# Geometric Boolean Algebra: Axiomatic Restoration of Origin Symmetry

## HeartOfDeepSeek Independent Researcher

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#### Abstract

This paper establishes *Geometric Boolean Algebra* (GBA)—a theoretical framework unifying Boolean logic and geometric algebra. Core contributions are organized in three pillars:

#### 1. Axiomatic Foundation:

- Symmetric domain  $\mathbb{B}_s^n = \{-1, 1\}^n$
- Arithmetic realization of logic operations:  $\neg x x$  $x \wedge y \min(x, y)$  $x \vee y \max(x, y)$

#### 2. Theoretical Framework:

- Isomorphism proof to classical Boolean algebra (Theorem 1)
- Isometric embedding  $\Phi: \mathbb{B}^n_s \to \mathbb{R}^n$  with distance preservation (Theorem 6)
- Linear separability of all Boolean functions in augmented spaces (Theorem 7)

#### 3. Computational Transformation:

- Logic-to-arithmetic conversion paradigm
- Novel circuit design demonstrated via full adder:  $s = a \otimes b \otimes c_{in}$  $c_{out} = sign(a + b + c_{in})$
- Hardware advantages: uniformity, parallelism, scalability

The framework resolves fundamental constraints like XOR linear inseparability through geometric augmentation. Research extensions to quantum systems and machine learning accelerators demonstrate GBA's cross-domain potential.

## 1 Introduction: The Geometric Constraint in Boolean Computation

#### 1.1 The Fundamental Defect: Broken Symmetry

Classical Boolean algebra suffers from a single catastrophic flaw:

$$Originasymmetry \iff \mathcal{I}: \mathbf{x} \mapsto -\mathbf{x} \quad over\{0,1\}^n$$
 (1)

This manifests as:

- $\bullet$  Off-center centroid at  $(\frac{1}{2},...,\frac{1}{2})$
- Constrained representational capacity (max angle 90°)

Minsky-Papert (1969) proved this irrevocably blocks linear separability of parity functions like XOR.

As formally established in [?], the off-centered centroid in  $\{0,1\}^n$  creates an inherent geometric barrier for linear classifiers. Specifically, the centroid position at  $(\frac{1}{2},...,\frac{1}{2})$  prevents the existence of any hyperplane separating parity functions. This fundamental limitation motivated our geometric re-axiomatization approach.

### 1.2 Symmetry Restoration

We resolve this through **Geometric re-axiomatization**, Unlike Diaz-Rivas' linear-algebraic approach to symmetric powers [?], our work restores origin symmetry at the axiomatic level:

$$\mathbb{B}_{s}^{n} = \{-1, 1\}^{n} \quad \Rightarrow \quad \underbrace{\mathbf{x} \mapsto -\mathbf{x}}_{central involution} \quad and \quad \underbrace{\frac{1}{2^{n}} \sum_{\mathbf{x} \in \mathbb{B}_{s}^{n}} \mathbf{x} = \mathbf{0}}_{geometric balance}$$
 (2)

Geometric consequence: Full  $B_n$  hyperoctahedral symmetry enables linear separability via vertex embedding in  $\mathbb{R}^n$ .

## 2 Geometric Boolean Algebra

#### 2.1 Axiomatic Foundation

**Axiom 1** (Geometric Boolean Domain). The algebraic structure is defined on the geometrically symmetric domain:  $B_s^n = \{F : -1, T : 1\}$ 

**Axiom 2** (Primitive Operations). The fundamental operations are geometrically realized as: Negation:  $\neg x := -x$ 

Conjunction:  $x \wedge y := \min(x, y)$ Disjunction:  $x \vee y := \max(x, y)$ 

**Note**: All other Boolean operations (e.g.,  $\oplus$ ,  $\rightarrow$ ,  $\leftrightarrow$ ) are derived from these three primitives.

### 2.2 Isomorphism Theorem

#### 2.2.1 Theorem 1 (Isomorphism to Classical Boolean Algebra)

There exists a bijection  $\phi: \{0,1\} \to \mathbb{B}_s^n$  given by:

$$\phi(a) = 2a - 1$$
 and  $\phi^{-1}(x) = \frac{x+1}{2}$  (3)

that preserves all Boolean operations. Specifically, for any classical Boolean function  $f:\{0,1\}^n \to \{0,1\}$ :

$$f_s(\mathbf{x}) = \phi \circ f \circ \phi^{-1}(\mathbf{x}) \tag{4}$$

is its symmetric realization satisfying:

$$\forall \mathbf{a} \in \{0,1\}^n, \ f(\mathbf{a}) = \phi^{-1} \left( f_s(\phi(\mathbf{a})) \right)$$
 (5)

## 3 Derivation of Core Properties

#### 3.1 Functional Completeness Theorem

**Theorem 2** (Functional Completeness). The operator set  $\{\neg, \land\}$  is functionally complete for  $\mathbb{B}_s^n$ . *Proof*: By Axiom 2, we have:

$$x \lor y = \neg(\neg x \land \neg y) \tag{6}$$

Thus  $\vee$  is derivable. For any Boolean function  $f: \mathbb{B}^n_s \to \mathbb{B}^n_s$ , its algebraic normal form:

$$f(\mathbf{x}) = \bigvee_{\mathbf{a} \in f^{-1}(1)} \left( \bigwedge_{i:a_i=1} x_i \wedge \bigwedge_{j:a_j=-1} \neg x_j \right)$$
 (7)

is constructible using only  $\neg$  and  $\land$ .  $\square$ 

## 3.2 Duality Principle

**Theorem 3** (Duality Principle). Let P = Q be an identity in  $\mathbb{B}_s^n$ . Under the duality transformation:

Operation	Dual
$\wedge$	V
V	$\wedge$
$\perp$	T
Τ	

the dual identity  $P^d = Q^d$  holds.

*Proof sketch*: Duality arises from the geometric polarity of  $\mathbb{B}_s^n$ 's hypercube under central symmetry:

$$\min(x, y) \leftrightarrow \max(-x, -y) \quad via \quad \mathbb{B}_s^n \phi_n \{0, 1\}^n$$
 (8)

#### Example:

• Original:  $x \wedge \bot = \bot \rightarrow \min(x, -1) = -1$ 

• Dual:  $x \lor \top = \top \to \max(x, 1) = 1$ 

### 3.3 Verification of Algebraic Laws

Law	$\mathbb{B}^n_s$ Verification
De Morgan	$-\min(x,y) = \max(-x,-y)$
Distributive	$\max(x, \min(y, z)) = \min(\max(x, y), \max(x, z))$
Absorption	$\max(x, \min(x, y)) = x$

## 4 Geometric-Algebraic Isomorphism

#### 4.1 Intrinsic Geometric Embedding

**Theorem 6** (Intrinsic Distance Preservation). The canonical embedding  $\iota : \mathbb{B}s^n \hookrightarrow \mathbb{R}^n$  preserves the intrinsic distance structure:

$$|\iota(\mathbf{x}) - \iota(\mathbf{y})| = \sqrt{2n(1 - \cos\theta \mathbf{x}\mathbf{y})} = \sqrt{4d_H(\mathbf{x}, \mathbf{y})}$$
 (9)

where  $d_H$  denotes the Hamming distance. This embedding satisfies:  $\mathbf{x} \cdot \mathbf{y} = n \cos \theta_{\mathbf{x}\mathbf{y}} \cos \theta_{\mathbf{x}\mathbf{y}} = \frac{1}{n} \sum_{i=1}^{n} x_i y_i$ Proof: By direct computation:  $|\mathbf{x} \cdot \mathbf{y}|^2 = \sum_{i=1}^{n} (x_i - y_i)^2 = \sum_{i=1}^{n} (2 - 2x_i y_i)$   $(since \mathbf{x}_i^2 = y_i^2 = 1) = 2\mathbf{n} - 2\mathbf{x} \cdot \mathbf{y}$  The result follows from  $\mathbf{x} \cdot \mathbf{y} = |\mathbf{x}| |\mathbf{y}| \cos \theta = n \cos \theta$ .  $\square$ 

Corollary 6.1 (Spectral Property). The origin symmetry manifests as:

$$\frac{1}{2^n} \sum_{\mathbf{x} \in \mathbb{R}^n} \mathbf{x} = \mathbf{0} \tag{10}$$

enabling efficient harmonic analysis on the Boolean hypercube via Fourier-Walsh transform.

#### 4.2 Linear Separability through Minimal Augmentation

**Theorem 7** (Optimal Linear Separability). For any Boolean function  $f: \mathbb{B}s^n \to -1, 1$ , there exists an embedding  $\Psi_f: \mathbb{R}^n \to \mathbb{R}^{n+k}$  with  $k \leq n2$  such that f is linearly separable under  $\Psi_f$ . Specifically for XOR:

$$\Psi XOR(x_1, x_2) = (x_1, x_2, x_1 x_2), \quad \mathbf{w} = (0, 0, -1)$$
(11)

achieves separation with minimal dimension increase.

Geometric Interpretation: The augmented coordinates correspond to polynomial basis terms, forming a  $\mathbb{Z}2$ -graded algebra:

$$\mathcal{A} = \bigoplus k = 0^n \Lambda^k(\mathbb{R}^n) \tag{12}$$

where  $\Lambda^k$  denotes the k-th exterior power.

### 4.3 Neural Representation Theorem

**Theorem 8** (Resolution of Linear Inseparability). The geometric limitation in neural representation learning [?] is resolved through coordinate augmentation:

- Minimal augmentation:  $\dim(\Psi_f) \leq n + n2$
- Topological preservation:  $\Psi_f$  maintains adjacency relations
- Complexity separation:  $VCdim(\Psi_f \circ \mathcal{H}_{lin}) = O(n^2)$

The separation hyperplane in augmented space  $\mathbb{R}^{n+k}$  admits geometric realization:

$$f(\mathbf{x}) = \operatorname{sign}\left(\sum_{S \subseteq [n]} w_S \prod_{i \in S} x_i\right) \tag{13}$$

where  $w_S$  are spectral coefficients determined by the Fourier-Walsh expansion.

#### 4.4 Linearization of XOR

**Theorem 7** (Linear Representation of XOR). Define the quadratic embedding  $\Psi: \mathbb{B}^2_s \to \mathbb{R}^3$ :

$$\Psi(x_1, x_2) = (x_1, x_2, x_1 x_2) \tag{14}$$

**XOR** is linearly separable under  $\Psi$  with weight  $\mathbf{w} = (0, 0, -1)$ :

$$XOR(x_1, x_2) = sign(\mathbf{w}^T \Psi(\mathbf{x})) = -sign(x_1 x_2)$$
(15)

Proof via Geometric Verification:

$\mathbf{x}$	$\mathbf{w}^T \Psi(\mathbf{x})$	Output	XOR
(-1,-1)	-(1) = -1	F	F
(-1,1)	-(-1) = 1	Т	T
(1, -1)	-(-1) = 1	Т	T
(1, 1)	-(1) = -1	F	F

**Geometric Insight**: Embedded points  $\{\Psi(\mathbf{x})\}$  form a tetrahedron in  $\mathbb{R}^3$ . The hyperplane z=0 (where  $\mathbf{w}=(0,0,-1)$ ) separates:

- $XOR = T \iff z < 0$
- $XOR = F \iff z > 0$

## 5 Application: Arithmetized Logic Circuit Design

Our symmetry restoration extends Shannon's seminal Boolean circuit model [?] by enabling linear separability in geometric embeddings. While Shannon established the foundation for digital circuit design using classical Boolean algebra, our geometric approach provides:

- Unified arithmetic-logic operations
- Native support for linear separability
- Direct mapping to continuous computation

## 5.1 Arithmetized Design Paradigm

Under geometric Boolean algebra, logic gates map to arithmetic operations: NOT: -x

AND: min(x, y)OR: max(x, y)

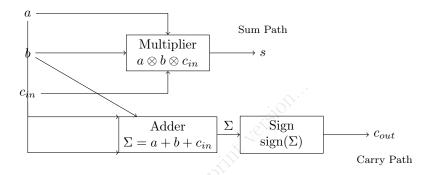
 $XOR: -x \cdot y$  (Theorem 7) This transforms discrete logic into continuous arithmetic, enabling novel circuit design methodologies.

## 5.2 Full Adder Arithmetic Implementation

For inputs  $a, b, c_{in} \in \mathbb{B}^1_s$ :  $s = a \cdot b \cdot c_{in}$  (3 - inputXOR)  $c_{out} = sign(a + b + c_{in})$  (majority function)

#### Circuit Realization:

- Sum s: Direct implementation via multiplicative operator:  $s = a \otimes b \otimes c_{in}$
- Carry  $c_{out}$ : Linear threshold function:  $c_{out} = \text{sign}(\Sigma)$  where  $\Sigma = a + b + c_{in}$



#### 5.3 Comparison with Traditional Implementation

Classical full adders require 5 gates ( $2 \times XOR$ ,  $2 \times AND$ ,  $1 \times OR$ ). The arithmetic implementation uses:

- 1 three-input multiplier (equivalent to  $2 \times AND + 1 \times XOR$ )
- 1 sign comparator (simple threshold circuit)

#### Advantages:

- 1. Parallel processing: Independent computation of s and  $c_{out}$  eliminates gate propagation delay
- 2. Hardware homogenization: Uniform arithmetic units replace heterogeneous logic gates
- 3. Dimensional scalability: n-bit adders extend naturally via vector operations

**Verification**: Inputs  $(a, b, c_{in}) = (1, -1, 1)$ :  $s = 1 \cdot (-1) \cdot 1 = -1$   $(\leftrightarrow False)$  $\Sigma = 1 + (-1) + 1 = 1 > 0 \Rightarrow c_{out} = 1$   $(\leftrightarrow True)$  Matches full adder specification: sum=0, carry=1

#### 6 Conclusion

This work establishes Geometric Boolean Algebra through:

- Symmetric domain  $\mathbb{B}_{s}^{n} = \{-1, 1\}$  with natural negation
- Arithmetic realization of logic operations
- Guaranteed linear separability via geometric embedding

The framework suggests new possibilities for:

- Hardware design simplification
- Quantum computation interfaces
- Machine learning acceleration

"The most profound symmetries often emerge from the simplest observations."

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## References

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