

■ OBJECT AND MATERIAL EIGENFUNCTIONS - Cont'd.

Laplacian eigenfunctions: ... THE INPUT SEGMENT CAN BE RECOGNIZED - at least in a statistical sense - AS CHECKBOARD-CLASS OBJECT, WITH m-VALUE AND M-VALUE PIXELS, WHEN THE CONVOLUTION GENERATES THE EXPECTED 18 VALUES IN 5-DIMENSIONAL FEATURE SPACE WITH THE EXPECTED MULTIPLICITIES  $N_1, N_2, \dots, N_{18}$  FOR THE RESPECTIVE SPECTRAL RESPONSES. Otherwise, the input segment is not recognized as a material/object of this "checkboard class."

• Note: The "ideal scenario" discussed here is used for explaining fundamental aspects of the approach. When adapting the fundamental principles to a "real-world, applied setting" one must adopt concepts including "being very close, nearly the same, approximately equal, within a limited distance, ..." In other words, the approach must be generalized further to be useful when one must handle more general distributions and probabilities.

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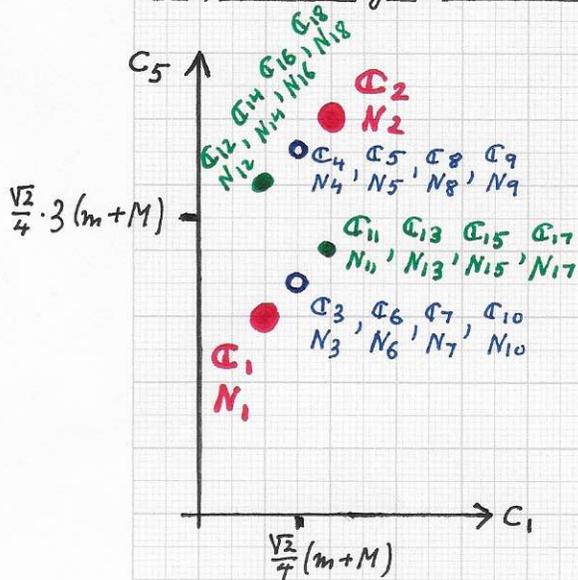
Laplacian eigenfunctions: 5) IF ONE WERE GIVEN A "PERFECT COPY" OF THE IDEAL CHECKERBOARD-TEXTURED SEGMENT AND IF "NO NUMERICAL ERRORS" WERE POSSIBLE, THEN ONE WOULD BE ABLE TO COMPUTE ONLY THE EXACT  $\mathcal{C}$ -TUPLE VALUES OF THE IDEAL CHECKERBOARD-TEXTURED SEGMENT. FURTHER, A "PERFECT COPY" WOULD ALSO LEAD TO THE SAME NUMBERS OF CASE INSTANCES, I.E., THE NUMBERS OF THE 18 POSSIBLE SPECTRAL MASK RESPONSES:  $N_1, N_2, \dots, N_{18}$ .

- Note: When analyzing a given segment that is NOT a "perfect copy" of the ideal segment, applying the convolution mask to all  $t^2 p^2$  pixels of the given segment, one can still obtain the exact (nearly exact) values for the stored ideal  $\mathcal{C}_i$ -tuples AND their exact numbers  $N_1, N_2, \dots, N_{18}$ . This is a consequence of the fact that the analysis and characterization of the given segment is still "merely a statistical characterization."

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■ OBJECT AND MATERIAL EIGENFUNCTIONS - cont'd.

° Laplacian eigenfunctions: ... As far as "case counts" / "numbers



of observed case instances" are concerned, one could keep track of recognized cases in a given segment being analyzed, producing a "count tuple  $IN$ ," i.e.,  $IN = (N_1, \dots, N_{18})$ . One could then determine whether this tuple  $IN$  is "very similar" to a tuple  $IN^{IDEAL}$  of a class to be recognized. In this case one would compare  $\|IN\|$  and  $\|IN^{IDEAL}\|$

The 18 unique  $\alpha$ -tuples  $\alpha_1, \dots, \alpha_{18}$  with their associated counts  $N_1, \dots, N_{18}$ . Feature points  $\alpha_i$  have been projected here to 2D  $(C_1, C_5)$ -space.

and consider the tuples' directional difference, i.e.,  $\frac{(IN \cdot IN^{IDEAL})}{(\|IN\| \cdot \|IN^{IDEAL}\|)}$ .

The corresponding DIRAC function defining this spectral analysis function can be written as

$$N(\alpha) = \begin{cases} N_i, & \text{if } \alpha = \alpha_i \\ 0, & \text{otherwise} \end{cases}$$

$N(\alpha)$  can serve as detector for the  $(m, M)$ -checkboard segment class.

6) The  $\alpha$ -tuples defining the 18 unique spectral responses of the convolution mask determine 18 points / positional vectors in 5-dimensional feature space.

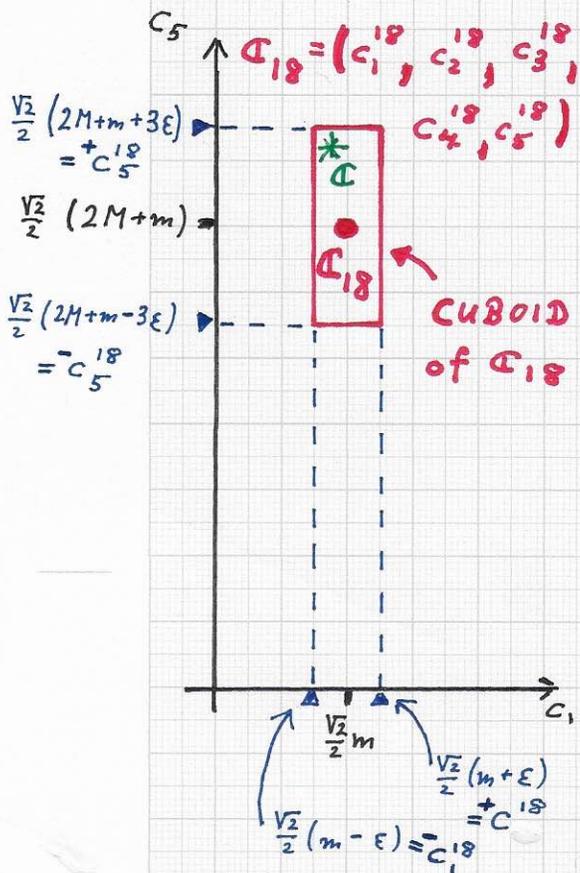
Again, only in the ideal setting precisely 18 points exist. One could view this definition of the perfect characterization of the checkboard segment as a "Dirac function," where 18 "pulses  $\alpha_i$ " with associated numbers  $N_i$  exist in the 5-dimensional feature space.

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OBJECT AND MATERIAL EIGENFUNCTIONS - Cont'd.

Laplacian eigenfunctions: 7) IN THE IDEAL SITUATION CON-

SIDERED HERE, IT IS POSSIBLE TO  
DEFINE THE EXACT 18  $C_i$ -VALUES  
PERFECTLY, TOGETHER WITH THEIR  
COUNTS  $N_i$ . (It will be necessary to  
 use DISTRIBUTIONS of "slightly different  
 $C$ -values" to determine a region in  
 5-dimensional  $C$ -space that is viewed  
 as region associated with one of the  
 18 cases.) WHEN A NEW SEGMENT  
MUST BE ANALYZED, ONE MUST  
CONSIDER ERROR (OR UNCER-  
TAINTY): "m" SHOULD BE " $m \pm \epsilon$ ";  
"M" SHOULD BE " $M \pm \epsilon$ ." Specifically,



Interval arithmetic and  
case recognition. Case 18  
 has the associated tuple  
 $\frac{\sqrt{2}}{2} \cdot (m, M-m, M-m, m-M,$   
 $2M+m) = C_{18}.$

Considering the first  
 and fifth tuple compo-  
 nents,  $m$  and  $2M+m$ ,  
 and replacing  $m$  by  
 $m \pm \epsilon$  and  $M$  by  $M \pm \epsilon$ , the  
 point  $\bullet, C_{18}$ , is "ex-  
panded" to the cuboid

$$[-C_1^{18}, +C_1^{18}]$$

$$\times [-C_5^{18}, +C_5^{18}].$$

The point/tuple  $\bullet, C$ , lies  
inside this cuboid and  
thus represents case 18-  
 when only considering  $C_1$  and  $C_5$ .

the values of the components of a  
 $C$ -tuple of one of the 18 possible  
 cases involve terms like these:

$$m, M, 3m, 3M, m+M, m-M, 3(m-M),$$

$$5m+M, 2m+M \text{ etc.}$$

One must  
 use INTERVAL ARITHMETIC when  
determining whether a  $C$ -tuple  
computed for a new segment to  
be classified represents one of the  
specific 18  $C_i$ -tuple cases or not.

The figure (left) sketches this issue  
 for case 18. ...

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■ OBJECT AND MATERIAL EIGENFUNCTIONS - Cont'd.

• Laplacian eigenfunctions: ... The exact 18  $\mathcal{C}_i$ -tuples are assumed

to be known, i.e.,  $\mathcal{C}_i = (c_1^i, c_2^i, c_3^i, c_4^i, c_5^i)$ ,

$i = 1 \dots 18$ . For the purpose of analyzing

and classifying a new given segment,

we associate "uncertainty intervals" with

each component of all  $\mathcal{C}_i$ -tuples, using

the principle described on the previous

page for case 18. We view  $\mathcal{C}_i$  as

the center point of a hyper-cuboid

in 5-dimensional space defining

the uncertainty region  $\mathcal{C}_i$  for case  $i$ :

$$\mathcal{C}_i = [-c_1^i, +c_1^i] \times [-c_2^i, +c_2^i] \times \dots \times [-c_5^i, +c_5^i],$$

i.e.,  $\mathcal{C}_i$  is the Cartesian product

of the five uncertainty intervals

associated with  $\mathcal{C}_i$ 's exact compo-

nent values. For example,  $[-c_1^i, +c_1^i]$

is the uncertainty interval associated

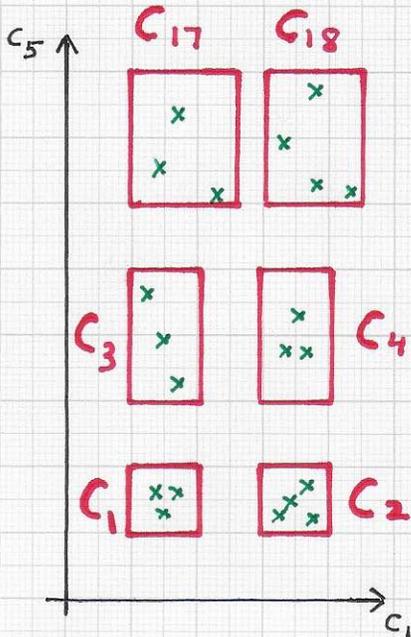
with  $c_1^i$ . When classifying a  $\mathcal{C}$ -tuple

of a new segment the TUPLE  $\mathcal{C}$

REPRESENTS A "CASE  $\mathcal{C}_i$ " WHEN

$\mathcal{C} \in \mathcal{C}_i$ , i.e., when  $\mathcal{C}$  lies inside

the cuboid associated with  $\mathcal{C}_i$ :  $\mathcal{C}_i$ .



Sketch of cuboids of six of the 18 possible cases in our example. The cuboids are projected into  $(c_1, c_5)$ -space. The points 'x' represent  $\mathcal{C}$ -tuples of pixels of a new segment that have passed the "INSIDE-CUBOID" test.

Whenever a pixel of a new segment being analyzed has an associated  $\mathcal{C}$ -tuple inside cuboid  $\mathcal{C}_i$ , the COUNTER for  $\mathcal{C}_i$  is incremented.