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PERCEPTION, PICTURE PROCESSING AND COMPUTERS

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INTRODUCTION

The ability to interpret and respond to the significant content of pictures is shared by a large proportion of the animal kingdom and a small but growing number of machines. The machines (usually digital computers) will classify simple shapes and printed letters and digits represented (by means of a suitable television scanner) as a matrix of 1s and 0s. Animals, and specifically man, will do this and more; that is, not only will a human subject correctly recognise letters written badly but also comment, 'That's a funny way to write a two', or 'It's a good portrait, but the eyes are a little too far apart'. No machine can as yet approach this sort of performance. We may well ask, therefore, what if anything we can learn from a study of 'machine perception' which could conceivably help us to understand human perception. The answer lies in the fact that any realistic account of the mechanism underlying perceptual or any other human skills will be complex. The virtue of a computer lies in its ability to capture in a definite form processes of indefinite complexity and subtlety, and moreover, to permit an *evaluation* of the efficacy of the proposed description by trying it out on the actual task.

This paper will be concerned with a description of a new method of interpreting pictures, now under development. The method arises directly out of earlier attempts (Clowes & Parks 1960, Clowes 1962) to develop letter recognition machines for commercial use. However, it now appears that the approach is capable of a wider interpretation in terms both of psychological observations and recent physiological evidence.

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HIERARCHIES OF DESCRIPTION

Psychologists concerned with analysing complex behaviour often have to select an appropriate level of description. For example, driving a motor car can be characterised in terms such as 'overtaking', 'giving way', 'filtering', or at a more detailed level in terms of 'changing from 2nd to 3rd gear', 'turning the wheel anti-clockwise', 'depressing the clutch pedal' or yet more detailed still in terms of the positions of the limbs as a function of time. Choosing the correct level of description might be vital to the assessment of the relative skills of two drivers, or the design of two different types of car. On the other hand perhaps all these levels of description should be employed simultaneously—if so, how are they related? This question forms the basis of a book *Plans and the structure of behaviour* (Miller, Galanter & Pribram 1960) in which it is shown that the problem arises not only for manual skills, but also in many cognitive processes and especially in language. It is in this latter case that perhaps the clearest expression of a hierarchy of description is evident—phonemes, syllables, words, phrases, clauses, sentences . . .—and it is also the area in which the most progress has been made in answering these questions.

LINGUISTIC GRAMMARS

The problem is to formulate a reasonably precise set of rules by means of which we can take an utterance, e.g., 'the boy kicked the red ball' and, by

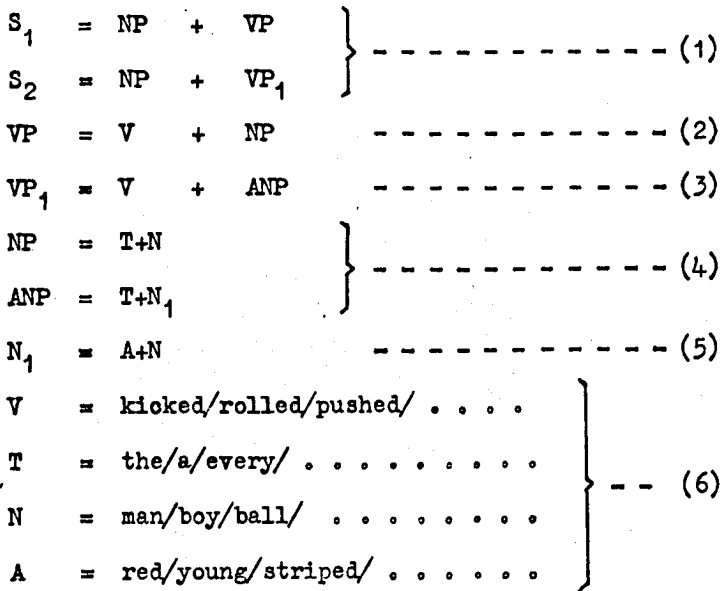


FIG. 1. A set of ordered substitution rules specifying the grammar of a simple sentence.

applying the rules in the prescribed manner, can obtain a description of this utterance at a number of different levels.

Let us confine our attention to three levels: words, phrases, sentences. Fig. 1 shows a set of rules sub-divided into numbered groups (1, 2, 3, . . . , 6).

We can regard each rule as defining a permissible substitution. Thus group 6 rules state that we may replace any of the words red, young, striped (i.e., any adjective) by the symbol A, and similarly for verbs, nouns and articles. So that the sentence

the boy kicked the red ball

becomes by application of group 6

(6) T N V T A N

Group 5 states that a pair of symbols A, N appearing *contiguously and in that order* can be replaced by N_1 so that the string becomes

(5) T N V T N_1

and so by application of successive rules, in descending order of their group number, we obtain

(4) NP V ANP

(3) NP VP_1

(2) NP VP_1

(1) S_2

If we now combine these different strings of symbols into one diagram (Fig. 2) we have a description of the utterance at a number of levels in a fashion which indicates the relation between the levels.

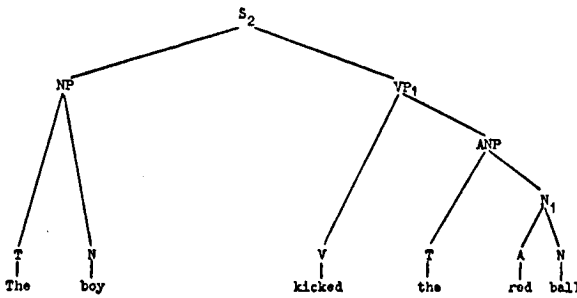


FIG. 2. The structural analysis of a simple sentence as obtained by applying the rules of FIG. 1.

The rules may be used to generate or analyse sentences. A typical generative sequence starting from the schema S_2 would be $S_2 \rightarrow NP VP_1 \rightarrow T N VP_1 \rightarrow the N VP_1 \rightarrow the boy VP_1 \rightarrow the boy V ANP \rightarrow \dots \rightarrow the boy kicked the red ball$. The generative path can be diagrammed in tree form as in Fig. 2.

This diagram—a so-called *tree*—indicates not only that the utterance is a sentence, but also what type of sentence it is (S_2 rather than S_1), as well as indicating what features (i.e., phrases like ANP, etc.) it had which caused

the label S_2 to be applicable. Let me hasten to add at this point that this is not presented as a serious piece of linguistics (there are grounds for believing it to be incorrect in several ways) but merely as an introduction to the notion of a set of rules (the grammar) and a procedure for applying them (a parsing algorithm) by which we can rigorously derive a *structural description* of a complex object like a sentence. A mechanism or mechanisms of this type undoubtedly underlie our ability to understand sentences.

PICTURE PROCESSING

Comparable mechanisms for processing pictures have not appeared to any significant extent either in Artificial Intelligence or in Psychology. Instead the emphasis has been on classifying pictures, i.e., upon the equivalent of deciding the type (S_1 or S_2) of a sentence. The possibility that a structural description of a pictorial object is necessary has only recently emerged in Artificial Intelligence (Kirsch 1964). It appears to be crucial to the automation of many picture processing tasks (e.g., interpreting bubble-chamber photographs, recognising fingerprints). That perceptual behaviour involves description as well as classification could be supported by innumerable examples beyond the two already quoted in the introduction to this paper. However, it seems to have received scant attention in recent psychological literature. This may well be because overtly descriptive (rather than classificatory) behaviour has a large verbal element. There is a strong temptation to avoid this 'complication' by designing experiments which merely require simple YES/NO responses and which have, therefore, no significant verbal component.

It follows from the foregoing remarks that our own objective is the development of a formal (i.e., rigorous) theory of picture processing which aims to provide a structural description of pictorial objects as well as a classification of them (wherever appropriate). The particular system I shall describe is by no means the only possible approach, both Ledley (1964) and Narasimhan (1964) have proposed alternatives. I shall illustrate the technique in terms of a limited class of pictures—the handwritten numerals. There are several reasons for choosing this class. (1) They are (hopefully) relatively simple objects; (2) at least two levels of description naturally occur, e.g., the symbol 7 is readily described as 'a horizontal stroke above a diagonal stroke . . .'; (3) psycholinguistic skills, e.g., reading (with which the Unit† is primarily concerned), involve pictorial objects of this type.

A PICTURE GRAMMAR FOR HANDWRITTEN NUMERALS: NOTATION

The substitution rules in the linguistic grammar (Fig. 1) contain on their right-hand side (typically) two elements, e.g., T, N or NP, VP, linked by a relational operator '+' meaning 'followed by'. This operator expresses the notion of contiguity as it arises in a *string* of symbols; contiguity in a

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geometrical sense is the more general relationship 'next to'. Thus a simple pictorial property like 'Edge' (i.e., boundary between black and white regions of the picture) might be defined as 'black area next to white area'. Specifically we might define a North Edge as 'white area *above* black area', which can be expressed in a rigorous (computable) form with the aid of a suitable notation. For example, in a pictorial notation the definition of North Edge for a picture composed only of black or white elements might take the form illustrated in Fig. 3. This asserts that the point arrowed

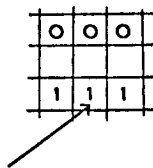


FIG. 3.

will be said to have the property N EDGE if it and the points to the left and the right of it are black (i.e., *True*) and the three points denoted '0' are white (i.e., *False*). If we introduce a Cartesian co-ordinate system we can write this definition as a Boolean expression:

$$\begin{aligned} \text{N EDGE } (x, y) = & \text{ PICTURE } (x-1, y) \ \& \ \text{ PICTURE } (x, y) \\ & \ \& \ \text{ PICTURE } (x+1, y) \ \& \ \sim \text{ PICTURE } (x-1, y+2) \\ & \ \& \ \sim \text{ PICTURE } (x, y+2) \ \& \ \sim \text{ PICTURE } (x+1, y+2) \end{aligned} \quad (1)$$

where (x, y) is the 'arrowed' point, a black point has the value *True* and a white point the value *False*.

The assignment of a label (e.g., N EDGE) to a picture point involves a decision based upon the state of a number of neighbouring points. Fig. 3 is a picture of the decision criterion. The values of only 6 specified points are considered. All other points are ignored, for the purpose of assigning this label to the arrowed point.

IMPLEMENTATION

This notation is used in the picture analysis program with some small differences for reasons of economy and ease of programming. Fig. 4 shows a system of such definitions as they are input to the computer. The first two lines specify various program options relating to print out, etc., and also define the size (48×48) of the input picture. The third line is the definition for N EDGE given above (1), but specified in Reverse Polish (i.e., parenthesis free) form.† Repetitions of the label PICTURE have been suppressed and the label N EDGE appears at the end of the definition rather than preceding it. The x, y appearing on both sides of the expressions (1) above have also been omitted, it being understood that this rule is applied to the picture (Fig. 5(a)) for all values of (x, y) so as to derive another picture called (Z) N EDGE (Fig. 5(b)). The axes have been rotated so that x is measured down the

† The Reverse Polish form of the Boolean expression $(A \ \& \ B) \vee (\sim C \ \& \ D)$ is $A, B, \ \&, C, \ \sim, D, \ \&, \vee$.

FIG. 4. A partial 'picture grammar' for handwritten numerals. Each group of rules is preceded by the control instruction + CODE>NNL + 02BY02.

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+LINEYX+48BY48
+CODE>NNL+02BY02
ZPICTURE.(00,00) NOT .(00,01) NOT AND .(00,02) NOT AND .(02,00) AND .(02,01) AND .(02,02) AND
Z(02,01)Z# EDGEX048,48 .(00,01) NOT AND .(00,02) NOT AND .(02,00) AND .(02,01) AND .(02,02) AND
ZPICTURE.(00,00) NOT .(00,01) NOT AND .(00,02) NOT AND .(02,00) AND .(02,01) AND .(02,02) AND
Z(02,01)Z# BLOBX000,72 .(00,01) NOT AND .(00,02) NOT AND .(02,00) AND .(02,01) NOT AND .(01,02) NOT AND
ZPICTURE.(00,00).(01,00) AND .(02,00) AND .(02,01) NOT AND .(02,02) NOT AND .(02,00) NOT AND
0(01,00)Z# EDGEX024,00 .(00,02) AND .(00,03) NOT AND .(02,00) AND .(02,01) NOT AND .(02,02) NOT AND
0(01,00)Z# BLOBX000,72 .(00,02) AND .(00,03) NOT AND .(02,00) AND .(02,01) NOT AND .(02,02) NOT AND
ZPICTURE.(00,00).(00,01) AND .(00,02) AND .(02,00) NOT AND .(02,01) NOT AND .(02,02) NOT AND
Z(00,01)Z# EDGEX024,48 .(00,01) AND .(00,02) AND .(02,00) NOT AND .(02,01) NOT AND .(02,02) NOT AND
ZPICTURE.(00,00).(00,01) AND .(00,02) AND .(02,00) NOT AND .(02,01) NOT AND .(02,02) NOT AND
Z(00,01)Z# BLOBX000,72 .(01,00) NOT AND .(01,00) NOT AND .(02,00) NOT AND .(02,01) NOT AND .(02,02) NOT AND
ZPICTURE.(00,00) NOT .(01,00) NOT AND .(01,00) NOT AND .(02,00) NOT AND .(02,01) NOT AND .(02,02) NOT AND
Z(01,00)Z# EDGEX048,00 .(01,00) NOT AND .(01,00) NOT AND .(02,00) NOT AND .(02,01) NOT AND .(02,02) NOT AND
ZPICTURE.(00,00,72 .(00,02) AND .(02,00) AND .(02,01) NOT AND .(02,02) NOT AND .(02,00) NOT AND
Z(01,01)Z# BLOBX000,72 .(00,02) AND .(02,00) AND .(02,01) NOT AND .(02,02) NOT AND .(02,00) NOT AND
ZPICTURE.(00,02).(01,01) AND .(02,00) AND .(02,01) NOT AND .(02,02) NOT AND .(02,00) NOT AND
Z(01,01)Z# EDGEX024,24 .(02,00) AND .(02,01) NOT AND .(02,02) NOT AND .(02,00) NOT AND .(02,01) NOT AND
ZPICTURE.(00,02).(01,01) AND .(02,00) AND .(02,01) NOT AND .(02,02) NOT AND .(02,00) NOT AND
Z(01,01)Z# BLOBX000,72 .(00,03) AND .(02,00) AND .(02,01) NOT AND .(02,02) NOT AND .(02,00) NOT AND
ZPICTURE.(00,02).(01,03) AND .(02,00) AND .(02,01) NOT AND .(02,02) NOT AND .(02,00) NOT AND
Z(01,03)Z#W EDGEX024,72 .(00,03) AND .(02,00) AND .(02,01) NOT AND .(02,02) NOT AND .(02,00) NOT AND
ZPICTURE.(00,02).(01,03) AND .(02,00) AND .(02,01) NOT AND .(02,02) NOT AND .(02,00) NOT AND
Z(01,03)Z# BLOBX000,72 .(01,02) NOT AND .(01,02) NOT AND .(02,00) AND .(02,01) NOT AND .(02,02) NOT AND
ZPICTURE.(00,03) NOT .(01,02) NOT AND .(01,02) NOT AND .(02,00) AND .(02,01) NOT AND .(02,02) NOT AND
Z(03,03)Z#W EDGEX048,24 .(01,02) NOT AND .(01,02) NOT AND .(02,00) AND .(02,01) NOT AND .(02,02) NOT AND
ZPICTURE.(00,03) NOT .(01,02) NOT AND .(01,02) NOT AND .(02,00) AND .(02,01) NOT AND .(02,02) NOT AND
Z(03,03)Z# BLOBX000,72 .(01,02) NOT AND .(01,02) NOT AND .(02,00) AND .(02,01) NOT AND .(02,02) NOT AND
ZPICTURE.(00,01) NOT .(01,02) NOT AND .(01,02) NOT AND .(02,00) AND .(02,01) NOT AND .(02,02) NOT AND
Z(03,01)Z#W EDGEX048,72 .(01,02) NOT AND .(01,02) NOT AND .(02,00) AND .(02,01) NOT AND .(02,02) NOT AND
ZPICTURE.(00,01) NOT .(01,02) NOT AND .(01,02) NOT AND .(02,00) AND .(02,01) NOT AND .(02,02) NOT AND
Z(03,01)Z# BLOBX000,72
    
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Fig. 4. (continued)

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*CODEHNL*02BY02
Z90 LINE.(00,00).(01,00) OR ZNW END.(00,00)ZN END.(00,00) OR ZNE END.(00,00) OR AND ZN LIMX072,00
Z00 LINE.(00,00).(00,01) OR ZNE END.(00,01)ZE END.(00,01) OR ZSE END.(00,01) OR AND Z(00,01)ZE LIM
X072,12
Z90 LINE.(00,00).(01,00) OR ZSE END.(01,00)ZE END.(01,00) OR ZSW END.(01,00) OR AND Z(01,00)ZE LIM
X072,24
Z00 LINE.(00,00).(00,01) OR ZSW END.(00,00)ZN END.(00,00) OR ZNW END.(00,00) OR AND ZN LIMX072,36
Z00 LINE.(00,01)Z45 LINE.(00,00).(01,00) OR AND Z135LINE.(00,02).(01,02) OR AND Z(00,01)ZN CURVX072,48
Z90 LINE.(01,01)Z135LINE.(00,00).(00,01) OR AND Z45 LINE.(02,00).(02,01) OR AND Z(01,01)ZE CURVX072,60
Z00 LINE.(01,01)Z135LINE.(00,00).(01,00) OR AND Z45 LINE.(00,02).(01,02) OR AND Z(01,01)ZE CURVX072,72
Z90 LINE.(01,00)Z45 LINE.(00,00).(01,00) OR AND Z135LINE.(02,00).(02,01) OR AND Z(01,00)ZE CURVX072,84
ZN EDGE.(00,00)ZE EDGE.(00,00).(01,00) OR AND Z00 LINEX060,00
ZNW EDGE.(00,00)ZE EDGE.(00,00).(01,01) OR AND Z45 LINEX060,72
ZW EDGE.(00,00)ZE EDGE.(00,00).(00,01) OR AND Z90 LINEX060,48
ZNE EDGE.(00,00)ZN EDGE.(00,01).(01,00) OR AND Z135LINEX060,24
Z00 LINE.(02,00).(01,01).(00,02) AND Z(00,01)Z00 LINEX060,00
Z45 LINE.(02,00).(01,00).(02,00) AND Z(01,01)Z45 LINEX060,72
Z90 LINE.(00,00).(01,00).(02,00) AND Z(01,00)Z90 LINEX060,48
Z135LINE.(00,00).(01,01).(02,02) AND Z(01,01)Z135LINEX060,24

*CODEHNL*02BY02
*(00,00)ZZERO X180,00
*(00,00)ZONE X180,09
Z45 LINE.(02,01)Z5 LIMX.(00,00).(01,00) OR AND ZN CURV.(00,01)ZE' CURV.(01,02) OR AND ZE LIMX.(03,01)
*(03,02) OR AND Z(02,01)ZTW0 X180,18
*(00,00)ZTHREE X180,27
*(00,00)ZFOUR X180,36
*(00,00)ZFIVE X180,45
*(00,00)ZSIX X180,54
*(00,00)ZSEVEN X180,63
*(00,00)ZEIGHT X180,72
*(00,00)ZNINE X180,81

```



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A LOST 2      -012345678911234567892123456789312345678941234567895123456789612345678971234567898123456789912345-
Z PICTURE
000 000=
001 001=
002 002=
003 003=
004 004=
005 005=
006 006=
007 007=
008 008=
009 009=
010 010=
011 011=
012 012=
013 013=
014 014=
015 015=
016 016=
017 017=
018 018=
019 019=
020 020=
021 021=
022 022=
023 023=
024 024=
025 025=
026 026=
027 027=
028 028=
029 029=
030 030=
031 031=
032 032=
033 033=
034 034=
035 035=
036 036=
037 037=
038 038=
039 039=
040 040=
041 041=
042 042=
043 043=
044 044=
045 045=
046 046=
047 047=
-012345678911234567892123456789312345678941234567895123456789612345678971234567898123456789912345-

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456701
34567012
12345670123
012345670123
701234 123
670123 123
67012 123
6701 0123
670 0123
670 67012
670 4567012
456701
12345670
123456
7012345
70123
7012
67012
6701
6701234
567012345
567012345
45670123456
456701234567012
456701234567 12
56701 456701
567 5670
6

```

Fig. 5(a). The picture to be analysed. The origin of coordinates is the cell immediately beneath the Z of Z PICTURE. The picture is defined to be only 48 x 48 elements hence in the direction (down the page) we can only print lines 0 to 47. Each 'black' element is represented internally as a binary digit. For display purposes, however, the 'black' elements are numbered (on a scale of 8) in the horizontal direction.

A LOST 2 -012345678911234567892123456789312345678941234567895123456789612345678971234567898123456789912345-

	Z	8LOB		ZS	EDGE	ZSE	EDGE	ZSN	EDGE
000 000-									
001 001-									
002 002-									
003 003-		234							
004 004-		12345							
005 005-		7012345							
006 006-		701 45							
007 007-		70 45							
008 008-		70 2345							
009 009-		1234							
010 010-		0123							
011 011-		701							
012 012-		70							
013 013-		7 1							
014 014-		6 12							
015 015-		6 1234							
016 016-		6701234							
017 017-		7 23							
018 018-									
019 019-									
020 020-									
021 021-									
022 022-									
023 023-									
000 024-									
001 025-									
002 026-									
003 027-									
004 028-									
005 029-									
006 030-									
007 031-									
008 032-									
009 033-									
010 034-									
011 035-									

Fig. 5(b). The outcome of the first group of rules when applied to Fig. 5(a). Note that the 'origin cell' of the array ZN EDGE is at the absolute location x=48, y=48 as stated in the picture grammar.

PATTERN RECOGNITION

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A LOST 2      -012345678911234567892123456789312345678941234567895123456789612345678971234567898123456789912345-
000 036-
001 037-      6
002 038-      6
003 039-      0 2
004 040-      12
005 041-
006 042-
007 043-      4
008 044-
009 045-
010 046-
011 047-

ZE EDGE ZSE EDGE ZS EDGE ZSW EDGE ZV EDGE ZNW EDGE ZN EDGE ZNE EDGE ZNE EDGE
      6      0      3      3      3
      0      2      3      3      3
      12
      4
      5

000 048-
001 049-
002 050-
003 051-
004 052-
005 053-
006 054-
007 055-      6
008 056-
009 057-
010 058-
011 059-

ZE END ZW END ZSE END ZSW END ZNW END ZS END ZN END ZNE END ZNE END
      5      2      6      3      3
      5      6      3      3      3
      70
      12
      34
      1

000 060-
001 061-      5
002 062-      5
003 063-      5
004 064-
005 065-      34
006 066-
007 067-      6
008 068-
009 069-
010 070-
011 071-

Z00 LINE Z103LINE Z20 LINE Z45 LINE
      5      2      6      70
      5      3 6      12
      34
      1

-0123456789112345678921234567893123456789412345678951234567896123456789712345678981234567899123456-
A LOST 2
000 000
001 001
002 002
003 003
004 004
005 005
006 006
007 007
008 008
009 009
010 010
011 011

-36 000
-37 001
-38 002
-39 003
-40 004
-41 005
-42 006
-43 007
-44 008
-45 009
-46 010
-47 011

-48 000
-49 001
-50 002
-51 003
-52 004
-53 005
-54 006
-55 007
-56 008
-57 009
-58 010
-59 011

-60 000
-61 001
-62 002
-63 003
-64 004
-65 005
-66 006
-67 007
-68 008
-69 009
-70 010
-71 011
    
```

Fig. 5(c). The outcome of the second group of rules applied to the contents of Fig. 5(b).

page and y across it, following the numbering appearing in the picture frame. Following the label (Z) N EDGE a further pair of coordinates (048, 48) specify the location in computer memory at which the resultant picture (Z) N EDGE is to be stored. The first group of definitions in the grammar expresses eight EDGE definitions corresponding to eight points of the compass, and a further definition (Z) BLOB which merely collects together the results of all the EDGE definitions in a single picture called (Z) BLOB. The next set of rules in the grammar contains concatenations of EDGES to define LINES having four orientations 00, 45, 90, 135 and line ENDS with eight orientations. These in turn are concatenated to form CURVES (combinations of appropriately oriented lines), LIMBS (combinations of lines and ends) and LINES (combinations of lines). Each set of rules in the picture grammar is separated by a line beginning +CODE. . . .

The action of the program is to apply each set of rules to the outcome of the previous set starting with a picture stored in computer memory (cf. the linguistic grammar). The performance of the program is monitored by having it print out those portions of its memory which currently contain a picture after each set of rules has been applied. The area of memory is—in effect—a matrix of 192×96 elements. At any one stage only a portion of this is in use, the portions being allocated by the coordinates associated with picture name given in the grammar, i.e., Z (N) EDGE (048, 48) is stored starting from location 48 (down), 48 (across). Thus the output for the first set of rules is as shown in Fig. 5(b).

The size of the initial picture (Z) PICTURE is 48×48 cells, the EDGE versions are, however, reduced in size to occupy only 24×24 cells. The necessity for this reduction in size is intimately connected with the form of picture grammar we have chosen to implement. A feature of the linguistic grammar was that substitutions were always defined for pairs of *adjacent symbols*, nowhere do we find a rule which calls for the replacement of symbols widely separated in the string being analysed. In spite of this immediate constituent or phrase structure restriction it is still possible to reflect with a single label (S) the structure of an extended utterance, because as we proceed with the analysis the string automatically shrinks by the very nature of the substitution operation. In the geometrical analysis we perform on a picture, the same immediate constituent limitation is imposed. If, however, we are able to label an extended property of a picture, e.g., a long line, with 'next to' concatenations, we need to draw together these local properties—hence this reduction in the size of the EDGE versions of the picture.

We can summarise the outcome of this 'parsing' process in a different form by collecting together the contents of several pictures, e.g., ENDS and LINES (Fig. 5(c)), and combining them into a single picture with a suitable pictorial convention to indicate the type of picture from which the entry originates (Fig. 6(b)). The results of the next level of description (Fig. 5(d)) containing CURVES, LIMBS and LINES can similarly be displayed pictorially (Fig. 6(c)).

PATTERN RECOGNITION

Hubel & Wiesel 1962) which indicate whether a particular local property occurs anywhere over a small area of the retinal image. Such cells are precisely what one would expect to find mediating the 'reduction' of scale (and resolution) injected into the picture grammar.

The detailed nature of the correspondence means that we can usefully consider the possible significance of discrepancies. For example, there are grounds for believing that the picture grammar is premature in beginning with EDGE definitions. In the visual system a considerable amount of processing—so called 'lateral inhibition'—precedes EDGE operations and is performed in the retina. Conversely we may ask of the physiologists what might correspond to certain features of our picture grammar which are designed to provide size invariance. This type of detailed cross-fertilisation is in many ways quite novel, at least with this subject matter.

The notation can also be used to express physiological results, e.g., the properties of a receptive field unit.

RELATION TO PSYCHOLOGY

Ultimately, the psychological relevance of this form of picture analysis will be in its ability to derive structural descriptions of pictorial objects which reflect our intuitions of their shape. To a limited extent the labelling already demonstrated achieves this: it will be more convincing, however, if we can, for example, show that the same grammar assigns adequate descriptions to objects which are generically different, e.g., fingerprints, chromosomes. It can already be shown, however, that by virtue of the way in which it computes picture labels, the system has some interesting weaknesses. Given a numerical value for the notion of local property, e.g., that no definition can refer to array elements separated by more than 10 units, the system is unable to judge the positions of an object relative to some frame, to better than 10 per cent (i.e., 1/10). This limitation appears in many forms of sensory judgment (Miller 1956). Again the system is unable to 'pick out' (i.e., uniquely label) one member of a large matrix of similar objects, e.g., a matrix of A's. This is a familiar experience. We overcome the limitation by 'tracing' or counting along a line with finger or pencil point. Significantly this use of a *pointer* changes the structure of the picture in a way which would also greatly help the computer system.

CONCLUSION

We can now return to a general remark made in the introduction. The essential characteristic of the system we have described is the volume of operations it involves. To calculate by hand the structural analysis illustrated earlier would occupy perhaps 2-3 hours for each numeral and would inevitably contain many mistakes. The validation of a picture grammar in terms of a whole alphabet and subsequently for other classes of picture would, therefore, be unthinkable except by computer. In that we already

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sense directions in which our approach requires further complications due to over-simplifications like assuming pictures to be only black or white, the claim that psychological theories are necessarily complex, and *require a computer to test them*, will need no further elaboration. In one way it is unfortunate that Artificial Intelligence has been so called, since it suggests that it is concerned with problems quite different from those traditionally studied by the biologist. In fact, however, both disciplines are concerned with the study of very complex systems and collaboration is becoming increasingly fruitful.

ACKNOWLEDGEMENTS

I would like to acknowledge with gratitude the use of the computing facilities at the Culham Fusion Research Laboratory, U.K.A.E.A., the assistance of Douglas Brand and Barry Astbury, and valuable discussions with John Marshall, Professor N. S. Sutherland and Professor R. C. Oldfield.

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