torres et al., 2008) has a uniform direction of the three dimensions of definition, theory and design, whereas the paper (Giunchiglia and

Chenu-Abente, 2009) presents quite an opposite view in terms of definitorial vs theoretical stances; finally, the paper (Massey and Montoya-Weiss, 2006) lays “somewhere in the middle”, presenting with a more emphasized socially-situated stance along the theoretical dimension with respect to the definitorial dimension. The design-related aspects of both the papers (Giunchiglia and Chenu-Abente, 2009) and (Massey and Montoya-Weiss, 2006) seem to go in the opposite directions along the theoretical dimension, although the design aspect seems to be less technical in the latter work than in the former one, which covers the technical details of the knowledge artifact at hand for approximately half of its length.

By composing together these three vectors for each of the papers under consideration, we obtained their overall vector representation on the objectivity-situativity vector space, as depicted in Figure 3.

This figure shows that our content analysis was such that paper (Salazar-Torres et al., 2008) looks *less situativity-oriented* than paper (Massey and Montoya-Weiss, 2006), and that paper (Giunchiglia and Chenu-Abente, 2009) looks much more *objectivity-oriented* than the first two papers⁵. Figure 2 can also suggest considerations about the relative (conceptual) proximity and mutual alignment between works: (Salazar-Torres et al., 2008) is maybe closer to (Massey and Montoya-Weiss, 2006) than to (Giunchiglia and Chenu-Abente, 2009), but even more importantly, they look more “aligned” to each other.

4 DISCUSSION AND CONCLUSIONS

Terrain is not a territory. For the dictionary, a terrain is any piece of ground with reference to its physical character. It is therefore the element of a territory that human beings have to cope with. We intend this term metaphorically to intend the scholarly landscape encompassing the diverse literature contributions regarding any concept, and in our specific case the concept of “knowledge artifact”. In this light, we intend the term terrain used in the title to be semantically closer to the French *terroir*, a term that traditionally “refers to the complex ecology in which a given vineyard is located [and that] evokes the unique

⁵We believe that trying to distinguish if the papers actually *are* or rather *look* according to the researcher interpretative lens would be out of the scope of this work. The maps provided by our method, like those depicted in Figures 3 and 2 are given as complementary visual resources to content analysis and literature interpretation.

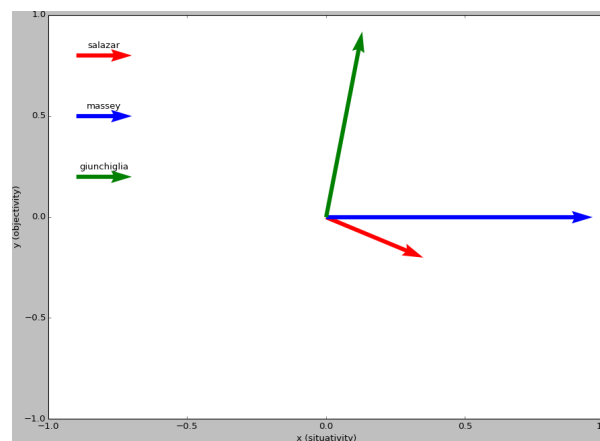


Figure 3: Visual mapping of the composite vectors for the three papers examined.

qualities of the soil, weather patterns, situated wine-making practices, sunshine, and irrigation that yields a particular and recognizable character to the wines that result” (McNely and Rivers, 2014). If we take the metaphor seriously, this means that the body of literature contributions on a particular topic “exists outside human control” (ibid.) but it also bears fruit (like knowledge, insights, new ideas) only in the actual practice of the researchers that explore and, so to say, harvest it.

Following the metaphor again: in order to exploit, take possession of, and also orient themselves in, terrains, human beings construct and use maps. One important point many cartography enthusiasts know well is that maps do not necessarily depict or represent (figuratively) terrains; obviously maps *relate to* terrains, but they are rather “resources for action” (Suchman, 2007), that is means to understand, explore and exploit terrains. A typical practice where maps are used is wayfinding: in this case, maps are just like marked pathways in the wood, or signposts affixed at relevant crossroads. Another practice is also *contemplation*: maps can be consulted just for the aesthetic pleasure to find them accurate, complete, up-to-date, clean and elegant. This should not be considered a lazy activity, as maps can also act as triggers for detecting relationships between terrain elements, as well as to reflect on them and discuss them.

In this paper we have presented a way to map a metaphorical terrain of a portion of the scholarly literature found to be related to a specific topic. This terrain unfolds conceptually along *discursive dimensions*, that is dimensions that are characterized in terms of positive, negative and neutral terms. The mapping method we devised, although simple, allows for a cursory “locating” of scholarly works in this space according to multiple criteria, like the definitions that they contain; the relevant concepts that can

be extracted by means of a content analysis; and relevant passages that researchers can extract in studying their content. Some of the insights that researchers may pull out from the visual representations given in all of the Figures of this paper have been illustrated in Section 3 as an outcome of our analysis and classification of the papers. Once again, we stress the fact that the task of placing single contributions in the terrain of interest (that is a discursive terrain) should then not be taken as a task of *representation* of each source’s position (even supposing such a thing really exists or can be pinpointed in any metaphorical space) within this landscape. Indeed, our method is highly dependent on qualitative content analysis in regard to all of its inputs: both the dimensional lexicons, and the strings (set of words) representing each literature source.

The resulting maps that our quantitative method allows to build are not aimed (nor built) to represent a body of works, nor to extract any objective essence of a scientific contribution, if such a thing exists. Far from it, those maps are intended as resources for the interpretation of selected papers by the qualitative researcher, as an aid to literature reviews to allow for qualitative comparisons and evaluations, and a trigger for discussion and idea exchange between scholars about what they study.

That is why we claim that these maps are “knowledge artifacts” in themselves; indeed, they visualize, so to say, the interpretation of a set of works by qualitative researchers, and allow them to build a visual comprehension of topological and qualitative relationships between the considered literature contributions.

In particular, we applied the method to the case of the “knowledge artifact” literature. As such the paper contributes in the literature regarding the concept of knowledge artifact in regard to two main aspects.

- First, we have defined pairs of lexicons (positive and negative ones) for the dimensions of situativity and objectivity (of knowledge artifacts), as these had been defined in (Cabitza and Locoro, 2014). These lexicons have been both produced and reconciled manually, after a comprehensive review of a set of relevant papers and an iterative process of inter-rater agreement that involved the authors. These vocabularies are offered to the community of interested researchers to be progressively maintained, and to enable further research on these topics along the same research strand advocated in (Cabitza and Locoro, 2014).
- Then, we propose an algorithm for the mapping and visualization of arbitrary sets of words (either directly taken or derived from the original literature sources) into the objectivity-situativity bi-dimensional vector space. In this paper, to validate the method we produced the resulting outputs:
 1. the set of the main definitions of knowledge artifact explicitly given by the authors of 22 papers selected from the review presented in (Cabitza and Locoro, 2014);
 2. the sets of definitions and relevant categories that we extracted from three relevant papers selected from the literature review mentioned above;
 3. the sets of all of the relevant design-oriented passages that we extracted from each of these papers, having in mind the concrete applications mentioned in each contribution.
- Each paper has then be mapped in terms of a graphical representation within a vector space, by considering its definition-, theory- and application-oriented aspects (respectively, the word sets of its definitions, categories and relevant passages). The vector-like representation should be also appraised for the related affordance of a “tension” and for evoking a “tendency” rather than a mere position, which eludes any too rigid pinpointing of the characteristics of a research contribution.

If the interpretation of a set of literature sources and the discussion of the related topics can be fostered by looking at the vector space that we propose to build with our method as a map for qualitative literature reviews, the main goals of our study would be reached. In this case, our method could complement the study of scholarly sources, and facilitate the qualitative researcher in extracting insights from the literature.

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