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Some Notes on Polyadic Concept Analysis

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Abstract. Despite the popularity of Formal Concept Analysis (FCA) as a mathematical framework for data analysis, some of its extensions are still considered arcane. Polyadic Concept Analysis (PCA) is one of the most promising yet understudied of these extensions. This formalism offers many interesting open questions but is hindered in its dissemination by complex notations and a lack of agreed-upon basic definitions. In this paper, we discuss in a mostly informal way the fundamental differences between FCA and PCA in the relation between contexts, conceptual structures, and rules. We identify open questions, present partial results on the maximal size of concept n-lattices and suggest new research directions.

 $\textbf{Keywords:} \ \ Formal\ Concept\ Analysis \cdot Polyadic\ Concept\ Analysis \cdot Conceptual\ Structures.$

1 Unde Venis Et Quo Vadis?

Formal Concept Analysis (FCA [1]) is a formalism that establishes a connection between classical binary data (crosstables) and the structure of concepts and rules that can be found in said data. It is very powerful, if underutilised, as it offers well-studied mathematical structures to be exploited by algorithms.

As crosstables are a rather limiting way of representing data, various extensions of the formalism have been proposed to deal with more complex data, such as Pattern Structures [2], Relational Concept Analysis [3], fuzzy FCA [4] or graph FCA [5]. Just like FCA, they are based on lattice theory [6].

Triadic Concept Analysis [7] and Polyadic Concept Analysis (PCA) [8] aim to extend FCA to data in the form of *n*-ary relations (*i.e.* multidimensional crosstables) and have the peculiarity of involving *n*-lattices instead of lattices. Such structures are considerably less known and studied, and results that would be considered basic in lattice theory are missing. While some would consider this an opportunity to pioneer a whole new field, the heavy notations and lack of clear definitions and impressive applicative results repels researchers. The opportunities are however numerous as multidimensional data is now ubiquitous: RDF datasets, folksonomies, pharmacogenomical knowledge

are all inherently at least triadic and transforming them to fit dyadic crosstables only results in lost information. As such data are also understudied in data mining in general, this is a unique occasion to position FCA and its extensions as a leading formalism. Some work has already been done in this direction in the field of multidimensional association rule mining [9,10].

In this paper, we aim at promoting Polyadic Concept Analysis by discussing the differences with FCA introduced by n-lattices, suggesting future research directions, and presenting recent results and open questions on the size of n-lattices. Section 2 contains definitions, including a proposition for a stable definition of implications. Section 3 discusses the impact of the loss of duality between orders in n-lattices on the relation between FCA structures, as well as a suggestion for a multidimensional generalisation of Boolean lattices. Section 4 presents recent results on the maximal size of concept n-lattices found in cubic contexts.

2 The Basics of Boxes

We assume that the reader is at least somewhat familiar with FCA. If this is not the case, we refer the reader to the usual book [1]. Polyadic concept analysis is a natural extension of FCA in which the underlying relation is n-ary. It was first introduced in its triadic form by Lehman and Wille [7], and then generalised to the n-dimensional case by Voutsadakis [8]. In this setting, an n-context is an (n+1)-tuple $(S_1,\ldots,S_n,\mathcal{R})$ where the S_i are sets called dimensions and $\mathcal{R} \subseteq \prod_{i=1}^n S_i$ is an n-ary relation between them. We shall call *objects* the elements of the first dimension S_1 while the (n-1)-tuples (s_2,\ldots,s_n) such that $(s_1,s_2,\ldots,s_n)\in\mathcal{R}$ form the *description* of the object s_1 . An example of a triadic context is depicted in Fig. 1.

	a	b	c	$\mid a$	b	c	a	b	c
1	×			X		×			X
2	× × ×			×					
3	×	X			×			X	×
		α			β			γ	

Fig. 1: A triadic context $(\{\alpha, \beta\}, \{1, 2, 3\}, \{a, b, c\}, \mathcal{R})$.

An n-concept is then a maximal n-dimensional box full of crosses, i.e. an n-tuple (X_1,\ldots,X_n) such that $\prod_{i=1}^n X_i\subseteq \mathcal{R}$ and no X_i is such that

$$\prod_{j=1}^{i-1} \mathcal{S}_j \times X_i \cup \{x\} \times \prod_{k=i+1}^n \mathcal{S}_k \subseteq \mathcal{R}$$

with $x \notin X_i$. For instance, $(\{\alpha, \beta\}, \{1, 2\}, \{a\})$ is an n-concept in Fig. 1. The set \mathcal{C} of all n-concepts in an n-context together with the n quasi-orders induced by the inclusion relation on their n components form an n-lattice $\mathcal{L} = (\mathcal{C}, \lesssim_1, \ldots, \lesssim_n)$. Note that an n-lattice is an n-ordered set, i.e. it respects:

– if
$$\forall i \in \{1,\ldots,n\} \setminus \{j\}, A \lesssim_i B$$
 then $A \gtrsim_j B$ (antiordinal dependency) and – if $\forall i \in \{1,\ldots,n\}, A \sim_i B$, then $A = B$ (uniqueness condition).

Different definitions of implications in triadic and polyadic contexts have been proposed through the years such as "Biedermann's implications" [11], attribute×condition, conditional attribute or attributional condition implications [12]. In [13], it was proposed to consider all the implications that hold in dyadic contexts resulting from combinations of two transformations of an n-context \mathcal{C} (see Fig. 2):

-
$$\mathcal{C}^{(A,B)} = (\prod A, \prod B, \mathcal{R}^{(A,B)})$$
, where A and B form a bipartition of the set of all dimensions and $((s_{a_1},\ldots,s_{a_k}),(s_{b_1},\ldots,s_{b_l})) \in \mathcal{R}^{(A,B)}$ iff $(s_1,\ldots,s_n) \in \mathcal{R}$ with $\{a_1,\ldots,a_k\} \cup \{b_1,\ldots,b_l\} = \{1,\ldots,n\}$

- $C_D = (S_1, \dots, S_{d-1}, S_{d+1}, \dots, S_n, \mathcal{R}_D)$, where $D \subseteq S_d$ and

$$(s_1, \ldots, s_{d-1}, s_{d+1}, \ldots, s_n) \in \mathcal{R}_D \text{ iff } \forall x \in D, (s_1, \ldots, s_{d-1}, x, s_{d+1}, \ldots, s_n).$$

Fig. 2: Transformations $\mathcal{C}^{(\{greek\},\{numbers,latin\})}$ (top), $\mathcal{C}_{\{a\}}$ (bottom left) and $\mathcal{C}_{\{1,3\}}$ (bottom right) of Fig. 1's triadic context \mathcal{C} .

Such implications contain all previously defined types of implications. For instance, in Fig. 1's triadic context $\mathcal C$ and if we assume that $numbers=\{1,2,3\}$ is the set of attributes and $latin=\{a,b,c\}$ is the set of conditions,

- conditional attribute implications $A_2 \xrightarrow{C} A_2$ are the dyadic implications in the 2-context $\mathcal{C}_C^{(\{greek\},\{numbers\})}$, e.g. $\{3\} \xrightarrow{\{a\}} \{1,2\}$,
- attributional condition implications $C_1 \xrightarrow{A} C_2$ are the dyadic implications in the 2-context $\mathcal{C}_A^{(\{greek\},\{latin\})}$, e.g. $\emptyset \xrightarrow{\{3\}} \{b\}$,
- and attribute×condition implications $A \to B$ are simply the dyadic implications in the 2-context $\mathcal{C}^{(\{greek\},\{numbers,latin\})}$, e.g. $\{(1,a)\} \to \{(3,b)\}$.

Note that possible implications include those in $C^{(\{greek, numbers\}, \{latin\})}$ whose support is in a Cartesian product of dimensions. Such rules (under the more general umbrella of association rules) are already under consideration in the data mining community [14,15]. Additionally, if one is only interested in implications that do not contain

the first dimension – the objects – these implications are all derivable from the implications of $C^{(\{S_1\},\{S_2,...,S_n\})}$ through the application of Armstrong's axioms plus two other axioms, as discussed in [13]. This means that one only has to reason on a single type of implication.

Just as in the dyadic case, the set of all the implications that do not involve the first (object) dimension and that hold in an n-context (or any its implication bases) can be used to reconstruct the set of n-concepts restricted to their last n-1 components [13].

We conclude this first section with a small digression about graphical representation. In two dimensions, both partial orders of the concept lattice $(\mathcal{T}(\mathcal{C}),\subseteq_1,\subseteq_2)$ have the decency of being dual, and thus one can be omitted, allowing a concept lattice to be graphically represented by a Hasse diagram. Starting from three dimensions, concepts are ordered differently. In n dimensions, there are classes of equivalent concepts with the same ith component, and those concepts can then be differentiated in the n-1 other quasi-orders. A graphical representation that clearly shows the equivalence classes and the quasi-order relations is – at best – hard to attain. This is a first open question offered by PCA.

In three dimensions, a tentative graphical representation exists [7, Section 3], that combines a geometric representation of equivalence classes together with Hasse-like representation for each quasi-order. The few graphical representations of 4-lattices, for example in [8, figures 1 and 2] can only make one hope for another type of graphical representation, possibly *via* Virtual Reality, as is already proposed for FCA [16,17].

3 Loss of Duality and the FCA Trinity

3.1 Structural Equivalences

In the dyadic case, the formal context, the concept lattice and the set of implications are equivalent in the sense that they can be computed from one another. This is the reason why the FCA formalism is so useful for data analysis. In the multidimensional case, this equivalence softens, which introduces new challenges, constraints and open questions. Let us consider the two triadic contexts depicted in Fig. 3 in which the Greek letters form the first dimension (the objects), and their triadic concepts.

We observe that the concepts differ only on their first components. Hence, the second and third quasi-orders of both triadic concept lattices are isomorphic while the first quasi-orders are not. This is a most significant change from the bidimensional case: the knowledge of n-1 quasi-orders is not enough to know the last one. Let us call the last n-1 components of a concept its *feature* and the first component its *extent*, as usual. Then, given a set of concepts known only by their features, there are multiple non-isomorphic ways objects can belong to the extents. This clashes with the usual notions of subsumption.

Let us say that two concept n-lattices are equivalent if and only if their kth quasiorders are isomorphic for all $k \in \{2, \ldots, n\}$. We denote by $[\mathcal{L}]$ the equivalence class of the concept n-lattice \mathcal{L} . As mentioned in the previous section, implications in ncontexts can be used to construct (exactly) the features of the associated n-concepts. Hence, all the implication bases of all the n-contexts of the lattices in $[\mathcal{L}]$ allow for

$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	$\begin{array}{c ccccccccccccccccccccccccccccccccccc$
3 ×	3 ×
$\alpha \parallel \beta$	$\begin{array}{c ccccccccccccccccccccccccccccccccccc$
$(\emptyset, \{1, 2, 3\}, \{a, b, c\})$ $(\{\alpha, \beta\}, \emptyset, \{a, b, c\})$ $(\{\alpha, \beta\}, \{1, 2, 3\}, \emptyset)$ $(\{\alpha\}, \{1\}, \{a, b, c\})$ $(\{\alpha\}, \{1, 2, 3\}, \{a\})$ $(\{\alpha, \beta\}, \{1\}, \{a\})$ $(\{\beta\}, \{2\}, \{b\})$	$(\emptyset, \{1, 2, 3\}, \{a, b, c\})$ $(\{\alpha, \beta, \gamma\}, \emptyset, \{a, b, c\})$ $(\{\alpha, \beta, \gamma\}, \{1, 2, 3\}, \emptyset)$ $(\{\alpha\}, \{1\}, \{a, b, c\})$ $(\{\beta\}, \{1, 2, 3\}, \{a\})$ $(\{\alpha, \beta, \gamma\}, \{1\}, \{a\})$ $(\{\gamma\}, \{2\}, \{b\})$

Fig. 3: Two formal contexts $(\{\alpha, \beta, \gamma\}, \{1, 2, 3\}, \{a, b, c\}, \mathcal{R}_1)$ (left) and $(\{\alpha, \beta\}, \{1, 2, 3\}, \{a, b, c\}, \mathcal{R}_2)$ (right) and their associated 3-concepts.

the construction of the same features. These implications can, however, differ from n-context to n-context. For instance, the implication $\{(2,a)\} \to \{(1,b)\}$ holds in Fig. 3's first triadic context (as both crosses appear together in the description of the object α) but not in the second. From this, we deduce that some implications have no influence on the construction of the features. We thus propose to identify two types of implications:

- Structural implications that are used to construct the features of all the n-concepts of an n-context
- Contextual implications that are not structural but still hold in an n-context

Structural implications carry information about the features of n-concepts while contextual implications carry information about the distribution of the objects in the n-concepts. Thus, structural implications are common to all the n-contexts in an equivalence class $[\mathcal{L}]$ while contextual implications are not. In Fig. 3's contexts, the implications $\{(1,b),(1,c)\} \to \{(1,a)\}$ and $\{(2,a)\} \to \{(3,a)\}$ are structural while $\{(2,b)\} \to \{(1,a)\}$ is contextual in both contexts and $\{(1,b),(2,a)\} \to \{(3,a),(1,c)\}$ is contextual in the first context only. In [13] is explained in an overly formal way, for which the author is very sorry, that constructing the features of all the n-concepts of an n-context only requires implications between "boxes", *i.e.* implications of the form $\prod_{k=2}^n X_k \to \prod_{k=2}^n Y_k$ with $X_k \subseteq Y_k \subseteq \mathcal{S}_k$. The structural implications are all the implications entailed by these implications between "boxes".

In an equivalence class $[\mathcal{L}]$, there is one n-lattice/n-context that seems to be of particular interest as it minimises the number of contextual implications with a non-empty support, which could have some use in data mining. It is fairly easy to construct this n-context: if two n-concept features are such that their intersection is not the feature of an n-concept, then the two features must appear in the descriptions of the same objects. This defines equivalence classes of features. Then, each such equivalence class of features is used to describe a different object. Fig. 4 illustrates this. The rectangles $(\{1\}, \{a,b\})$ and $(\{1,2\}, \{a\})$ are features of triadic concepts while $(\{1\}, \{a\})$, their

intersection, is not so both rectangles have to appear in the description of the same object. The intersection $(\{3\},\{c\})$ of the rectangles $(\{2,3\},\{c\})$ and $(\{3\},\{b,c\})$ is the feature of the triadic concept $(\{\alpha,\beta\},\{3\},\{c\})$ so the three rectangles are put in the descriptions of different objects.

	$\begin{vmatrix} a & b & c \end{vmatrix}$	$\begin{vmatrix} a & b & c \end{vmatrix}$			a	b	c	$\mid a$	b	c	a	b	c	a	b	c
1	××			1	×	×										
2	×	×	\Longrightarrow	2	×								×			
3	×	× ×		3					X	\times			×			×
	α	β				α			β			γ			δ	

Fig. 4: Transformation of a triadic context in a way that minimises the number of contextual implications with a non-empty support without changing the equivalence class of the associated concept lattice.

3.2 A Generalisation of Boolean Concept Lattices

We believe that studying such equivalence classes of concept n-lattices, instead of individual n-context/n-lattice pairs, is the way to go. We suggest to start with classes that seem to be particularly interesting as they generalise Boolean concept lattices. Let $\mathcal{B}_{j_1,\ldots,j_{n-1}}$ be the class of concept n-lattices such that every possible feature on n-1 dimensions of size j_1,\ldots,j_{n-1} appear in an n-concept. For instance, Fig. 5 depicts the 3-context and 3-concepts of a member of $\mathcal{B}_{3,3}$. All rectangles in a 3×3 table appear as the feature of a 3-concept (the empty rectangle appears twice as $(\{1,2,3\},\emptyset)$ and $(\emptyset,\{a,b,c\})$.

The n-lattices in $\mathcal{B}_{j_1,\dots,j_{n-1}}$ contain $(\prod_{i=1}^{n-1} 2^{j_i} - 1) + n - 1$ concepts. They generalise the Boolean concept lattices \mathcal{B}_{j_1} as they are extremal in the sense that all possible features appear in concepts, they do not contain any structural implications (contextual implications may exist) and one of the corresponding n-contexts resembles the contranominal scale as, for each object, an element of a dimension is completely missing (see Fig 5). As such, it is always possible, in a context in which the number of objects is at least equal to the sum of the sizes of all the other dimensions, to produce such an extremal n-lattice. It appears that $\mathcal{B}_{2,2}$ can exist on 3 objects (Fig 6) while $\mathcal{B}_{3,3}$ cannot exist on 5 (see Fig. 7 for another such extremal 3-context on 6 objects). It is currently unknown whether $\mathcal{B}_{2,2}$ is an exception or other extremal n-lattices can exist on one less object. This problem of finding "maximally compact" n-contexts producing specific n-lattices seems relevant as it is tied to other potentially interesting problems such as finding the maximal number of n-concepts in an n-context of a given size or maximising the number of contextual implications.

Fig. 5: 3-context and 3-concepts of a concept 3-lattice in $\mathcal{B}_{3,3}$.

	a b		$\mid a \mid$	b	a	b				
1	××	T		×	×					
2	×		×	×		×				
	α	T	1	3	1	γ				

Fig. 6: A 3-context producing a member of $\mathcal{B}_{2,2}$ with only three objects.

														a		
														×		×
2	×	\times												×		
3	×	X	×	×	×	×	X	×	X	×	×	X	×	×		×
		α		β			γ		δ			ϵ			ζ	

Fig. 7: Another 3-context producing a member of $\mathcal{B}_{3,3}$ with six objects.

4 The Counter-Curse of Dimensionality

A natural question one can ask themself about such structures as lattices and n-lattices is their maximal size. As discussed in the previous subsection, n-contexts in which a dimension is as big as the sum of all the others can contain an extremal n-lattice, the size of which is known. The question remains open for all other n-contexts.

Given an n-context $(S_1, \ldots, S_n, \mathcal{R})$ with dimensions of equal size s, how many n-concepts might that context contain, at most? We call that number $f_n(s)$. In this section, we discuss this question, first in the bidimensional case where it is already elegantly solved, and then in the general case where it is not. The state of the art on this question is presented in Fig. 11, at the end of the section.

4.1 All is Clear in 2 Dimensions

It is well known that 2-lattices with 2^s elements – powerset lattices, or Boolean lattices – can be constructed from 2-contexts of size $s \times s$, (2-)contranominal scales. The 2-contranominal scale with s objects and s attributes is denoted $\mathbb{N}_2^c(s)$. An example of $\mathbb{N}_2^c(4)$ and its concept lattice is shown in Figure 8.

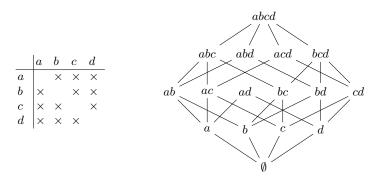


Fig. 8: The contranominal scale $\mathbb{N}_2^c(4)$ and its associated 2-lattice, a *tesseract* (only the extents are represented).

In [18], Albano and Chornomaz show that not only do 2-dimensional contranominal scales $\mathbb{N}_2^c(s)$ give rise to powerset lattices, but that they are the sole culprit in the large size a concept 2-lattice can have. A *large* 2-lattice has as standard context a context in which lurks a *large* contranominal scale. In their paper, Albano and Chornomaz quantify exactly what *large* means in both its uses.

4.2 General Case: What About Dimension n?

First, we can identify a naive upper bound on the maximal number of concept in a n-context of size $s \times \cdots \times s$. A n-concept is always uniquely described by (n-1) of its components, and as such $(2^s-1)^{n-1}+n-1$ is an upper bound for $f_n(s)$.

Before giving a lower bound for $f_n(s)$, it is necessary to recall some information on the construction of contexts.

The direct sum of two contexts $\mathcal{C}^1=(\mathcal{S}^1_1,\ldots,\mathcal{S}^1_n,\mathcal{R}^1)$ and $\mathcal{C}^2=(\mathcal{S}^2_1,\ldots,\mathcal{S}^2_n,\mathcal{R}^2)$ where the S^1_i,S^2_i are disjoint is the context $\mathcal{C}=(\mathcal{S}_1,\ldots,\mathcal{S}_d,\mathcal{R})$ constructed in the following way:

$$S_i = S_i^1 \cup S_i^2, \forall i \in \{1, \dots, n\}$$

and

$$\mathcal{R} = \mathcal{R}^1 \cup \mathcal{R}^2 \cup \{(x_1, \dots, x_d) \mid \exists i, j \in \{1, \dots, n\}, i \neq j \text{ such that } x_i \in \mathcal{S}_i^1, x_j \in \mathcal{S}_i^2\}.$$

The number of concepts in C is then the product of the number of concepts in C^1 and in C^2 . This construction is illustrated in two and three dimensions in Fig. 9.

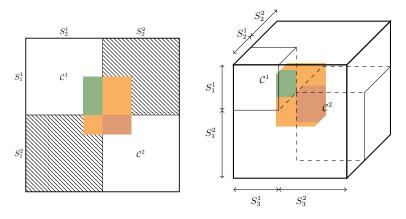


Fig. 9: Illustration of a combination of concepts under a direct sum of two contexts \mathcal{C}^1 and \mathcal{C}^2 in two dimensions (left) and three dimensions (right). Any two concepts of \mathcal{C}^1 and \mathcal{C}^2 – a maroon and a green one – can be extended into a new concept of \mathcal{C} – through the gray areas –, thus multiplying their number of concepts.

An empty one-cell n-context $\mathcal{C}_{ell}=(1,\ldots,1,\emptyset)$ gives rise to exactly n n-concepts. Adding together s such empty n-contexts gives rise to the n-contranominal scale $\mathbb{N}_n^c(s)$, that is an n-context of the form $(\mathcal{S},\ldots,\mathcal{S},R)$ with $R=\prod \mathcal{S}\setminus\{(x,\ldots,x)\mid x\in\mathcal{S}\}$ that has $n^{|\mathcal{S}|}$ concepts. This is a lower bound for $f_n(s)$.

Those two bounds are identical in two dimensions, but are increasingly distant with higher dimensions.

4.3 In Three Dimensions

There is no clear answer yet for the value of $f_3(s)$. The two bounds discussed in the previous section amount to looking for $f_3(s)$ between 3^s and 4^s .

The well known triadic contranominal scale and the associated powerset trilattice were studied by Biedermann [19]. Powerset trilattices are presented as a triadic generalisation of powerset lattices, can be constructed from 3-contranominal scales $\mathbb{N}_3^c(s) = (\mathcal{S}, \mathcal{S}, \neq)$ and have 3^s elements. No explicit claim that this class of 3-lattices is extremal is made, other than the use of the name powerset 4 .

Although the 3-contexts responsible for powerset trilattices are good generalisations of contranominal scales, the 3-lattices themselves are not that good of a generalisation of the extremal 2-dimensional case, since they are not extremal w.r.t. the size. This is discussed in Section 3.2.

In [20], the authors present bounds for the maximum number of 3-concepts in a 3-context of size $s \times s \times s$. Using a small example of a $5 \times 5 \times 5$ 3-context that contains more 3-concepts than a 3-contranominal scale and the direct sum of contexts, they provide a construction of arbitrarily large contexts with 3.359^s concepts. Using a measure and conquer approach, they also provide an upper bound of 3.384^s for 3-contexts of that size. The improvement from the naive bounds is represented in Figure 10.



Fig. 10: While either naive bounds would be a satisfying answer for the maximum number of 3-concepts in a 3-context, the real answer seems to be in a small corner near 3.36^s .

4.4 In Four Dimensions

Using the context C_{rook} depicted in Table 1, we can build arbitrarily large 4-contexts that have $c4.82^s$ concepts, with c a constant.

Let s be a integer greater than 3. Then, there exists two integers k and r such that s=3k+r, with $r\in[0,2]$. To build a 4-context $\mathcal C$ of size $s\times s\times s\times s$, we add k versions of $\mathcal C_{rook}$ with the aforementioned procedure, and then add a 4-dimensional contranominal scale of size r. The resulting context has $4^r\times 4.82^{s-r}=(\frac{4}{4.82})^r\times 4.82^s$ concepts. By setting $c=(\frac{4}{4.82})^2$, we have that $f_4(s)\geq c4.82^s$.

⁴ In [19], the name powerset trilattice is justified by some algebraic properties of the powerset trilattice.

⁵ A data file containing the complement of this contexts (its holes) is available at http://giacomo.kahn.science/resources/Crook.dat

	1	2	3	1	2	3	1	2	3	1	2	3	1	2	3	1	2	3	1	2	3	1	2	3	1	2	3
1		X	×	×		X	×	X		×		X	×	×			X	×	×	×			X	×	×		X
2	×		X	×	X			X	×	×	×			X	X	X		\times		X	X	X		X	×	X	
3	×	×			X	X	×		Χ		×	X	×		X	×	×		×		X	×	×			X	×
	1 2 3						1		2 3						1			2		3							
1									2 3																		

Table 1: This is a $3 \times 3 \times 3 \times 3$ 4-context that we call \mathcal{C}_{rook} . Empty cells are coloured blue for readability. This 4-context has 112 4-concepts. This amounts to around 4.82^3 concepts. This is, just as in the three dimensional case, a solution to a multidimensional chess rook problem.

4.5 Summary

Figure 11 summarises the known results related to the size of n-lattices, taking into account the contexts from Section 3.2 and Section 4.

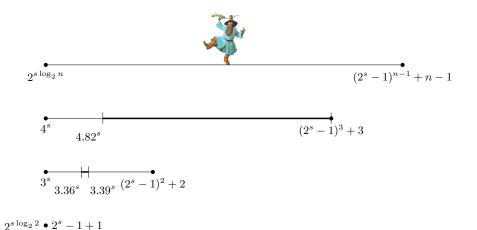


Fig. 11: "Old Tom Bombadil is a merry fellow! Bright Blue his jacket is, and his boots are yellow!". In the general case, we do not really know where the maximum number of n-concepts lives. This uncertainty is represented here by a dancing Tom Bombadil. This figures summarizes the known bounds for $f_n(s)$. In two dimensions the bound is well known and reachable, in three dimension the gap is small.

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