



Sensitivity conjecture

... or a 1-page proof of an ≈ 30 y.o. problem

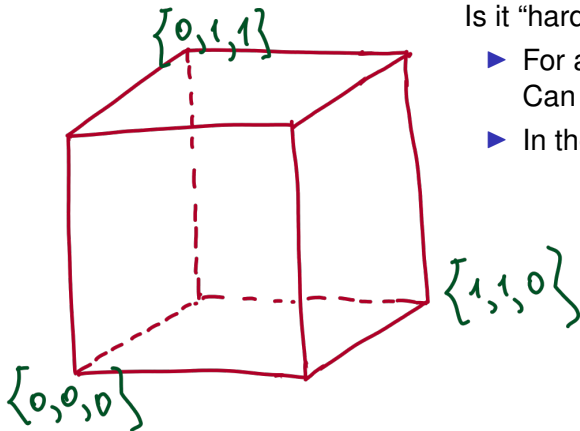
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Sensitivity

Let $f: \{0, 1\}^n \rightarrow \{0, 1\}$ be a Boolean function.



Is it “hard” to flip its value?

- For a *fixed* input:
Can we change it by flipping only one bit?
- In the worst case?

Sensitivity

Let $f: \{0, 1\}^n \rightarrow \{0, 1\}$ be a Boolean function.
For $x \in Q$ let x^J be a vector with all x_j flipped.

$$x = (x_1, x_2, x_3, x_4, x_5, x_6, x_7, x_8)$$
$$x^{\{3,6\}} = (\dots, \bar{x}_3, \dots, \bar{x}_6, \dots)$$

► For a fixed input $x \in Q^n$:

$$s(f, x) = \max \left\{ |J| \mid J \subseteq [n] : \forall i \in J. f(x) \neq f(x^{\{i\}}) \right\}$$

► In the worst case: $s(f) = \max_{x \in Q^n} s(f, x)$

Block sensitivity



What if we can flip not only one bit, but a block of bits?

Block sensitivity

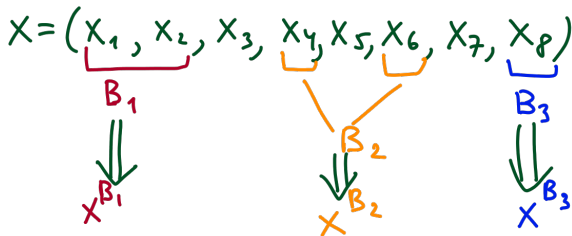
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$$x^{\{3,6\}} = (\dots \bar{x}_3 \dots \bar{x}_6 \dots)$$

► For a fixed input $x \in Q^n$:

$$\text{bs}(f, x) = \max \left\{ k \mid B_1 \sqcup B_2 \sqcup \dots \sqcup B_k \subseteq [n] : \forall i \in [k]. f(x) \neq f(x^{B_i}) \right\}$$

► In the worst case: $\text{bs}(f) = \max_{x \in Q^n} \text{bs}(f, x)$



Block sensitivity



How are they related?

Block sensitivity



How are they related?

Obviously, $s(f) \leq bs(f)$.

$J = \{i_1, \dots, i_k\}$ is optimal for $s(f, x)$



Consider $B_j = \{i_j\} \Rightarrow bs(f, x) \geq k$

Block sensitivity



How are they related?

Obviously, $s(f) \leq bs(f)$.

Sensitivity Conjecture [Nisan, Szegedy] (now Theorem [Huang])

$$bs(f) \leq s(f)^C, \text{ for a fixed constant } C \geq 1$$

(It holds for $C = 4$)

Why?



Low-sensitivity (or “smooth”) functions:

Why?



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- ▶ (**Computational**) are easy to compute even in the simplest models (like decision trees).

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- ▶ (**Algebraical**) have low degree as real polynomials.

Why?

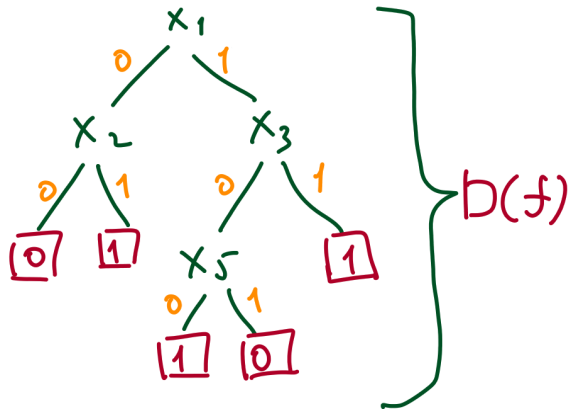


Low-sensitivity (or “smooth”) functions:

- ▶ (**Computational**) are easy to compute even in the simplest models (like decision trees).
- ▶ (**Algebraical**) have low degree as real polynomials.
- ▶ Combinatorial applications.
- ▶ Randomized and quantum query complexity.
- ▶ Certificate complexity.
- ▶ ...

Computational application: Decision trees

$$f: \{0, 1\}^n \rightarrow \{0, 1\}$$



Relation with decision trees

$$\text{bs}(f) \leq D(f) \leq \text{bs}(f)^3$$

(the height of a Christmas tree
with presents)

Algebraic application



Polynomial $p: \mathbb{R}^n \rightarrow \mathbb{R}$ represents f if

$$\text{For all } x \in \{0, 1\}^n. p(x) = f(x).$$

The degree $\deg(f)$ of f is the degree of a unique multilinear p that represents f .

Relation with $\deg f$

$$\sqrt{\text{bs}(f)} \leq \deg(f) \leq \text{bs}(f)^3$$

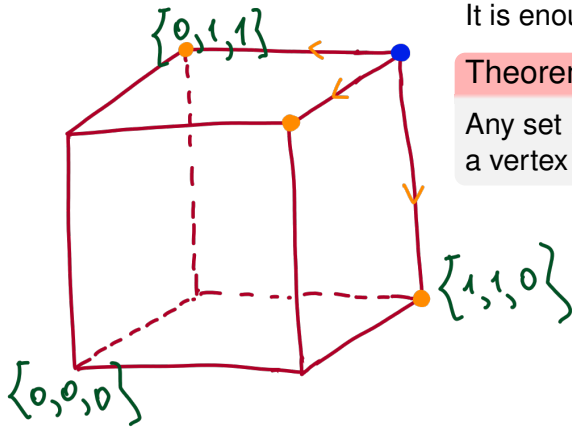
Supplementary Theorem

Let $f: \{0, 1\}^n \rightarrow \{0, 1\}$ be a Boolean function.

It is enough to prove:

Theorem: vertex with large degree

Any set H of $2^{n-1} + 1$ vertices of $\{0, 1\}^n$ contains a vertex with degree $\geq \sqrt{n}$.



What we need



Folklore

The linear system of m equations and $m + 1$ variables has a non-trivial solution.

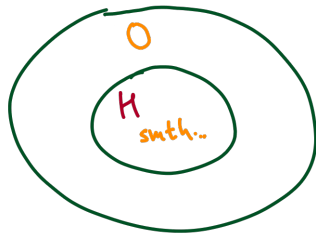
Proof (a simplification by Fedya Petrov)

Not-identically-zero function $f: \{0, 1\}^n \rightarrow \mathbb{R}$:

- ▶ $f(x), x \in \{0, 1\}^n$ are variables.
- ▶ f vanishes outside H : $2^{n-1} - 1$ equations.
- ▶ f satisfies **some** other 2^{n-1} equations.

Exists by the theorem above.

$$\left\{ \begin{array}{l} f(\bar{H}) = 0 \\ f(y) = \dots, \text{ for all } y \in H \end{array} \right.$$



Proof: Weight function



For $x = (x_1, \dots, x_k)$ define

$$w_i(x) = (-1)^{x_1 + \dots + x_{i-1}}.$$

Clearly:

$$w_i(x) = w_i(x^i).$$

Also:

$$w_i(x)w_j(x)w_i(x^j)w_j(x^i) = -1 \text{ for } i \neq j.$$

Proof: Relations



$$\sqrt{n} \cdot f(y) = \sum_{i=1}^n w_i(y) f(y^i), \text{ for all } y \in \{0, 1\}^n.$$

They're linearly dependent:

For $y = (x_1, \dots, x_{n-1}, 0) = x_0$:

$$\sqrt{n} \cdot f(x_0) = f(x_1) + \sum_{i=1}^{n-1} w_i(x) f(x^i_0), \text{ for all } y \in \{0, 1\}^n,$$

$$\sqrt{n} \cdot f(x_1) = f(x_0) - \sum_{i=1}^{n-1} w_i(x) f(x^i_1), \text{ for all } y \in \{0, 1\}^n.$$

Now substitute the former into the latter.

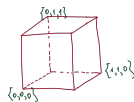
Proof: QED



Choose $y \in \{0, 1\}^n$ s.t. $|f(y)|$ is maximal. From

$$\sqrt{n} \cdot f(y) = \sum_{i=1}^n w_i(y) f(y^i), \text{ for all } y \in \{0, 1\}^n.$$

it follows that for at least \sqrt{n} elements $f(y^i) \neq 0$, hence they are in H .



Proof: Motivation



There is only 2^{n-1} linearly independent equations since the operator

$$f(y) \mapsto \sum_{i=1}^n w_i(y) \cdot f(y^i)$$

has an eigensubspace of dimension 2^{n-1} for the eigenvalue \sqrt{n} .

Possibly Part 2



- Why we defined such an operator?

Possibly Part 2



- ▶ Why we defined such an operator?
- ▶ It can be explained using expander graphs.