

■ OBJECT AND MATERIAL EIGENFUNCTIONS - Cont'd.

• Laplacian eigenfunctions and neural networks...

The determination of an optimal set of orthonormal basis

vectors  $u_{1,1}, u_{1,2}, \dots$  is motivated by the goal of defining an optimal set of orthogonal basis vectors  $b_{1,1}, b_{1,2}, \dots$  for a (hyper-) box

in  $D$ -dimensional space. (The box will subsequently be used for classification decision-making calculations instead of a parallelepiped implied by non-orthogonal basis vectors.)

The left figure shows the eight acceptable ways to define a pair of orthogonal basis vectors  $b_1$  and  $b_2$

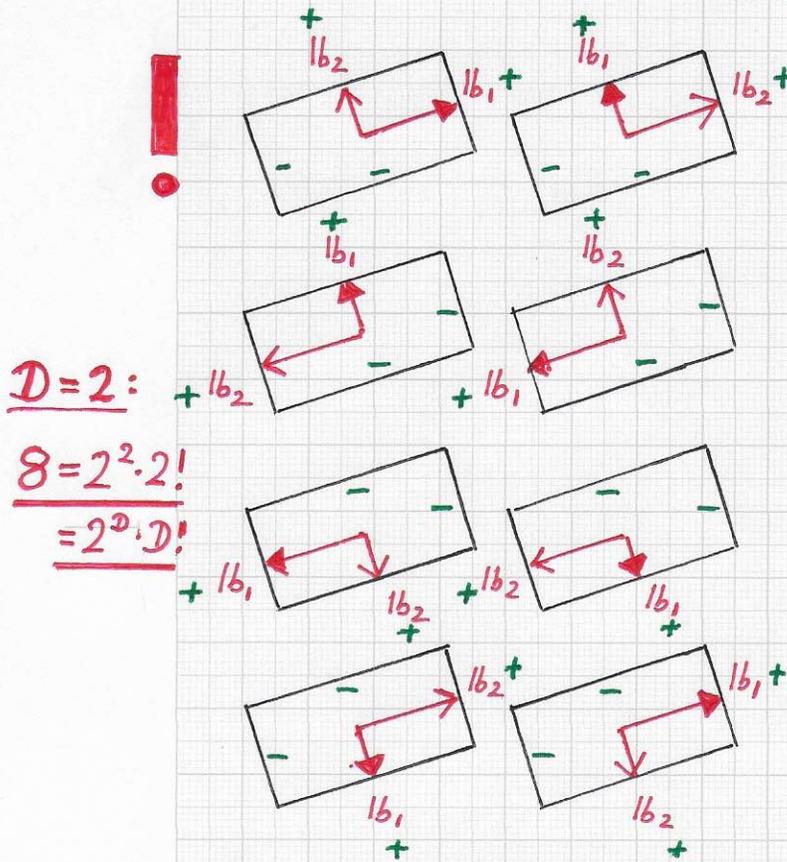
Eight possible and "equivalent pairs" of a pair of orthogonal basis vectors  $b_1, b_2$ .

with a  $2D$  box. The fact that eight basis vector pairs are possible for a  $2D$  box results

$D$	$2^D \cdot D!$
1	2
2	8
3	48
4	384
5	3840

from the facts that (i) four ( $2^2$ ) ways exist for designating a box edge as a positive (+) or negative (-) edge and (ii) two ( $2!$ ) possibilities exist to

assign indices (1 or 2) to a pair of basis vectors. The table shows how rapidly  $2^D \cdot D!$  values increase.



$D=2:$   
 $8 = 2^2 \cdot 2!$   
 $= 2^D \cdot D!$



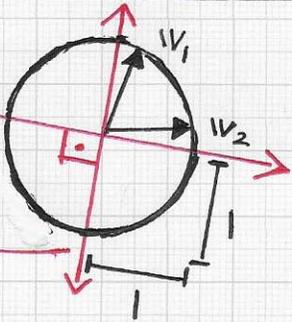
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Considering the fact that one can choose one of all the com-

binatorially possible basis vector sets for a  $D$ -dimensional (hyper-) box, one does not need to worry about the positive or negative sign (orientation) of a basis vector or the indices used for  $D$  basis vectors; they only must be orthogonal. The left figure

Orthogonal line pair



Optimal "axis line pair" constructed for two vectors  $v_1$  and  $v_2$ .

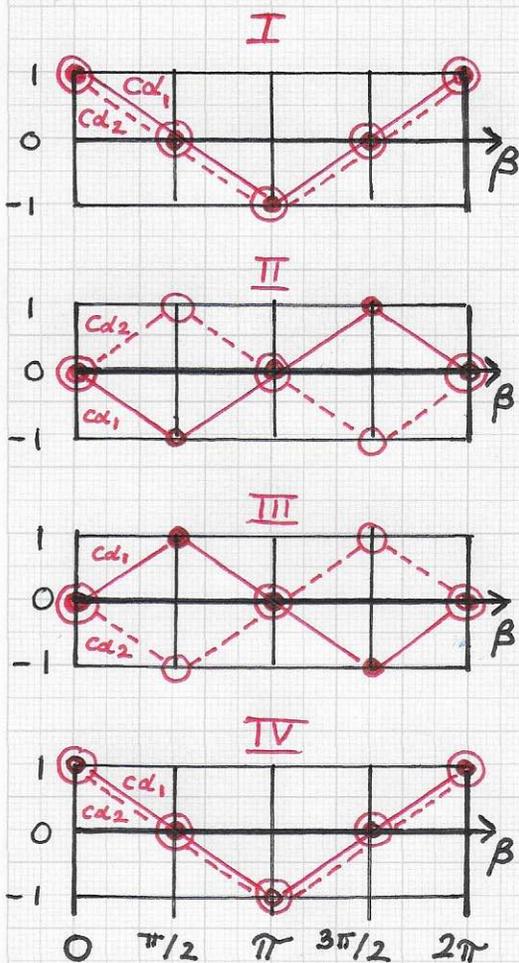
sketches an orthogonal line pair that, when optimally oriented (rotated) fully captures the eight equivalent and acceptable basis vector pairs for two given vectors  $v_1$  and  $v_2$ . The table lists the 16 possible and acceptable orthogonal basis vector pairs  $u_1, u_2$  for orthonormal vectors

$v_2$ 	I	 $u_2$ (up), $u_1$ (right)	1	0	-1	0	$\cos \alpha_1$ $\cos \alpha_2$
	II	 $u_1$ (up), $u_2$ (right)	0	-1	0	-1	
$v_1$	III	 $u_2$ (up), $u_1$ (right)	0	-1	0	1	$\langle v_1, u_1 \rangle$ $\langle v_2, u_2 \rangle$
	IV	 $u_1$ (up), $u_2$ (right)	1	0	-1	0	

$v_1, v_2$ . Each entry lists the values of  $\langle v_1, u_1 \rangle$  and  $\langle v_2, u_2 \rangle$ . . . .

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—  $\cos \alpha_1 = \langle W_1, u_1 \rangle$   
 - - -  $\cos \alpha_2 = \langle W_2, u_2 \rangle$

Cases I, IV:

$$\begin{aligned} C(\beta) &= \cos(\alpha_1(\beta)) \\ &\quad + \cos(\alpha_2(\beta)) \\ &= \underline{2\cos \alpha_1} = \underline{2\cos \alpha_2} \end{aligned}$$

Cases II, III:

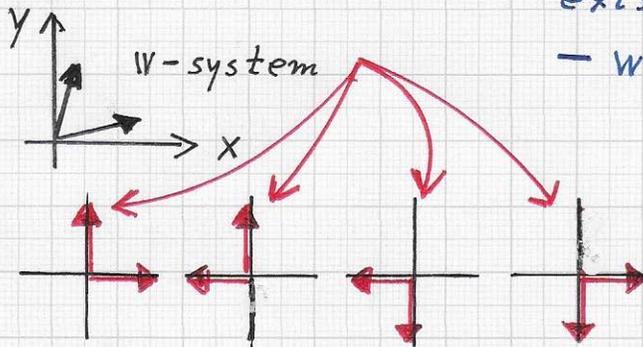
$$\begin{aligned} C(\beta) &= \cos(\alpha_1(\beta)) \\ &\quad + \cos(\alpha_2(\beta)) = \underline{0} \end{aligned}$$

The table provided on the previous page analyzes the special case of a given pair of vectors:  $W_1$  and  $W_2$  are orthonormal — and the two possible index choices are considered. The  $u_1$ -system is rotated relative to the given  $W$ -system by rotation angles  $0, \pi/2, \pi$  and  $3\pi/2$ . Regarding cases I, II, III and IV, the table lists for each of these cases, and each rotation angle, the resulting values of  $\cos \alpha_1$  and  $\cos \alpha_2$ . The left figure shows the same information by providing approximating poly-line plots that use straight line segments for connecting two consecutive  $\cos$ -values, terminating the plots at the value  $\beta = 2\pi$  due to the periodicity of the  $\cos$  function.

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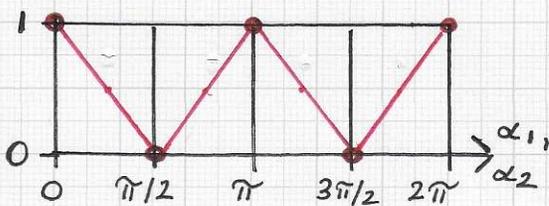


Given W-system and its resulting four optimal ui-systems.

For a given W-system, there exist four optimal ui-systems

- When ignoring the issue of all possible ways to assign indices to the vectors of a pair of vectors. The left figure provides a sketch.

The goal is to formulate and solve this optimization in an algebraic way that yields these four optimal solutions.



The squared functions cos^2(alpha\_1) and cos^2(alpha\_2). They have the same extrema.

• Note. The left figure shows qualitative sketches of the graphs of the functions  $c^2 \alpha_1 = \cos^2 \alpha_1 = (\cos \alpha_1)^2$  and  $c^2 \alpha_2 = \cos^2 \alpha_2 = (\cos \alpha_2)^2$  - which are equal for the considered

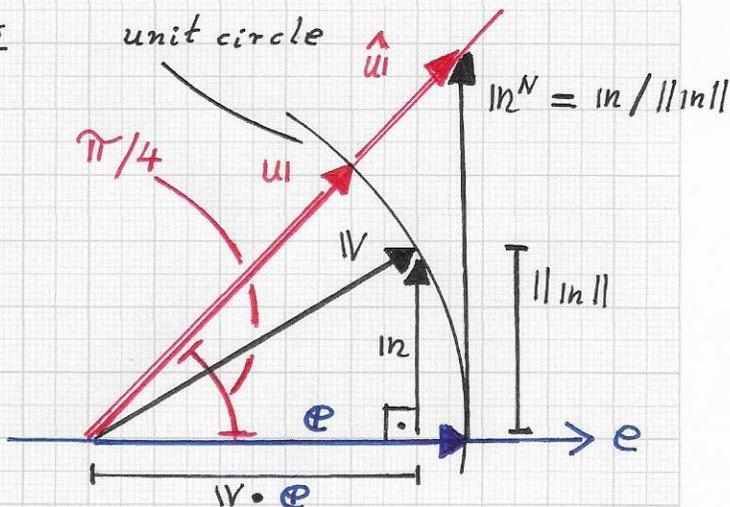
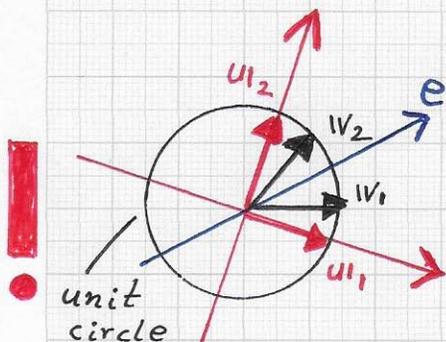
example. The first derivative of  $\cos^2 \alpha_1$  is the function  $-2 \cos \alpha_1 \cdot \sin \alpha_2$ , which has value 0 at  $0, \pi/2, \pi, 3\pi/2, 2\pi$  in the interval  $[0, 2\pi]$ . For these  $\alpha_1$ -values, one obtains the allowable four possible optimal solutions shown in the top figure. (The same holds for  $\alpha_2$ .) This optimization problem is subject to the constraint  $x^2 + y^2 = 1$

concerning all vectors; Lagrange multipliers can be used.

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• Note. The top-left figure shows the desired transformation-mapping two non-orthogonal vectors  $w_1$  and  $w_2$  in an "optimal" way to two orthogonal vectors  $u_1$  and  $u_2$ . The top-right figure illustrates how one can solve

$$\begin{aligned}
 l_n &= w - (w \cdot e) e \\
 \hat{u}_1 &= e + l_n^N \\
 u_1 &= \hat{u}_1 / \|\hat{u}_1\| = \hat{u}_1 / \sqrt{2} \\
 (\|\hat{u}_1\| = \sqrt{2} : \|e\|^2 + \|l_n^N\|^2 &= 2 \\
 \Rightarrow \|\hat{u}_1\|^2 &= 2) \\
 \|l_n\|^2 + (w \cdot e)^2 &= 1 \\
 \Rightarrow \|l_n\| &= \pm \sqrt{1 - (w \cdot e)^2}
 \end{aligned}$$

$$\begin{aligned}
 \Rightarrow u_1 &= \hat{u}_1 / \sqrt{2} = (e + l_n^N) / \sqrt{2} = \\
 &= (e + l_n / \|l_n\|) / \sqrt{2} \\
 &= \left( e \pm \frac{w - (w \cdot e) e}{\sqrt{1 - (w \cdot e)^2}} \right) / \sqrt{2}
 \end{aligned}$$

this transformation problem in the 2D case, using a geometrical approach. For simplicity, the figure and formulas do not use the indices of the  $w$ - and  $u$ -vectors. The "e-direction" and vector  $e$  can be understood as "average" or "eigen" direction implied by  $w_1$  and  $w_2$ .