

# A Journey from Curiosity to Higher-Dimensional Geometry

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## Abstract

The first draft of this work was completed in June 2017, when I was turning 18 in my final year of high school. At the time, I was learning a lot of fundamental mathematics and had a strong desire to understand dimensions beyond what we can see with our own eyes, in other words, trying to visualize higher-dimensional objects in my mind. Motivated by this curiosity while feeling I had acquired enough knowledge on the subject, I wrote a manuscript called “Dimensions”, which explored some mathematical concepts but primarily focused on interpreting and visualizing such new objects. During that period, I regarded the connection between geometry and combinatorics as the fundamental truth of mathematics, much like the mathematicians of earlier eras [2]. With this in mind, my essay aimed to bridge the gap between abstract mathematical concepts and intuitive visualization. This current version serves as a translation, reinterpretation, and self-review of the original draft, featuring refined explanations and updated insights. Adopting a bottom-up approach, it simultaneously serves as both a complement to and detailed revision of Coxeter’s foundational work [1].

## 1 Introduction

We focus on the study of hyper-cubes, one of the many geometric structures extensively explored by Coxeter [1]. His work presents a highly effective top-down approach to geometry, though some inductive proofs are left as exercises for the reader. In this paper, we introduce this area of mathematics through a bottom-up, detail-oriented approach, providing complete proofs where they were previously omitted to build intuition for non-expert readers.

This chapter begins by providing the motivation for our study, defining key terms, and outlining our approach to the subject. In Chapter 2, we explore the elements introduced, aiming to identify fundamental patterns that relate the number of components to the dimensionality of hyper-cubes. These patterns are then generalized and their geometric interpretations analyzed in detail in Chapter 3 to enhance our ability to visualize the elements. In Chapter 4, we discuss potential applications and properties derived from our findings. Finally, in Chapter 5, we conclude the paper and propose directions for further research.

### 1.1 Motivation

Let’s start with an exercise, let  $S$  be a square [2] with sides of length  $a$  and let  $d$  be the length of its face diagonal. By Pythagoras [2] we can establish the relation  $d = a\sqrt{2}$ . Let  $d'$  be the length of the space diagonal on  $C$ , a cube [2] with edges of length  $a'$ . Under a similar argument the length of the larger diagonal, which forms a right triangle jointly with the diagonal of one of the faces and an edge, can be calculated as  $d' = a'\sqrt{3}$ .

We now define as  $d(0) = a\sqrt{0}$  the diagonal of a point (no diagonal) and as  $d(1) = a\sqrt{1}$  the diagonal of a line (the line itself). Then, we can generalize the analysis by the formula  $d(n) = a\sqrt{n}$ , where  $d(n)$  outputs the length of the larger diagonal of an  $n$ -dimensional ( $n$ -dim from now on) cube, a polytope  $P$  such that all sides are of size  $a$  and all internal angles are right angles [1] (see Section 1.2).

*Proof.* We will show the statement by the principle of induction [7].

1. Regarding the base cases, we assumed  $d(n) = a\sqrt{n}$  for  $n = 0$  and  $n = 1$ .

2. Assume  $d(n) = a\sqrt{n}$ . Then, by geometry, the larger diagonal of an  $(n + 1)$ -dim polygon forms a right triangle with a  $n$ -dim diagonal and an edge of the polygon. Therefore, by Pythagoras we have that  $d(n + 1) = \sqrt{d(n)^2 + a^2}$ . Finally, by hypothesis,  $d(n + 1) = \sqrt{(a\sqrt{n})^2 + a^2} = a\sqrt{n + 1}$ .

By the base and inductive cases, we conclude the property holds.  $\square$

Can we uncover additional patterns that depend on the dimension of a polygon? Will this exercise help us visualize and understand higher-dimensional structures? To explore these questions, we will apply tools from combinatorics and geometry.

## 1.2 Definitions

How many physical dimensions can we observe? Our eyes can perceive only two dimensions, as the light reflected from objects in front of us enters our eyes and stimulates the retina at the back of the eyeball. This results in a 2-dim projection of 3-dim objects. While we do perceive depth, it is actually an illusion created by our brain, which interprets the two different images provided by our eyes.

This concept may not be immediately intuitive. To clarify, if we could truly see in three dimensions, we would be able to observe all six faces of a cube simultaneously. However, we can only perceive up to three faces at once and must rely on our imagination to visualize the entire object.

First, let us define the fundamental elements, much like Euclid did at the beginning of his books [2]. A point is a 0-dimensional object, and a straight line is a 1-dim object, defined as the shortest path between two points (called line from now on). A square is a 2-dim object with four lines of equal length that connect four points in a closed loop, with all angles being right angles. And a cube is a 3-dim object composed of six squares of equal size, such that every line is shared by two squares and all angles between adjacent squares are right angles.

We refer to the components of a higher-dimensional object as sub-objects, for example, a square is a sub-object of a cube. To construct the next object from a given object, we perform a translation of a copy of the object perpendicular to all the lines on the object, and connect each original point to its corresponding point in the copy with a straight line. Each new translation is of the same length as the translation in the previous dimension. This particular generalization of a regular polytope given by Coxeter [1] is called hyper-cube or measure polytope (or  $\gamma_n$ ). An  $n$ -dim cube is a particular case of an orthotope, where all translations are not necessarily the same length; which in turn is a particular case of a parallelotope, where translations are not necessarily perpendicular [1].

When deriving formulas to describe objects, it is essential to demonstrate how they relate to this described construction. We recommend to analyze the visualization at [https://commons.wikimedia.org/wiki/File:From\\_Point\\_to\\_Tesseract\\_\(Looped\\_Version\).gif](https://commons.wikimedia.org/wiki/File:From_Point_to_Tesseract_(Looped_Version).gif).

The number of sub-objects for each object is listed in Table 1.

Sub-objects \ Objects	Point	Line	Square	Cube
0-dim	1	<b>2</b>	4	8
1-dim	0	1	<b>4</b>	12
2-dim	0	0	1	<b>6</b>
3-dim	0	0	0	1

Table 1: Number of components in the constructed objects with the  $(n - 1)$ -dim diagonal highlighted [1]

## 2 Deducible formulas

Notice in Table 1 that interesting patterns begin to emerge once the information is organized. There is a clear relationship between the amount of sub-objects and the dimension of a particular object. We start inferring patterns bottom-up because deriving a general formula from the construction of the objects using intuition is a very challenging task for someone who seeks an introduction to the matter.

After discussing a pattern, we will mathematically demonstrate that the derived formula aligns with our method for constructing an object's next dimension. With this approach, every found pattern will be a mental exercise and a great way to develop an intuition. For instance, the number of points  $p(n)$  given a dimension  $n$  can be simply calculated as  $p(n) = 2^n$ . The geometric interpretation of this formula arises from the duplication of an object as it is extended into the next dimension, with the generation of points occurring exclusively in this process.

For the number of lines, we have to shift the row to the right so the exponent will become  $n - 1$ . This second row differs from the first because it seems to be multiplied by a factor that depends on the dimension (the sequence is  $k_1 \cdot 2^0, k_2 \cdot 2^1, k_3 \cdot 2^2$ ). In this case this factor is clearly  $n$  and it follows that the deducted formula is  $l(n) = n \cdot 2^{n-1}$ .

The geometric interpretation for this formula is not as straightforward as in the case of the first row. In the duplication process, each line is doubled, while in the connection process, a new line is drawn between each original point and its corresponding copy. The combined process outputs the recursive formula  $f(n) = 2 \cdot f(n - 1) + p(n - 1)$  such that  $f(0) = 0$  and  $f(1) = 1$  (see Table 1).

*Proof.* It is evident that  $l(0) = 0$  and  $l(1) = 1$ , and if we plug in our formula for  $l(n)$  and  $p(n)$  into the recursion we obtain

$$\begin{aligned} f(n) &= 2 \cdot (n - 1) \cdot 2^{n-2} + 2^{n-1} \\ &= (n - 1) \cdot 2^{n-1} + 2^{n-1} \\ &= n \cdot 2^{n-1} \\ &= l(n), \end{aligned}$$

thereby solving the recursion using the principle of induction [7]. □

The previous analysis is key to understand where these formulas come from as we will show a generalization in Chapter 3. Each pattern has its own corresponding expression, interpretation, and recursive formula.

Regarding the third and fourth row in Table 1, we lack enough data and we must generate the next dimension to be capable of deducting further patterns. We define a tesseract following the same construction described in Chapter 1 with respect to a cube, such that we can use the formulas that we already derived to compute some of the cells. In consequence, a tesseract must have 16 points and 32 lines.

To determine how many cubes are in a tesseract, we can examine the  $(n - 1)$ -dim diagonal which is highlighted in Table 1. We observe: 2 points in a line, 4 lines in a square, and 6 squares in a cube [1]. This new pattern suggests that the  $n - 1$  diagonal is described by the expression  $2 \cdot n$ , and that a tesseract contains  $2 \cdot 4$  cubes (8 cubes). Its important to mention that intuition is not always reliable [1], therefore all formulas must be justified by a geometric interpretation and proof.

To calculate the number of squares in a tesseract, we can also use a diagonal as it provides more information than the row. Although this is more challenging to deduce, all indications suggest that we are dealing with a combinatorial problem, which points toward a factorial approach. The  $(n - 2)$ -dim diagonal in Table 1 seems to be described by the expression  $2 \cdot n \cdot (n - 1)$ , leading to the conclusion that a tesseract contains  $2 \cdot 4 \cdot (4 - 1)$  squares (24 squares). In summary, the number of all tesseract components are shown in Table 2.

Sub-objects \ Objects	Point	Line	Square	Cube	Tesseract [1]
0-dim	1	2	4	8	16
1-dim	0	1	4	<b>12</b>	32
2-dim	0	0	1	<b>6</b>	<b>24</b>
3-dim	0	0	0	1	8
4-dim	0	0	0	0	1

Table 2: Number of components in the constructed objects up to 4 dimensions with 3 cells highlighted

Finally, we can approach the third and fourth rows under a similar analysis as the  $l(n)$  formula:  $s(n) = n \cdot (n - 1) \cdot \frac{2^{n-2}}{2}$  for the amount of squares and  $c(n) = n \cdot (n - 1) \cdot (n - 2) \cdot \frac{2^{n-3}}{6}$  for the amount of cubes. We leave the task of finding the interpretation and recursion of these formulas as an exercise for the reader. In the following chapter we will show how to generalize these patterns, analyze a geometric interpretation and prove our derivation is consistent with the object construction from Chapter 1.

### 3 Generalization and interpretation

In Chapter 2, our approach to organizing the information allowed us to identify patterns in the structures. We now apply this same process to the formulas in order to derive a general form. Subsequently, we present a proof demonstrating that the formula aligns with the construction presented in Chapter 1.

$$\begin{aligned}
p(n) &= 2^n \\
l(n) &= n \cdot 2^{n-1} \\
s(n) &= n \cdot (n-1) \cdot \frac{2^{n-2}}{2} \\
c(n) &= n \cdot (n-1) \cdot (n-2) \cdot \frac{2^{n-3}}{2 \cdot 3}
\end{aligned}$$

Table 3: Compilation of formulas

### 3.1 Generalization

We show a compilation of the formulas in Table 3. From this, it is natural to conclude that  $t(n) = n \cdot (n-1) \cdot (n-2) \cdot (n-3) \cdot \frac{2^{n-4}}{2 \cdot 3 \cdot 4}$  describe the amount of 4-dim objects. Table 2 could be infinitely extended using these patterns, but our main task is to find a general formula that can model the object construction.

Let  $n$  and  $m$  be dimensions such that  $n \geq m \geq 0$ . We will denote the amount of  $m$ -dim sub-objects in a  $n$ -dim object as  $P(n, m)$  and the falling factorial of  $n$  over  $m$  factors as  $(n)_m$ . We infer from Table 3 that the pattern of the formulas is described by the expression [1]

$$P(n, m) = n \cdot (n-1) \cdot \dots \cdot [n - (m-2)] \cdot [n - (m-1)] \cdot \frac{2^{n-m}}{m!} = \frac{(n)_m}{m!} \cdot 2^{n-m} = \binom{n}{m} \cdot 2^{n-m}.$$

Notice that given an  $n$ -dim object and a point on the object, selecting a set of  $m$  adjacent points fully identifies an  $m$ -dim object. For example, considering a particular point in a cube, combining it with an adjacent point identifies one of the 12 lines. Moreover, when a point in a cube is combined with two other adjacent points, only one of the 6 squares is adjacent to all three points simultaneously. This selection enables us to count sub-objects and give many explanations for the formula.

A first interpretation of the formula is based on a combinatorial analysis. Given a dimension  $n$ , the object has  $2^n$  points by construction. Then, a single point is adjacent to  $n$  points and for each set of  $m$  adjacent points a different  $m$ -dim object is counted. Thus, every point is adjacent to  $\binom{n}{m}$   $m$ -dim objects. Finally, since an  $m$ -dim object has  $2^m$  points, the formula is divided by that factor to avoid counting each  $m$ -dim object more than once [3].

A second interpretation comes from a known property of binomial coefficients [5]. It can be proven that  $P(n, m) = \sum_{k=m}^n \binom{n}{k} \binom{k}{m}$  [3], where this expression is derived by fixing a point on the object and counting all adjacent structures connected to that point. In more explicit terms,  $P(n, m)$  can be calculated as selecting a point and computing the sum, for every  $k \geq m$ , of: the number of  $m$ -dim sets adjacent to a point in  $k$  dimensions, multiplied by the number of  $k$ -dim sets adjacent to a point in  $n$  dimensions. Our derived formula counts the same: there are  $\binom{n}{m}$  ways to identify  $m$ -dim sets adjacent to a point in  $n$  dimensions, and there are  $2^{n-m}$  ways to complete a singular  $m$ -dim set of points, in order to enumerate all intermediate  $k$ -dim sets present in the object summation.

Lastly, using the binomial theorem [5], another interesting interpretation is that  $P(n, m)$  can be defined as the coefficient of the term  $X^m$  in the expansion of the expression  $(2 + X)^n$  [1]. This interesting fact can be used when proving properties over these objects, an example is shown in Section 4.1.

### 3.2 Proof of the formula's geometric interpretation

Let  $n$  and  $m$  dimensions, we expand on the geometric interpretations by giving the recursion [1]

1.  $f'(0, 0) = 1$ ,
2.  $f'(n, m) = 0$  if  $n < m$ ,
3.  $f'(n, m) = 0$  if  $n < 0$  or  $m < 0$ , and
4.  $f'(n, m) = 2 \cdot f'(n-1, m) + f'(n-1, m-1)$  on any other case.

To count the amount of  $m$ -dim objects in an  $n$ -dim object, generate an  $(n-1)$ -dim object, duplicate its amount of  $m$ -dim objects and add its amount of  $(m-1)$ -dim objects. In other words, each cell's value in Table 2 is determined by doubling its immediate left neighbor and adding the upper-left diagonal cell. For example, in the case of the 3 cells highlighted on the table:  $24 = 2 \cdot 6 + 12$ . Similar to  $f$ , the previously presented recursion,  $f'$  also has a geometric interpretation related to the construction of objects (see Chapter 1).

*Proof.* To generate the next dimension, we duplicate an object, perform a translation, and connect the duplicated points (while Coxeter's treatment remains concise [1], our following analysis provides a detailed argument). Each time a translation and connection are performed: a line is generated between a point and its duplicate, points from one line are connected to points from its duplicate to form a square, points from one square are connected to points from its duplicate to form a cube, and so on. Thus, from every  $m$ -dim object a new  $(m + 1)$ -dim object is additionally generated in the connection.

The duplication and connection are described by the recursion: double the number of  $m$ -dim objects and add the number of  $(m - 1)$ -dim objects to account for the connection process.

Thus, now we show that the derived formula is a solution for the recursion  $f'$  when  $m \leq n$ . It is clear that base case  $f'(0, 0) = 1 = P(0, 0)$  holds. Then, consider  $1 < n$  and assume that  $f'(k, m) = P(k, m)$  for every  $k < n$ . In the case  $m = 0$ , if we plug in  $P(n, m)$  into the recursion we obtain

$$\begin{aligned} f'(n, 0) &= 2 \cdot f'(n - 1, 0) + f'(n - 1, -1) \\ &= 2 \cdot \binom{n - 1}{0} \cdot 2^{(n-1)-0} + 0 \\ &= \binom{n - 1}{0} \cdot 2^{n-0} \\ &= \binom{n}{0} \cdot 2^{n-0} \\ &= P(n, 0). \end{aligned}$$

While in the general case we obtain

$$\begin{aligned} f'(n, m) &= 2 \cdot f'(n - 1, m) + f'(n - 1, m - 1) \\ &= 2 \cdot \binom{n - 1}{m} \cdot 2^{(n-1)-m} + \binom{n - 1}{m - 1} \cdot 2^{(n-1)-(m-1)} \\ &= \binom{n - 1}{m} \cdot 2^{n-m} + \binom{n - 1}{m - 1} \cdot 2^{n-m} = \left[ \frac{(n - 1)_m}{m!} + \frac{(n - 1)_{m-1}}{(m - 1)!} \right] \cdot 2^{n-m} \\ &= \left[ \frac{(n - 1)_m}{m!} + \frac{m \cdot (n - 1)_{m-1}}{m \cdot (m - 1)!} \right] \cdot 2^{n-m} = \frac{(n - 1)_m + m \cdot (n - 1)_{m-1}}{m!} \cdot 2^{n-m} \\ &= \frac{[(n - 1) - m + 1] \cdot (n - 1)_{m-1} + m \cdot (n - 1)_{m-1}}{m!} \cdot 2^{n-m} \\ &= \frac{(n - m) \cdot (n - 1)_{m-1} + m \cdot (n - 1)_{m-1}}{m!} \cdot 2^{n-m} \\ &= \frac{n \cdot (n - 1)_{m-1}}{m!} \cdot 2^{n-m} = \frac{n_m}{m!} \cdot 2^{n-m} \\ &= \binom{n}{m} \cdot 2^{n-m} = P(n, m), \end{aligned}$$

thereby solving the recursion using the principle of induction [7]. □

We believe that the detailed explanations of the construction process assist in visualizing these high-dimensional elements, especially as an introduction to the topic. Once the mathematical concepts are internalized, we encourage readers to explore various projections of these objects to deepen their understanding of the spatial representations of these structures.

## 4 Applications

After fully grasping the concepts outlined in Chapter 3, we proceed to demonstrate various applications of the formula. We believe that our approach holds significant potential for modeling and solving problems not only in geometry but also in fields such as optimization, combinatorics, physics, and beyond.

## 4.1 Euler characteristic

A known property of polygons is the Euler characteristic [6], defined for 3-dim objects as the number of points, minus the number of lines, plus the amount of squares (using our name convention). For example, cubes have an Euler characteristic of  $8 - 12 + 6 = 2$ .

Given a  $n$ -dim object, let  $d_m$  be the amount of  $m$ -dim objects and  $E$  the Euler characteristic. We generalize the property similar to Schläfli [1]:  $E = d_0 - d_1 + d_2 - \dots \pm d_{n-3} \pm d_{n-2} \pm d_{n-1} = \sum_{m=0}^{n-1} (-1)^m d_m$ . If we apply the formula to the studied objects obtained by the construction described in Chapter 1 we arrive at

$$E_{obj} = \sum_{m=0}^{n-1} (-1)^m \cdot P(n, m) = \sum_{m=0}^{n-1} (-1)^m \cdot \binom{n}{m} \cdot 2^{n-m},$$

and if we complete the sum (from the fact that  $P(n, m)$  comes from the expansion of  $(2 + X)^n$  [1]) we obtain

$$E_{obj} = \left[ \sum_{m=0}^n \binom{n}{m} \cdot 2^{n-m} \cdot (-1)^m \right] - (-1)^n = (2 + (-1))^n - (-1)^n = 1 - (-1)^n.$$

Consequently, for every  $i \geq 0$ ,  $2i$  dimensional objects have an Euler characteristic of 2, whereas those of dimension  $2i + 1$  exhibit an Euler characteristic of 0 [1].

## 4.2 Measurements

Given an  $n$ -dim object, the distance  $a$  associated with each of the  $n$  translations is called length. In such an object, all lines measure  $a$  units, all squares measure  $a^2$  square units, and so on. The amount of  $m$ -dimensional space occupied by the object can be calculated by multiplying the number of objects by the space occupied by a single object:

$$M(n, m, a) = P(n, m) \cdot a^m = \binom{n}{m} \cdot 2^{n-m} \cdot a^m.$$

Similarly to the case of  $P(n, m)$ , this formula arises naturally from the terms of the expansion of  $(2 + a)^n$ . This observation further links the construction of these objects to the binomial theorem [5], exhibiting notable parallels between their geometric interpretations.

It is worth noticing that new units of measurement appear from the construction of higher-dimensional objects, for example  $cm^4$ . Or even new equivalencies like  $1 \text{ litre}^2 = 1000000 \text{ cm}^6$ . For example, a 5-dim object of length  $3cm$  occupies a 4-dim space of:  $\binom{5}{4} \cdot 2^{5-4} \cdot (3cm)^4 = 10 \cdot 81cm^4 = 810cm^4$ .

## 4.3 Similarity

Due to the construction, and the angles formed between the objects, all elements of the same dimension are similar to each other [2], meaning they all have the same shape. Given two  $n$ -dim objects, one with length  $a$  and the other with length  $a'$ , the ratio of similarity [2] is defined as  $k = a : a'$ . Consequently, all squares on both objects, and thus the total 2-dim space, scale with a ratio of  $k^2$ .

The generalization of this concept for comparing  $m$ -dim objects is straightforward: the ratio is simply a factor of  $k^m$ . For example, let  $\varphi$  and  $\phi$  be two objects of unknown dimensions. The only available information is that the three-dimensional spaces ratio between  $\varphi$  and  $\phi$  is  $1 : 8$ , and the length of  $\varphi$  is  $5cm$ . Using the generalization, we find that the ratio of similarity is  $k = 1 : 2$ , leading to a length of  $10cm$  for  $\phi$ .

Analogous to conic sections, one can consider the slicing or truncation of high-dimensional objects [1] with a plane or high-dimensional planes. This approach enables the study of numerous concepts from Euclidean geometry [2] within our framework, including congruence, Euclidean distance, Hamming distance, Thales's theorem, and many others.

## 4.4 Real dimensions

We can extend the formula  $P(n, m)$  to a domain of real positive inputs by considering the generalization of factorials using integrals as  $\Pi(n) = \int_0^\infty t^n e^{-t} dt$  [5]. Then, we arrive at

$$P(n, m) = \frac{\Pi(n) \cdot 2^{n-m}}{\Pi(m) \cdot \Pi(n-m)},$$

where  $n, m \in \mathbb{R}_0^+$  such that  $n \geq m$ . Given these restrictions, it follows that the output satisfies  $P(n, m) \geq 1$ .

While an interpretation and application of this formula have yet to be fully explored, we believe it may be linked to fractal dimensions [4], potentially offering real-world applications in fields such as biology and physics.

## 5 Conclusion and Future Directions

To summarize, we have outlined key concepts, identified fundamental patterns, and developed geometric interpretations to improve our understanding of the relationship between component numbers and structural dimensions. We believe our approach provides a detailed and valuable tool for mentally visualizing higher-dimensional objects.

Furthermore, our exploration has uncovered potential applications and suggested further avenues for investigation, including the examination of slicing objects and the study of fractional dimensions. Additionally, we propose the development of new constructions that differ from the one presented in this work, or study a similar analysis on other constructions proposed by Coxeter [1].

We hope this work serves as both an introduction to the subject and a valuable contribution to future research, paving the way for new applications in both theoretical and practical contexts.

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