

ON RANDOM CONVEX HULLS

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Let S^{d-1} be the sphere $x_0^2 + \dots + x_d^2 = 1$ in R^d . Let $A, B \subseteq S^{d-1}$ be two finite sets of "data points" given by some experiment. The perceptron learning procedure (see [2]–[5]) is an algorithm for finding a linear subspace L of R , of dimension $d-1$, which separates A from B , whenever such a subspace exists. The following problem arises. Suppose that the procedure succeeded and found a separating L ; is this a significant fact or merely a chance phenomenon? We will show that, under appropriate conditions, the existence of L is very improbable, whence the result can be quite significant.

In general, the answer depends on the probability distribution (in S^{d-1}) which rules the choice of the points. If, e.g., there was a constraint $A \cup B \subseteq \{(x_1, \dots, x_d): x_1 \geq \frac{1}{2}\}$, then the cardinalities of A and B and their relative sizes are essential and little can be said except to suggest that the probability in question could perhaps be estimated by a Monte Carlo test. However, we will settle the problem for the case when the probability distribution of the points of A and B is uniform over S^{d-1} , or, more generally, when

(*) every $(d-2)$ -dimensional great circle in S^{d-1} has probability 0 and the probability is invariant under reflection of S^{d-1} through its center.

A natural case of the problem, namely when B is empty, was solved under those conditions by J. G. Wendel [6]. Our theorem is an easy corollary of his, but, for convenience of the reader, we produce a complete proof (which is somewhat different from his).

THEOREM. *If a probability measure μ in S^{d-1} satisfies (*) and x_1, \dots, x_n are chosen in S^{d-1} independently according to μ , then the probability p that $\{x_1, \dots, x_i\}$ is linearly separable from $\{x_{i+1}, \dots, x_n\}$ does not depend on i and is given by*

$$(1) \quad p = \frac{1}{2^{n-1}} \sum_{k=0}^{d-1} \binom{n-1}{k}.$$

Moreover the probability that the center 0 of S^{d-1} is not in the convex hull of $\{x_1, \dots, x_n\}$ also equals p .

Remark. Of course, by (1) and the theorem of de Moivre–Laplace,

$$p \approx \Phi\left(\frac{d-n/2}{\sqrt{n/2}}\right),$$

where Φ is the standard normal distribution. Hence p tends to 0 iff $\frac{\sqrt{n}}{2} - \frac{d}{\sqrt{n}}$ tends to ∞ .

Proof of the Theorem. The fact that the probability does not depend on i follows immediately from three obvious facts:

(a) $\{x_1, \dots, x_i\}$ is linearly separable from $\{x_{i+1}, \dots, x_n\}$ iff the set $\{-x_1, \dots, -x_i, x_{i+1}, \dots, x_n\}$ is linearly separable from the empty set \emptyset .

(b) The distribution of $\{-x_1, \dots, -x_i, x_{i+1}, \dots, x_n\}$ is the same as the distribution of $\{x_1, \dots, x_n\}$. (This follows from (*).)

(c) 0 is not in the convex hull of $\{x_1, \dots, x_n\}$ iff $\{x_1, \dots, x_n\}$ is linearly separable from \emptyset .

At this point the theorem of Wendel [6] yields (1), but, for convenience of the reader, we continue the proof (in a slightly different way).

Let $S \subseteq \{1, \dots, n\}$ and p be the probability that $\{x_i: i \in S\}$ is linearly separable from $\{x_i: i \notin S\}$. By the above remarks p does not depend on S . Hence

$$p = \Pr[(x_1, \dots, x_n, S) \text{ is such that } \{x_i: i \in S\} \\ \text{is linearly separable from } \{x_i: i \notin S\}],$$

where x_i are chosen as in the Theorem and S is chosen uniformly among the 2^n subsets of $\{1, \dots, n\}$. Hence, by (*), the equation (1) of the Theorem follows immediately from the theorem of Fubini and the following Lemma.

LEMMA. Let $x_1, \dots, x_n \in S^{d-1}$ be in general position (i.e., no k of them are in the same $(k-2)$ -dimensional great circle) and let $f_d(n)$ be the number of partitions of $\{x_1, \dots, x_n\}$ into two sets linearly separable from each other. Then

$$(2) \quad f_d(n) = \sum_{k=0}^{d-1} \binom{n-1}{k}.$$

(This Lemma is an analog of Proposition 2.7 of T. H. Brylawski [1].)

Proof. If $n \leq d$, then it is easy to check that

$$(3) \quad f_d(n) = 2^{n-1}.$$

Let S be the $(d-2)$ -dimensional great circle in a hyperplane perpendicular to the vector x_1 . Let X be the projection of $\{x_2, \dots, x_n\}$ into S (along meridians

with pole x_1). Now we will prove that

$$(4) \quad f_d(n) = f_d(n-1) + f_{d-1}(n-1).$$

In fact $f_d(n)$ equals the number of all the partitions of $\{x_2, \dots, x_n\}$ which are linearly separable plus the number of those partitions of $\{x_2, \dots, x_n\}$ which are separable by a hyperplane containing 0 and x_1 . Clearly the first number is $f_d(n-1)$ and the second equals the number of partitions of X which are linearly separable, which is $f_{d-1}(n-1)$. This proves (4).

It is routine to derive (2) from (3) and (4).

PROBLEM (P 1298). Suppose that μ is uniform over S^{d-1} . Can one estimate the expected value of the volume of the convex hull of $\{x_1, \dots, x_n\}$?

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