

SIGNAL ANALYSIS: AN APPLICATION OF C-MATRIX

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A Calculus of Partition

We propose to present some application to Pattern recognition of a "mathematical game" started some years ago (1), which was named "C-calculus" for reasons which will be reminded in the sequel. We wish to state forthwith that it is simpler in principle than ordinary arithmetics; various fields can be envisaged in which it might prove to use: e. g. manifold topology, integration theory, fuzzy sets (where it might provide a natural tool for numerical computation), measure theory in physics, data bases structures, neural models, etc.

We begin therefore by reminding the game with which it all started. Take any integer positive numbers, and apply to them the rules of arithmetics, with the restrictions that only the direct operations, sum and multiplication, be allowed, the inverses, subtraction and division, forbidden; define furthermore the sum and the product of any two digit as follows

$$1) \quad a + b = \max (a, b) \\ \quad \quad a \times b = \min (a, b).$$

It would be an easy matter to demonstrate that, provided the "single digit operation" 1) are meaningful, one can operate in the same way on objects (subtraction and division being of course barred), such as vectors, matrices, etc., obtaining additivity and commutativity whenever they hold in arithmetics.

These "numbers" or "strings of digits", with the operation 1), form clearly a commutative semi-ring. As in arithmetics each "digit" plays two different roles: one intrinsic to it (cardinality), the other (position) relative to the string in which it belongs. The next remark is that standard set theory treats only intrinsic properties of sets. If in 1) we interpret

+ as union "U" and \times as intersection " \cap ", we can immediately transport all that was said thus far "strings of sets", or "composite sets", "C-set" for short.

Operating on C-set as before, with \cup and \cap in place of + and \times in 1) (a, b denote now the "simple" sets of which C-sets are strings), one has "C-calculus": a commutative semi-ring which permits, from same given C-sets, to generate any number.

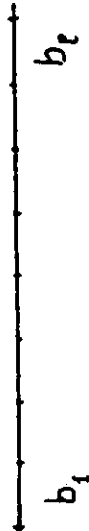
An example of C-operation of special relevance for our present purpose is the following:

Consider a segment S partitioned in segments a_1, a_2, \dots, a_k ; this partition $\Lambda = a_1, a_2, \dots, a_k$



Consider now the same segment partitioned in a different way

$$B = b_1, b_2, \dots, b_l$$



Consider now Λ and B as C-sets: the element of each partition, or string, are "simple" sets; C-multiplication of Λ and B gives

$$\Lambda \times B = \beta \times \alpha = a_1 a_2 \dots a_k \times b_1 b_2 \dots b_l = \\ = c_1 c_2 \dots c_p = C$$

and it is immediate to verify that the simple sets of the product are obtained, in order, by joining on the segments the terminal points of both partitions Λ and B .

The C-product of two partitions gives thus the refinement of one by the other: C-calculus is the natural way of composing partitions, or coverings. In fact, the same property holds true in any number of dimensions (2). This is the key property of C-calculus as regards its application to pattern recognition.

Under "suitable" circumstances (to be defined explicitly in the sequel) this procedure can be carried through to the extreme limit of perfect reconstruction of the original picture (1)

digitized at the finest possible level: e. g. with a $2^{10} \times 2^{10}$ grid for the original, it may be reconstructed by covering it stepwise with, say, a $2^k \times 2^k$ grid. During this process many things which one does with specific techniques, such as contour extraction, contrast enhancement, feature extraction, etc., can be performed by interpolating in it steps with "answer" questions of this sort and become part of the algorithm.

But the application of C-calculus will often fail: the original image may not be thus reconstructed. There is an element to be considered, which was before ignored through the adjective "suitable": the size of the window. It is a feature of our approach that the critical size, below which total reconstruction of the picture is impossible, is determined by the structure of the picture itself, and is not a matter of guesswork or trial and error.

One can arrange readings, and ways of analyzing them, from grids having sizes appropriate to constitute in fact filters that see some wanted features and are blind to others. (3, 4, 5,)

C-MATRIX

1.- Consider a pattern F in any number N of dimensions (for the sake of illustration we restrict here N = one or two), and consider only one of some K " attributes " of interest, e. g. the level of greyness (which we may suppose now digitized). An i -dimensional grid of (square) windows generates, as we have seen, a C-set in $N + 1$ ($K = 1$) space: C-multiplication of all C-sets obtained by displacing (according to some rule) the grid over the pattern yields, after a suitable number of iterations, a pattern F, which coincide with F if condition 2) is respected, differ from F otherwise. This process we have called C-filter; as such, it might deserve per se mathematical investigation. Our interest here is rather with concrete ways of exploiting the typical new feature of C-calculus, that of permitting either " precise " measurement of a whole through serial composition (C-multiplication of " coarser " partial parallel measurements) (each, a C-set from the grid), or " filtering operations " " Precise " and " coarse " are to be understood as in physics. We have found profitable, for this purpose, to introduce a mathematical object which (for any i , $K = 1$) is always 2-dimensional: the C-matrix.

2.- We define the C-matrix. Its element C_{ij}^k has as row label i (= 2, 3, ...) the linear size of the window w_i of the scan-

ning grid G_i (C-calculus will be more useful the larger can be kept the minimum h needed for a given analysis); the column label k (= 0, 1, 2, ...) denotes the possible values of the dynamic: $K = h - m$ as read through w_i . The grid G_i scans the pattern according to some criterion (in two dimensions, the best has proved to run down the main diagonal) by one step at a time (" 1 " means the original digitizing window, if any, or simply the " resolving power " of the system); G_i has thus h (= h , square grid h, linear) scanning positions; any value k of the dynamic is registered (= 0, 1, ...) times in the windows, for each position $i = 1, \dots, h$ of the grid G_i . In conclusion

The building of C-matrix seems, from the definition just given, a much more imposing task than actually is. In fact, the operations that lead to C_{ij}^k are performed in parallel for $k = 0, 1, \dots$ so that the C-matrix is built one row at a time, or in one piece with suitable hardware; also, the procedure has to be started from the bottom row, i. e. largest w_i , and it becomes an obvious matter to implement the algorithm with devices that suppress the need to explore, moving towards smaller i , territories where, e. g. $k = 0$ (uniform greyness) for a larger i . Nor need one proceed through the full sequence $h_{max}, h_{max} - 1, \dots, 3, 2$: large jumps may be made.

An element C_{ij}^k of C-matrix tell us thus: if vanishes, that the value j of the dynamic of the pattern can never be seen through windows of size w_i ; if not, how many times that dynamics seen by no matter which window w_i .

It is easy then to read from the C-matrix many features of the pattern; e. g. If only $C_{ij}^0 = 0$, all i , the pattern is a plateau, of constant grayness throughout; the max i for which $C_{ij}^0 = 0$ is the width of the largest plateau; a highest-slope of C_{ij}^1 -dimensional signals shows as the sequel $C_{ij}^1 = 0, j_{max}$, for each i .

3.- Contour extraction

This is a relevant step in every problem of P.D., with a wide literature and a large number of techniques: variations of gradients and threshold, thinning, joining or separating of broken or interpenetrating pieces, template matching, etc. (6, 7, 8,). We can extract contours with the following rule. Consider

the first among the columns of the C-matrix, let it be the column h , in which the elements c_{ij} go through a maximum, at some value i . Take the corresponding window w_i as an element of the grid of a C-filter, given by C-multiplication of C-sets whose simple sets vanish if smaller than the (hyper) rectangles $w_i \times h$ (i. e. h is a threshold on dynamic). Fig's 1, 2 illustrate the procedure.

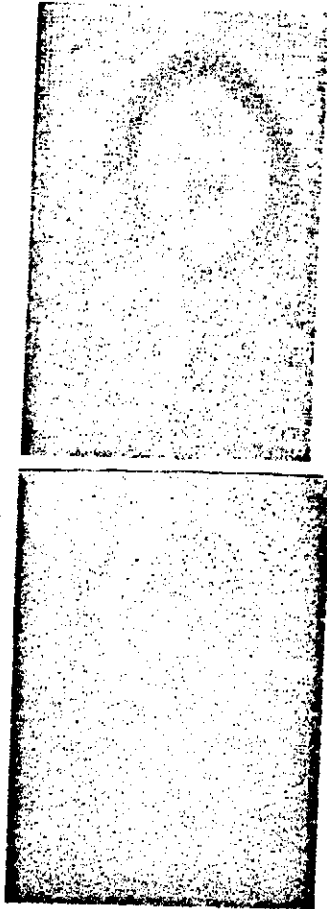


fig. 1

I. Matrix

fig. 2

O. Matrix

Comparison with other algorithms (10, 6, 11) was made; it appears that thresholds come out of the C-matrix and do not have to be guessed, and that computation time is here less than with RAG, LAG, PT.

TEXTURES.

The analysis of textures, both per se and as backgrounds, is almost a science within a science; see, e.g. Gibson (9), Koehler (10), Haralik (11).

Our views on this subject do not belong in this short summary of our work; more than "percepta" are "phenomena", in the sense of Kant, quantum mechanics or the Vedas, to which the "noumenon" and the "observer" equally concur. We wish only to report here, in conclusion, that C-calculus has proved especially "natural" in this context, leading to ready and elementary classification, analysis and discrimination.

We refer the reader to previous works (16), where some examples are reported. It is the "chessboard and saucer" game we were mentioning earlier, which is easily implemented through C-calculus into algorithmic simulations which may be

rather close in principle to the actual operation of neuronal tissues. (12, 13)

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