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Combinatorial Properties of One-Dimensional Arrangements

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Abstract

Arrangements are an omni-present topic in computational geometry, since many problems in computer graphics and robotics reduce to the study of such sets. Motivated by two problems from these areas —more precisely from ray-tracing and assembly planning, we study in this paper the combinatorial structure of arrangements of segments on a line and of cones on a circle. We show that the numbers of such arrangements are respectively $\frac{1}{3} \frac{3}{5} \ldots \left(2n - 1\right)$ and $\frac{(2n)!}{n!}$, that the probabilities for the $i^{th}$ vertex of a random arrangement to be a beginning point are $1 - \left(i - 1\right)/\left(2n - 1\right)$ and $1/2$, and that the average numbers of segments or cones the $i^{th}$ vertex is contained in are $\left(1 - i\right)(2n - 1)/(2n - 1)$ and $(n - 1)/2$. In addition to providing results for the analysis of the ray tracing related and assembly sequencing problems, the constructions used to prove these results provide sampling schemes for generating random inputs usable to test and validate the correctness of programs manipulating arrangements.

Along with the derivation of these identities, we also point out connections between arrangements, sub-diagonal random walks and the ballot problem, as well as other integer sequences.

Topics: Combinatorics, Computational Geometry, Algorithms and Data Structures.

1 Introduction

1.1 Assembly sequencing and arrangements

Assembly sequencing is a domain of robotics the purpose of which is, given a collection of mechanical parts and a class of motions these parts can be moved by, to compute a way (if any) to get the single parts from the whole assembly. Analyzing the disassembly sequences of a given assembly can lead to substantial improvements in many departments. For example to check that the product is disassemblable, or to enable the designers to make sure that the parts which may be serviced often are easily accessible, or to anticipate a better recycling by clustering parts of the same material. Of major practical interest, it also turns out that assembly sequencing is a difficult algorithmic problem since it is in its more general form intractable —see e.g. [Natarajan, 1988]. Nevertheless, restricted yet interesting versions of the problem have been shown to have polynomial time and space algorithms.

For example, let us consider the case of planar polygonal assemblies where the only class of motions allowed is infinite translations and where each split results in two sub-assemblies —[Wilson and Latombe, 1994, Latombe et al., 1996]. The space of motions is described by the circle $S^1$ since a translation corresponds to a unit vector in the plane. Given any two parts, the set of directions along which one can be translated without colliding the other is described by a cone on the circle $S^1$. This is the well known Minkowski difference depicted on figure 1(b) ([Latombe, 1991]). The blocking relations for all the pairs of parts are thus described by $\binom{n}{2}$ cones. Put together, this arrangement of cones partitions $S^1$ into vertices and open cells as shown on figure 2(a). This arrangement is called the Non-Directional Blocking Graph or NDBG since it gives the blocking relations for any pair of parts and any direction. To each vertex of the arrangement corresponds a directed graph called Directional Blocking Graph with a vertex for each part and an edge between vertices $i$ and $j$ if part $i$ collides part $j$ when translated along this direction. A topological sorting of the strong connected components of this graph gives the removable sub-assemblies.

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along this direction. Starting with the full assembly, the disassembly algorithm consists in recursively removing translatable sub-assemblies with the previous scheme.

Performing a worst-case analysis of this algorithm is pretty easy. Indeed, the NDBG has at most \( O(n^2) \) vertices and each DBG has size \( O(n^2) \), which gives a space requirement of \( O(n^4) \). As for the time complexity of the recursive disassembly, it is \( O(n^5) \) since there are at most \( n \) levels of recursion, and that each level requires examining at most \( O(n^2) \)DBGs for which the reduced graph (graph of the strongly connected components) and a topological sorting have to be computed.

On the other hand, coming up with an average-case analysis is much more challenging. Firstly, a precise understanding of the combinatorics of cone arrangements is required. Secondly, some random graph structure is needed for the directional blocking graphs. The latter question is difficult since the number of edges of a DBG depends on the geometric information encoded in the relative position of the pairs of parts, which require some definition of random assemblies. This goes beyond the scope of this paper and we skip it. On the opposite, the former problem is more well defined and raises precise questions such as the generation of a random arrangement — see also [Zimmermann, 1994], the probability for a given vertex to be a beginning point or an endpoint of a cone, the average number of cones a given vertex of an arrangement is contained in, etc. These questions are addressed in section 3.

![Figure 1: (a) A simple assembly (b) The Minkowski difference](image1.png)

![Figure 2: (a) Arrangement of cones on \( S^1 \) (b) Arrangement of line segments](image2.png)

1.2 Ray tracing and clustering

Ray tracing is a technique from computer graphics that consists in computing views of scenes defined by geometric primitives. Very often these primitives are polygons defined by their geometry and color, a given object of the scene being defined by a set of such polygons. As an example, consider fig. 3(a) where the whole kitchen model consists of about 25,000 polygon, and objects such as the bowl on the table or the teapot are made of about 1000 polygons. To sketch the ray-tracing algorithm — see [Foley et al., 1990] for the details, let a ray be defined by a point and a direction in 3D. Rays are used to simulate the light received by the observer’s eye, so that the key operation of the whole algorithm consists in finding, for a given ray, the closest object hit in order to plot the corresponding color on the screen of the computer where the algorithm is run.

Reducing the number of ray-polygon intersection tests has ever been a challenging issue. The main paradigm consists in partitioning the volume containing the scene in order to test for intersection those polygons only stored in the voxels of the partition crossed by the ray of interest. An example of such partitioning, the so called uniform grid, is done with a 3D axis aligned grid — see [Cazals et al., 1995]
for a discussion of grid-like data structures. The problem of this approach is that whenever too many polygons fall in the same voxel, the spatial partitioning does not result in data partitioning so that the number of ray-polygon intersection tests is not reduced significantly. To remedy this problem, it was observed in [Cazals et al., 1995] that using uniform grids for densely populated areas of the scene called clusters could partially solve the problem. Examples of clusters are the neighborhoods of the bowl, teapot, or door knobs, and are depicted on fig. 3(a)(b). More precisely, a cluster is defined as a subset of objects whose projection along the three axes $x, y$ and $z$ is almost-connected. And since the projection of a polygon on a line is a line-segment, the clustering algorithm analysis turns out to be closely related to the combinatorics of line-segments arrangement — see fig. 2(b). Also, the results presented thereafter in section 2.3 were recently used in [Cazals and Sbert, 1997] in conjunction with integral geometry techniques to define statistics aiming at characterizing standard scenes types such as natural models, architectural scenes, etc.

Figure 3: (a) Kitchen model (b) Clusters: far view (c) Clusters: close view

1.3 Notations, previous work and results

In the circular and linear cases just described, since we are only interested in the topological properties of the arrangements, we map the $2n$ extremities of the $n$ cones or segments of interest onto the integers $1, 2, 3, \ldots, 2n$, which we note $1..2n$. Let $S_n$ and $R_n$ be the sets of all the arrangements of segments and cones of size $n$, and $s_n$ and $r_n$ their cardinalities. For every vertex of the arrangement, the statistics we are interested in are the following:

- its probability to be a beginning point or an end point,
- its overlap number, defined as the average of the number of cones or line segments it is contained in.

The cone and the segment starting at index $i$ and ending at index $j$ are noted $(i, j)$ and $[i, j]$ respectively. Also, the beginning and ending points of a cone or a segment are labeled B and E. For a particular arrangement $a \in R_n$ or $S_n$ and $i \in 1..2n$, we denote by $a[i]$ the operation that returns the letter B or E stored at slot $i$ (this is correct since we suppose that the segments extremities do not overlap). The symbol $1_x$ with $x$ a boolean expression is the indicator of $x$, i.e. takes the value 1 if $x$ is true, 0 else.

At last, for a given slot, the number of B letters, the number of E letters and the overlap number are defined as follows:

$$\beta_i^{(n)} = \sum_{a \in S_n} 1_{a[i]=B}, \quad \epsilon_i^{(n)} = \sum_{a \in S_n} 1_{a[i]=E}, \quad \tau_i^{(n)} = \sum_{s \in a, s \in a} 1_{i \in s}.$$ 

The corresponding vectors for all the slots $1..2n$ are respectively $\vec{\beta}_n, \vec{\epsilon}_n, \vec{\tau}_n$. For $n = 2$ e.g., we have $(\vec{\beta}_2 = [3, 2, 1, 0], \vec{\epsilon}_2 = [0, 1, 2, 3], \vec{\tau}_2 = [0, 2, 2, 0])$.

Before proceeding, let us review the previous work. Actually, the numbers $s_n$ and $r_n$ have already appeared in the literature under several forms, in particular in the work of [Touchard, 1950] and [Riordan, 1975] related to the folding-stamps problems. In [Touchard, 1950] [section 4], it is shown that $s_n = 1.3\ldots(2n - 1)$. In [Riordan, 1975], it is stated that the number of pairings of $2n$ points on a circle is also $1.3\ldots(2n - 1)$, and the figure of page 216 –reproduced on the bottom part of our figure 4– shows a mapping from an arrangement on the circle to an arrangement on the line. Indeed, consider a labeling of the points on the circle, say in counterclockwise order. Any pairing $(i, j)$ on the
circle corresponds to the line segment \([i, j]\) on the line, and conversely. At last, a look in the very nice book [Sloane and Plouffe, 1995] shows that sequence \(s_n\) has also long been known in relation with the expression of Wallis integrals.

An interesting note from [Riordan, 1975] is the relation between the number of pairings on a circle and the Catalan numbers: pairings where the chords are not allowed to intersect give rise to the Catalan numbers \(C_n = \binom{2n}{n}/(2n+1)\), while pairings with crossings between the chords lead to \(s_n\). The author also cites a correspondence between the Catalan numbers and the ballot problems also known as the sub-diagonal random walks problem ([Comtet, 1974, Yaglom and Yaglom, 1964, Knuth, 1973]). Apart from a detailed analysis of the statistics listed above, an interested result of this paper is to show that exactly as in the case of the Catalan sequence \(C_n\), the sequences \(s_n\) and \(r_n\) can also be viewed in terms of random walks.

The overview of the paper is the following. In section 2 and 3 respectively we study the planar and circular cases. In section 4 we conclude and list interesting problems.

\section{Planar case}

\subsection{Number of arrangements}

As indicated in section 1, the number of line-segment arrangements is given by \(s_n = (2n-1)(2n-3) \ldots 3.1\). We first give a direct proof of this formula based on an inductive construction of all arrangements of size \(n\). This construction will be used in section 2.2. We then establish another expression for the number \(s_n\) based on a sub-diagonal random walk.

\begin{figure}[h]
  \centering
  \includegraphics[width=\textwidth]{pairings_arrangements.png}
  \caption{Pairings and arrangements}
\end{figure}

\subsection{Incremental construction of the arrangements}

Suppose we are given the set \(S_n\) of all arrangements of size \(n\) and that we want to construct \(S_{n+1}\) from it. We just have to find a match for the endpoint located at \(2n+2\). Let \(i \in 1..2n+1\) be this match. If \(i = 2n+1\), the newly added segment and the \(n\) previous segments do not overlap. If \(i < 2n+1\), we are left \(2n\) slots in \(1..2n+1\). But since the configurations with \(n\) segments were defined in \(1..2n\), to embed any of these configurations in \(1..2n+1 \setminus \{i\}\), we just have to shift by one unit to the right the extremities
of the segments whose index is $i$. We thus have $s_{n+1} = (2n + 1)s_n$. For an example, see figure 5(a) where the newly added segment is depicted in bold.

### 2.1.2 Arrangements and random walks

As already noticed, a segment can be viewed as two endpoints B and E. Constructing an arrangement of size $n$ therefore consists in assigning B and E letters to the 2$n$ slots and pairing each E with a B. This process can be seen as a random walk where at stage $i$ in 1,2$n$ an horizontal (vertical) move consists in putting a B (E) on slot $i$. But putting an E consists in closing a segment which means that some letter B must be available at a location $1 \leq k < i$. Moreover, if $j$ represents the number of such letters, the new E can be matched with any of them.

For an arrangement of size $n$, let $\varphi_n(i,j)$ the function equal to the number of ways we can be left with $j$ unmatched letters B after using $i$ of them, with $1 \leq i \leq n$ , $0 \leq j \leq i$. Since the first letter has to be a B, we start at $\varphi_n(1,1)$ and the total number of arrangements sought is $\varphi_n(n,0)$. But according to the above discussion, $\varphi$ satisfies the recurrence

$$\varphi_n(i,j) = \varphi_n(i+1,j+1) + j\varphi_n(i,j-1) \text{ with } 1 \leq i \leq n , \ 0 \leq j \leq i \quad (1)$$

Solving this recurrence is not that easy. The so-called reflection principle of André—[Comtet, 1974]—does not apply since at any time, we have to take into account the past history of the random walk. Also, the general machinery of bivariate generating functions—[Knuth, 1973]—exercise 2.2.1.4 and its solution—of little help. Indeed, let

$$G(x,z) = \sum_{i,j \geq 0} \varphi_{i,j} x^i z^j \quad \text{and} \quad g(x) = G(x,0) = \sum_{i \geq 0} \varphi_{i,0} x^i$$

Observing that $\varphi_{0,j} = \delta_{0,j}$ ($\delta_{\cdot,\cdot}$ being the Kronecker symbol), the above recurrence can be rewritten as follows using these generating functions:

$$G(x,z) = \frac{1}{xz}(G(x,z) - g(x)) + zg(x,z) + z^2 \frac{\partial G(x,z)}{\partial z}$$

which unfortunately does not give a straightforward asymptotic expansion of $G(x,z)$. We therefore derive a more ad-hoc solution.

**Theorem 1** The total number of configurations can be expressed as follows,

$$s_n = \sum_{r=1}^{n} a_r^{(n)} \varphi_n(n,r) = \sum_{r=1}^{n} a_r^{(n)} r! \quad (2)$$

where $\{a_r^{(p)}\}_{p=1..n,r=1..p}$ is a sequence of positive natural integers satisfying

$$a_j^{(p+1)} = a_{j-1}^{(p)} + \sum_{r=j}^{p} a_r^{(p)} r^{r-j+1}, \text{ with } a_1^{(1)} = 1 \text{ and } a_0^{(i)} = 0 \quad (3)$$

and $i!$ is the falling factorial $i(i-1) \cdots (i-j+1)$.

Proof: To see that $s_n = \sum_{r=1}^{n} a_r^{(n)} \varphi_n(n,r)$, observe that we can classify all the arrangements on the numbers of E letters they are finishing by, which is either 1, 2, ..., $n$. And $\varphi_n(n,r) = r!$ since there are respectively $r, r-1, \ldots , 1$ choices to close the 1$^{st}$, 2$^{nd}$, ..., $r^{th}$ segments corresponding to the $r$ letters E left. We therefore address the computation of the numbers $\{a_r^{(p)}\}_{p=1..n,r=1..p}$.

First observe that at the rank $p+1$, to compute $s_{p+1} = \sum_{r=1}^{p+1} a_r^{(p+1)} \varphi_{p+1}(p+1,r)$ one must beforehand have computed the sequence $\{a_r^{(p)}\}_{r=1..p}$. Indeed, before having used $p+1$ letters B, one must have used $p$ of them. Therefore,

$$\varphi_{p+1}(1,1) = a_1^{(p)} \varphi_{p+1}(p,1) + a_2^{(p)} \varphi_{p+1}(p,2) + \ldots + a_p^{(p)} \varphi_{p+1}(p,p)$$
But a straightforward computation leads to
\[ \varphi_{p+1}(p, j) = \varphi_{p+1}(p, j + 1) + \sum_{r=1}^{j} j_r \varphi_{p+1}(p + 1, j + 1 - r), \] (4)
that is
\[ \varphi_{p+1}(1, 1) = \sum_{j=1}^{p} a_j^{(p)} \left[ \varphi_{p+1}(p + 1, j + 1) + \sum_{r=1}^{j} j_r \varphi_{p+1}(p + 1, j + 1 - r) \right] \] (5)
Expanding and diagonally rearranging the terms yields to theorem 1. □

The table below shows the values of the previous sequences up to \( n = 10 \):

<table>
<thead>
<tr>
<th>( p )</th>
<th>( {a_i^{(p)}} )</th>
<th>( s_p )</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>1</td>
<td>1</td>
</tr>
<tr>
<td>2</td>
<td>1 1</td>
<td>3</td>
</tr>
<tr>
<td>3</td>
<td>3 3 1</td>
<td>15</td>
</tr>
<tr>
<td>4</td>
<td>15 15 6 1</td>
<td>105</td>
</tr>
<tr>
<td>5</td>
<td>105 105 45 10 1</td>
<td>945</td>
</tr>
<tr>
<td>6</td>
<td>945 945 420 105 15 1</td>
<td>10395</td>
</tr>
<tr>
<td>7</td>
<td>10395 10395 4725 1260 210 21 1</td>
<td>135135</td>
</tr>
<tr>
<td>8</td>
<td>135135 135135 62370 17325 3150 378 28 1</td>
<td>2027025</td>
</tr>
<tr>
<td>9</td>
<td>2027025 2027025 945945 270270 51975 6930 630 36 1</td>
<td>34459425</td>
</tr>
<tr>
<td>10</td>
<td>34459425 34459425 16216200 4729725 945945 135135 13860 990 45 1</td>
<td>654729075</td>
</tr>
</tbody>
</table>

Figure 6: Numerical values up to \( n = 10 \)

Some interesting facts are the following:
1. From the above recurrence, \( s_n \) is obviously super-factorial since \( n! \) is just the number of ways to close the \( n \) opened segments when the first \( n \) letters used are B,
2. We have \( a_1^{(p+1)} = s_p \). Indeed, on figure 5(b), the number of ways to reach point \( O_3 \) is exactly the number of ways to reach point \( O_1 \): while at the former the only possibility we have is to use a letter B,
3. Also, \( a_1^{(1)} = a_2^{(1)} \). On figure 5(b), this means that each path that joins \( O_2 \) gives rise to a path to \( O_3 \) and a path to \( O_4 \),
4. From [Sloane and Plouffe, 1995], it appears that the sequence \( \{s_p\}_{p=1,2,3,...} \) (M3002) has long been known as the 'sequence of double factorials' introduced to make easier the expression of Wallis integrals.

2.2 Beginning points and Endpoints
For any arrangement, the first letter must be a B while the last must be a E. The following theorem actually proves that there is a continuous variation over the indices of the vertices:

**Theorem 2** The number \( \beta_i^{(n+1)} = \sum_{a \in S_{n+1}} 1_{a[i]=B} \) satisfies the recurrence relation
\[ \forall \ i \in 1..2n+1 \ \beta_i^{(n+1)} = (i-1)\beta_i^{(n)} + s_n + (2n + 1 - i)\beta_{i+1}^{(n)}, \text{ and } \beta_{2n+2}^{(n+1)} = 0 \] (6)
and the corresponding closed forms are
\[ \forall \ i \in 1..2n \ \begin{cases} \beta_i^{(n)} = (2n - i)s_{n-1} - s_n = (i-1)s_{n-1} \\ \beta_i^{(n)} = s_n - \beta_i^{(n)} = (i-1)s_{n-1} \end{cases} \] (7)

The probability for the \( i^{th} \) vertex of a random arrangement to be a beginning point is therefore \( 1 - (i-1)/(2n-1) \)
Proof: To compute $\beta_i^{(n+1)}$, let us consider the construction of section 2.1.1: once we have fixed the slot $i \in 1..2n+1$ to be matched by the E located at $2n+2$, the number of letters B located in $S_n$ at the indices $1..2n$ are reported on $1..i-1 \cup i+1..2n+1$, and of course, the slot $i$ is credited of $s_n$ letters B. Thus, we have

$$\beta_{n+1} = \begin{bmatrix} s_n & \beta_1^{(n)} & \beta_2^{(n)} & \ldots & \beta_{2n}^{(n)} & 0 \end{bmatrix}^{i=1} + \begin{bmatrix} \beta_1^{(n)} & s_n & \beta_2^{(n)} & \ldots & \beta_{2n}^{(n)} & 0 \end{bmatrix}^{i=2} + \ldots + \begin{bmatrix} \beta_1^{(n)} & \beta_2^{(n)} & \beta_3^{(n)} & \ldots & s_n & 0 \end{bmatrix}^{i=2n+1}$$

The summation column by column gives equation (6). To get the closed form of $\beta_i^{(n)}$, let’s set up the recurrence hypothesis $\beta_i^{(n)} = s_{n-1}(2n-i)$ for $i = 1..2n$. With equation (6), we get for any $i \in 1..2n+1$ (notice that the indexes of $\beta_i^{(n)}$ are violated for $\beta_0^{(n)}$ and $\beta_{2n+1}^{(n)}$, but it does not matter since these terms are multiplied by null factors)

$$\beta_i^{(n+1)} = (i-1)s_{n-1}(2n-i+1) + s_n + (2n-i+1)s_{n-1}(2n-i) = s_n + s_{n-1}(2n-i+1)(2n-i+1)$$

But $s_n = s_{n-1}(2n-i)$, which completes the proof for $i = 1..2n+1$. The case $i = 2n+2$ is trivial. As for $\varepsilon_i^{(n)}$, we of course have $\beta_i^{(n)} + \varepsilon_i^{(n)} = s_n.$

As for the probability for the $i^{th}$ vertex of a random arrangement to be a beginning point, it’s the previous value divided by the number of arrangements, which is $1 - (i-1)/(2n-1)$ $\square$

2.3 Overlap number

For a given vertex, this number is defined as the average over all the possible arrangements of the number of segments it is contained in. We have the following theorem:

**Theorem 3** The number $\tau_i^{(n)} = \sum a \in S_n 1_{a \subseteq s}$ satisfies the recurrence relation

$$\forall i \in 1..2n+1 \quad \tau_i^{(n+1)} = (i-1)(\tau_{i-1}^{(n)} + \varepsilon_i^{(n)} + (2n+2-i)\tau_i^{(n)} + \tau_{2n+2}^{(n+1)} = 0 \quad (9)$$

More precisely, $\forall i \in 1..2n \quad \tau_i^{(n)} = (1-i)(i-2n)s_n$. Thus, the average overlap number of the $i^{th}$ vertex is $\frac{(1-i)(i-2n)}{2n-1}$.

Proof: Here again, we start with a recurrence coming from the incremental construction of the arrangements. We have seen that fixing the slot to be matched by the letter E located at $2n+2$, say the $i^{th}$ one, was giving rise to $s_n$, configurations. See figure 7 for the case $n+1 = 3$ and $i = 2$. In fact, there are three different cases:

- if $i = 1$, all the vertices of the segments from $S_n$ embedded in $2..2n+1$ are covered by the newly added segment,
- if $i = 2n+1$, neither of the segments from $S_n$ and the newly added segment overlap,
- the tricky case is if $i > 1$ and $i < 2n+1$. As we have seen, all the vertices of segments from $S_n$ the index of which were $i$ are shifted of one unit on the right side. Therefore, index $i$ is not only covered by the segments from $S_n$ that were covering it, but is also covered by all the segments of $S_n$ that were ending at index $i$. On figure 7 e.g., vertex 2 is now covered by the first segment of $S_2$ in the configuration 1.

We thus have the equality

$$\tau_{n+1} = \begin{bmatrix} 0 & \tau_1^{(n)} + s_n & \tau_2^{(n)} + s_n & \ldots & \tau_{2n-1}^{(n)} + s_n & \tau_{2n}^{(n)} + s_n & 0 \end{bmatrix}^{i=1} + \begin{bmatrix} \tau_1^{(n)} & \tau_2^{(n)} + \varepsilon_2^{(n)} & \tau_3^{(n)} + s_n & \ldots & \tau_{2n-1}^{(n)} + s_n & \tau_{2n}^{(n)} + s_n & 0 \end{bmatrix}^{i=2} + \ldots + \begin{bmatrix} \tau_1^{(n)} & \tau_2^{(n)} & \tau_3^{(n)} & \ldots & \tau_{2n}^{(n)} & 0 & 0 \end{bmatrix}^{i=2n+1}.$$
and the summation column by column gives equation (9). Let $\tau_n^{(n)} = (1 - i)(i - 2n)s_{n-1}$ be the recurrence hypothesis. Using equation (9) we get (transitions from line 1 to 2 and 2 to 3 corresponding to $s_n = (2n - 1)s_{n-1}$)

$$
\tau_i^{(n+1)} = (1 - i)[s_n + (2 - i)(i - 1 - 2n)s_{n-1} + (i - 1)s_{n-1} + (2n + 2 - i)(1 - i)(i - 2n)s_{n-1}]
$$

$$
= (i - 1)s_{n-1}[(2n - 1)(2n + 2 - i)]
$$

$$
= (1 - i)(i - 2n - 2)s_n
$$

(10)

which proves the theorem. □

It is interesting to note that this theorem, which was proved using the mechanical inductive construction of the arrangements, also admits a nicer direct proof:

Proof: The slot $i$ is covered by segments of the form $[j, k]$ for $j = 1..i - 1$ and $k = i + 1..2n$, and there are $(i - 1)(2n - i)$ different such segments. But each of these appears exactly $s_{n-1}$ times in the $s_n$ arrangements, since once we have fixed the segment $[j, k]$ we are left an arrangement of $n - 1$ segments.

Figure 7: Three configurations of $S_3$ with slot 2 used

3 Circular case

We now address the questions encompassed in the linear case for the circular one. Actually, due to the equivalent role played by any vertex on $S^1$, the situation is easier. We start with the number of arrangements:

**Theorem 4** The number $r_n$ of arrangements of $n$ cones on a circle is equal to $(2n)!/n!$

Proof: First observe that the set of all possible arrangements can be partitioned into subclasses corresponding to the different pairings of vertices ([Riordan, 1975]), and as recalled above this number is $(2n - 1)(2n - 3)\ldots 3.1$. But the pairing connecting vertices $i$ and $j$ leads to two different cones as indicated on figure 8(a). And since the $n$ pairings are independent, we get $r_n = (2n - 1)(2n - 3)\ldots 3.1.2^n = (2n)!/n!$

□

The numbers $r_n$ can also expressed as the result of a random walk as is section 2.1.2. Indeed, the only difference is that when closing a cone, we now have two possibilities as indicated on figure 8(a) instead of one, so that the recurrence is

$$
\varphi_n(i, j) = \varphi_n(i + 1, j + 1) + 2j\varphi_n(i, j - 1) \text{ with } 1 \leq i \leq n, \ 0 \leq j \leq i
$$

(11)

which we can solve as the linear one.

Also, the classification of the arrangements into pairings solves the question of computing the probability for a given vertex to be a beginning point or an endpoint. Because each pairing gives rise to two configurations as depicted on figure 8(a), these probabilities are equal to $1/2$. As for the overlapping number, we have the following theorem:

**Theorem 5** The average overlap number of a vertex in an arrangement of $n$ cones is $\frac{n-1}{2}$
Proof: Without loss of generality, let 0 be the index of the vertex we want to compute the overlapping number of, and 1, . . . , 2n − 1 the indices of the subsequent vertices in the counter-clock-wise order. Vertex 0 is covered by cones (i, j) with i < j and A ∈ (i, j). But vertex i is involved in exactly i − 1 cones, namely (i, 1), (i, 2), . . . , (i, i − 1), which is \( \sum_{i=1}^{2n-1} i - 1 = (2n-1)(2n-2)/2 \). Also cone (i, j) appears exactly \( r_{n-1} \) times in the \( r_n \) configurations, since the \( n-1 \) other cones are built from the \( 2n-2 \) remaining vertices. Therefore, the average over all the possible arrangements is \( (n-1)(2n-1)r_{n-1}/r_n = (n-1)/2 \). □

Figure 8: (a) Two cones for a pairing (b) Overlap numbers

4 Conclusions

In this paper we studied some combinatorial aspects of one-dimensional arrangements of segments on a line and of cones on a circle. We showed that the numbers of such arrangements are respectively 1.3.5 . . . (2n − 1) and \((2n)!/n!\), that the probabilities for the \( i^{th} \) vertex of a random arrangement to be a beginning point are \( 1 - (i - 1)/(2n - 1) \) and 1/2, and that the average numbers of segments or cones the \( i^{th} \) vertex is contained in are \( (1 - i)(i - 2n)/(2n - 1) \) and \((n - 1)/2\). The interests of the results are threefold.

Firstly, the analysis in the linear case is of major interest for computer graphics algorithms dealing with objects’ projections along lines. Also, the analysis presented about the circular case is the first step for an average case analysis of the \( NDBG \)-based algorithm computing assembly sequences in the simple case of polygons in the plane moved with infinite translations. Although this particular assembly sequencing problem might appear quite restrictive, it is actually one of the few for which it is reasonable to come up with an implementation for, so that any precise analysis would be of interest. Secondly, from the study of the arrangements combinatorial structure presented in this paper, it is easy to randomly generate such arrangements in order to test and validate geometric software. Thirdly, we pointed out connections between several problems where the sequences \( s_n \) and \( r_n \) arise, among which arrangements, random walks and ballot problems.

There are still many interesting issues remaining. In particular, getting higher moments for the statistics presented here is an open question. Also, coming up with results in the two-dimensional case would be very interesting, especially since the work done so far deals with arrangements of line in the plane, but not line segments (see [Edelsbrunner, 1986]).

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