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# On the Partition of Space by Hyperplanes 

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#### Abstract

We consider the problem of partitioning of space by hyperplanes that arises in many application areas, where the number of regions the space is divided into is required to be determined, such as speech/pattern recognition, various classification problems, data analysis. We obtain some relations for the number of divisions and establish a recurrence relation for the maximum number of regions in $d$-dimensional Euclidean space cut by $n$ hyperplanes. We also re-derive an explicit formula for the number of regions into which the space can be partitioned by $n$ hyperplanes.


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## 1. Introduction and preliminaries

The problem of partitioning of $d$-space (or, points in $d$-space) by a set of hyperplanes (a single hyperplane) has been the subject of multiple works. It is an interesting problem in its own right (see $[1,4,8,11]$ ), but the problem has been considered not only for its general mathematical interest. The question of how many such partitions are possible arises in various applications, for example, in the theory of pattern classification and machine learning $[3,5,10]$ which employs partitions using some number $k$ of parallel hyperplanes; in cluster analysis [6] dealing with the methods of partitioning of a set of objects on the basis of their characteristics or properties into clusters or groups so that the objects of a cluster are closely related according to certain criteria; data analysis and classification in which space partitioning by hyperplanes may be used together with linear regression techniques based on least absolute value estimates approach and its generalizations; in the theory of hybrid systems and control [9]. The problem has been generalized and analyzed in different ways, e.g. [7] considers the division of $d$-space by topological hyperplanes, subspaces of $\mathbb{R}^{d}$ or homeomorphs of it, that is topological equivalent to an ordinary straight hyperplane, and in [2] partitioning by the polynomial separating surfaces has been studied.

In this note we deal with the partitions of $\mathbb{R}^{d}$ that may naturally arise in the problems of clustering or classification considered in discrete form.

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Let us first introduce the notations. We use the term hyperplane in $\mathbb{R}^{d}$ to mean an affine subspace of codimension one in $\mathbb{R}^{d}$, i.e. a ( $d-1$ )-dimensional straight plane in $\mathbb{R}^{d}$ which need not necessarily pass through the origin. Let $\mathcal{H}$ be a family of $n$ distinct hyperplanes $\left\{\mathfrak{h}_{i}\right\}_{i=1, \ldots, n}$ in the $d$-dimensional space $\mathbb{R}^{d}$, where $d$ and $n$ are arbitrary positive integers. We denote by $G_{\notin \ell}(d, n)=G_{\mathfrak{h}_{1}, \mathfrak{h}_{2}, \ldots, \mathfrak{h}_{n}}(d, n)$ the number of regions into which the space $\mathbb{R}^{d}$ is partitioned by the hyperplanes $\mathfrak{h}_{1}, \mathfrak{h}_{2}, \ldots, \mathfrak{h}_{n}$, that is the number of connected components of the set $\mathbb{R}^{d} \backslash \mathscr{H}=\mathbb{R}^{d} \backslash\left(\mathfrak{h}_{1} \cup \mathfrak{h}_{2} \cup \ldots \cup \mathfrak{h}_{n}\right)$. We also set $G(d, n)=\max _{\mathcal{H}} G_{\mathscr{H}}(d, n)$.

Any nonvoid intersection of hyperplanes of the family $\mathscr{H}$ is referred to as an edge of the family. The edge of dimension $k$, which is a $k$-dimensional plane, is called a $k$-edge and denoted by $\mathcal{E}^{k}, k=0,1, \ldots, m$. We let $\mathcal{E}_{[t]}^{k}$ denote a $k$-edge made of intersections of any $j=1, \ldots, t$ hyperplanes, $\mathcal{E}_{[1, t]}^{k}$ a $k$-edge made of intersections of all of $\mathfrak{h}_{1}, \mathfrak{h}_{2}, \ldots, \mathfrak{h}_{t}$ hyperplanes, and $\mathcal{E}_{[1, t ; n]}^{k}$ a $k$-edge made of intersections of each $\mathfrak{h}_{i}$ with $\mathfrak{h}_{n}, i \neq n$. In these terms the space $\mathbb{R}^{d}$ can be considered as a $d$-dimensional edge of the family $\mathcal{H}$.

Each separating surface in $\mathbb{R}^{d}$, represented as a hyperplane, is given by a linear equation $\mathfrak{h}_{i}=\left\{x \in \mathbb{R}^{d} \mid a_{i} x=b_{i}\right\}$, or equivalently

$$
\begin{equation*}
\mathfrak{h}_{i}(x) \equiv \sum_{j=1}^{d} a_{i j} x_{j}-b_{i}=0, \quad i=1, \ldots, n \tag{1}
\end{equation*}
$$

Then the classification problem in discrete form can typically be formulated as follows. Given the probabilities $p_{\tau}$ of appearance of each of $N$ objects $\widetilde{x}_{1}=\left(x_{11}, \ldots, x_{1 d}\right), \ldots$, $\widetilde{x}_{\tau}=\left(x_{\tau 1}, \ldots, x_{\tau d}\right), \ldots, \widetilde{x}_{N}=\left(x_{N 1}, \ldots, x_{N d}\right)$ and information about belonging of each object to a certain class, one needs to distinguish $s$ classes $C_{1}, \ldots, C_{k}, \ldots, C_{s}$ in the space of parameters $\mathbb{R}^{d}$.

The rule that determines the assignment of the object $\widetilde{x}$ to a class can be defined as

$$
\widetilde{x} \in\left\{\begin{array}{ccc}
C_{1}, & \mathfrak{h}_{\alpha}(x) \geq 0, \alpha \in \mathbf{I}_{1} ; & \mathfrak{h}_{\beta}(x)<0, \beta \in \mathbf{J}_{1}  \tag{2}\\
C_{2}, & \mathfrak{h}_{\alpha}(x) \geq 0, \alpha \in \mathbf{I}_{2} ; & \mathfrak{h}_{\beta}(x)<0, \beta \in \mathbf{J}_{2} \\
\vdots & \vdots & \vdots
\end{array}\right.
$$

where $\mathbf{I}_{k}, \mathbf{J}_{k}, k=1, \ldots, s$, are the predefined index sets.
Hence, the points $\tilde{x}^{\prime}$ and $\tilde{x}^{\prime \prime}$ lie in different regions if and only if there exists an index $i$ for which $\mathfrak{h}_{i}\left(\tilde{x}^{\prime}\right)$ and $\mathfrak{h}_{i}\left(\tilde{x}^{\prime \prime}\right)$ have different signs. For this purpose the sign vector-function, the partition signature, $f: \mathbb{R}^{d} \rightarrow\{-,+\}^{n}$ defined as

$$
f_{i}(x) \in\left\{\begin{array}{lll}
-, & \text { if } & a_{i} x \leq b_{i} \\
+, & \text { if } & a_{i} x>b_{i}
\end{array}, \quad i=1, \ldots, n\right.
$$

can be employed for the analysis of the classification schemes with different rules.
Therefore, the problem of classification has two different formulations. Direct problem: find the minimum number of hyperplanes (1) that divide $\mathbb{R}^{d}$ into $s$ regions according to the rule (2); Inverse problem: determine the maximum number of regions into which $\mathbb{R}^{d}$ can be divided by means of $n$ hyperplanes (1) according to the rule (2).

## 2. Main Results

The main results of this paper are contained in the propositions presented below.
Theorem 1. Suppose that $\mathcal{H}(x)=\left\{\mathfrak{h}_{1}(x), \mathfrak{h}_{2}(x), \ldots, \mathfrak{h}_{n}(x)\right\}$ is a collection of hyperplanes in $\mathbb{R}^{d}$. Let $\mathscr{H}^{\prime} \subset \mathscr{H}$ be the partition of $\mathbb{R}^{d}$ by $n-1$ hyperplanes and $\mathscr{H}^{\prime \prime} \subset \mathcal{H}^{\prime}$ be the partition of $\mathbb{R}^{d-1}$ by $\ell$ hyperplanes, $\ell<n$. Then we have

$$
\begin{equation*}
G_{\not{H}}(d, n)=G_{\mathscr{H}^{\prime}}(d, n-1)+G_{\mathcal{H}^{\prime \prime}}(d-1, \ell) \leq G(d, n-1)+G(d-1, n-1) . \tag{3}
\end{equation*}
$$

Proof. Without loss of generality, we can assume that the coordinate system in $\mathbb{R}^{d}$ is chosen in such a way that $\mathfrak{h}_{n}(x)$ coincides with $\mathbb{R}^{d-1} \subset \mathbb{R}^{d}$ and is defined by the equation $x_{d}=0$. The hyperplanes $\mathfrak{h}_{1}(x), \ldots, \mathfrak{h}_{n-1}(x)$ divide $\mathbb{R}^{d}$ into $G_{\mathcal{H}^{\prime}}(d, n-1) \leq G_{\mathcal{H}}(d, n-1)$ regions. The hyperplane $\mathfrak{h}_{n}(x)$ divides some of these regions into two other regions. The boundaries of these latter regions are $(d-1)$-dimensional regions that lie in $\mathfrak{h}_{n}(x)$; or, if $\mathfrak{h}_{n}(x)$ is parallel to the all $\mathfrak{h}_{i}(x), i \neq n$, then the whole $\mathfrak{h}_{n}(x)$ is the boundary hyperplane. Hence, we have that the number of regions contributed by the hyperplane $\mathfrak{h}_{n}(x)$ coincides with the number of regions formed in $\mathfrak{h}_{n}(x)$ by some number $\ell$ of $k$-edges $\mathcal{E}_{[1, n-1 ; n], j}^{k}$, $j=1, \ldots, \ell, k \leq d-1$, of intersection of $\mathfrak{h}_{n}(x)$ with each $\mathfrak{h}_{i}(x), i=1, \ldots, n-1$, or, equivalently, generated by the "lines" of intersection of $\mathfrak{h}_{n}(x)$ with each $\mathfrak{h}_{i}(x)$; moreover, $\mathfrak{h}_{n}(x)$ contributes only one region if it is parallel to the all $\mathfrak{h}_{i}(x), i=1, \ldots, n-1$. The number of such lines is equal to $\ell \leq n-1$. Therefore, the number of new regions is $G_{\mathcal{H}^{\prime \prime}}(d-1, \ell) \leq G(d-1, n-1)$, from which the inequality (3) immediately follows. The proof is completed.

Corollary 1. Suppose that for the partition $\mathscr{H}(x)$ it holds that $G_{\mathcal{H}}(d, n)=G(d, n)$. Then we have

$$
\begin{equation*}
G(d, n) \leq G(d, n-1)+G(d-1, n-1) . \tag{4}
\end{equation*}
$$

Remark 1. Since $G_{\mathscr{H} "}(d-1,0)=1$, the relation (3) holds also in "parallel" case, that $i$ s, in the case when $\mathfrak{h}_{n}(x)$ is parallel to the all $\mathfrak{h}_{i}(x), i \neq n$.

In the next theorem we obtain the recurrence formula for the number of regions.
Theorem 2. Let $A=\left[a_{i j}\right], i=1, \ldots, n, j=1, \ldots, d$ be the matrix of coefficients of the equation (1). Further, suppose that the partition $\mathcal{H}(x)=\left\{\mathfrak{h}_{i}(x), i=1, \ldots, n\right\}$ of $\mathbb{R}^{d}$ is such that every set of $k$ rows, $k \leq d$, of $A$ has the rank $k$, i.e. $\operatorname{rank}_{k \leq d}\left[a_{i j}\right]=k$.

Then we have

$$
\begin{equation*}
G_{\mathcal{H}}(d, n)=G(d, n-1)+G(d-1, n-1), \tag{5}
\end{equation*}
$$

that is, $G_{\mathcal{H}}(d, n)$ takes on the maximal value.
Proof. We prove the theorem by induction over $n$. It is straightforward to check that for $n=1$ the statement of theorem is true, since $G(d, 1)=2$ and $G(d, 0)=G(d-1,0)=1$.

Suppose that the theorem holds true for the partition by $(n-1)$ hyperplanes. Now let $\mathfrak{h}_{1}(x), \mathfrak{h}_{2}(x), \ldots, \mathfrak{h}_{n}(x)$, where $\mathfrak{h}_{n}(x)=\mathbb{R}^{d-1}$, be the hyperplanes satisfying the theorem's
condition. Then it is obvious to see that the partition $\mathscr{H}^{\prime}(x)=\left\{\mathfrak{h}_{1}(x), \mathfrak{h}_{2}(x), \ldots, \mathfrak{h}_{n-1}(x)\right\}$ satisfies the condition of the theorem as well, so that by induction we have $G_{\mathscr{H}^{\prime}}(d, n-1)=$ $G(d, n-1)$. The equations of $k$-edges, $k \leq d-1, \mathfrak{h}_{i}^{\prime}(x)$ as the "lines" of intersection of $\mathfrak{h}_{i}(x)$ with $\mathfrak{h}_{n}(x), i \neq n$, has the following form

$$
\begin{equation*}
\mathfrak{h}_{i}^{\prime}(x) \equiv \sum_{j=1}^{d-1} a_{i j} x_{j}-b_{i}=0, \quad i=1, \ldots, n-1, \tag{6}
\end{equation*}
$$

regarding $\mathfrak{h}_{i}^{\prime}(x)$ as the hyperplanes in $\mathbb{R}^{d-1}$.
Now it is sufficient to show that the partition $\mathscr{H}^{\prime \prime}(x)=\left\{\mathfrak{h}_{1}^{\prime}(x), \mathfrak{h}_{2}^{\prime}(x), \ldots, \mathfrak{h}_{n-1}^{\prime}(x)\right\}$ of $\mathfrak{h}_{n}(x)=\mathbb{R}^{d-1}$ satisfies the condition of the theorem. Indeed, assume that the set of $k \leq d-1$ rows $\left(a_{\nu_{i}, 1}, a_{\nu_{i}, 2}, \ldots, a_{\nu_{i}, d-1}\right), i=1,2, \ldots, k$, of coefficients $\left\{a_{i j}\right\}$ in (6) has the rank $r<k$. Then we get that the set of rows $\left(a_{\nu_{i}, 1}, a_{\nu_{i}, 2}, \ldots, a_{\nu_{i}, d-1}, a_{\nu_{i}, d}\right), i=1,2, \ldots, k$, together with the row $(0, \ldots, 0,1)$ of coefficients of the equation $\mathfrak{h}_{n}(x)=0$ has the rank $r+1 \leq k$, less than the number of rows $k+1$, which contradicts the theorem's condition. Therefore, by the induction we again have that $G_{\not \mathscr{ユ}^{\prime \prime}}(d-1, n-1)=G(d-1, n-1)$, from which the statement of the theorem follows. The proof is completed.

Remark 2. The existence of the partition $\mathscr{H}$ that satisfies the condition of Theorem 2 is provided by and can be deduced from the fact that $\mathbb{R}^{d}$ contains infinitely many hyperplanes.

The following result is well-known $[1,4,12]$. Here, we give an alternative proof of this result by re-deriving the statement using the formula (5) of Theorem 2.

Theorem 3. It holds that

$$
\begin{equation*}
G(d, n)=\sum_{k=0}^{d}\binom{n}{k}, \tag{7}
\end{equation*}
$$

where $\binom{n}{k}=\frac{n!}{k!(n-k)!}, k \leq n$ and $\binom{n}{k}=0$ if $k>n$.
Proof. Consider the recurrence relation

$$
\begin{equation*}
G(d, n)=G(d, n-1)+G(d-1, n-1) \tag{8}
\end{equation*}
$$

for the maximum number of regions that can be obtained by partitioning of $\mathbb{R}^{d}$ by $n$ hyperplanes (see eq. (5)), together with the initial conditions

$$
\begin{equation*}
G(d, 0)=1, \quad d=1,2, \ldots . \tag{9}
\end{equation*}
$$

The number $G(d, n)$ is uniquely defined by (8), (9). Indeed, the values of $G(d, 0)$ are known for all $d$. Hence, if the value of $G(d, n)$ is known for all pairs $(d, n)$ for which $n=k-1$, then $G(d, k)$ can be found from (8) for any $d$.

Now it is sufficient to show that (7) satisfies (8) and (9).
Substituting (7) into (8), we get

$$
\begin{equation*}
\sum_{k=0}^{d}\binom{n}{k}=\sum_{k=0}^{d}\binom{n-1}{k}+\sum_{k=1}^{d}\binom{n-1}{k-1} . \tag{10}
\end{equation*}
$$

The equality (10) is in fact the identity which can be established using the combinatorial formula

$$
\binom{n}{k}=\binom{n-1}{k}+\binom{n-1}{k-1}
$$

It is easy to see that (7) satisfies the condition (9), which completes the proof.

## 3. Conclusions and Further Work

In this paper we have investigated the problem of space partitioning by hyperplanes. We obtained some relations regarding the number of divisions and derived the recurrence formula for the maximum number of regions in $d$-dimensional Euclidean space cut by $n$ hyperplanes in arbitrary position. An explicit formula for the number of regions into which the space can be partitioned by $n$ hyperplanes can also be found. Using the recurrence formula, we gave an alternative derivation of the well known explicit formula for the number of regions that $n$ hyperplanes in general position divide the $d$-dimensional space. In our subsequent works we plan to continue studying the combinatorics of hyperplane configurations in $\mathbb{R}^{d}$. Further work may address the investigation of the number of $k$-edges of hyperplane configurations, the study of partitioning problem of $\mathbb{R}^{d}$ by hyperplanes with the use of Möbius function defined on finite posets connected with the rank of homology group $H_{d}\left(\mathbb{C}^{n} \backslash \mathscr{H}\right)$, the study of separating subgroups of the homology group $H_{d}\left(\mathbb{C}^{n} \backslash \mathcal{H}\right)$, and finding the formulas for Möbius function of posets of edges of hyperplane configurations. These results may find applications in toric geometry, singularity theory and multidimensional residues, and the theory of hypergeometric functions. Yet another direction of future research may be related to the application of the obtained results in data analysis and classification, and experimental data processing using linear regression models, and in the problems of identifiability of linear dynamical systems in state space.

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