

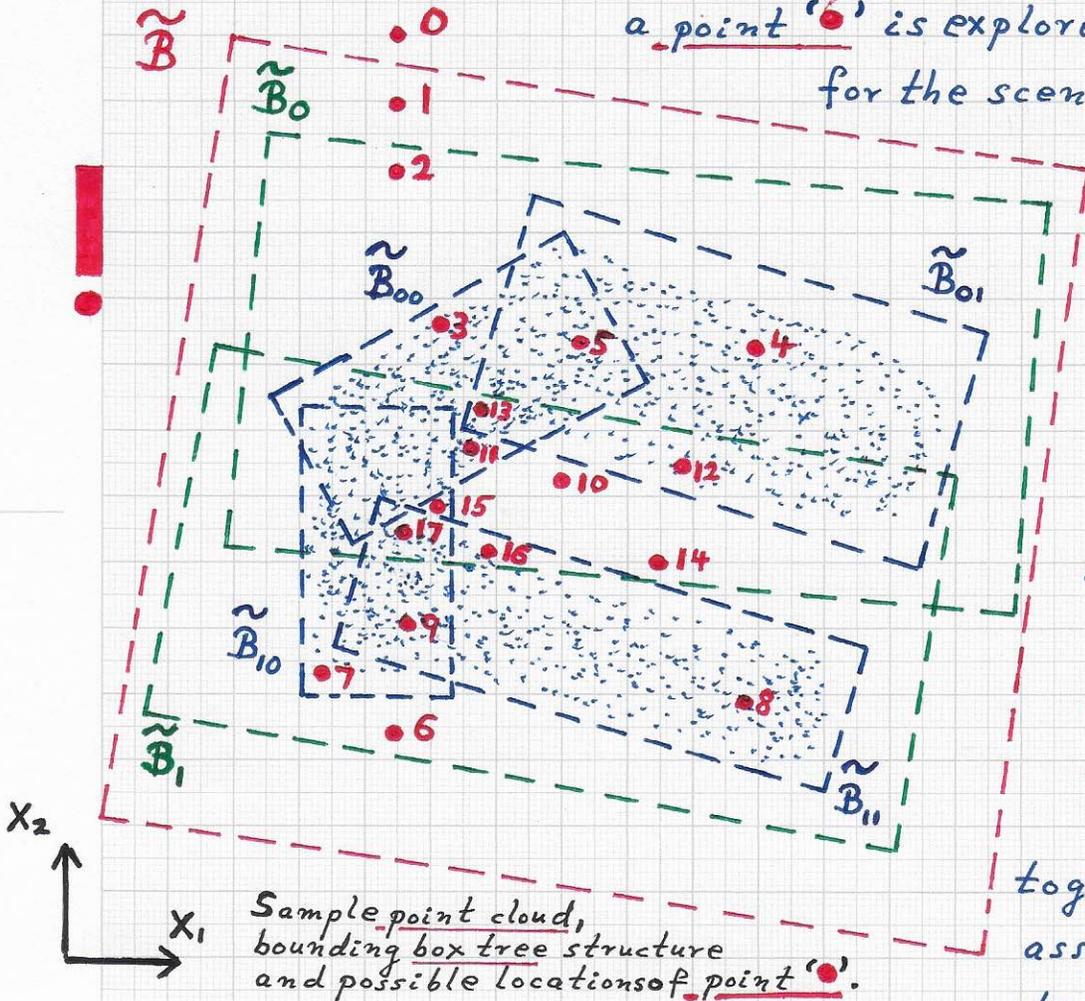
Stratoran

OBJECT AND MATERIAL EIGENFUNCTIONS - Cont'd.

• Laplacian eigenfunctions and neural networks:...

The described method for determining a probability value P for a point '•' is explored in detail for the scenario shown in the left

figure. The point cloud '•••' represents a set of sample points that all belong to the same (material) class. The point cloud is shown together with its associated bounding box tree, consisting of partially intersecting / overlapping expanded local bounding boxes \tilde{B} , \tilde{B}_0 , \tilde{B}_1 , \tilde{B}_{00} , \tilde{B}_{01} , \tilde{B}_{10} and \tilde{B}_{11} .



Sample point cloud, bounding box tree structure and possible locations of point '•'.

The figure includes specific locations of '•' used to exemplify the 18 distinct cases one must consider when calculating the class-membership probability (P) value for point '•'. First, some of these 18 cases are discussed in detail, before summarizing all in a table.

■ OBJECT AND MATERIAL EIGENFUNCTIONS - Cont'd.

• Laplacian eigenfunctions and neural networks:... In the following, we discuss the decision-making and probability calculation for some of the prototypical cases:

- 0 The point is outside \tilde{B} . $P := 0$.
- 1 The point is outside \tilde{B}_0 AND outside \tilde{B}_1 AND inside \tilde{B} . $P := 0$.
- 2 The point is outside \tilde{B}_{00} AND outside \tilde{B}_{01} AND outside \tilde{B}_1 AND inside \tilde{B}_0 AND inside \tilde{B} . $P := 0$.
- 3 The point is inside \tilde{B}_{00} AND outside \tilde{B}_{01} AND outside \tilde{B}_1 AND inside \tilde{B}_0 AND inside \tilde{B} . $P := P_{00}$.
- 5 The point is inside \tilde{B}_{00} AND inside \tilde{B}_{01} AND outside \tilde{B}_1 AND inside \tilde{B}_0 AND inside \tilde{B} . Calculate P_{00} and P_{01} . $P := \max \{P_{00}, P_{01}\}$.
- 10 The point is inside \tilde{B}_0 AND inside \tilde{B}_1 AND inside \tilde{B} . Calculate P_0 and P_1 . $P_0 \geq P_1$. Use sub-tree of \tilde{B}_0 . The point is outside \tilde{B}_{00} AND outside \tilde{B}_{01} . $P := 0$.
- 11 The point is inside \tilde{B}_0 AND inside \tilde{B}_1 AND inside \tilde{B} . Calculate P_0 and P_1 . $P_0 \geq P_1$. Use sub-tree of \tilde{B}_0 . The point is inside \tilde{B}_{00} AND outside \tilde{B}_{01} . $P := P_{00}$.
- 13 The point is inside \tilde{B}_0 AND inside \tilde{B}_1 AND inside \tilde{B} . Calculate P_0 and P_1 . $P_0 \geq P_1$. Use sub-tree of \tilde{B}_0 . The point is inside \tilde{B}_{00} AND inside \tilde{B}_{01} . Calculate P_{00} and P_{01} . $P := \max \{P_{00}, P_{01}\}$.

• Note. The probability values P_0 and P_1 (P_{00} and P_{01}) are the result of mapping \tilde{B}_0 and \tilde{B}_1 (\tilde{B}_{00} and \tilde{B}_{01}) to the normalized domain $[-1, 1]^D$ and computing '•'s probability.

■ OBJECT AND MATERIAL EIGENFUNCTIONS - Cont'd.

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We can now summarize the probability value computations for all 18 cases.

case	\tilde{B}	\tilde{B}_0	\tilde{B}_1	\tilde{B}_{00}	\tilde{B}_{01}	\tilde{B}_{10}	\tilde{B}_{11}	P
0	y							0
1	y	n	n					0
2	y	y	n	n	n			0
3	y	y	n	y	n			P_{00}
4	y	y	n	n	y			P_{01}
5	y	y	n	y	y			$\max\{P_{00}, P_{01}\}$
6	y	n	y			n	n	0
7	y	n	y			y	n	P_{10}
8	y	n	y			n	y	P_{11}
9	y	n	y			y	y	$\max\{P_{10}, P_{11}\}$
10	y	y	y	n	n			0
11	y	y	y	y	n			P_{00}
12	y	y	y	n	y			P_{01}
13	y	y	y	y	y			$\max\{P_{00}, P_{01}\}$
14	y	y	y			n	n	0
15	y	y	y			y	n	P_{10}
16	y	y	y			n	y	P_{11}
17	y	y	y			y	y	$\max\{P_{10}, P_{11}\}$

The left table defines how to calculate the class-membership probability P for a to-be-classified point in 2D (X_1, X_2) -space. The point cloud representing sample points all belonging to the same class C has an associated binary (expanded) bounding box tree obtained after applying repeated box splitting two times to the root box \tilde{B} . The resul-

$P_0 \geq P_1$

$P_1 > P_0$

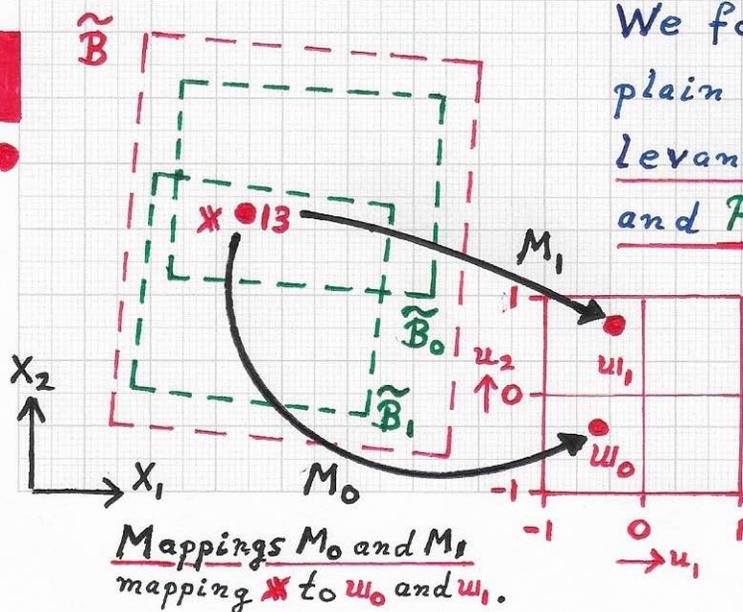
ting boxes are $\tilde{B}, \tilde{B}_0, \tilde{B}_1, \tilde{B}_{00}, \tilde{B}_{01}, \tilde{B}_{10}$ and \tilde{B}_{11} — and they are allowed to intersect. The table uses 'y' and 'n' to indicate that the point belonging to the specific case lies inside or does not lie in a particular box.

■ OBJECT AND MATERIAL EIGENFUNCTIONS - Cont'd.

• Laplacian eigenfunctions and neural networks:...

The bounding box scenario shown in the figure on the previous page includes intersecting expanded bounding boxes \tilde{B}_0 and \tilde{B}_1 . In this case, the described paradigm for box tree traversal ensures that the sub-tree of at most one of the tree nodes \tilde{B}_0 and \tilde{B}_1 is used for the potentially necessary consideration of the children of these nodes for P -value calculation. Thus, when a to-be-classified point lies inside \tilde{B} AND inside \tilde{B}_0 AND inside \tilde{B}_1 , (cases 10-17)

one must compute the box-specific probability values P_0 for \tilde{B}_0 and P_1 for \tilde{B}_1 , to uniquely determine the sub-tree to be used for tree traversal: if $P_0 \geq P_1$, then the sub-tree of \tilde{B}_0 will be used (cases 10-13); if $P_1 > P_0$, then the sub-tree of \tilde{B}_1 will be used (cases 14-17).



We focus on case 13 to explain the computation and relevance of the values of P_0 and P_1 in detail. The left

figure illustrates the essential aspects of a point x representing case 13. One must decide whether to traverse the sub-tree of \tilde{B}_0 or \tilde{B}_1 .

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■ OBJECT AND MATERIAL EIGENFUNCTIONS - Cont'd.

- Laplacian eigenfunctions and neural networks:... The decision to be made concerning sub-tree choice must be made based on a quantitative measure. Such a measure is then used to determine whether the to-be-classified point $*$ should be assigned its class-membership probability based on the subset of the sample point cloud inside \tilde{B}_0 or \tilde{B}_1 . It has been described how one can construct a linear mapping (transformation) M for mapping a bounding box in X -space to the normalized u -space domain $[-1, 1]^D$. The figure on the previous page refers to two such mappings, M_0 and M_1 , associated with \tilde{B}_0 and \tilde{B}_1 , respectively. Point $*$ — lying inside both boxes — is mapped to its corresponding normalized u -space locations/coordinates via M_0 and M_1 . The two results are u_0 and u_1 . Further, it is assumed that a class-membership probability function is defined for the u -space domain square $[-1, 1]^2$. When evaluating this function for u_0 and u_1 , one obtains the needed values for P_0 and P_1 , i.e., $P_0 = P_0(u_0)$ and $P_1 = P_1(u_1)$. For example, if the probability function decreased exponentially with increasing distance from the origin $u = (0, 0)^T$, then P_0 would be greater than P_1 , in the illustrated scenario since $\|u_0\|^2 < \|u_1\|^2$ (u_0 being closer to the origin). Thus, the decision is to choose the sub-tree of \tilde{B}_0 .