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CSCI4230 Computational Learning Theory
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Based on Rocco Servedio's notes

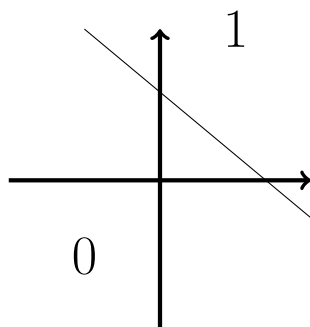
Notes 3: Winnow algorithms

1. LINEAR THRESHOLD FUNCTIONS (LTF)

Let $w \cdot x \stackrel{\text{def}}{=} \sum_{1 \leq i \leq n} w_i x_i$ (inner product between $w \in \mathbb{R}^n$ and $x \in \mathbb{R}^n$)
An **LTF** $f: \mathbb{R}^n \rightarrow \{0, 1\}$ has the form

$$f(x) = \begin{cases} 1 & \text{if } w \cdot x \geq \theta \\ 0 & \text{otherwise} \end{cases}$$

for some weight vector $w \in \mathbb{R}^n$ and threshold $\theta \in \mathbb{R}$



Every disjunction is LTF, e.g. for $x \in \{0, 1\}^n$

$$x_1 \vee x_2 \vee \bar{x}_3 \text{ true} \iff x_1 + x_2 + (1 - x_3) \geq 1 \iff x_1 + x_2 - x_3 \geq 0$$

Every 1-DL is LTF (why?)

2. WINNOW1

Update weights **multiplicatively**

Learn **k -sparse** (i.e. involves k literals) monotone DNF using LTF hypothesis

$O(k \log n)$ mistakes

When k really small (e.g. 5) and n really big, $O(k \log n)$ is better than n (in Elimination Algorithm)

Winnow1

Initialize: $w_1 = \dots = w_n = 1$, θ fixed to be n
On input x , output hypothesis $h(x) = \mathbb{1}(w \cdot x \geq \theta)$ and get $c(x)$
False positive ($h(x) = 1, c(x) = 0$): For every i s.t. $x_i = 1$
Set $w_i = 0$ (demotion, in fact elimination)
False negative ($h(x) = 0, c(x) = 1$): For every i s.t. $x_i = 1$
Double w_i (promotion)

In fact non-zero w_i is always $1, 2, 4, 8, \dots$ (power of 2)

Observation: no w_i is ever negative

Observation: in every promotion step, some x_i in c has its w_i doubled

Claim: Each w_i always $< 2n$

Reason: When w_i is doubled, x_i must be 1 and $w \cdot x < n$

Claim: #promotion steps $\leq k \log(2n)$

Reason: No x_i in c is ever eliminated, and is promoted $\leq \log(2n)$ times (k many such x_i)

Lemma 2.1. #elimination steps \leq #promotion steps + 1

Proof. Let W = total weight = $\sum_{1 \leq i \leq n} w_i$ (initially n)

Each elimination step W decreases by $w \cdot x \geq n$ (w_i becomes 0 iff $x_i = 1$)

Each promotion step W increases by $w \cdot x < n$ (w_i doubled iff $x_i = 1$)

After e elimination steps and p promotion steps, $0 \leq W \leq n - en + pn$, so $e \leq p + 1$. □

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Winnow1 makes $\leq 2k \log(2n) + 1 = O(k \log n)$ mistakes on k -sparse monotone DNF

Variation: During promotion, instead of doubling w_i , can multiply w_i with $\alpha > 1$; Threshold θ need not be n ; See Littlestone if interested

Can Winnow1 learn non-monotone DNF? (False positive kills Winnow1 e.g. $c(x) = \bar{x}_1, x^1 = 11$)
 Or LTF with nonnegative weights? (Not without new ideas such as Winnow2)

3. WINNOW2

Can assume threshold $\theta = 1$ (by rescaling w)

An LTF $x \in \{0, 1\}^n \mapsto \mathbb{1}(w \cdot x \geq 1)$ is δ -separated if

$$\forall x \in \{0, 1\}^n, \quad \text{either } w \cdot x \geq 1 \text{ or } w \cdot x \leq 1 - \delta$$

e.g. r -out-of- k threshold function

$$c(x) = \mathbb{1}(x_{i_1} + \dots + x_{i_k} \geq r) = \mathbb{1}\left(\frac{1}{r}x_{i_1} + \dots + \frac{1}{r}x_{i_k} \geq 1\right)$$

is $1/r$ -separated

Winnow2

Initialize: $w_1 = \dots = w_n = 1$, θ fixed to be n , α fixed to be $1 + \delta/2$

On input x , output hypothesis $h(x) = \mathbb{1}(w \cdot x \geq \theta)$ and get $c(x)$

False positive ($h(x) = 1, c(x) = 0$): For every i s.t. $x_i = 1$

Divide w_i by α (demotion)

False negative ($h(x) = 0, c(x) = 1$): For every i s.t. $x_i = 1$

Multiply w_i by α (promotion)

Claim 3.1. Winnow2 can learn δ -separated LTF with nonnegative weights $w \in \mathbb{R}^n$ with $O((\log n)\delta^{-2} \sum_{1 \leq i \leq n} w_i)$ mistakes

Proof in Littlestone §5

k -sparse monotone DNF are 1-out-of- k threshold functions

Winnow2 also learns k -sparse monotone DNF with $O(k \log n)$ mistakes (direct proof in Blum §3.2)

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