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Abstract

An universe U, a hypothesis space $S = \{L_1, \dots, \}$ of susbets of U. S can be infinite. An unknown susbet L_* . After some queries, learn L_* .

Quesries include:

Memebership query:

Equivalence query

Subset query

Supserset query

Disjointness query

Exhaustiveness query

When an input of a query is a subset of U, then it must be in S. poker hand

a card: a number and a suit(S,H,D,C))

A hand is a set of five cards, without any order

A pair of hands is an ordered pair of hands with no card in common

The universe U is the set of all pairs of hands.

Exact identification: after some queries to find an index i such that the target notion is exactly L_i in S.

Probablistic identification (by L.G. Valiant 1988).

a disributuion D on U. Pr(x) is the probability of element x wrt D

Sampling oracle EX() which has no input. When EX() is called it returns an element x with an identification of whether x is in the target set

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difference of two sets L_1, L_2 : $d(L_1, L_2) := \sum_{x \in L_1 \otimes L_2} Pr(x)$

two parameters ϵ accuracy, δ confidence.

probably approximately correctly(pac) identification: always halts and output an index i such that $Pr(d(L_*, L_i) \leq \epsilon) \geq 1 - \delta$

pac identification is used if $\mathrm{EX}(\)$ is available. Otherwise we use exact identification

Equivalence query

exhaustive search: enuerate indeices $i = 1, \dots$, quering each L_i until one gets an answer of yes, and halts at this point.

If hypothesis space is too large, then exhaustive search may use too many times of equivalence queries.

U binary strings of length n. S = U. L_* is a unknown string want to learn. Only equivalece, membership, subset, disjoint queries are available, then in the worst case, may need $2^n - 1$ queries.

Majority strategy (me want to learn some thing from an adversary). an input of the equivalent query can be not a hypothesis. Then $\log N$ queries is sufficient.

Lower bound, and techniques.

input must be a hypothsis.

hypothesis L_1, \dots, L_N such that $L_i \cap L_j = L_{\cap}$ for $i \neq j$.

To learn L_{\cap}

Stochastic equivalence

Inequality

$$\sum_{i=1}^{n} (1 - \epsilon)^{q_i} \le \sum_{i=1}^{n} e^{-\epsilon q_i}$$