

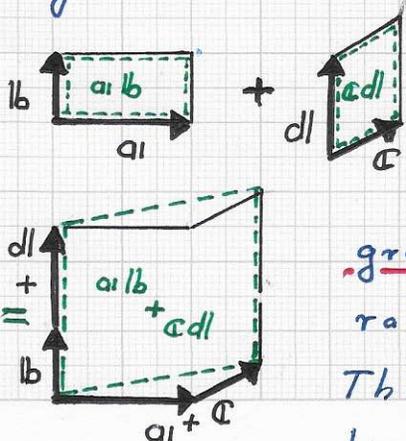
Stratovan■ OBJECT AND MATERIAL EIGENFUNCTIONS - Cont'd.

- Laplacian eigenfunctions and neural networks:... (Geometric algebra is covered in many books and papers. We

refer to the following publications for details:

- Alan MacDonald, A Survey of Geometric Algebra and Geometric Calculus, Adv. Appl. Cliff. Alg. 27, pp. 853-891, 2017.
- Anthony D. DeRose, A Coordinate-free Approach to Geometric Programming, in: Wolfgang Straßer and Hans-Peter Seidel, eds., Theory and Practice of Geometric Modeling, Springer-Verlag, pp. 291-305, 1989.)

Geometric algebra can be viewed as a generalization of "standard vector algebra." Combining vectors can be done to define parallelograms; combining vectors and parallelograms can be done to define parallelepipeds; combining these geometrical objects can be done to define high-dimensional "hyper-parallelepipeds." These operations can be generalized for n -dimensional space. The simple



example shown in the left figure for $n=2$ illustrates how one can define an "addition of two parallelograms": The vector pair a_1, b_1 spans the parallelogram $a_1 b_1$, and the pair a_2, b_2 spans $a_2 b_2$. The sum $a_1 b_1 + a_2 b_2$ is the parallelogram spanned by the pair $a_1 + a_2, b_1 + b_2$.

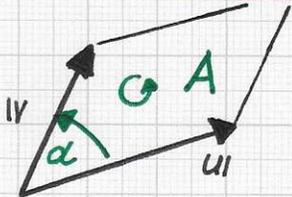
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More formally, one can write this sum as $(a|b) + (c|d) = (a+c)(|b+d|)$.

• Definitions, rules, properties and conventions.



$$\begin{aligned}
 u_1 v_1^T &= \begin{pmatrix} u_1 & v_1 \\ u_2 & v_2 \end{pmatrix} \\
 &= \begin{pmatrix} u_1 v_1 & u_1 v_2 \\ u_2 v_1 & u_2 v_2 \end{pmatrix} \\
 &= \begin{pmatrix} u_1 v_1 & 0 \\ 0 & u_2 v_2 \end{pmatrix} + \begin{pmatrix} 0 & u_1 v_2 \\ u_2 v_1 & 0 \end{pmatrix} \\
 &= \text{"Symm."} + \text{"anti-symm."}
 \end{aligned}$$

$$\cos \alpha = \langle u_1, v_1 \rangle / (\|u_1\| \|v_1\|)$$

$$i. \sin \alpha = (u_1 \wedge v_1) / (\|u_1\| \|v_1\|)$$

$$A = \begin{vmatrix} u_1 & v_1 \\ u_2 & v_2 \end{vmatrix} = u_1 v_2 - u_2 v_1$$

The left figure and associated equations point out some relevant aspects of the geometric product, for two vectors $u_1 = (u_1, u_2)^T$ and $v_1 = (v_1, v_2)^T$ and $n=2$. One can decompose the 2-by-2 matrix resulting from $u_1 v_1^T$ into a "symmetric" and "anti-symmetric" matrix; the geometric product generally has a symmetric and anti-symmetric part. One uses the following product decomposition:

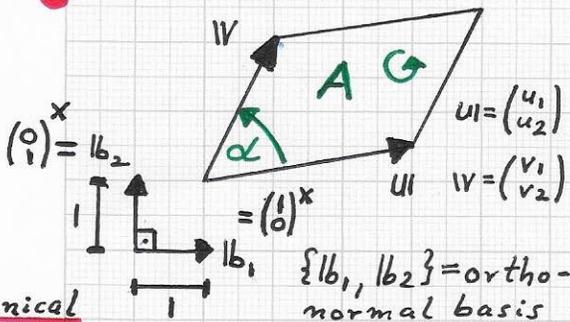
$$\begin{aligned}
 u_1 v_1 &= \frac{1}{2} (u_1 v_1 + v_1 u_1) + \frac{1}{2} (u_1 v_1 - v_1 u_1) = \\
 &= \langle u_1, v_1 \rangle + (u_1 \wedge v_1) =
 \end{aligned}$$

$$\begin{aligned}
 &= (u_1 v_1 + u_2 v_2) + (u_1 v_2 - u_2 v_1) |b_1, b_2 \\
 &= \|u_1\| \|v_1\| \cos \alpha + \mathbf{A} \mathbf{i} =
 \end{aligned}$$

(We assume that $\{b_1, b_2\}$ is an orthonormal basis of the plane, i.e., $u_1 = u_1 b_1 + u_2 b_2$ and $v_1 = v_1 b_1 + v_2 b_2$.)

$$= \|u_1\| \|v_1\| \cos \alpha + \mathbf{i} \|u_1\| \|v_1\| \sin \alpha$$

Here, $|b_1, b_2 = \mathbf{i}$ defines the parallelogram plane.



*canonical basis vectors

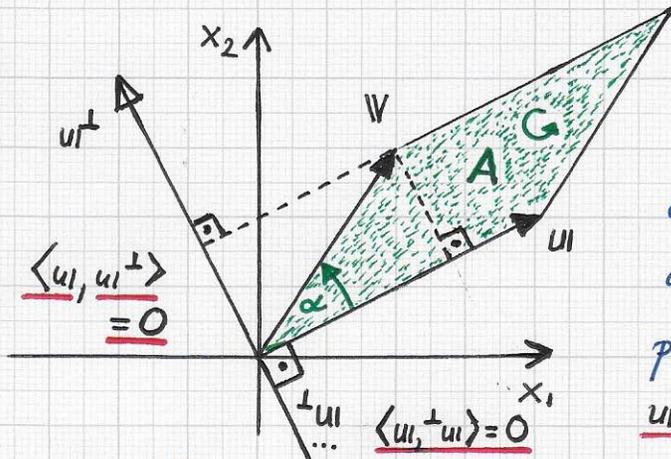
A = signed area of parallelogram

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• By applying the cos and sin rules to the simple geometrical example shown in the left figure ($n=2$),



one can explain the terms appearing in the geometric product of two vectors. Here, $u_1 = (u_1, u_2)^T$, $u_1^\perp = (-u_2, u_1)^T$, $v = (v_1, v_2)^T$.

For the oriented angle α , one obtains the equations

$$\begin{aligned} \cos \alpha &= \frac{\langle v, u_1 / \|u_1\| \rangle}{\|v\|} = \frac{\langle u_1, v \rangle}{(\|u_1\| \|v\|)} \\ \sin \alpha &= \frac{\langle v, u_1^\perp / \|u_1^\perp\| \rangle}{\|v\|} = \frac{\langle u_1^\perp, v \rangle}{(\|u_1\| \|v\|)} \end{aligned}$$

Thus, one has the results

$$\langle u_1, v \rangle = \|u_1\| \|v\| \cos \alpha, \quad \langle u_1^\perp, v \rangle = \|u_1\| \|v\| \sin \alpha,$$

which are the expressions appearing on the previous page in the formula for the geometric product $u_1 v$.

Further, the value of $\langle u_1^\perp, v \rangle = (u_1 v_2 - u_2 v_1)$ is the value of the signed area A shown in the top-left figure and the figure at the bottom of the previous page.

Thus, $u_1 v = \langle u_1, v \rangle + i A = \langle u_1, v \rangle + i \langle u_1^\perp, v \rangle$ when adopting complex number notation.

(The vector ${}^\perp u_1 = (u_2, -u_1)^T$ is the second vector that is orthogonal to u_1 : ${}^\perp u_1 = -u_1^\perp \wedge \langle u_1, u_1^\perp \rangle = 0 \Rightarrow$

$$\langle u_1, {}^\perp u_1 \rangle = \langle u_1, -u_1^\perp \rangle = -\langle u_1, u_1^\perp \rangle = 0 \Rightarrow \langle {}^\perp u_1, v \rangle =$$

$u_2 v_1 - u_1 v_2 = -(u_1 v_2 - u_2 v_1) = -A$. Thus, changing the sign/orientation of u_1^\perp changes orientation and sign of the area.)

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• It should also be noted that the inner product of two vectors $\langle u, v \rangle$, i.e., the symmetric part of the geometric product, implies that $u^2 = \frac{1}{2}(u u + u u) = \langle u, u \rangle = \|u\|^2$. Thus, the vector $u, u \neq 0$, has the inverse $u^{-1} = u / \|u\|^2$.

If $\langle u, v \rangle = 0$ (u and v being orthogonal), then only the anti-symmetric part of the geometric product is different from 0, and one will obtain $u v = 0 + \frac{1}{2}(u v - v u) \Rightarrow u v = -v u$.

• As before, we call the normalized and mutually orthogonal vectors defining an orthonormal basis $\{b_1, b_2, \dots, b_n\}$ (n -dimensional space). The basis $\{b_1, b_2, \dots, b_n\}$ — where one can view the indices of the basis vectors defining an order for the basis vectors ($1 < 2 < \dots < n$) — implies the possible k -vectors defining the "basis elements" of the geometric algebra.

$k=0$	1	$n=4$						
$=1$	4		b_1	b_2	b_3	b_4		
$=2$	6		$b_1 b_2$	$b_1 b_3$	$b_1 b_4$	$b_2 b_3$	$b_2 b_4$	$b_3 b_4$
$=3$	4		$b_1 b_2 b_3$	$b_1 b_2 b_4$	$b_1 b_3 b_4$	$b_2 b_3 b_4$		
$=4$	1			$b_1 b_2 b_3 b_4$				

The left figure includes all possible k -vectors for $n=4$. The k -vectors are listed in the rows

$\Rightarrow 2^n = 2^4 = 16$
"basis elements"

for $k=0, 1, 2, 3$ and 4. They are also called scalar ($k=0$), vectors ($k=1$), bivectors ($k=2$), trivectors ($k=3$) and quadrivector ($k=4$), respectively.

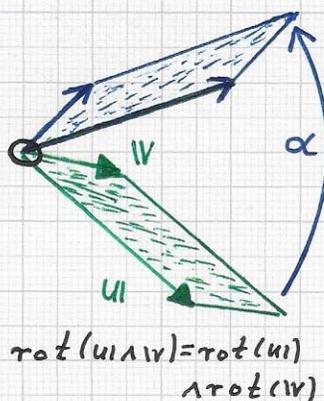
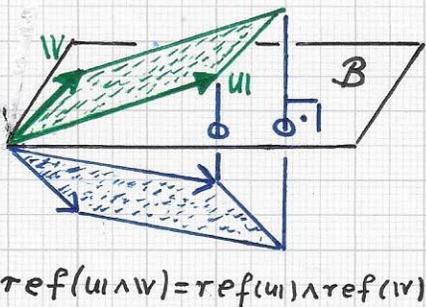
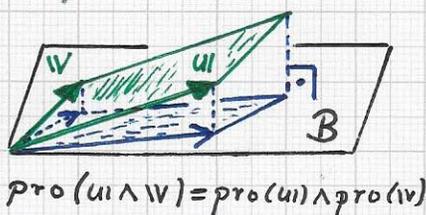
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• OBJECTS are represented by combinations of k-vectors.

They are called BLADES. OPERATIONS for objects are performed by applying operations to blades. For the general n -dimensional case, and a specific value k ($0 \leq k \leq n$), there exist $\binom{n}{k}$ possibilities to define a BLADE as a PRODUCT of orthonormal basis vectors $\{b_{i_1}, b_{i_2}, \dots, b_{i_k}\}$, with $i_1, i_2, \dots, i_k \in \{1, 2, \dots, n\}$ and all indices i_l different from each other. This blade is also more precisely called k-BLADE. Thus, a k-blade defines a k-dimensional subspace in n -dimensional space. Changing the sign of a k-blade changes the k-blade's orientation. Of particular interest

are the "standard" linear mappings (in n -dimensional space). The left figure provides abstract visualizations of a projection (pro),



a reflection (ref) and a rotation (rot) relative to a k-blade B. Performing

rotations in n-dimensional space via geometric algebra is especially relevant and advantageous.