Enhancement of Knowledge Resources and Discovery by Computation of Content Factors

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Abstract—This paper presents a methodology for data description and analysis, the Content Factor (CONTFACT). The Content Factor method can be applied to arbitrary data and content and it can be adopted for many purposes. Normed factors and variants can also support data analysis and knowledge discovery. This paper presents the algorithm, introduces into the norming of Content Factors, and discusses examples and a practical case study and implementation based on long-term knowledge resources, which are continuously in development. The methodology is used for advanced processing and also enables methods like data rhythm analysis and characterisation. It can be integrated with complementary methodology, e.g., classification and allows the application of advanced computing methods. The goal of this research is to create a general and flexible methodology for data description and analysis that can be used with huge structured and even unstructured data resources, allows an automation, and can therefore also be used for long-term multi-disciplinary knowledge.

Keywords–Data-centric Knowledge Processing; Content Factor (CONTFACT) method; Data Rhythm Analysis; Universal Decimal Classification; Advanced Computing.

I. INTRODUCTION

Information systems handling unstructured as well as structured information are lacking means for data description and analysis, which is data-centric and can be applied in flexible ways. In the late nineteen nineties, the concept of in-text documentation balancing has been introduced with the knowledge resources in the LX Project. Creating knowledge resources means creating, collecting, documenting, and analysing data and information. This can include digital objects, e.g., factual data, process information, and executable programs, as well as realia objects. Long-term means decades because knowledge is not isolated, neither in space nor time. All the more, knowledge does have a multi-disciplinary context.

Therefore, after integration knowledge should not disintegrate, instead it should be documented, preserved, and analysed in context. The extent increases with growing collections, which requires advanced processing and computing. Especially the complexity is a driving force, e.g., in depth, in width, and considering that parts of the content and context may be continuously in development. Therefore, the applied methods cannot be limited to certain algorithms and tools. Instead there are complementary sets of methods.

The methodology of computing factors [1] and patterns [2] being representative for a certain part of content was considered significant for knowledge resources and referred material. Fundamentally, a knowledge representation is surrogate. It enables an entity to determine consequences without forcing an action. For the development of these resources a definition-supported, sortable documentation-code balancing was created and implemented.

The Content Factor (CONTFACT) method advances this concept and integrates a definition-supported sortable documentation-code balancing and a universal applicability. The Content Factor method is focussing on documentation and analysis The Content Factor can contain a digital ‘construction plan’ or a significant part of digital objects, like sequenced Deoxyribonucleic Acid (DNA) does for biological objects [3]. Here, a construction plan is what is decided to be a significant sequence of elements, which may, e.g., be sorted or unsorted. Furthermore, high level methods, e.g., “rhythm matching”, can be based on methods like the Content Factor.

Classification has proven to be a valuable tool for long-term and complex information management, e.g., for environmental information systems [4]. Conceptual knowledge is also a complement for data and content missing conceptual documentation, e.g., for data based on ontologies used with dynamical and autonomous systems [5].

Growing content resources means huge amounts of data, requirements for creating and further developing advanced services, and increasing the quality of data and services. With growing content resources content balancing and valuation is getting more and more important.

This paper is organised as follows. Section II summarises the state-of-the-art and motivation, Sections III and IV introduce the Content Factor method and an example for the application principle. Section V shows implemented Content Factor examples, explains flags, definition sets, and norming. Section VI provides the results from an implementation case study, showing complementary properties and complex scenarios. Section VII discusses aspects of processing and computation. Sections VIII and IX present and evaluation and main results, summarise the lessons learned, conclusions and future work.

II. STATE-OF-THE-ART AND MOTIVATION

Most content and context documentation and knowledge discovery efforts are based on data and knowledge entities. Knowledge is created from a subjective combination of different attainments, which are selected, compared and balanced
against each other, which are transformed, interpreted, and used in reasoning, also to infer further knowledge. Therefore, not all the knowledge can be explicitly formalised.

Knowledge and content are multi- and inter-disciplinary long-term targets and values [6]. In practice, powerful and secure information technology can support knowledge-based works and values. Computing goes along with methodologies, technological means, and devices applicable for universal automatic manipulation and processing of data and information. Computing is a practical tool and has well defined purposes and goals.

Most measures, e.g., similarity, distance and vector measures, are only secondary means [7], which cannot cope with complex knowledge. Evaluation metrics are very limited, and so are the connections resulting from co-occurrences in given texts, e.g., even with Natural Language Processing (NLP), or clustering results in granular text segments [8].

Evaluation can be based on word semantic relatedness, datasets and evaluation measures, e.g., the WordSimilairity 353 dataset (EN-WS353) for English texts [9]. The development of Big Data amounts and complexity up to this point show that processing power is not the sole solution [10]. Advanced long-term knowledge management and analytics are on the rise.

Value of data is an increasingly important issue, especially when long-term knowledge creation is required, e.g., knowledge loss due to departing personnel [11]. Current information models are not able to really quantify the value of information. Due to this fact one of the most important assets [12], the information, is often left out [13]. Today a full understanding of the value of information is lacking. For example, free Open Access contributions can bear much higher information values than contributions from commercial publishers or providers.

For numberless application scenarios the entities have to be documented, described, selected, analysed, and interpreted. Standard means like statistics and regular expression search methods are basic tools used for these purposes.

Anyhow, these means are not data-centric, they are volatile methods, delivering non-persistent attributes with minimal descriptive features. The basic methods only count, the result is a number. Numbers can be easily handled but in their solenity such means are quite limited in their descriptiveness and expressiveness.

Therefore, many data and information handling systems create numbers of individual tools, e.g., for creating abstracts, generating keywords, and computing statistics based on the data. Such means and their implementations are either very basic or they are very individual.

The pool of tools requires new and additional methods of more universal and data-centric character – for structured and unstructured data.

New methods should not be restricted to certain types of data objects or content and they should be flexibly usable in combination and integration with existing methods and generally applicable to existing knowledge resources and referenced data. New methods should allow an abstraction, e.g., for the choice of definitions as well as for defined items.

III. THE CONTENT FACTOR

The fundamental method of the Basic Content Factor (BCF), \( \kappa_B \), “Kappa-B”, and the Normed Basic Content Factor (NBCF), \( \pi_B \), can be described by simple mathematical notations. For any elements \( o_i \) in an object \( o \), holds

\[
o_i \in o.
\]

The organisation of an object is not limited, e.g., a reference can be defined an element. For \( \kappa_B \) of an object \( o \), with elements \( o_i \) and the count function \( c \), holds

\[
\kappa_B(o_i) = c(o_i).
\]

For \( \pi_B \) of an object \( o \), for all elements \( n \), with the count function \( c \), holds

\[
\pi_B(o_i) = \frac{c(o_i)}{\sum_{i=1}^{n} c(o_i)}.
\]

All normed \( \kappa \) for the elements \( o_i \) of an object \( o \) sum up to 1 for each object:

\[
\sum_{i=1}^{n} \pi_B(o_i) = 1.
\]

For a mathematical representation counting can be described by a set \( o \) and finding a result \( n \), establishing a one to one correspondence of the set with the set of ‘numbers’ \( 1, 2, 3, \ldots, n \). It can be shown by mathematical induction that no bijection can exist between \( 1, 2, 3, \ldots, n \) and \( 1, 2, 3, \ldots, m \) unless \( n = m \). A set can consist of subsets. The method can, e.g., be applied to disjoint subsets, too. It should be noted that counting can also be done using fuzzy sets [14].

IV. APPLICATION EXAMPLE

The methodology can be used with any object, independent if realia objects or digital objects. Nevertheless, for ease of understanding the examples presented here are mostly considering text and data processing. Elements can be any part of the content, e.g., equations, images, text strings, and words. In the following example, “letters” are used for demonstrating the application. Given is an object with the sample content of 10 elements:

\[
\text{A T A H C T O A R Z}
\]

For this example it is suggested that \( A \) and \( Z \) are relevant for documentation and analysis. The relevant elements, \( \text{AAA}AZ \), in an object of these 10 elements for element \( A \) means \( 3/10 \) normed so the full notation is

\[
\text{AAA}AZ/10 \text{ with } \pi_B(A) = 3/10 \text{ and } \pi_B(Z) = 1/10.
\]

In consequence, the summed value for \( \text{AAA}AZ/10 \) is

\[
\pi_B(A, Z) = 4/10.
\]

\( \text{AAA}AZ \) in an object of 20 elements, for element \( A \) means \( 3/20 \) normed, which shows that it is relatively less often in this object. \( 3/22 \) for element \( A \) for this object means this object or
an instance in a different development stage, e.g., at a different time or in a different element context. The notation
\[
\{i_1, i_2, i_3, \ldots, i_n\}/n
\]

of available elements holds the respective selection where \(\{i_1, i_2, i_3, \ldots, i_n\}\) refers to the definitions of element groups. Elements can have the same labels respectively values. From this example it is easy to see that the method can be applied independent from a content structure.

V. CONTENT FACTOR EXAMPLES

The following examples (Figures 1, 2, 4, 3, 5) show valid notations of the Normed Basic Content Factor \(\pi_B\), which were taken from the LX Foundation Scientific Resources [15]. The LX Project is a long-term multi-disciplinary project to create universal knowledge resources. Application components can be efficiently created to use the resources, e.g., from the Geo Exploration and Information (GEXI) project. Any kind of data can be integrated. Data is collected in original, authentic form, structure, and content but data can also be integrated in modified form. Creation and development are driven by multi-fold activities, e.g., by workgroups and campaigns. A major goal is to create data that can be used by workgroups for their required purposes without limiting long-term data to applications cases of a specific scenario. The usage includes a targetted documentation and analysis. For the workgroups, the Content Factor has shown to be beneficial with documentation targeted documentation and analysis. There are countless fields to use the method, which certainly depend on the requirements of the workgroups. For the majority of use cases, especially, selecting objects and comparing content have been focus applications. With these knowledge resources multi-disciplinary knowledge is documented over long time intervals. The resources are currently already developed for more than 25 years. A general and portable structure was used for the representation.

<table>
<thead>
<tr>
<th>Purpose</th>
<th>Flag</th>
<th>Meaning</th>
</tr>
</thead>
<tbody>
<tr>
<td>Content Factor quality</td>
<td>U</td>
<td>Unsorted</td>
</tr>
<tr>
<td></td>
<td>S</td>
<td>Sorted</td>
</tr>
<tr>
<td>Content Factor source</td>
<td>M</td>
<td>Manual</td>
</tr>
<tr>
<td></td>
<td>A</td>
<td>Automated</td>
</tr>
<tr>
<td></td>
<td>H</td>
<td>Hybrid</td>
</tr>
</tbody>
</table>

The CONFACT core entries can have various qualities, e.g., unsorted (U) or sorted (S). Unsorted means in the order in which they appear in the respective object. Sorted means in a different sort order, which may also be specified. CONFACT entries can result from various workflows and procedures, e.g., they can be created on manual base (M) or on automated base.
(A). If nothing else is specified the flag refers to the way object entries were created. Content Factor quality refers to core entries, source also refers to the definitions and information.

The Content Factor method provides the specified instructions. The required features with an implementation can, e.g., implicitly require large numbers of comparisons, resulting in highly computationally intensive workflows on certain architectures. It is the choice of the user to weighten between the benefits and the computational efforts, and potentially to provide suitable environments.

B. Definition sets

Definition sets for object elements can be created and used very flexibly, e.g., word or string definitions. Therefore, a reasonable set of elements can be defined for the respective purpose, especially:

- Definition sets can contain appropriate material, e.g., text or classification.
- Groups of elements can be created.
- Contributing elements can be subsummarised.
- Definition sets can be kept persistent and volatile.
- Definition set elements can be weighted, e.g., by parameterisation of context-sensitive code growth.
- Context sensitive definition sets can be referenced with data objects.
- Content can be described with multiple, complementary definition sets.
- Any part of the content can be defined as elements.

The Content Factors can be computed for any object, e.g., for text and other parts of content. Nevertheless, the above definition sets for normed factors are intended to be used with one type of elements.

C. Normed application

$\pi_B$ is a normed quantity. Norming is a mathematical procedure, by which the interesting quantity (e.g., vector, operator, function) is modified by multiplication in a way that after the norming the application of respective functionals delivers 1. The respective $\pi_B$ Content Factor can be used to create a weighting on objects, e.g., multiplying the number of elements with the respective factor value.

VI. Implementation

The implementation has been created for the primary use with knowledge resources’ objects (lxcontfact). This means handling of any related content, e.g., documentation, keywords, classification, transliterations, and references. The respective objects were addressed as Content Factor Object (CFO) (standard file extension .cfo) and the definition sets as Content Factor Definition (CFD) (standard file extension .cfd).

A. Case study: Computing complementation and properties

The following sequence of short examples shows a knowledge resources object (Figure 7), and three pairs of complementary CONTFACT definition sets and the according $\pi_B$ computed for the knowledge resources object and respective definition sets (Figures 8 and 9; 10 and 11; 12 and 13).

Figure 7. Artificial knowledge resources object (LX Resources, excerpt).

The right parts are entry and keywords. Here, the algorithm can count in object entry name (right “object A”), keywords (in brackets), and object documentation (lower right block).

Figure 8. CONTFACT definition set 1 of 3 (LX Resources, excerpt).

The definition set defines $\{\text{A}\}$ and $\{\text{O}\}$. The definitions are case sensitive for this discovery. We can compute $\pi_B$ (Figure 9) according to the knowledge resources object and definition set.

Figure 9. NBCF $\pi_B$ computed for knowledge resources object and definition set 1 (LX Resources, excerpt).

The result is shown in a line-oriented representation, each line carrying the respective date-time code for all the core, statistics, and additional information. The second complementary set (Figure 11) defines $\{\text{B}\}$ and $\{\text{D}\}$.

Figure 10. CONTFACT definition set 2 of 3 (LX Resources, excerpt).

The resulting $\pi_B$ is shown in the excerpt (Figure 12).

Figure 11. NBCF $\pi_B$ computed for knowledge resources object and definition set 2 (LX Resources, excerpt).

Figure 12. CONTFACT definition set 3 of 3 (LX Resources, excerpt).
The third complementary set (Figure 13) defines \{c\}.

The sum of all elements considered for $\pi_B$ by the respective CONFACT algorithm in an object is 100 percent. Here, the overall number of

- definitions is $2 + 2 + 1 = 5$,
- elements is 32,
- matches is $13 + 12 + 6 = 31$.

The sum of aggregated $\pi_B$ values for all relevant elements results in $0.40625000 + 0.37500000 + 0.18750000 + 1/32 = 1$.

B. Case study: Complex resources and discovery scenario

The data used here is based on the content and context from the knowledge resources, provided by the LX Foundation Scientific Resources [15]. The LX knowledge resources’ structure and the classification references [16] based on UDC [17] are essential means for the processing workflows and evaluation of the knowledge objects and containers.

Both provide strong multi-disciplinary and multi-lingual support. For this part of the research all small unsorted excerpts of the knowledge resources objects only refer to main UDC-based classes, which for this part of the publication are taken from the Multilingual Universal Decimal Classification Summary (UDCC Publication No. 088) [18] released by the UDC Consortium under the Creative Commons Attribution Share Alike 3.0 license [19] (first release 2009, subsequent update 2012).

The excerpts (Figures 14, 15, 16), show a CFO from the knowledge resources a CFD and the computed CONFACT.

Labels, language fields, and spaces were stripped. A knowledge object can contain any items required, e.g., including storing data, documentation, classification, keywords, algorithms, references, implementations, in any languages and representations, allowing support tables and algorithms. Examples of application scenarios for the Content Factor method range from libraries, natural sciences and archaeology, statics, architecture, risk coverage, technology to material sciences [20].

C. Case study: Rhythm matching and core sequences

As soon as Content Factors have been computed for an object the patterns can be compared with pattern of other objects. The Content Factor method allows to compare the occurrences of relevant elements in objects in many ways. The following example shows the “rhythm matching” method for two computed unsorted CONFACT core sequences (Figures 18, 19) for an object and a definition set (Figure 17).
The Content Factor method does not have built-in or intrinsic information. 

Application and purpose with respective data. Sorting the core data for a certain comparison, is a matter of combining several methods in a workflow is possible. Formalised content maybe candidates for sorted CONTFACT. Unsorted CONTFACT are more likely to describe objects, e.g., with factual, reduced focus on their internal organisation. Sorted CONTFACT can be modularised, which means that not only the CONTFACT core, e.g., in collections and containers, containing millions of entries each. The computation routines allow a modularised and dynamical use. Everything can be considered a set, e.g., an object, a collection, and a container. Therefore, an implementation can scale from single on the fly objects to millions of objects, which may also be kept non invasive, the results can be created dynamically and persistent. Content Factors can be kept volatile as well as persistent. Definitions can be kept with the Content Factor, together with additional Content Factor data, e.g., statistics and documentation. Any of this Content Factor information has been successfully used with long term knowledge resources and with unstructured data. The Content Factor method can describe arbitrary data in a unique form and supports data analysis and knowledge discovery in many ways, e.g., complex data comparison and tracking of relevant changes. Application scenarios may allow to compute Content Factors for many objects in parallel. Content Factors can be computed dynamically as well as in batch mode or "pre-computed". Content Factors can be kept volatile as well as persistent. Everything can be considered a set, e.g., an object, a collection, and a container. Therefore, an implementation can scale from single on the fly objects to millions of objects, which may also be kept non invasive, the results can be created dynamically and persistent. Content Factors can be kept volatile as well as persistent. Definitions can be kept with the Content Factor, together with additional Content Factor data, e.g., statistics and documentation. Any of this Content Factor information has been successfully used with long term knowledge resources and with unstructured and dynamical data. The Content Factor method can describe arbitrary data in a unique form and supports data analysis and knowledge discovery in many ways, e.g., complex data comparison and tracking of relevant changes.

The case study has shown that the formal description can be implemented very flexibly and successful. Application scenarios may allow to compute Content Factors for many objects in parallel. Content Factors can be computed dynamically as well as in batch mode or "pre-computed". Content Factors can be kept volatile as well as persistent. Everything can be considered a set, e.g., an object, a collection, and a container. Therefore, an implementation can scale from single on the fly objects to millions of objects, which may also be kept non invasive, the results can be created dynamically and persistent. Content Factors can be kept volatile as well as persistent. Definitions can be kept with the Content Factor, together with additional Content Factor data, e.g., statistics and documentation. Any of this Content Factor information has been successfully used with long term knowledge resources and with unstructured and dynamical data. The Content Factor method can describe arbitrary data in a unique form and supports data analysis and knowledge discovery in many ways, e.g., complex data comparison and tracking of relevant changes.

The algorithms can be used for single objects as well as for large collections and containers, containing millions of entries each. The computation routines allow a modularised and dynamical use. The parts required for an implementation computing a Content Factor can be modularised, which means that not only the Content Factor computation can be implemented as a module but even core, definitions, and additional parts can be computed by separate modules. A sequence of routine calls used for examples in this case study shows the principle and modular application of respective functions (Figure 20).

The modules create an entity for the implemented Content Factor (begin to end). They include labels, date, unsorted elements and so on as well as statistics and additional information.

### VII. Processing and Computation

#### A. Scalability, modularisation, and dynamical use

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#### B. Parallelisation and persistence

There is a number of modules supporting computation based on persistent data, e.g., in collections and containers. The architecture allows task parallel implementations for multiple instances as well as highly parallel implementations for core routines. Examples are collection and container decollators, collection and container slicers, collection and container atomisers, formatting modules, and computing modules for (intermediate) result matrix requests. Content Factor data can easily be kept persistent and dynamically. The algorithms and workflows allow the flexible organisation of data locality, e.g., central locations and with compute units, e.g., in groups or containers.

### VIII. Evaluation

The case study has shown that the formal description can be implemented very flexibly and successful. Application scenarios may allow to compute Content Factors for many objects in parallel. Content Factors can be computed dynamically as well as in batch mode or "pre-computed". Content Factors can be kept volatile as well as persistent. Everything can be considered a set, e.g., an object, a collection, and a container. Therefore, an implementation can scale from single on the fly objects to millions of objects, which may also be kept non invasive, the results can be created dynamically and persistent. Content Factors can be kept volatile as well as persistent. Definitions can be kept with the Content Factor, together with additional Content Factor data, e.g., statistics and documentation. Any of this Content Factor information has been successfully used with long term knowledge resources and with unstructured and dynamical data. The Content Factor method can describe arbitrary data in a unique form and supports data analysis and knowledge discovery in many ways, e.g., complex data comparison and tracking of relevant changes.

Definition sets can support various use cases. Examples were given from handling single characters to string elements. Definitions can be kept with the Content Factor, together with additional Content Factor data, e.g., statistics and documentation. Any of this Content Factor information has been successfully used to analyse data objects from different sources. The computation of Content Factors is non invasive, the results can be created dynamically and persistent. Content Factors can be automatically computed for elements and groups of large data resources. The integration with data and knowledge resources can be kept non invasive to least invasive, depending on the desired purposes. Knowledge objects, e.g., in collections and
containers, can carry and refer to complementary information and knowledge, especially Content Factor information, which can be integrated with workflows, e.g., for discovery processes.

The benefits and usability may depend on the field of application and the individual goals. The evaluation refers to the case context presented, which allows a wide range of freedom and flexibility. The benefits for the knowledge resources are additional means for documentation of objects. In detail, the benefits for the example workflows were improved data-mining pipelines, due to additional features for comparisons of objects, integrating developing knowledge resources, and creating and developing knowledge resources. In practice, the computation of Content Factors has revealed significant benefits for the creation and analysis of large numbers of objects and for the flexibility and available features for building workflows, e.g., when based on long-term knowledge objects. In addition, creators, authors, and users of knowledge and content have additional means to express their views and valuation of objects and groups of objects. From the computational point of view, the computation of Content Factors can help minimise the recurrent computing demands for data.

IX. Conclusion

This paper introduced a methodology for data description and analysis, the Content Factor (CONTFACT) method. The paper presents the formal description and examples, a successful implementation, and a practical case study. It has been shown that the Content Factor is data-centric and can describe and analyse arbitrary data and content, structured and unstructured. Data-centricity is even emphasized due to the fact that the Content Factor can be seamlessly integrated with the data. The data locality is most flexible and allows an efficient use of different computing, storage, and communication architectures.

The method can be adopted for many purposes. The Content Factor method has been successfully applied for knowledge processing and analysis with long-term knowledge resources, for knowledge discovery, and with variable data for system operation analysis. It enables to specify a wide range of precision and fuzziness for data description and analysis and also enables methods like data rhythm analysis and characterisation, can be integrated with complementary methodologies, e.g., classifications, concordances, and references. Therefore, the method allows weighting data regarding significance, promoting the value of data. The method supports the use of advanced computing methods for computation and analysis with the implementation. The computation and processing can be automated and used with huge and even unstructured data resources. The methodology allows an integrated use with complementary methodologies, e.g., with conceptual knowledge like UDC. It will be interesting to see various Content Factor implementations for individual applications, e.g., dynamical classification and concordances. Future work concentrates on advanced analysis and automation for different application scenarios, e.g., object comparisons, multi-lingual discovery, and concordance discovery.

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