

A ROBUST VERSION OF HEGEDUS'S LEMMA

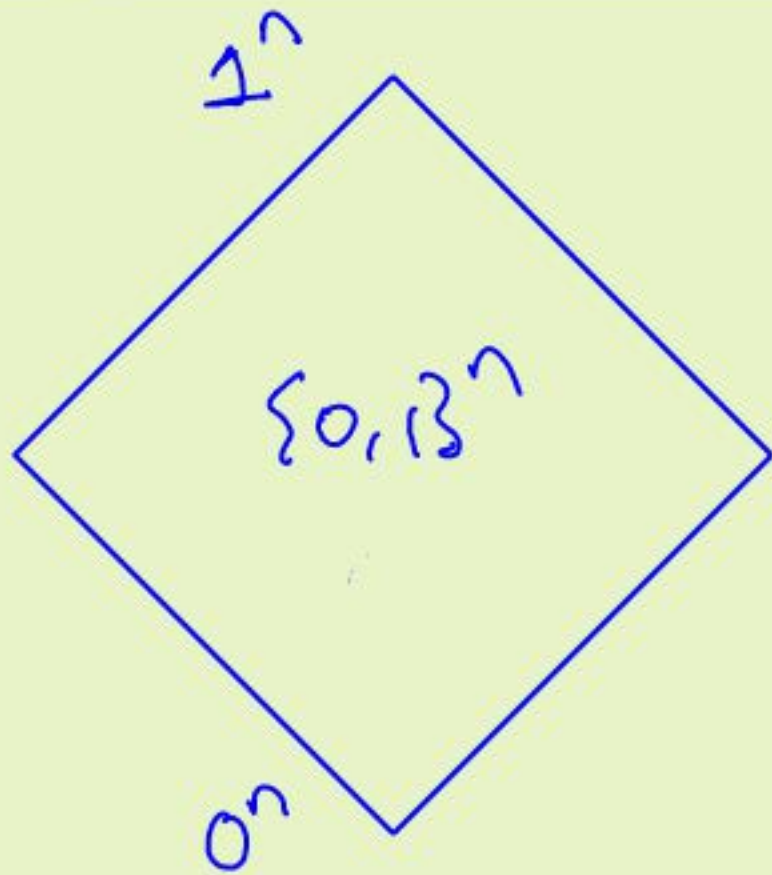
WITH APPLICATIONS

SRIKANTH SRINIVASAN

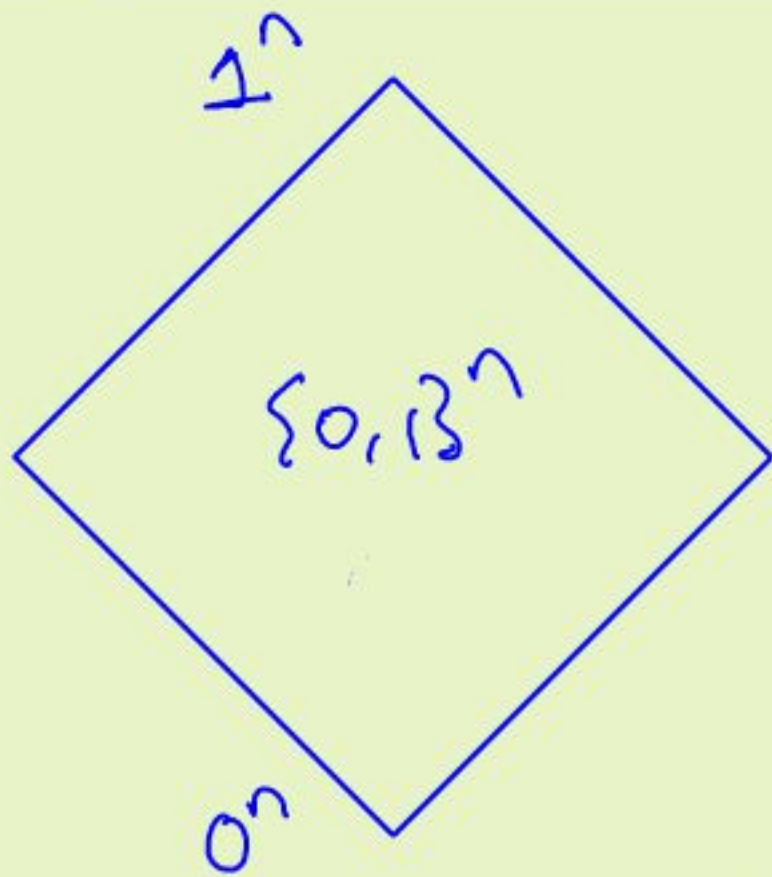
(DEPT. OF MATH., IIT BOMBAY)

OXFORD-WARWICK COMPLEXITY MEETINGS

# Polynomials over the hypercube



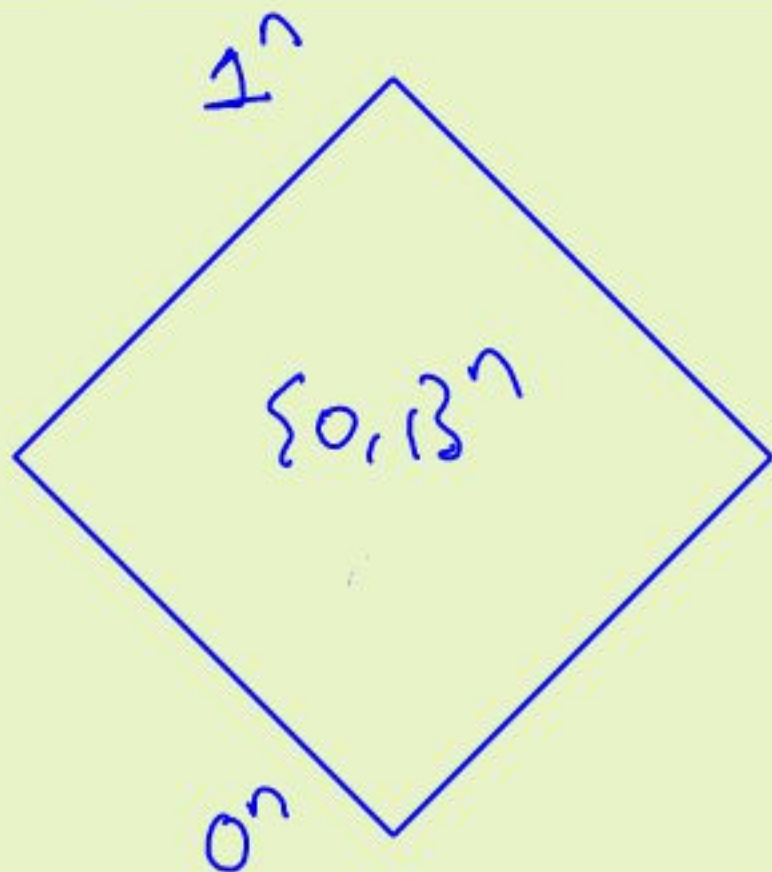
# Polynomials over the hypercube



Combinatorial / algebraic  
properties of low degree  
polynomials over  $\{0,1\}^n$

$$R \in \mathbb{F}[x_1, \dots, x_n]$$

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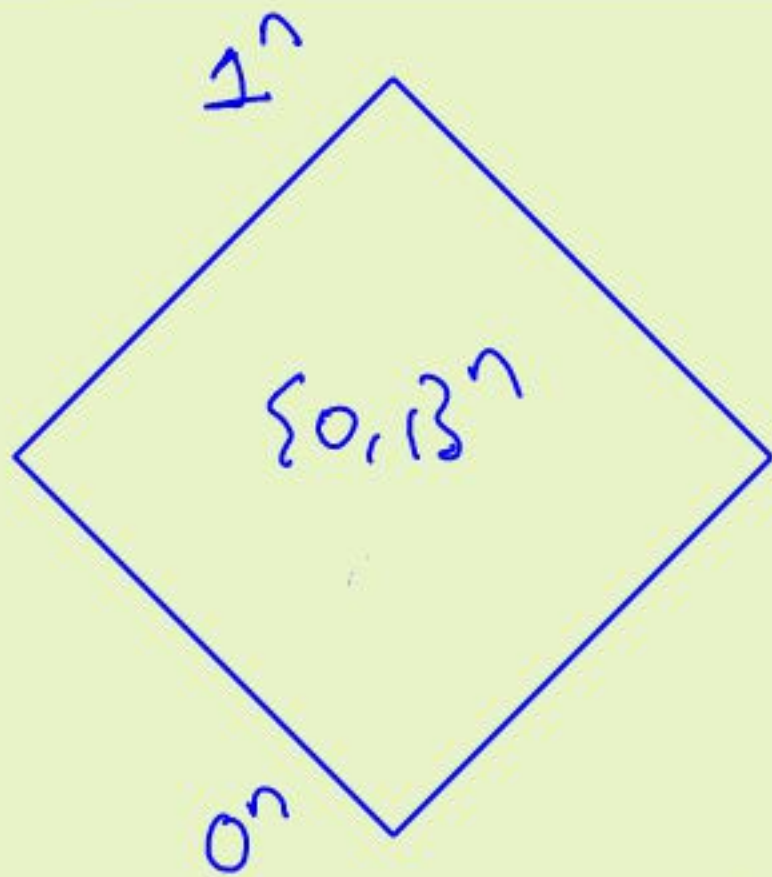


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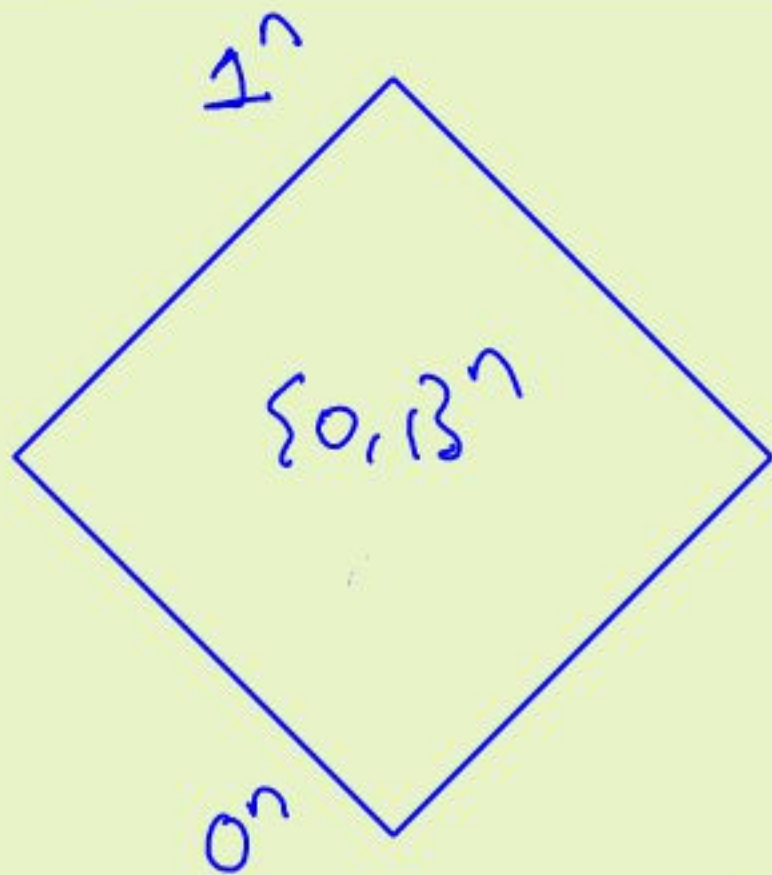
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1. Complexity theory
  - Circuit complexity [Razborov '87]
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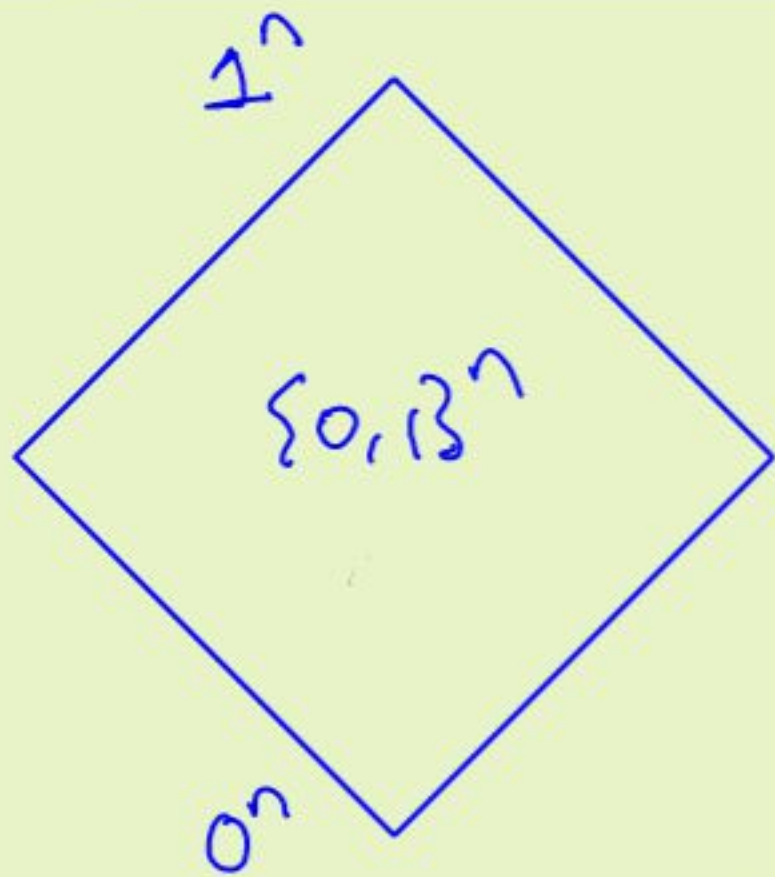
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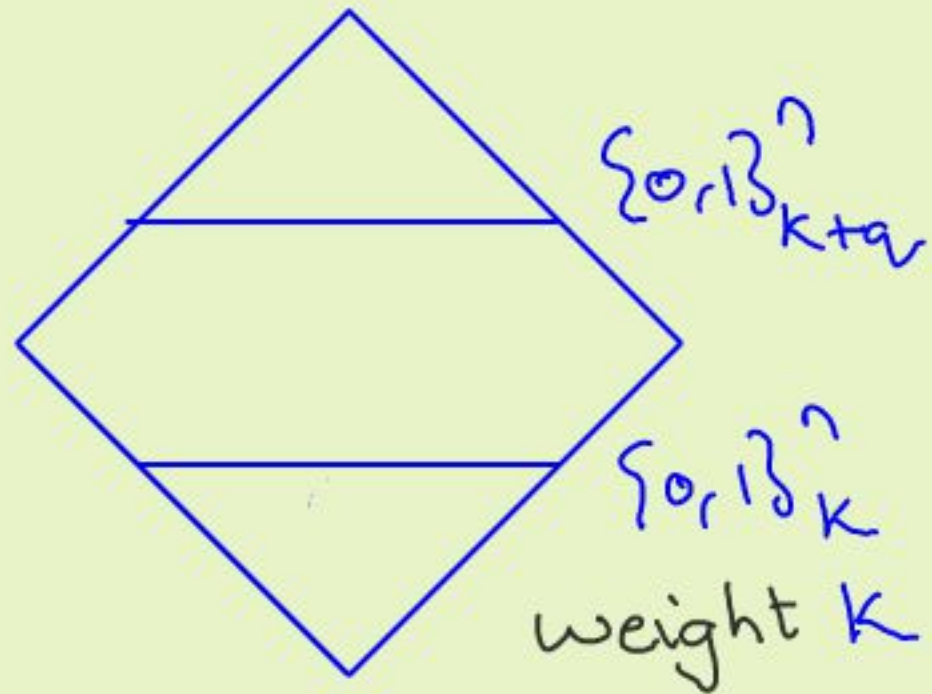
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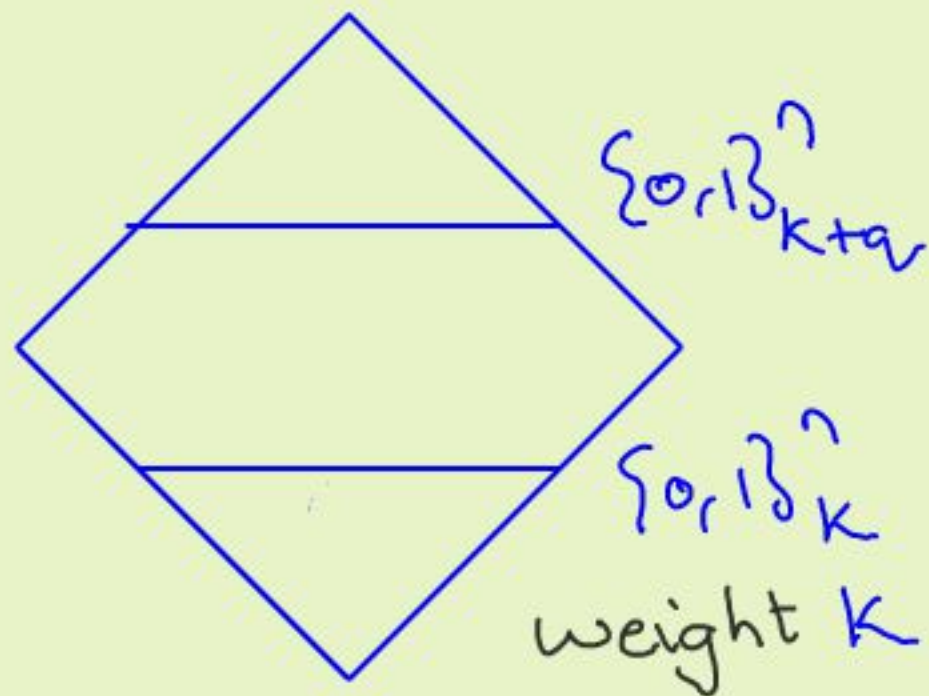
Hegedüs's Lemma



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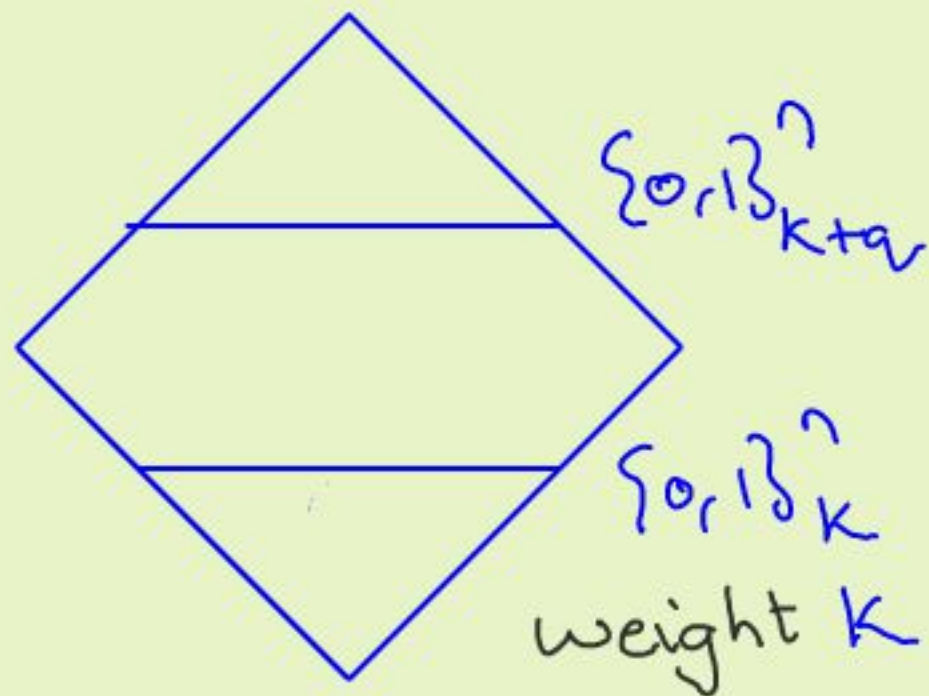
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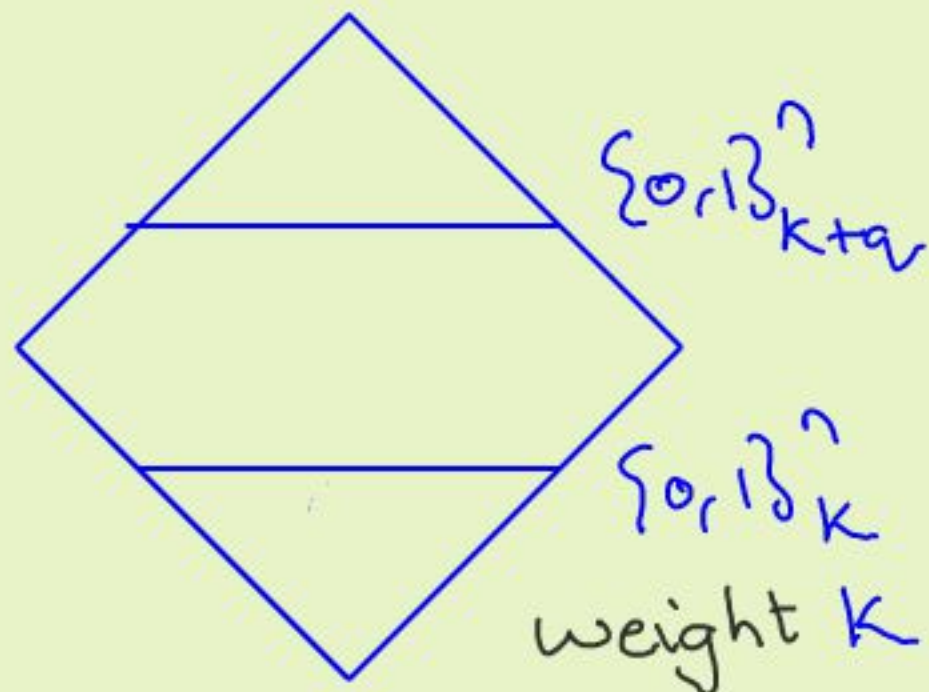
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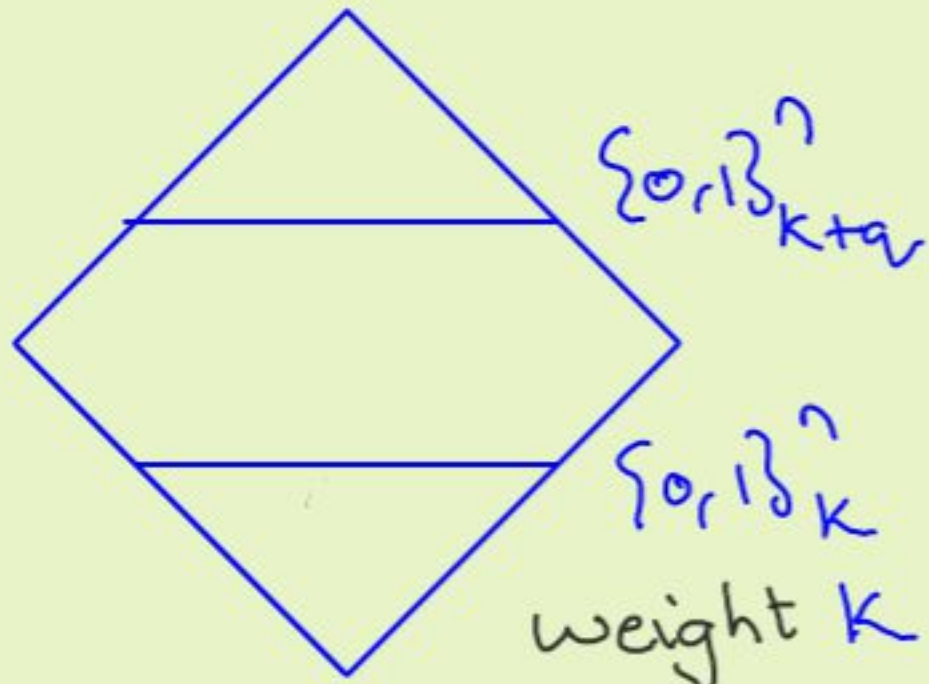
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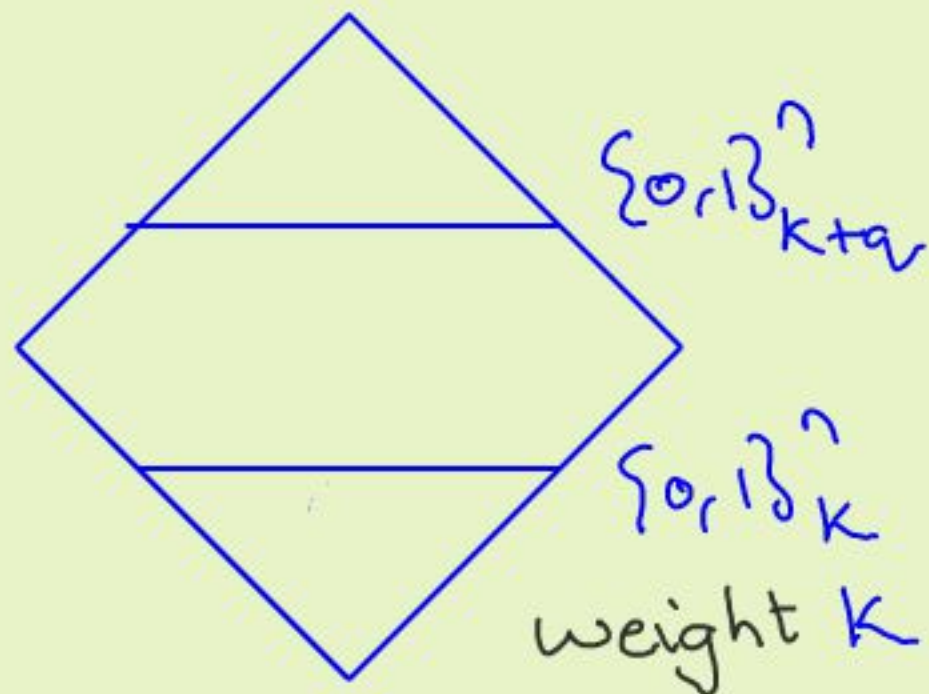
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$\mathbb{F} = \mathbb{Q}$ :  $R(x_1, \dots, x_n) = \left( \sum_i x_i - k \right)$

$\text{char}(\mathbb{F}) = p$ : Same construction if  $p \nmid q$ .

# Hegedüs's Lemma



$\text{char}(\mathbb{F}) = p, q = p^s :$

$\mathbb{F}[x_1, \dots, x_n]$   
 $\Downarrow$

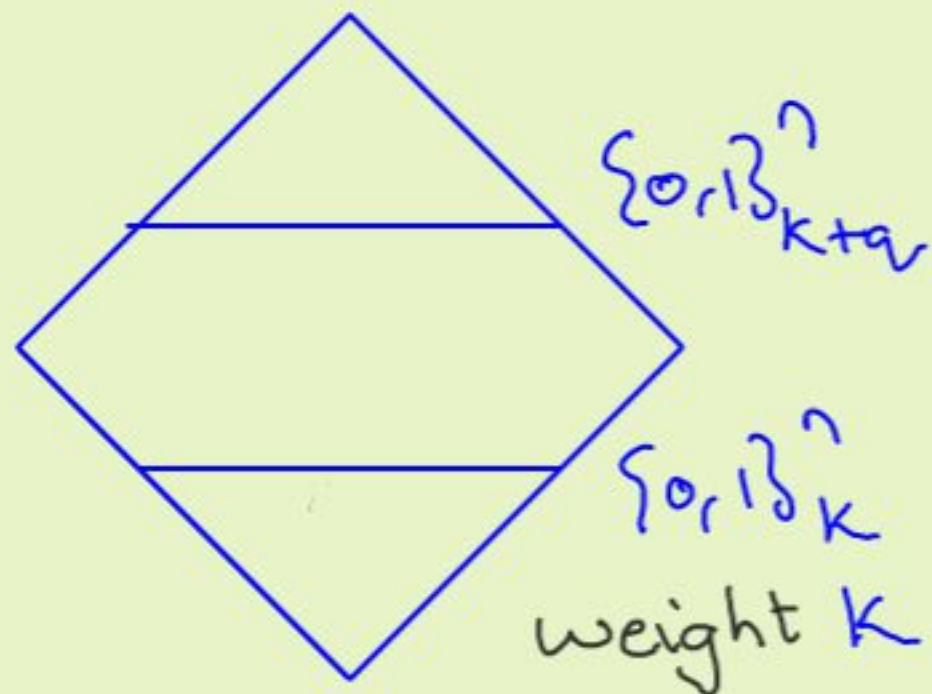
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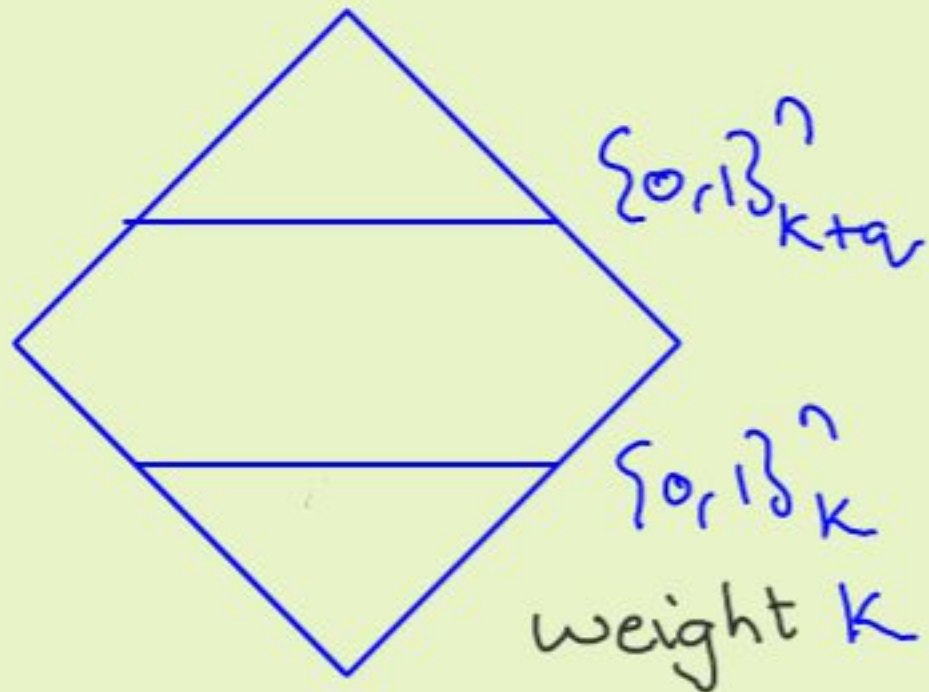
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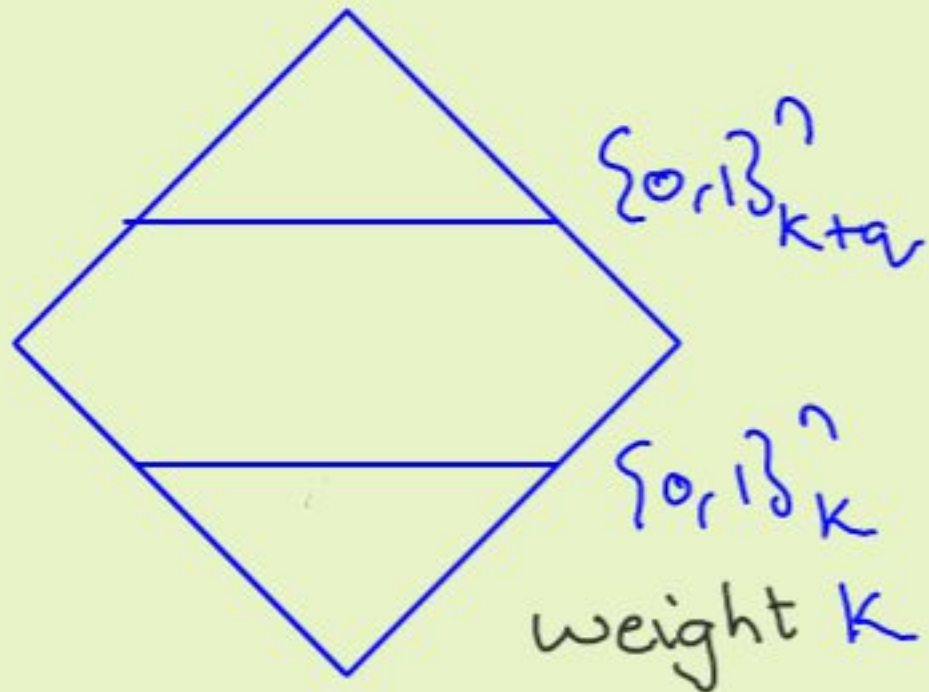
Lucas's thm





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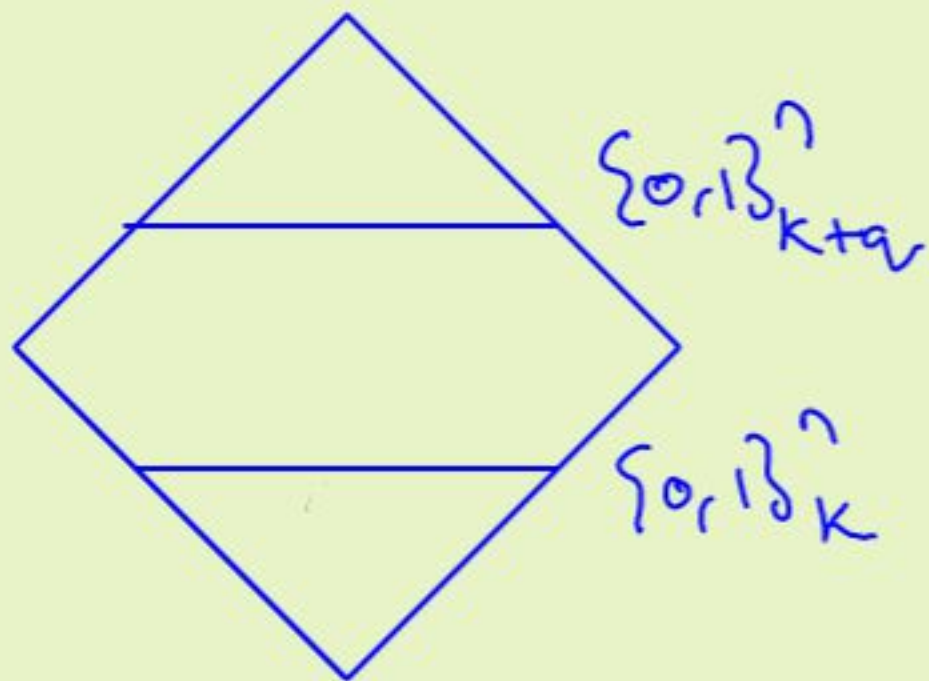
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Hegedüs's lemma: This is tight.

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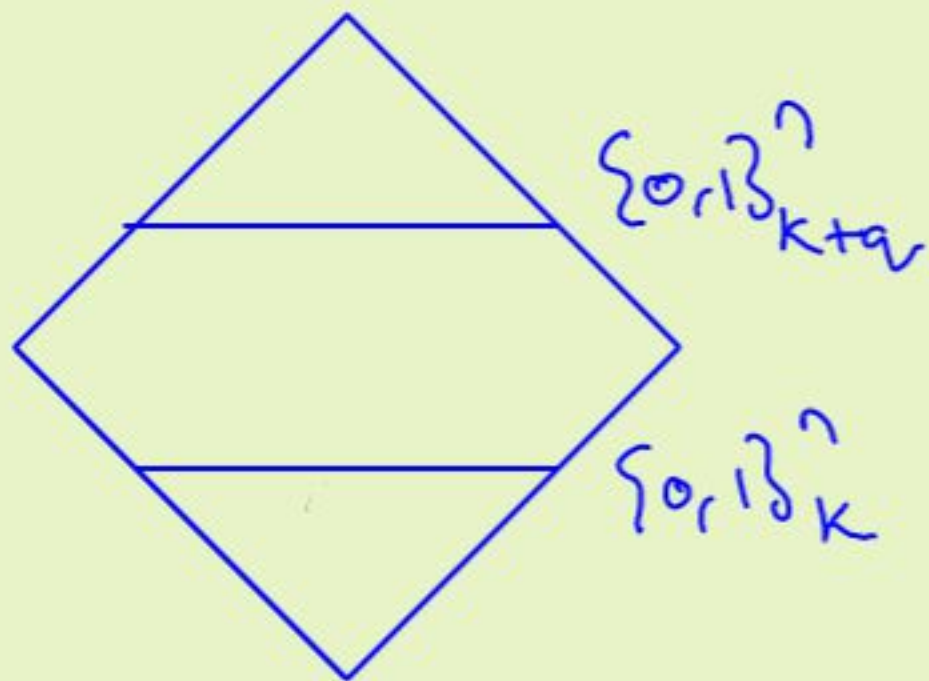
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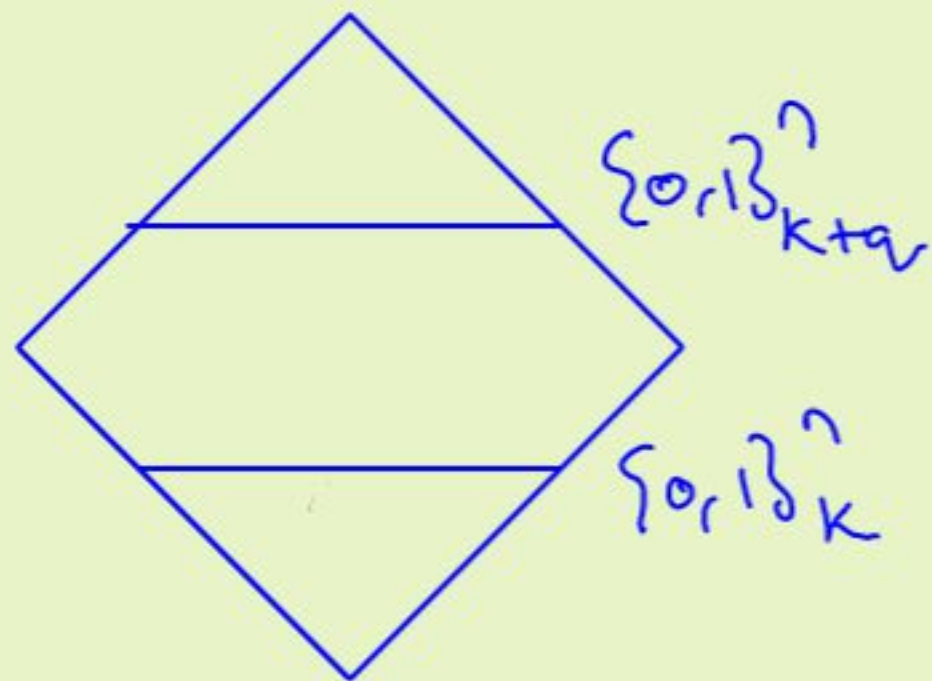
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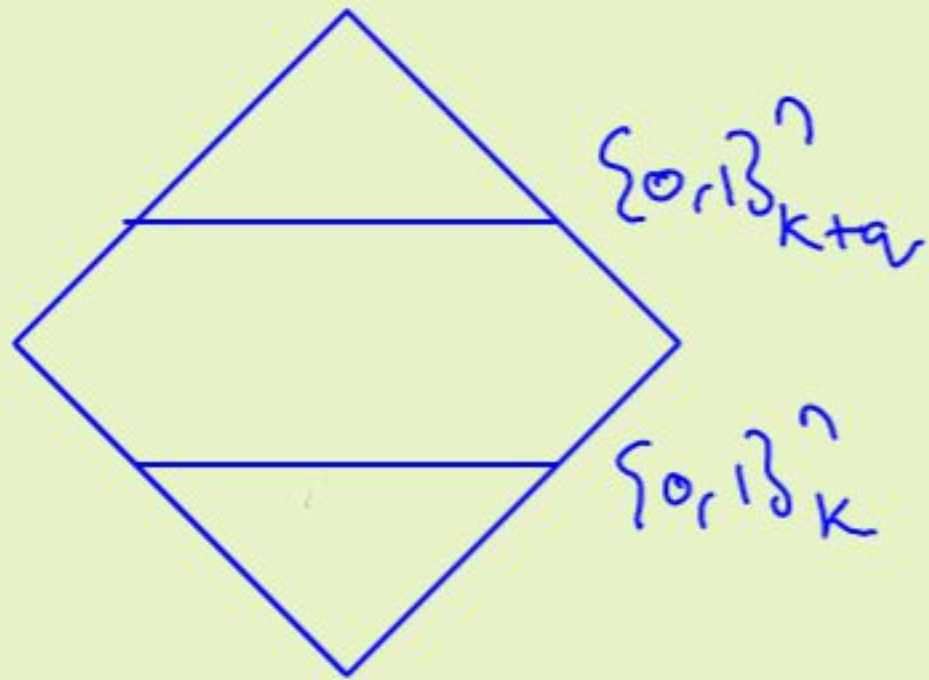
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[Hegedüs '10] - Gröbner bases

[S'18, Alon '19] - Elementary

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## Applications

[Hegedüs '10] - Gröbner bases

- Algebraic ckts

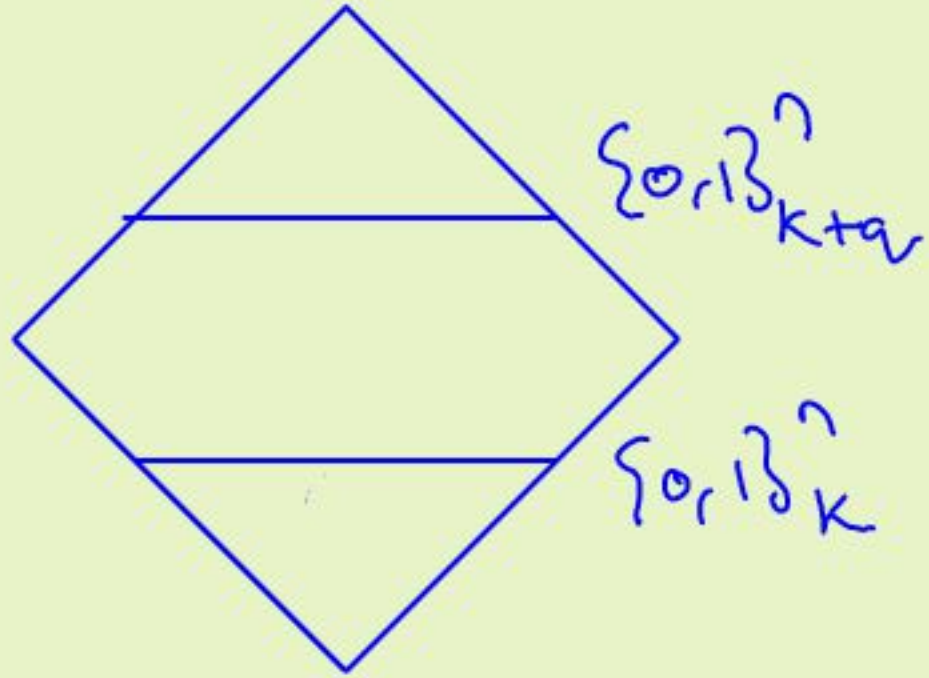
[S'18, Alon '19] Elementary

- Threshold ckts

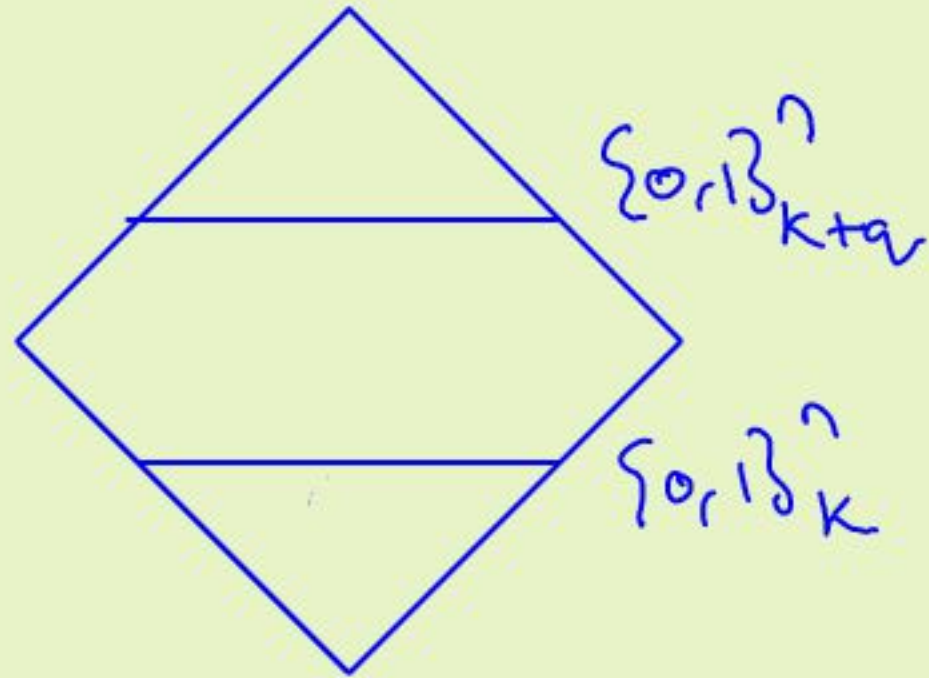
[Alon-Kumar-Volk '18]

[Krubes et al. '19]

# Robust Hegedűs



# Robust Hegerdus



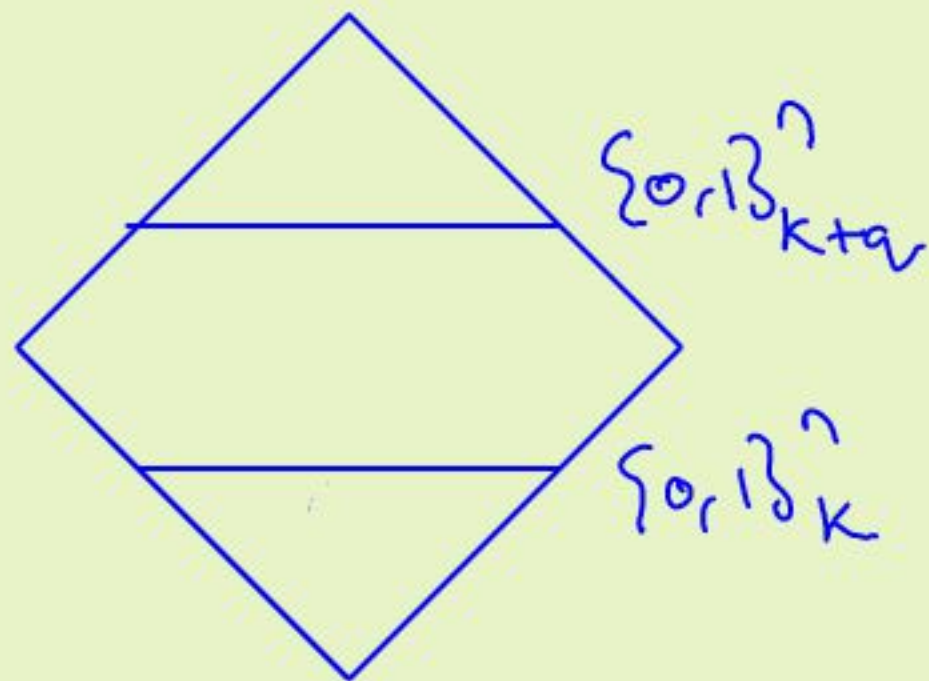
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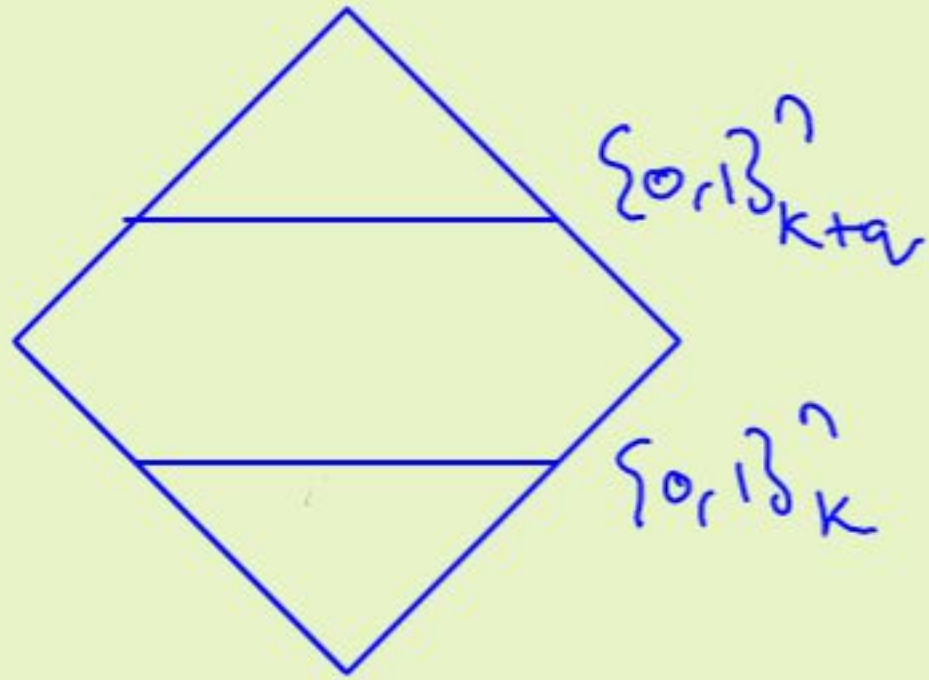
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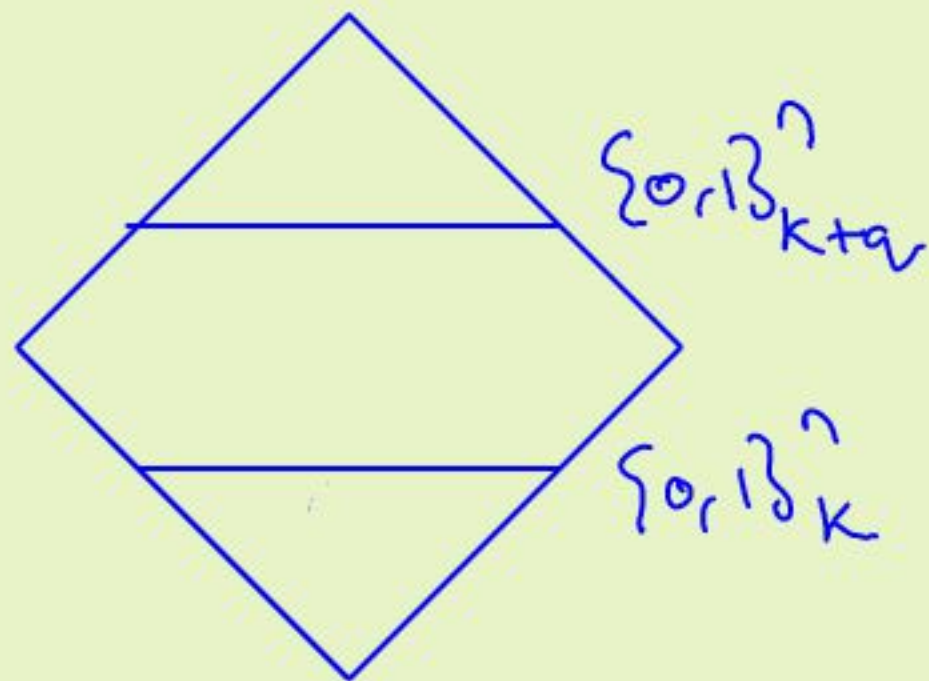
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Same lower bound (i.e.  $q$ )?

No!

# Sampling construction



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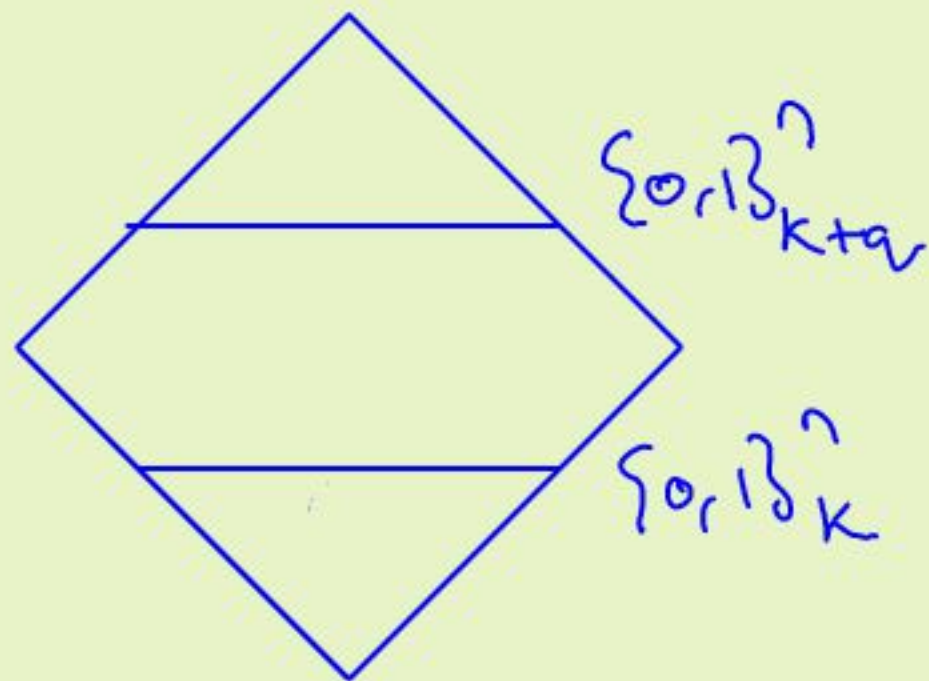
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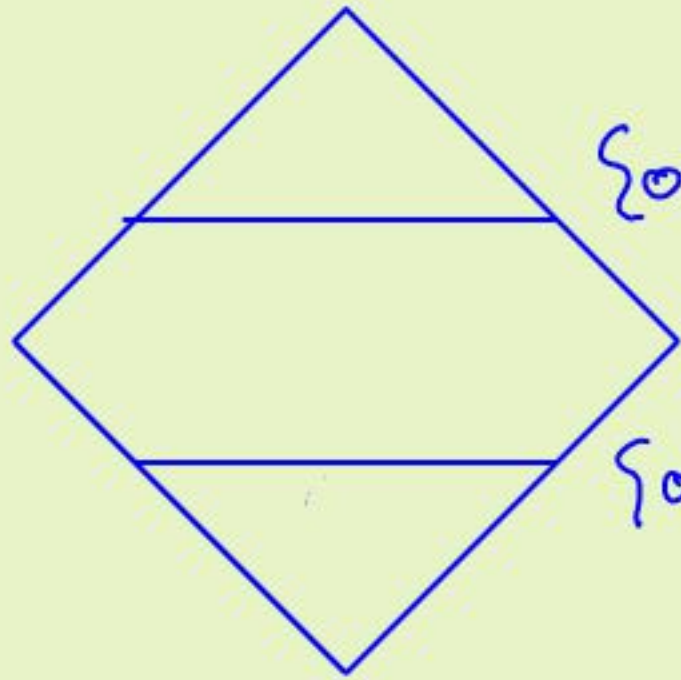
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$\{0, 1\}^{n_{k+q_v}}$

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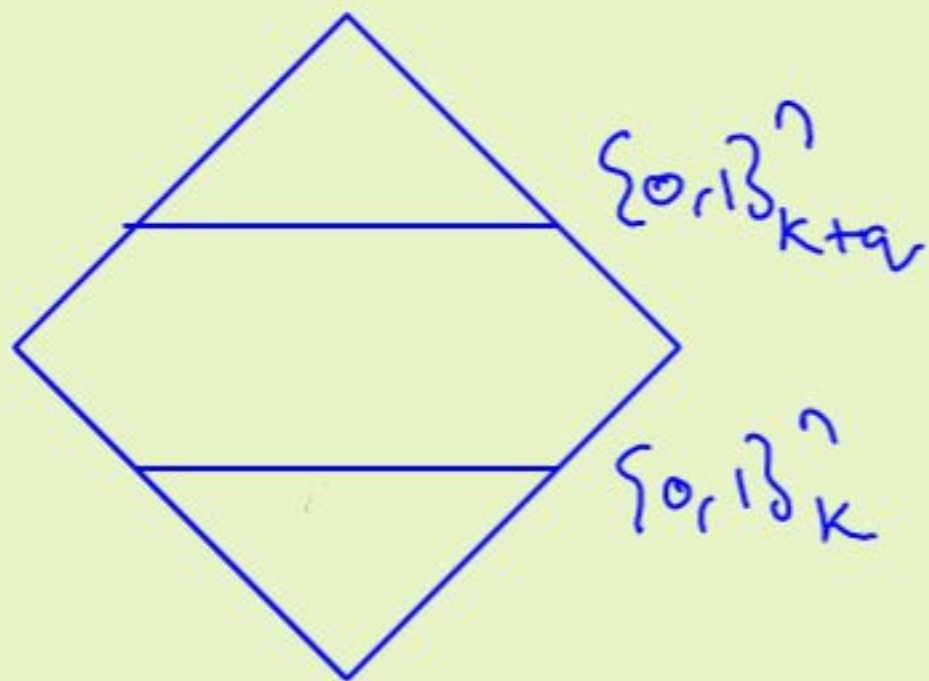
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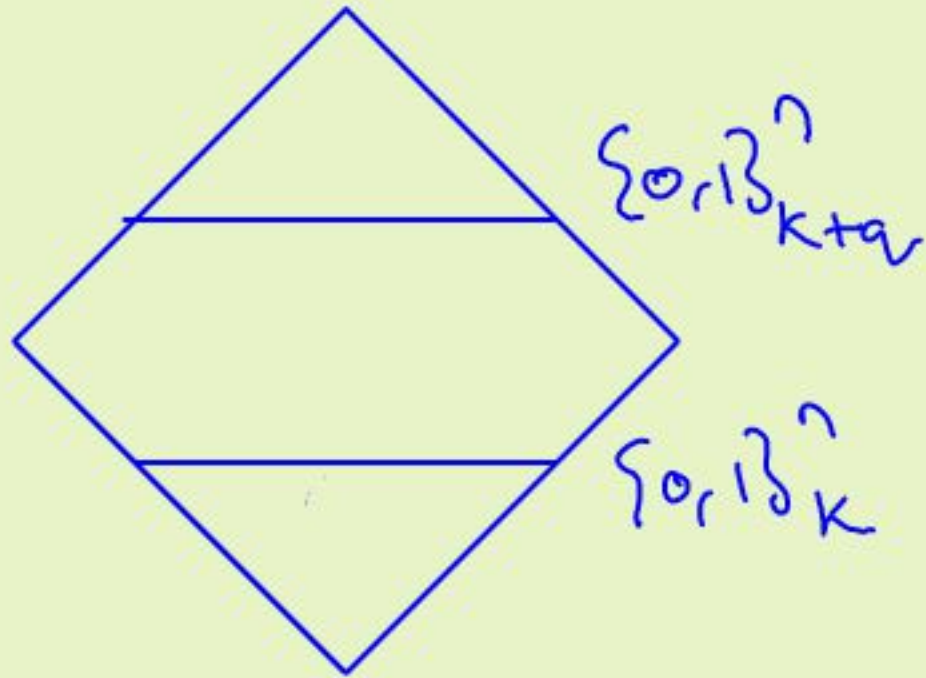
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(B)  $\text{deg } t = \Theta(\log^{1/\varepsilon} n)$ .

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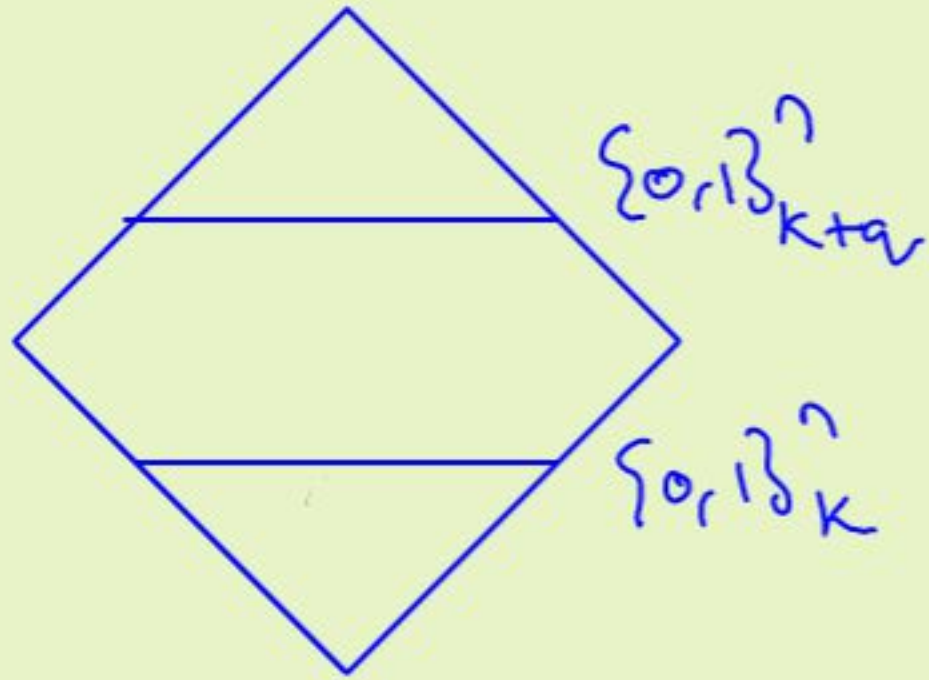
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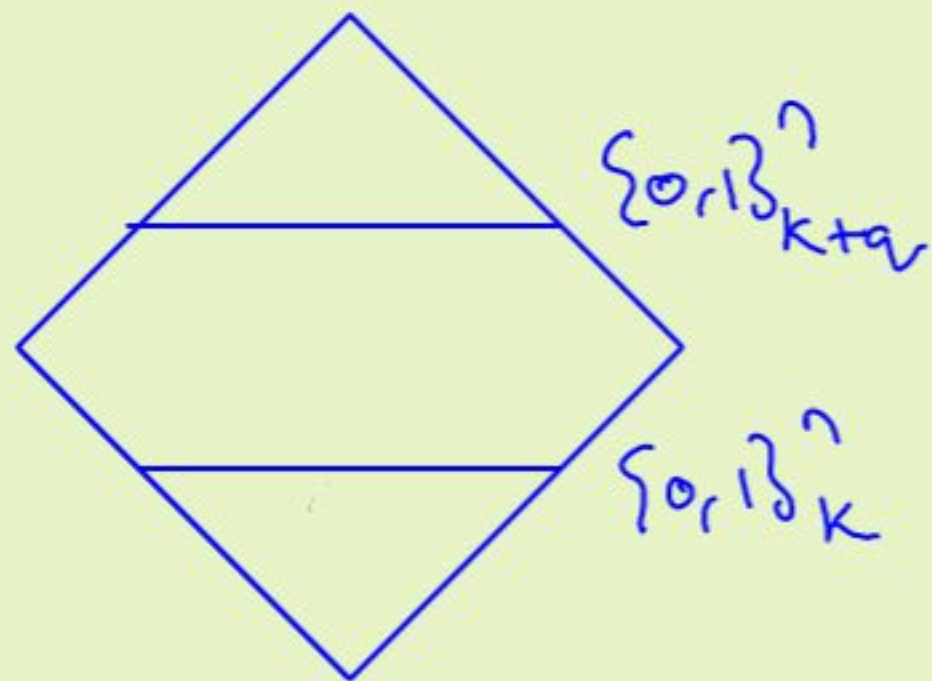
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# Probabilistic polynomials

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→ Used to prove lower bounds for  $AC^0(\text{MOD}_p)$  cks.

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→ Smolensky '87:  $\text{pdeg}_{1/10}(\text{Maj}_n), \text{pdeg}_{1/10}(\text{MOD}_n^2) = \Omega(\sqrt{n})$

→ S. '13:  $\text{pdeg}_{1/10}(\text{any symmetric } f) = \tilde{O}(\sqrt{n})$

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## Upper & lower bounds on pdeg

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→ Algorithmic applications



Tight bounds on  $\rho_{\text{deg}}$  of any symmetric  $f$ ?

Tight bounds on  $\text{pdeg}$  of any symmetric  $f$ ?

S. - Tripathi - Venkitesh '19: Upper bounds for any symmetric  $f$

**Theorem 8** (Known upper bounds on probabilistic degree of symmetric functions [STV19]). Let  $\mathbb{F}$  be a field of constant characteristic  $p > 0$  and  $n \in \mathbb{N}$  be a growing parameter. Let  $f \in \mathcal{SB}_n$  be arbitrary and let  $(g, h)$  be a standard decomposition of  $f$ . Then we have the following for any  $\varepsilon > 0$ .

1. If  $\text{per}(g) = 1$ , then  $\text{pdeg}_\varepsilon(g) = 0$ .

If  $\text{per}(g)$  is a power of  $p$ , then  $\text{pdeg}_\varepsilon^{\mathbb{F}}(g) \leq \text{per}(g)$ ,

2.  $\text{pdeg}_\varepsilon(h) = O(\sqrt{B(h) \log(1/\varepsilon)} + \log(1/\varepsilon))$  if  $B(h) \geq 1$  and 0 otherwise, and

3.  $\text{pdeg}_\varepsilon(f) = \begin{cases} O(\sqrt{n \log(1/\varepsilon)}) & \text{if } \text{per}(g) > 1 \text{ and not a power of } p, \\ O(\min\{\sqrt{n \log(1/\varepsilon)}, \text{per}(g)\}) & \text{if } \text{per}(g) \text{ a power of } p \text{ and } B(h) = 0, \\ O(\min\{\sqrt{n \log(1/\varepsilon)}, \text{per}(g) + \sqrt{B(h) \log(1/\varepsilon)} + \log(1/\varepsilon)\}) & \text{otherwise.} \end{cases}$

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Also: nearly matching lower bounds.

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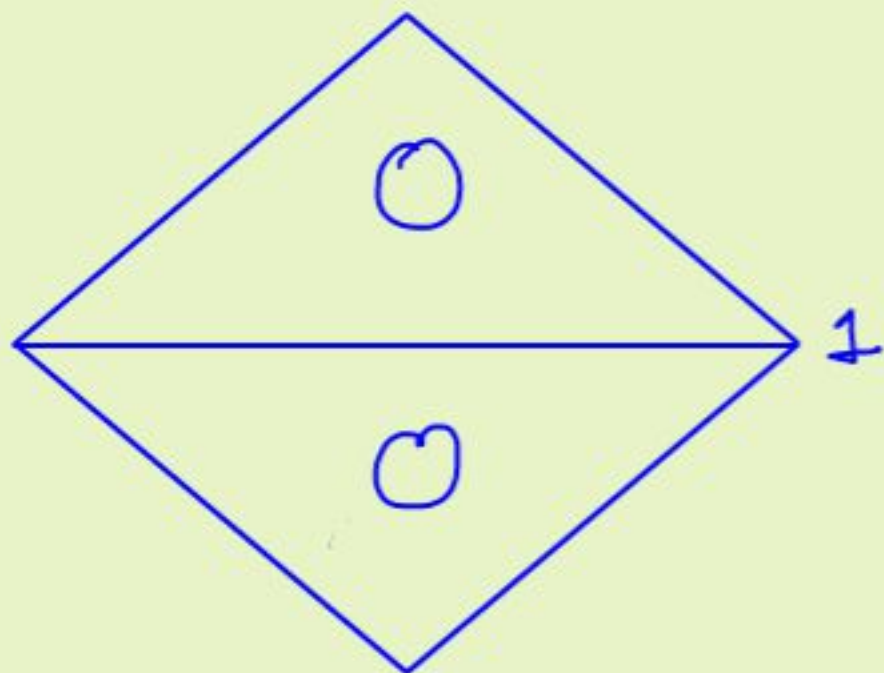
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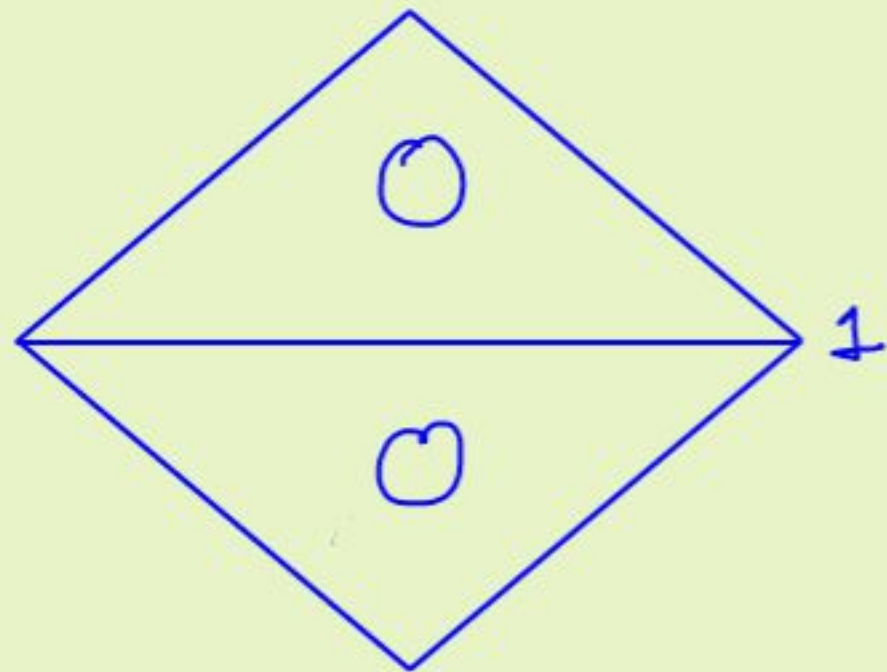
This paper: Tight bounds for all symmetric fns

Example:  $ETHR_n^{n/2}$



$$pdeg_{\varepsilon}(ETHR_n^{n/2}) = ?$$

Example:  $\text{ETHR}_n^{n/2}$

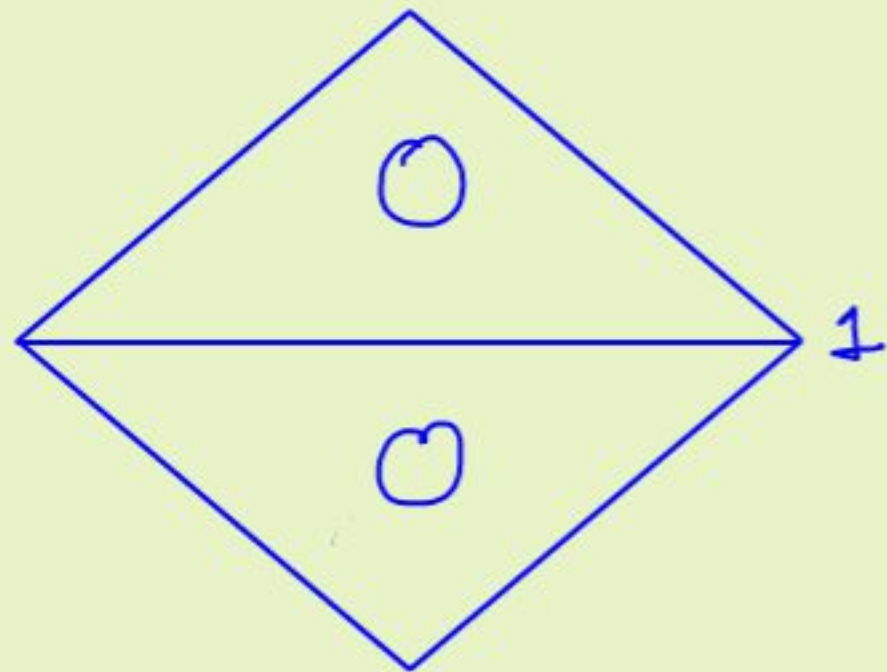


$$\text{pdeg}_\varepsilon(\text{ETHR}_n^{n/2}) = ?$$

[STV'19]:

$$\text{Maj}_{n/2}(x) = \sum_{i=0}^{\frac{n/4-1}{2}} \text{ETHR}_n^{n/2}(x \perp 0^{n/2-i})$$

Example:  $\text{ETHR}_n^{n/2}$



$$\text{pdeg}_\varepsilon(\text{ETHR}_n^{n/2}) = ?$$

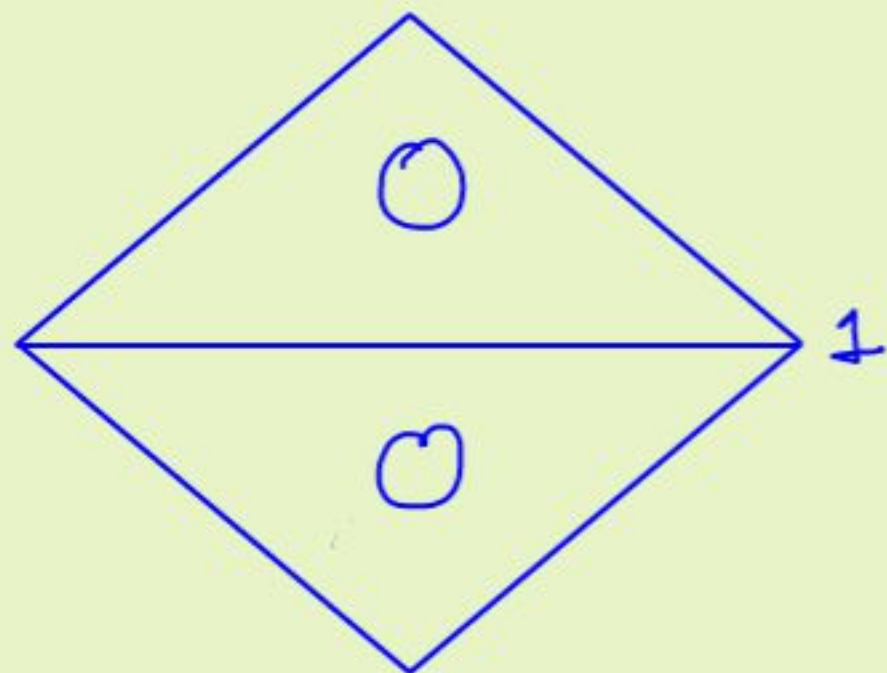
[STV'19]:

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$$\begin{aligned} \Rightarrow \text{pdeg}_\varepsilon(\text{Maj}_{n/2}(x)) &\leq \text{pdeg}_{\varepsilon/n}(\text{ETHR}_n^{n/2}) \\ &\leq \text{pdeg}_\varepsilon(\text{ETHR}_n^{n/2}) - \log n \end{aligned}$$



Example:  $\text{ETHR}_n^{n/2}$



$$\text{pdeg}_\varepsilon(\text{ETHR}_n^{n/2}) = ?$$

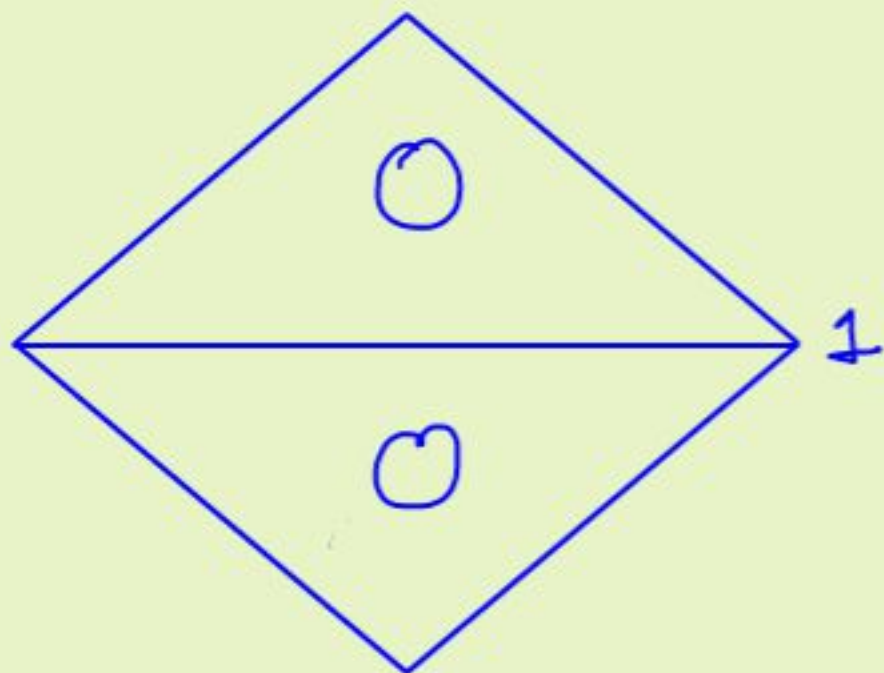
[STV'19]:

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$$\leq \text{pdeg}_\varepsilon(\text{ETHR}_n^{n/2}) - \log n$$

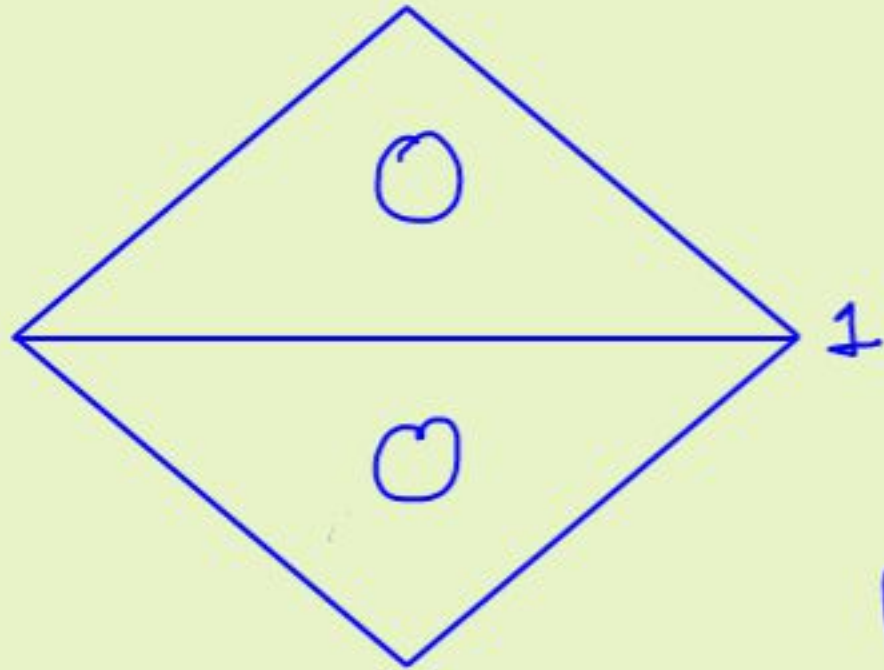
$$\Rightarrow \text{pdeg}_\varepsilon(\text{ETHR}_n^{n/2}) = \tilde{\Omega}(\sqrt{n})$$

Example:  $ETHR_n^{n/2}$



$$pdeg_{\varepsilon}(ETHR_n^{n/2}) = ?$$

Example:  $ETHR_n^{n/2}$



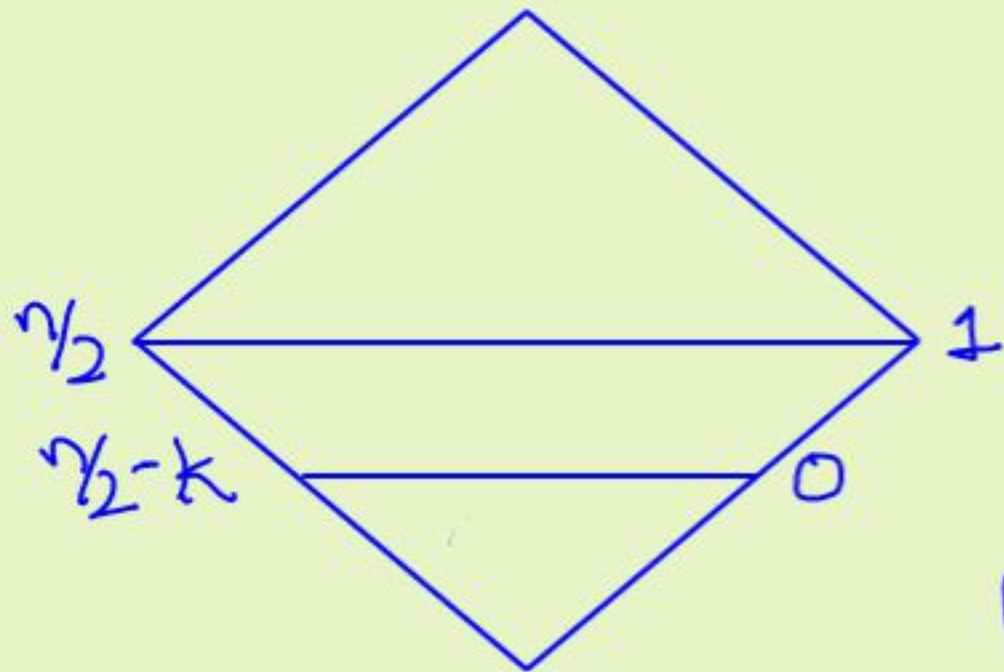
$$pdeg_{\varepsilon}(ETHR_n^{n/2}) = ?$$

$$\mathbb{P} = \{P_1, \dots, P_r\}$$

$$P_{\gamma} [P(x) \neq ETHR_n^{n/2}(x)] \leq \varepsilon$$

$P \sim \mathbb{P}$

Example:  $ETHR_n^{n/2}$



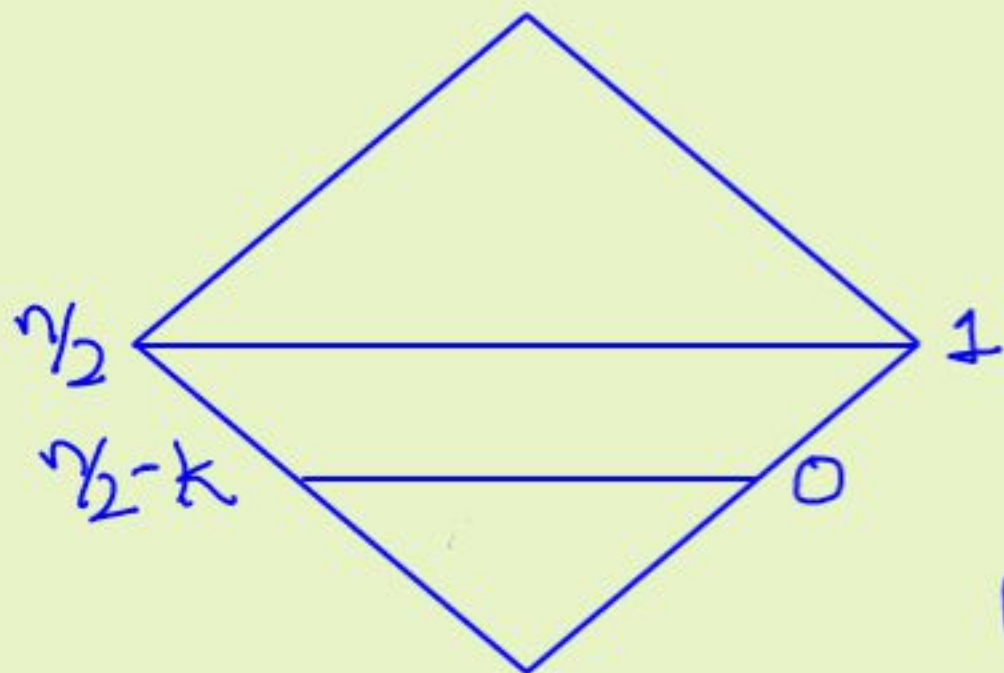
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Example: ETHR<sub>n</sub><sup>n/2</sup>



$$pdeg_{\epsilon}(\text{ETHR}_n^{n/2}) = ?$$

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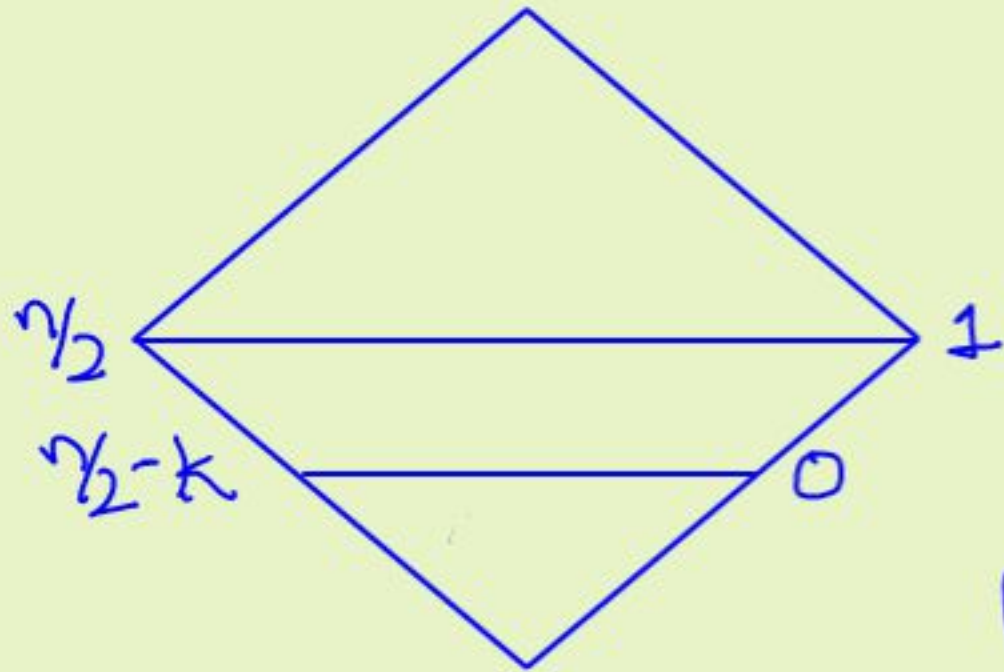
$$P_r [P(x) \neq \text{ETHR}_n^{n/2}(x)] \leq \epsilon$$
$$P \sim \mathbb{P}$$

By averaging,  $\exists P \in \mathbb{P}$  such that

$$P_r [P(x) \neq 1] < 2\epsilon$$
$$x \sim \{0,1\}_{n/2}^n$$

$$P_r [P(x) \neq 0] < 2\epsilon$$
$$x \sim \{0,1\}_{n/2-k}^n$$

Example: ETHR<sub>n</sub><sup>n/2</sup>



$$pdeg_{\epsilon}(ETHR_n^{n/2}) = ?$$

$$\mathbb{P} = \{P_1, \dots, P_r\}$$

$$Pr [P(x) \neq ETHR_n^{n/2}(x)] \leq \epsilon$$

$$P \sim \mathbb{P}$$

By averaging,  $\exists P \in \mathbb{P}$  such that

$$Pr_{x \sim \{0,1\}^{n/2}} [P(x) \neq 1] < 2\epsilon \quad k = p^8 = \Theta(\sqrt{n})$$

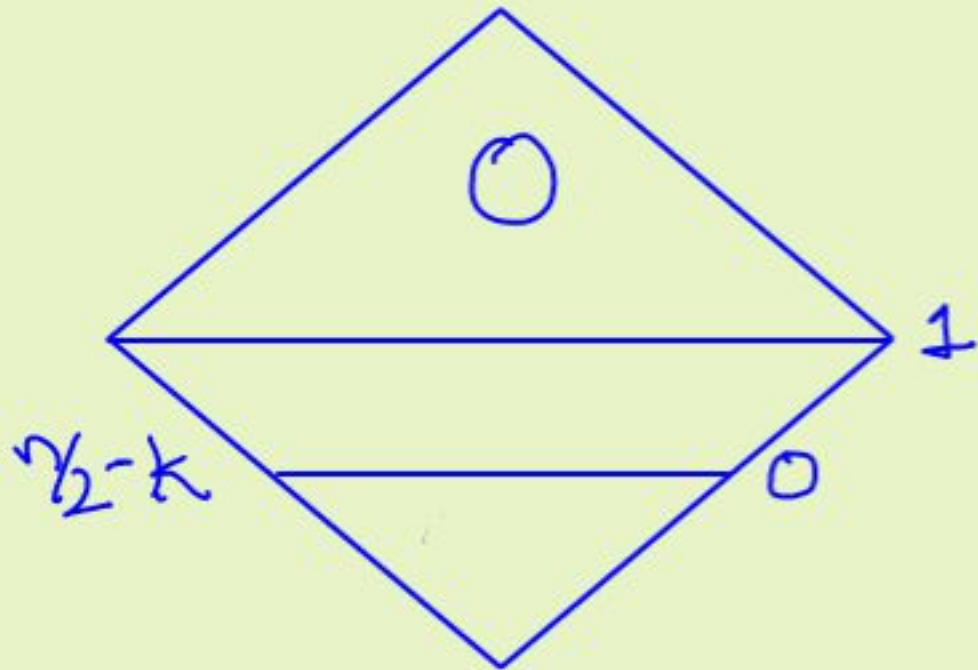
$$Pr_{x \sim \{0,1\}^{n/2-k}} [P(x) \neq 0] < 2\epsilon$$



Robust  
Hegedüs

$$deg(P) = \Omega(\sqrt{n})$$

Example:  $ETHR_n^{n/2}$



Message:

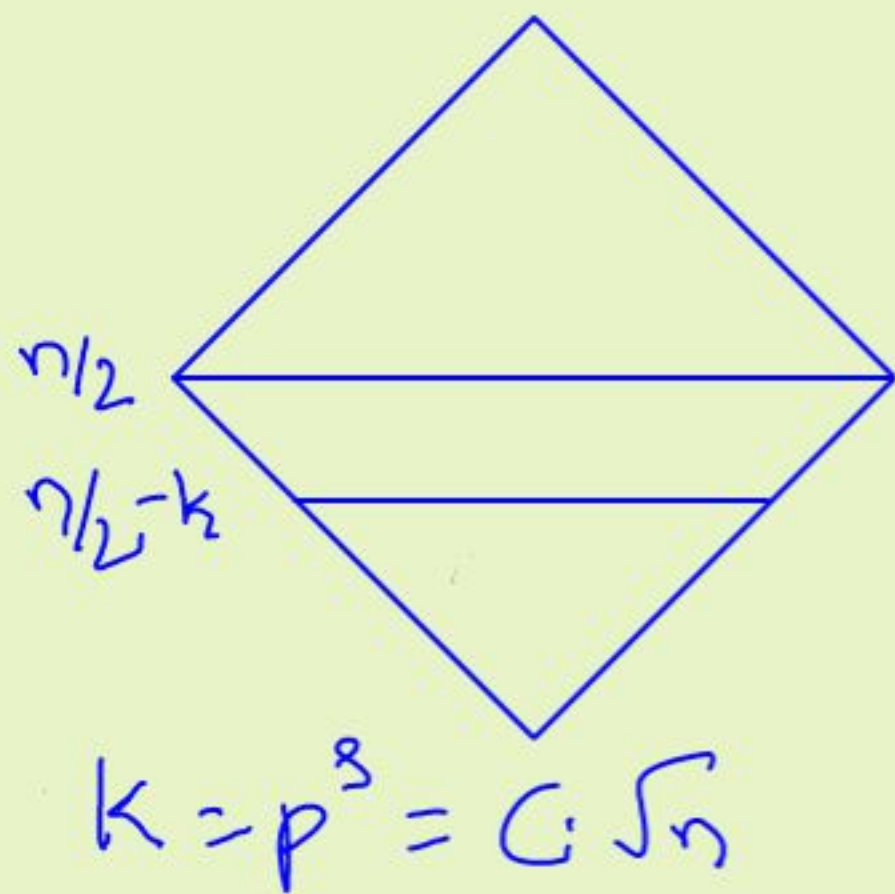
→ Robust Megedüs  $\Rightarrow$   
lbs on pdeg of simple  
promise symmetric fns.

→ Easier to reduce to other  
symmetric functions.

# Proof of Robust Hegedüs's Lemma



# Proof of Robust Hegedüs's Lemma

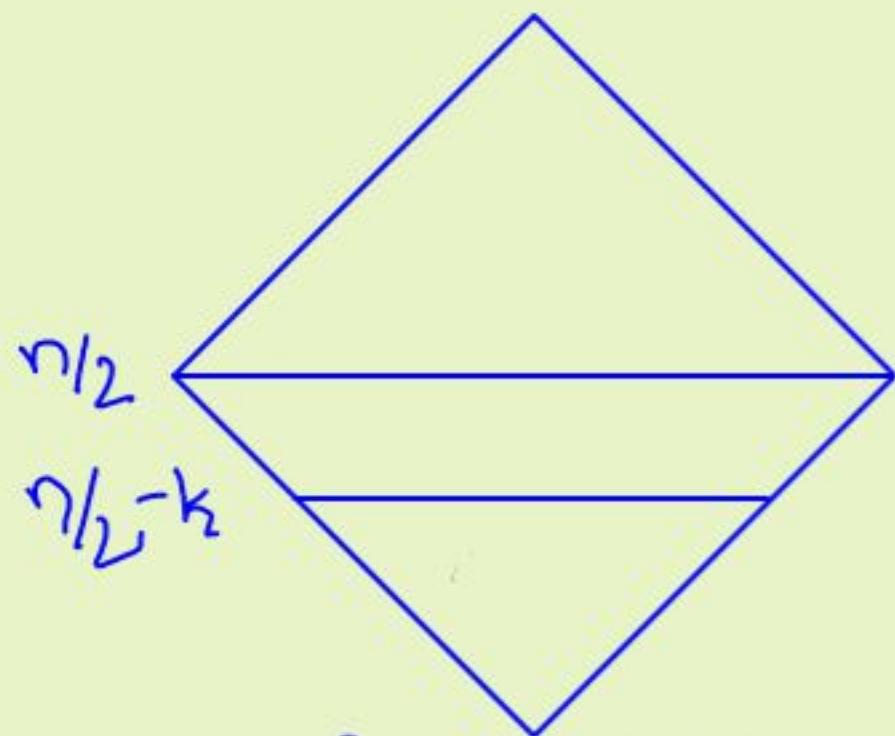


$R(x_1, \dots, x_n) \in \mathbb{F}_p[x_1, \dots, x_n]$  s. t.

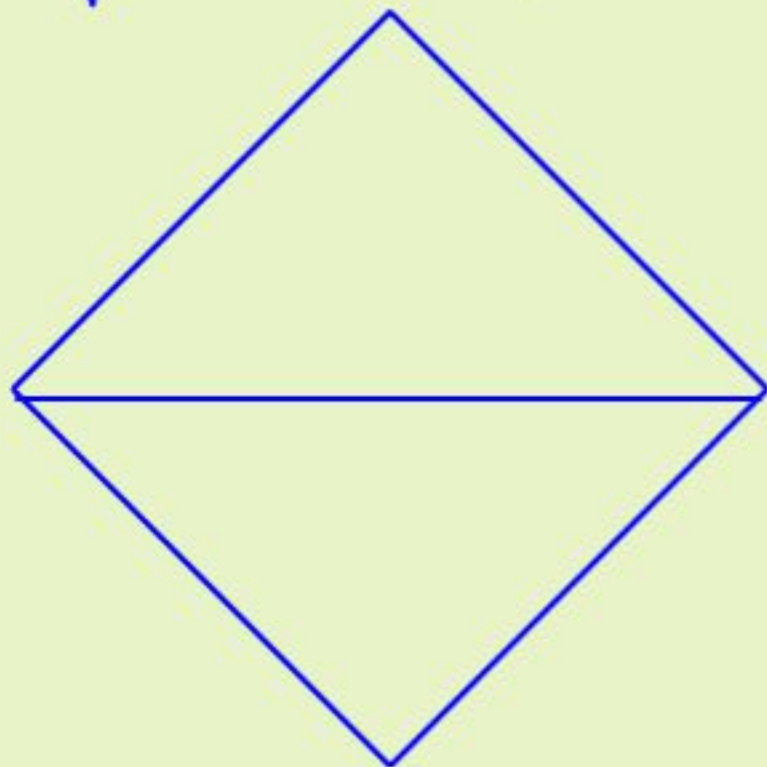
(A)  $\Pr [R(a) = 0] \leq \epsilon$   
 $a \sim \{0, 1\}^{n/2}$

(B)  $\Pr [R(b) \neq 0] \leq \epsilon$   
 $b \sim \{0, 1\}^{n/2} \Rightarrow \deg(R) = \Omega(\sqrt{n})$

# Proof of Robust Hegedüs's Lemma



$$k = p^3 = O(\sqrt{n})$$



$$R(x_1, \dots, x_n) \in \mathbb{F}_p[x_1, \dots, x_n] \text{ s.t.}$$

$$(A) \Pr [R(a) = 0] \leq \epsilon$$

$$a \sim \{0, 1\}^{n/2}$$

$$(B) \Pr [R(b) \neq 0] \leq \epsilon$$

$$b \sim \{0, 1\}^{n/2} \Rightarrow \deg(R) = \Omega(\sqrt{n})$$

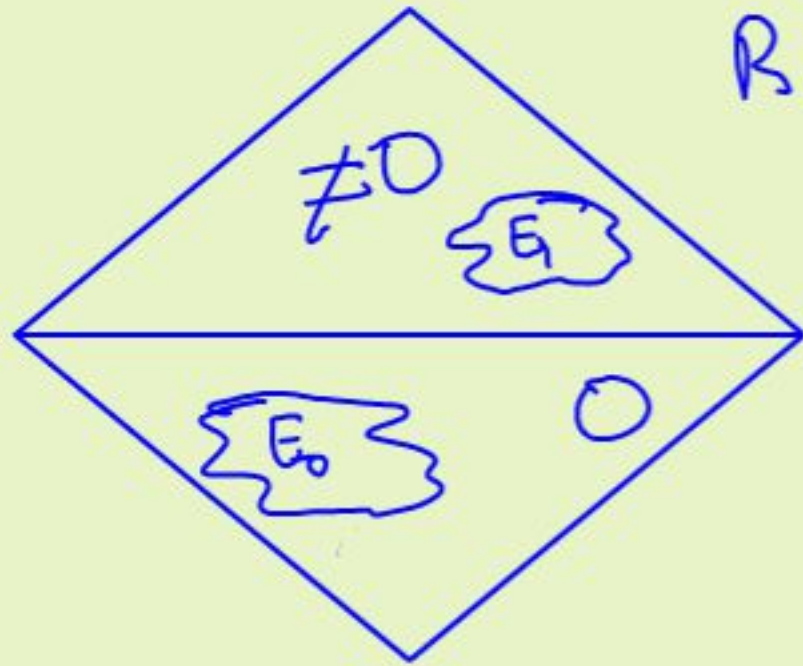
Smolensky's thm:

$$(A) \Pr_{a \sim \{0, 1\}_{\geq n/2}^n} [R(a) = 0] \leq \epsilon$$

$$(B) \Pr_{b \sim \{0, 1\}_{< n/2}^n} [R(a) \neq 0] \leq \epsilon$$

$$\Rightarrow \deg(R) = \Omega(\sqrt{n})$$

# Proof of Smolensky's thm

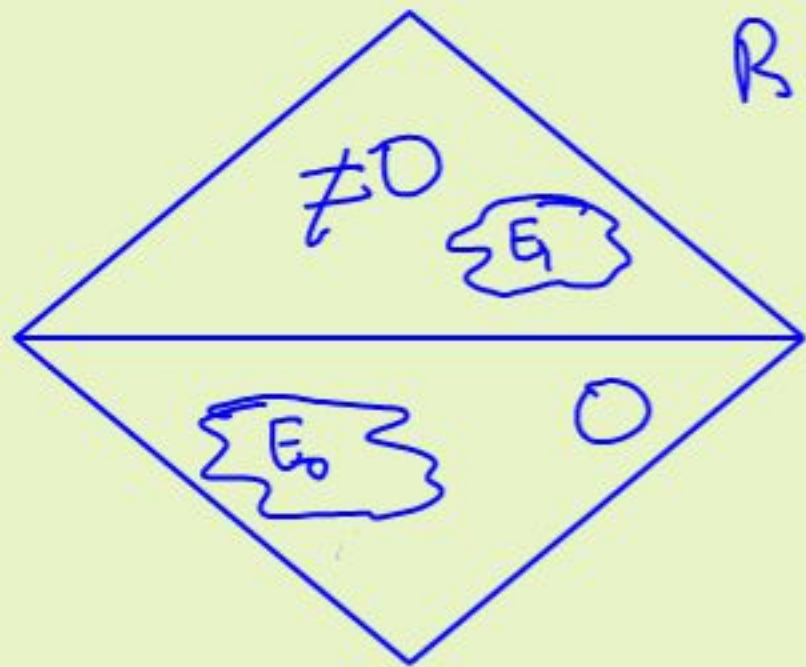


$R(x_1, \dots, x_n)$

$$|E_0|, |E_1| \leq \epsilon \cdot 2^n.$$

$$\Rightarrow \deg(R) = \Omega(\sqrt{n})$$

# Proof of Smolensky's thm



$R(x_1, \dots, x_n)$

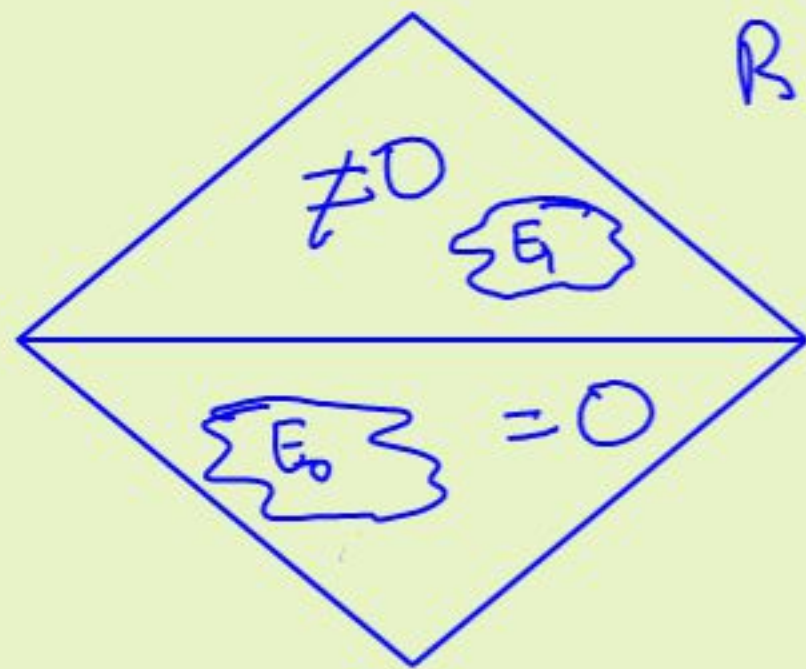
$$|E_0|, |E_1| \leq \epsilon \cdot 2^n.$$

$$\Rightarrow \deg(R) = \Omega(n)$$

Clm 1:  $\exists Q, \deg(Q) \leq n/2 - \Omega(n)$  s.t.

$$Q|_{E_0 \cup E_1} \equiv 0 \quad \& \quad Q \cdot R \neq 0$$

# Proof of Smolensky's thm



$R(x_1, \dots, x_n)$

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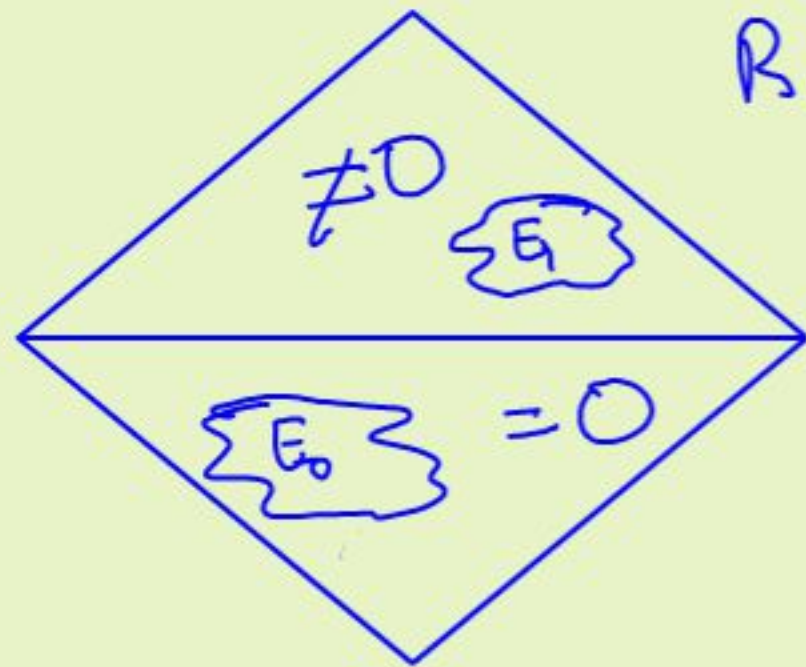
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Clm 2:  $P \neq 0$  &  $P|_{\Sigma_0, B_{\epsilon \cdot 2^n}^n} \equiv 0 \Rightarrow \deg(P) \geq n/2.$

# Proof of Smolensky's thm



$R(x_1, \dots, x_n)$

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$$\Rightarrow \deg(R) = \Omega(n)$$

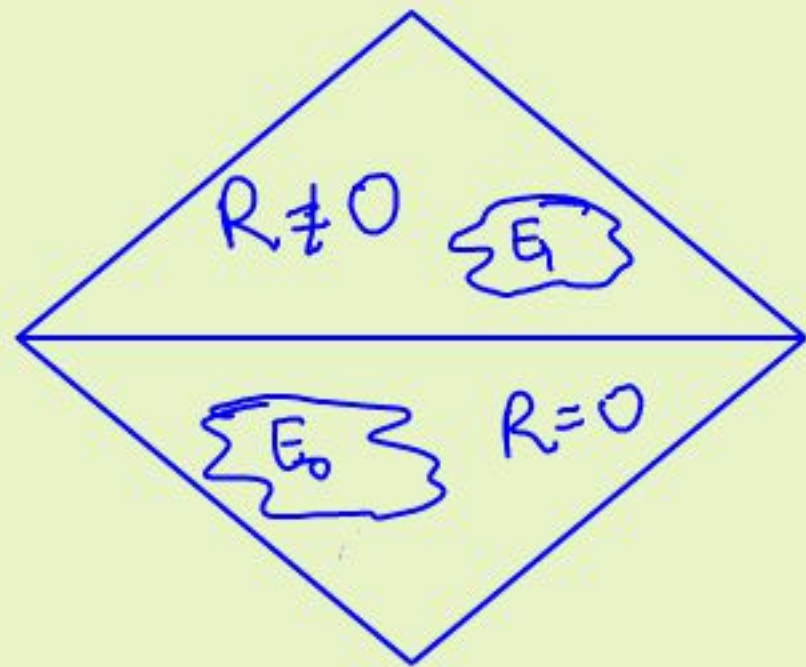
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Clm 2:  $P \neq 0$  &  $P|_{\Sigma_0, B_{\epsilon \cdot 2^n}^n} \equiv 0 \Rightarrow \deg(P) \geq n/2$ .

Pf of Smolensky:  $P = Q \cdot R \Rightarrow \deg(P) \geq n/2$ .  $\square$

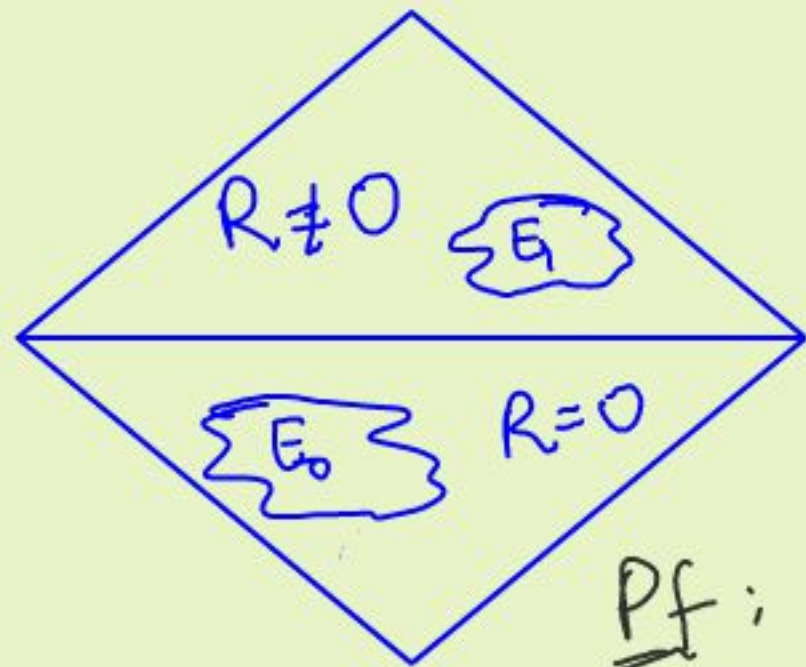
# Proof of Smolensky's thm



$$E = E_0 \vee E_i, \quad |E_i| \leq 2\epsilon \cdot 2^n.$$

Claim 1:  $\exists Q, \deg(Q) \leq n/2 - \Omega(\sqrt{n})$   
s.t.  $Q|_E = 0$  &  $Q.R \neq 0$ .

# Proof of Smolensky's thm



$$E = E_0 \vee E, \quad |E| \leq 2\epsilon \cdot 2^n.$$

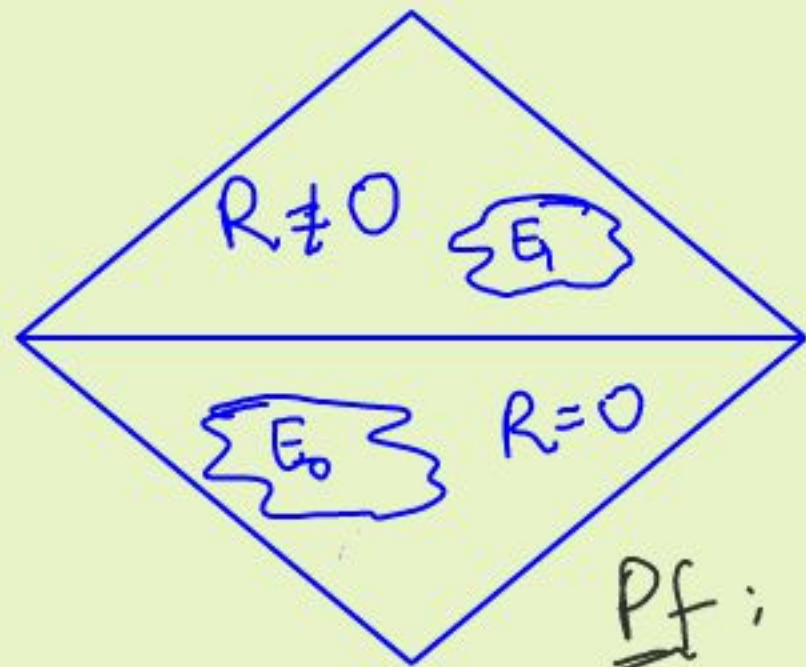
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Pf: Assume  $Q$  of  $\deg \leq D$ .

$$Q(x_1, \dots, x_n) = \sum_{I \subseteq [n]: |I| \leq D} \alpha_I x^I$$



# Proof of Smolensky's thm



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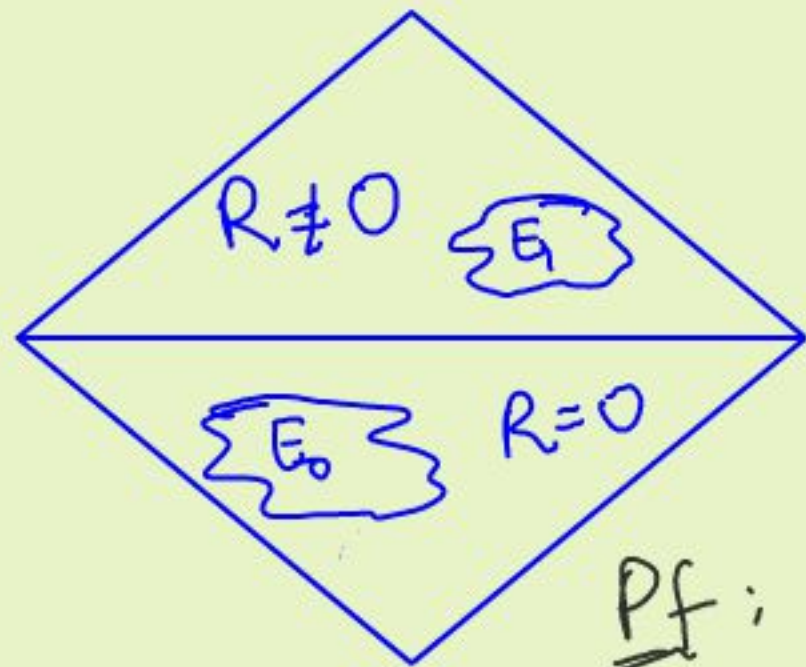
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# Proof of Smolensky's thm



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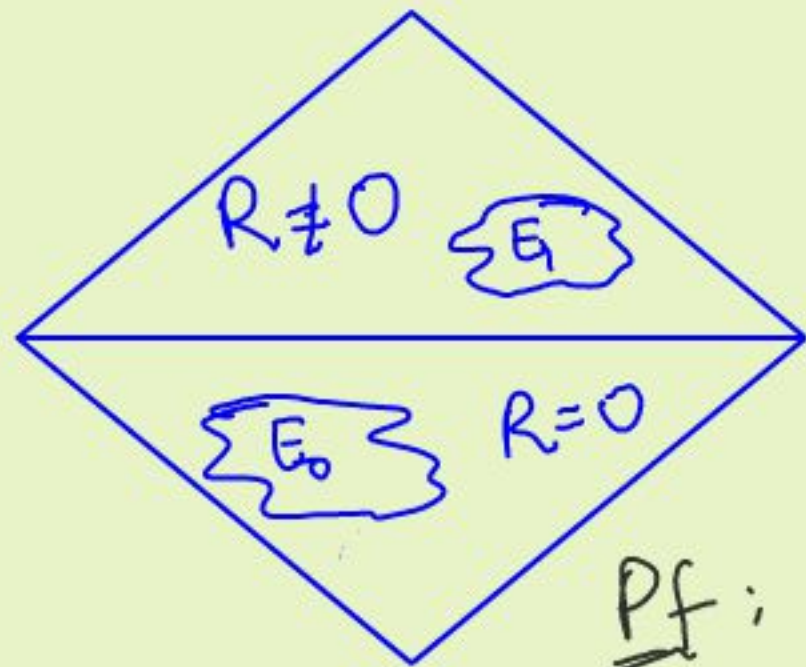
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# of variables =  $\binom{n}{\leq D}$

# Proof of Smolensky's thm



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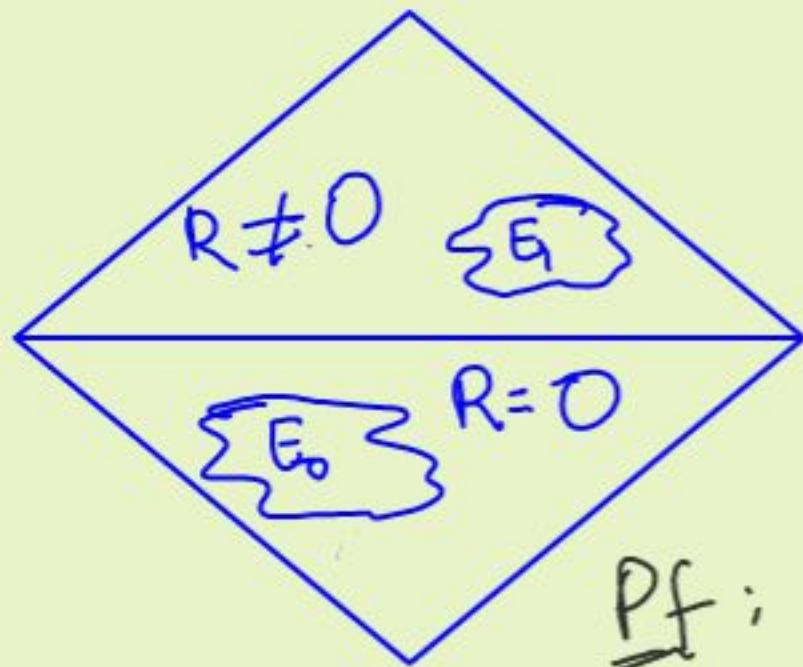
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# of constraints =  $|E| \leq 2\varepsilon \cdot 2^n$

# of variables =  $\binom{n}{\leq D} > 2\varepsilon \cdot 2^n$  if  $D \geq \frac{n}{2} - \Omega(\sqrt{n})$ .

# Proof of Smolensky's thm



$$E = E_0 \vee E, \quad |E| \leq 2\varepsilon \cdot 2^n.$$

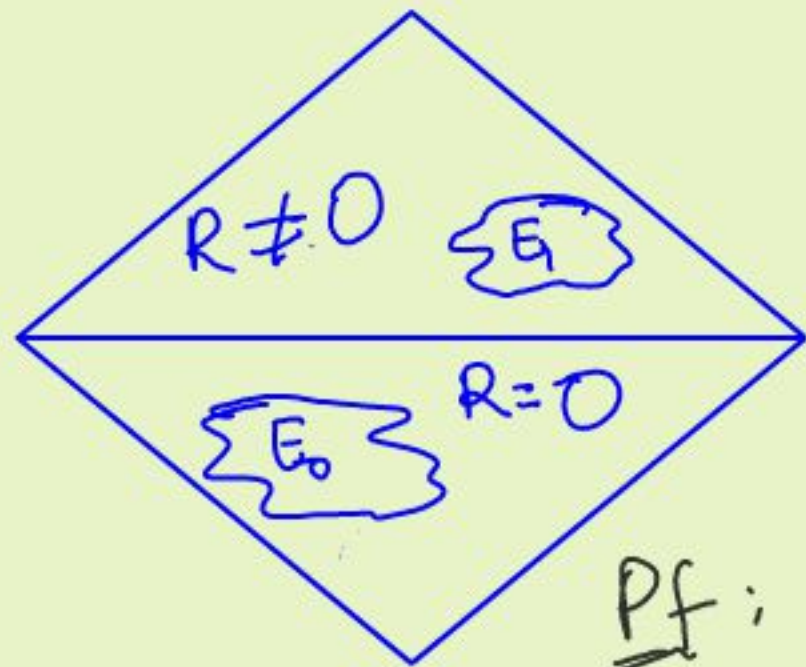
$$\underline{\text{Claim 1}}: \exists Q, \deg(Q) \leq n/2 - \Omega(\sqrt{n})$$

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$$\underline{\text{Pf}}: Q \text{ of } \deg \leq n/2 - \Omega(\sqrt{n}).$$

$$Q \neq 0, \quad Q|_E \equiv 0$$

# Proof of Smolensky's thm



$$E = E_0 \cup E_1, \quad |E_1| \leq 2\epsilon \cdot 2^n.$$

$$\underline{\text{Claim 1}}: \exists Q, \deg(Q) \leq n/2 - \Omega(\sqrt{n})$$

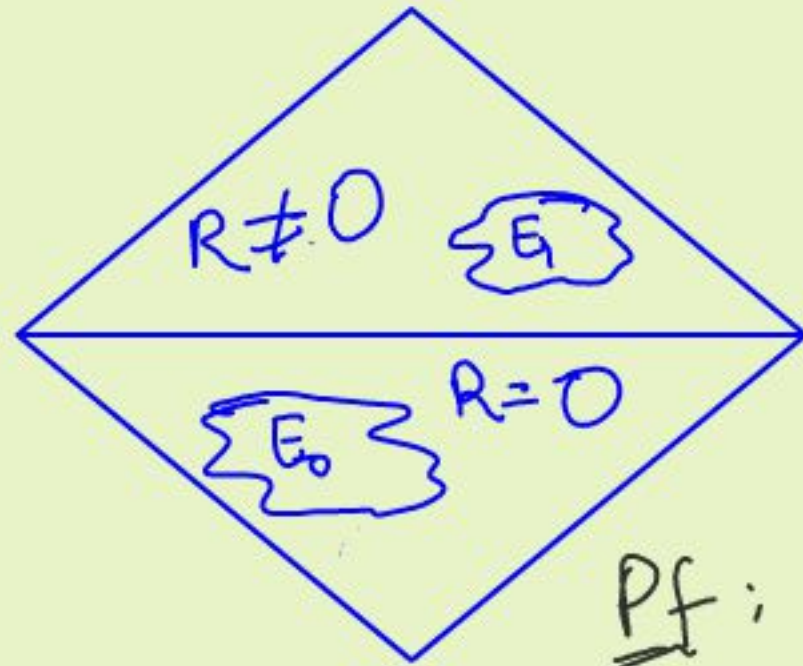
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$$\underline{\text{Pf}}: Q \text{ of } \deg \leq n/2 - \Omega(\sqrt{n}).$$

$$Q \neq 0, \quad Q|_E \equiv 0$$

$$\deg(Q) < n/2 \Rightarrow Q|_{\{0,1\}^n} \neq 0$$

# Proof of Smolensky's thm



$$E = E_0 \cup E, \quad |E| \leq 2\varepsilon \cdot 2^n.$$

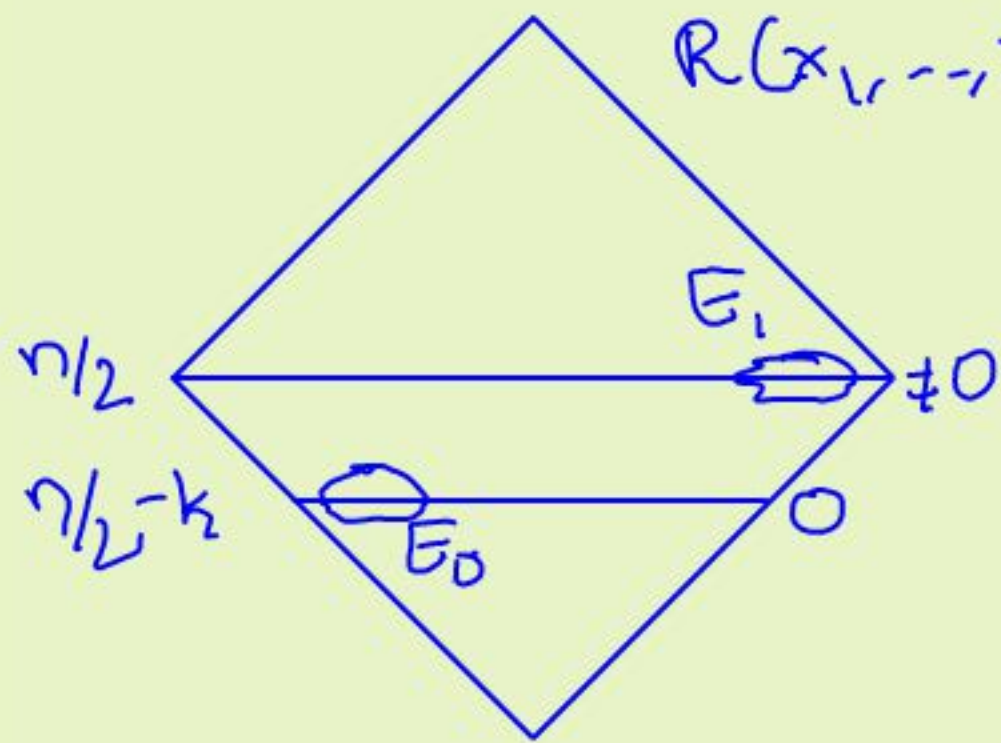
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Pf:  $Q$  of  $\deg \leq n/2 - \Omega(\sqrt{n})$ .

$$Q \neq 0, \quad Q|_E \equiv 0$$

$$\deg(Q) < n/2 \Rightarrow Q|_{\{0,1\}^n} \neq 0 \Rightarrow Q \cdot R \neq 0. \quad \square$$

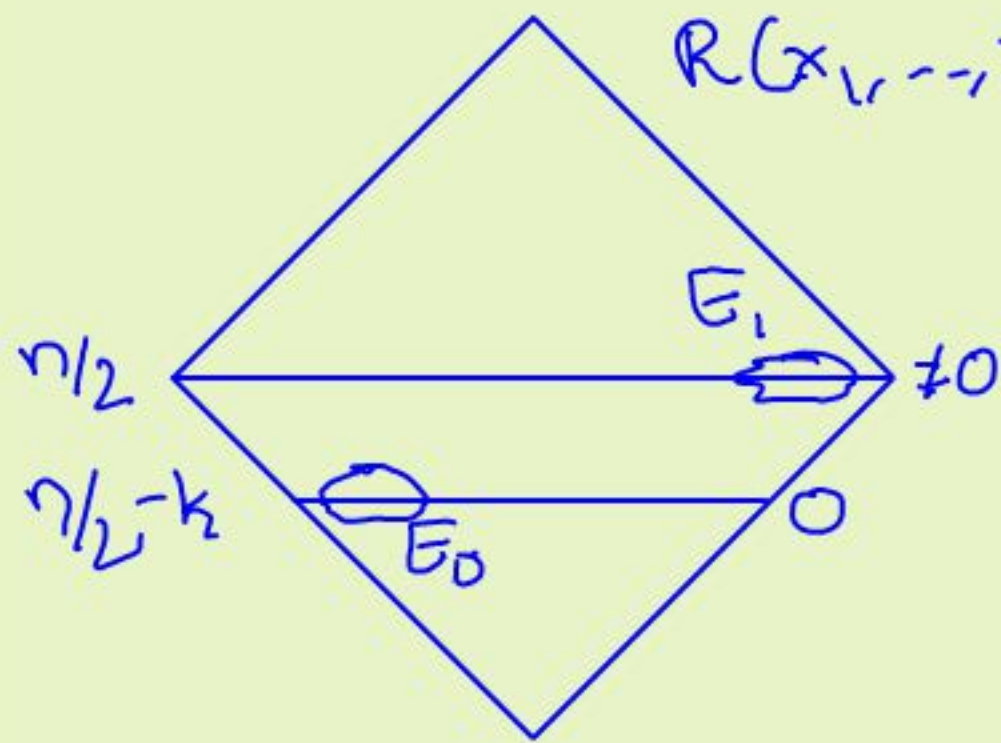
# Proof of Robust Hoeffding's Lemma



$$k = \rho^2 = C \cdot \sqrt{n}, \quad C \rightarrow \infty$$

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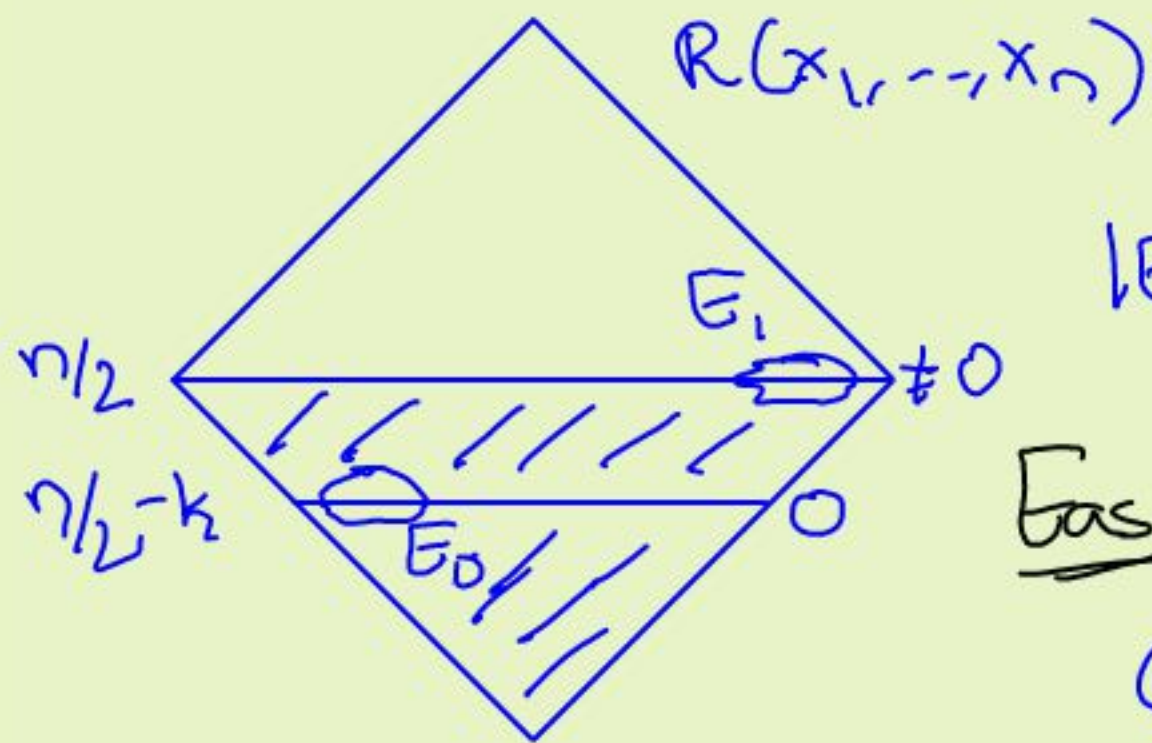
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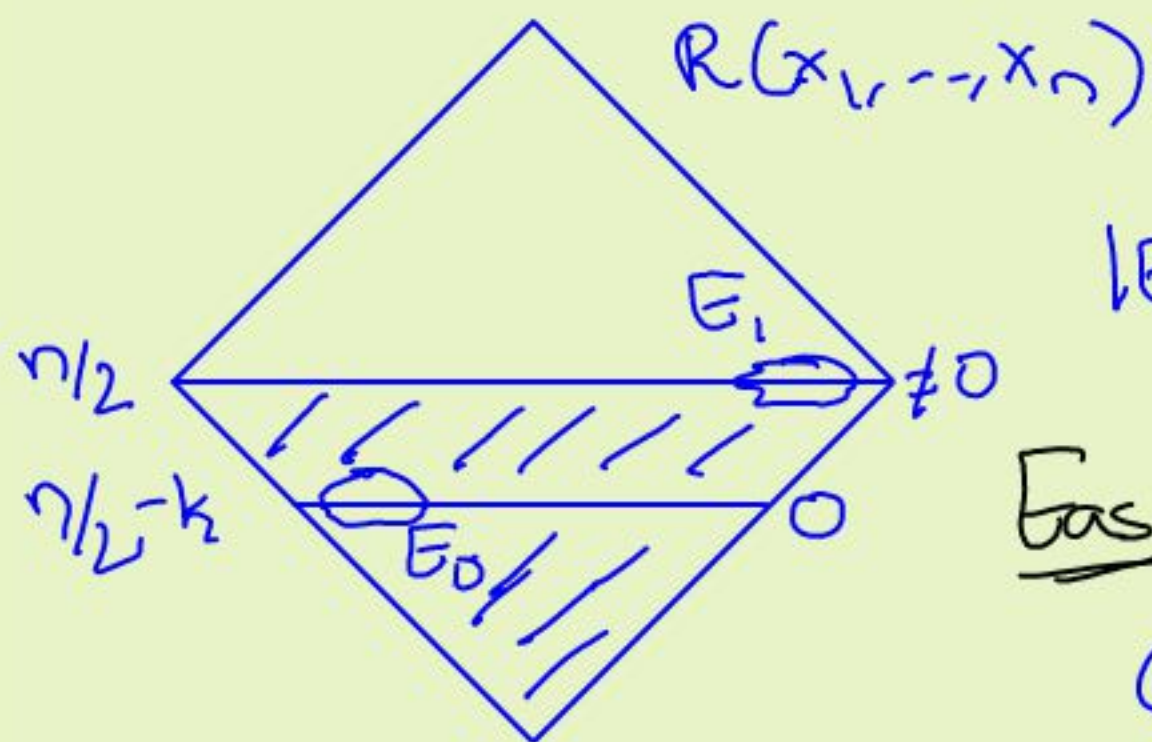
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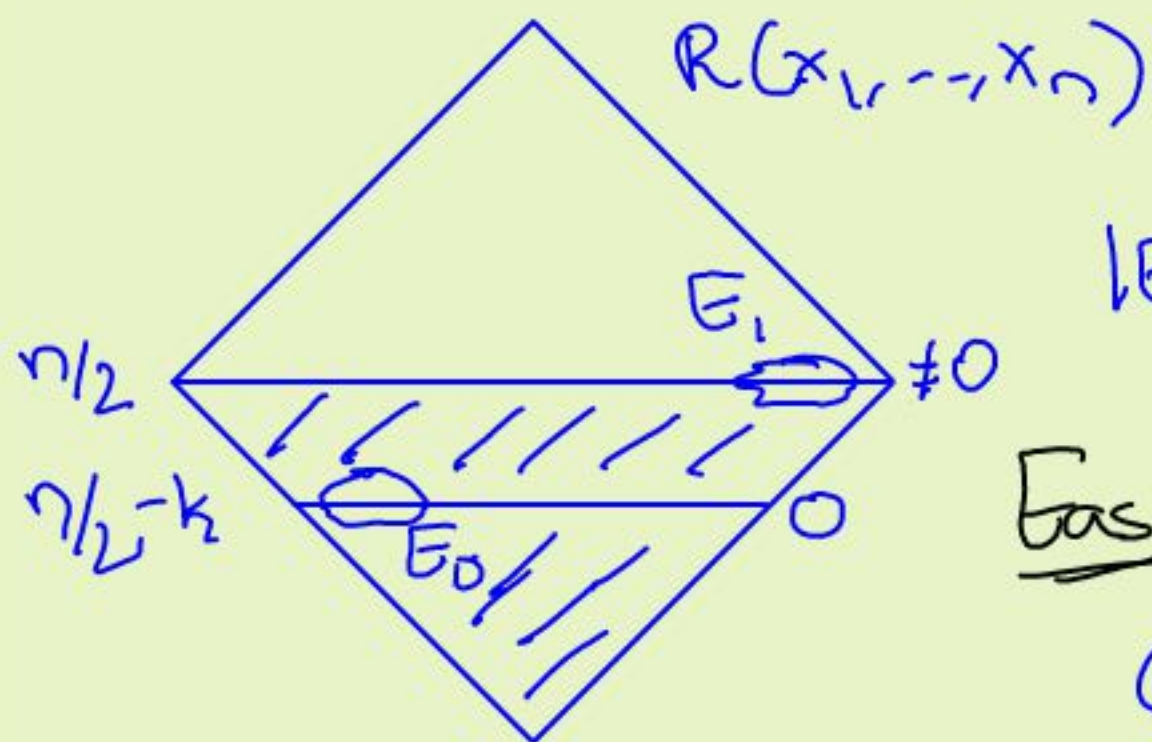
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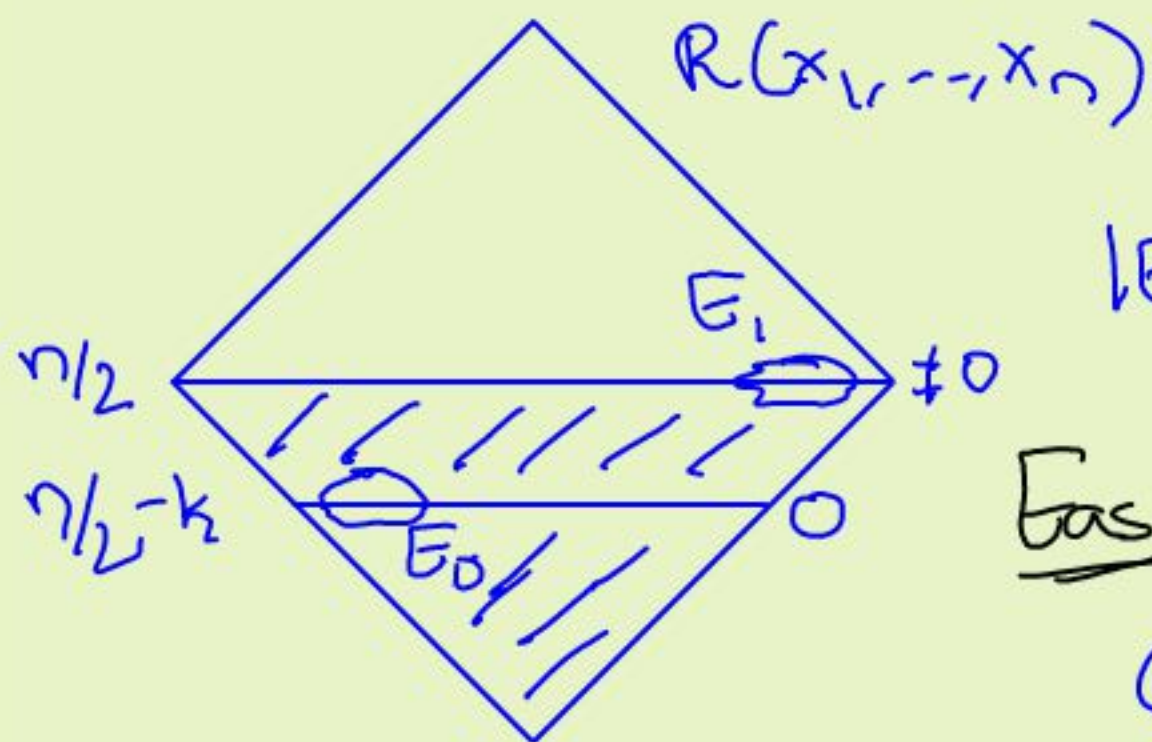
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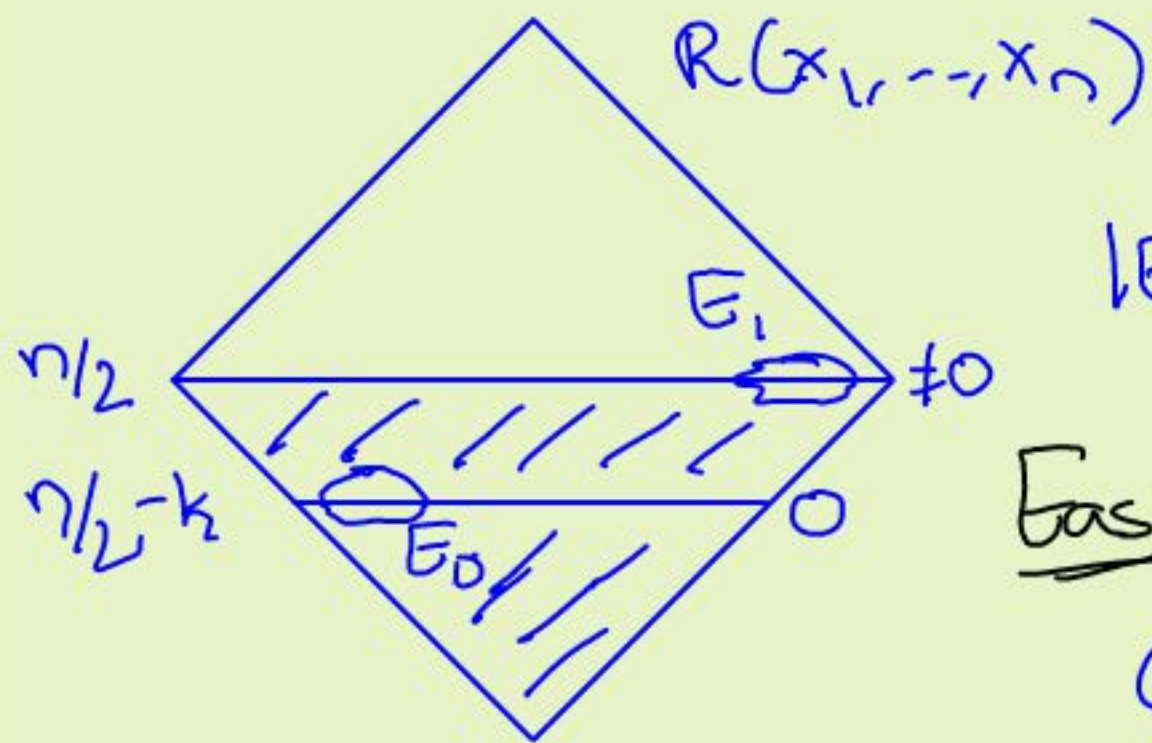
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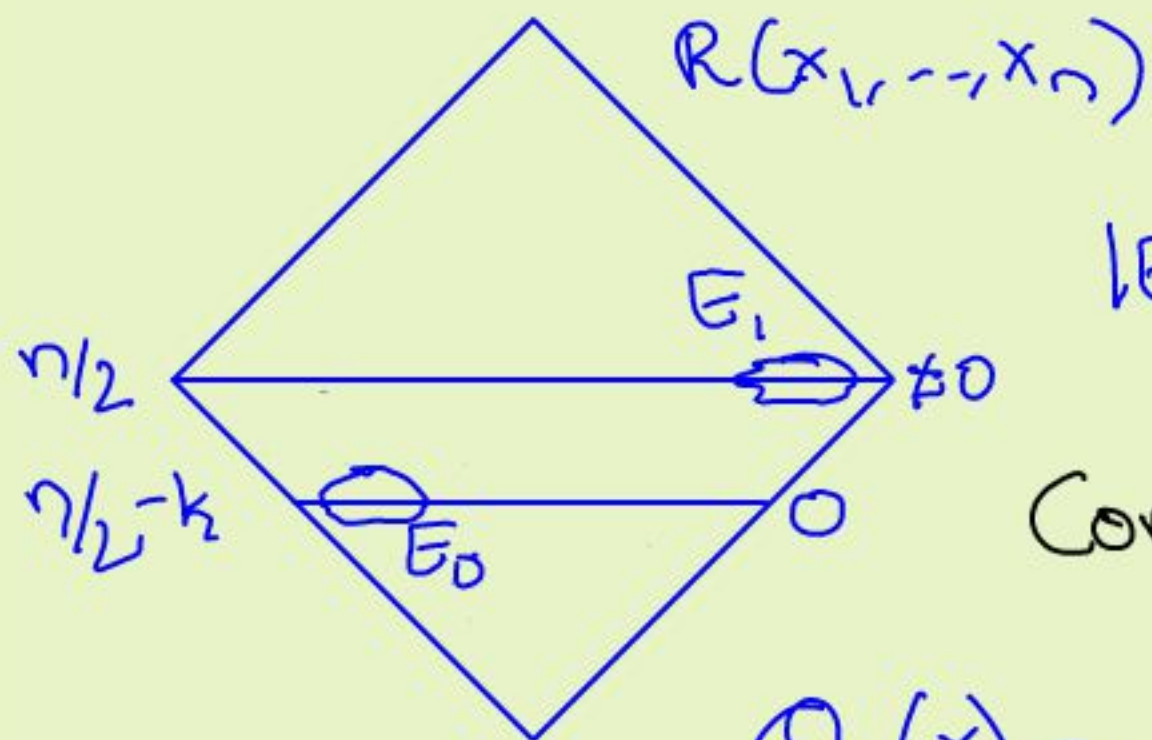
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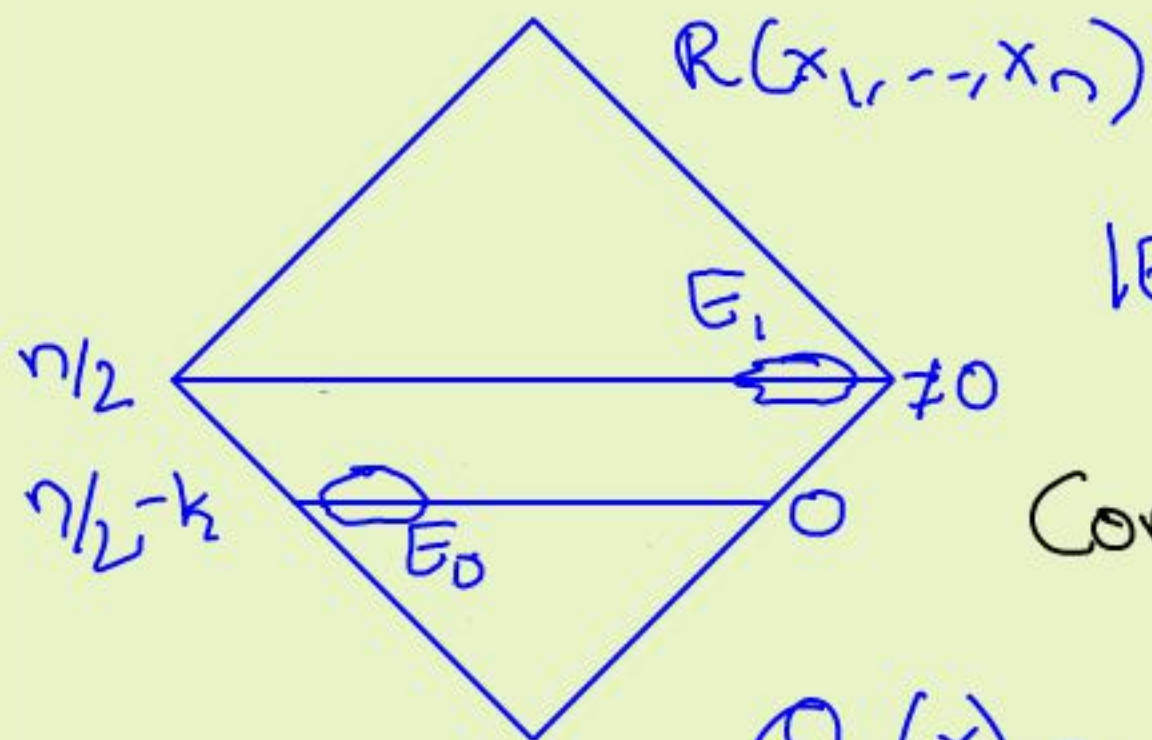
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$$Q_1(x) = \prod_{j=1}^{s-1} \left( \left( \sum_{|H|=p} x^H \right)^{p-1} - 1 \right)$$

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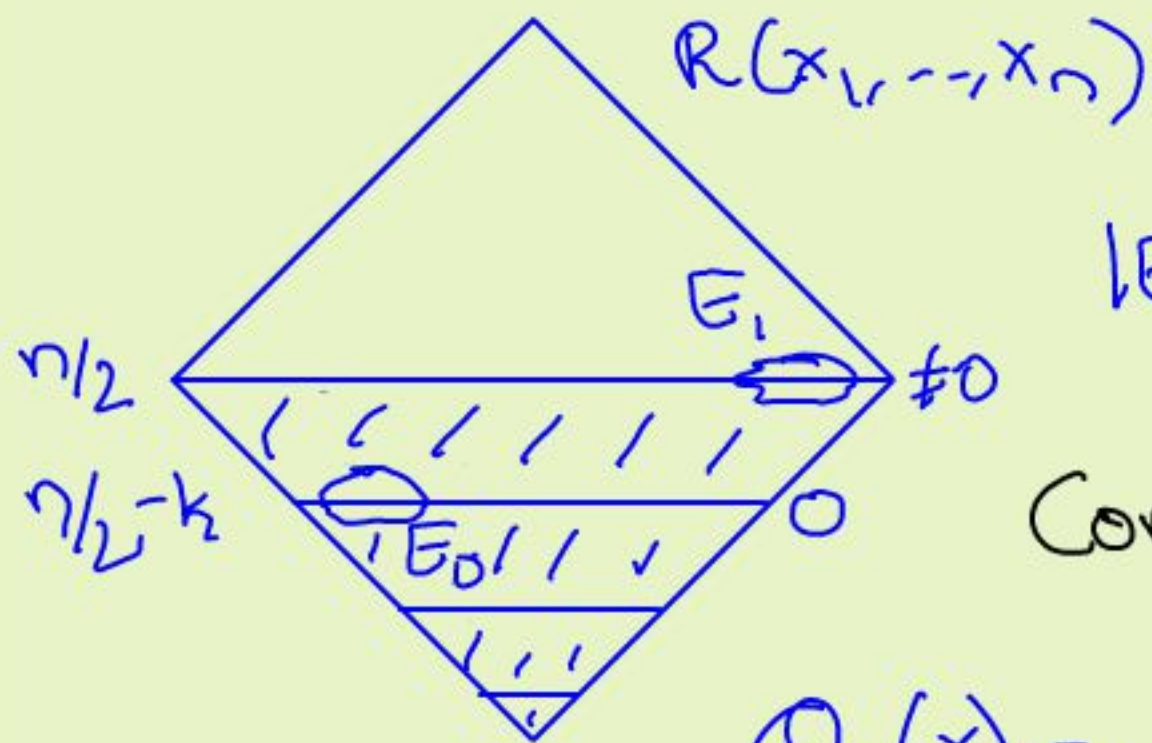
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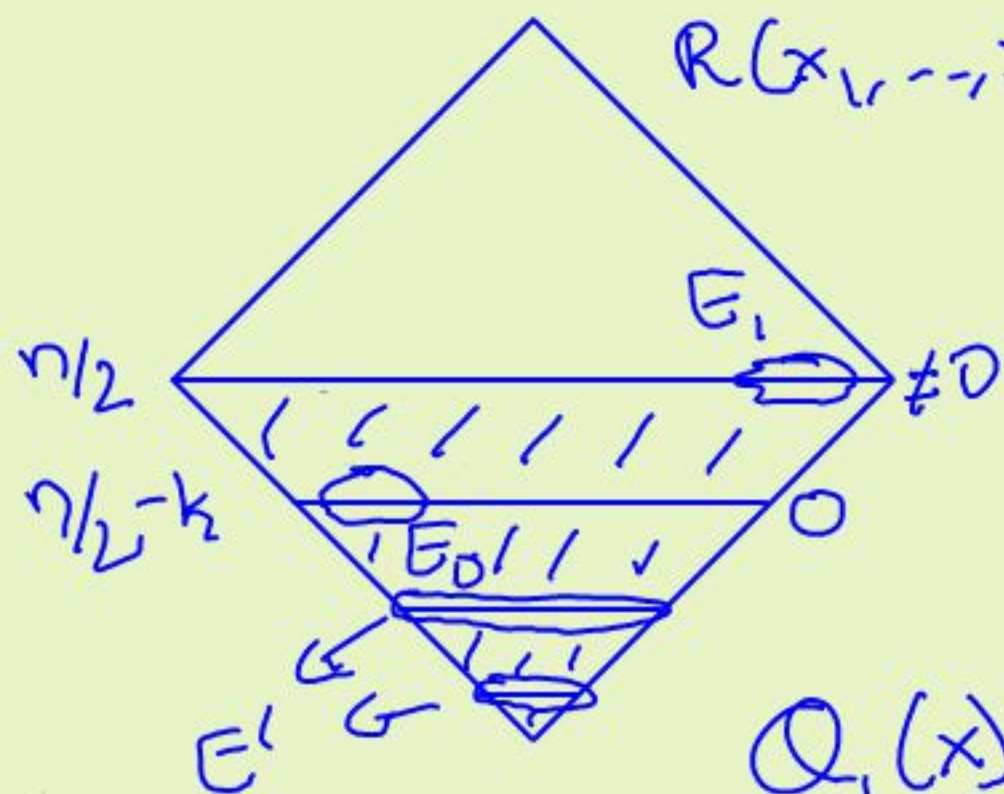
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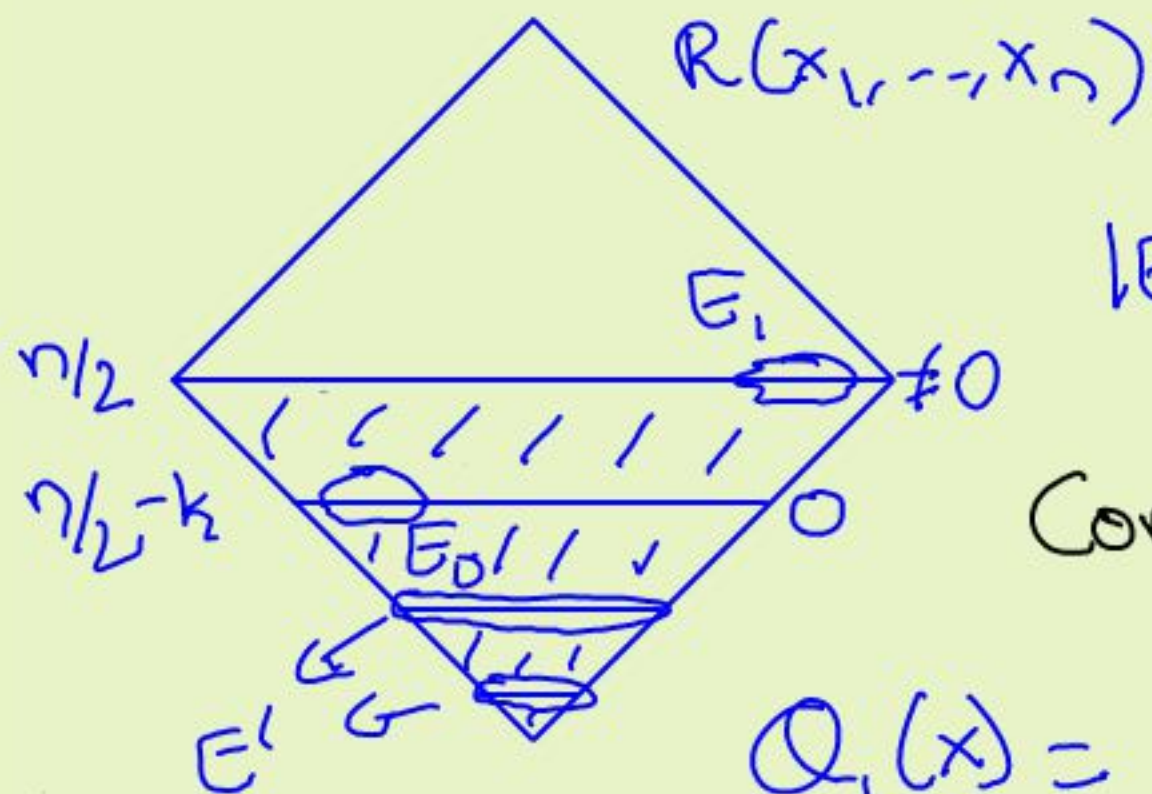
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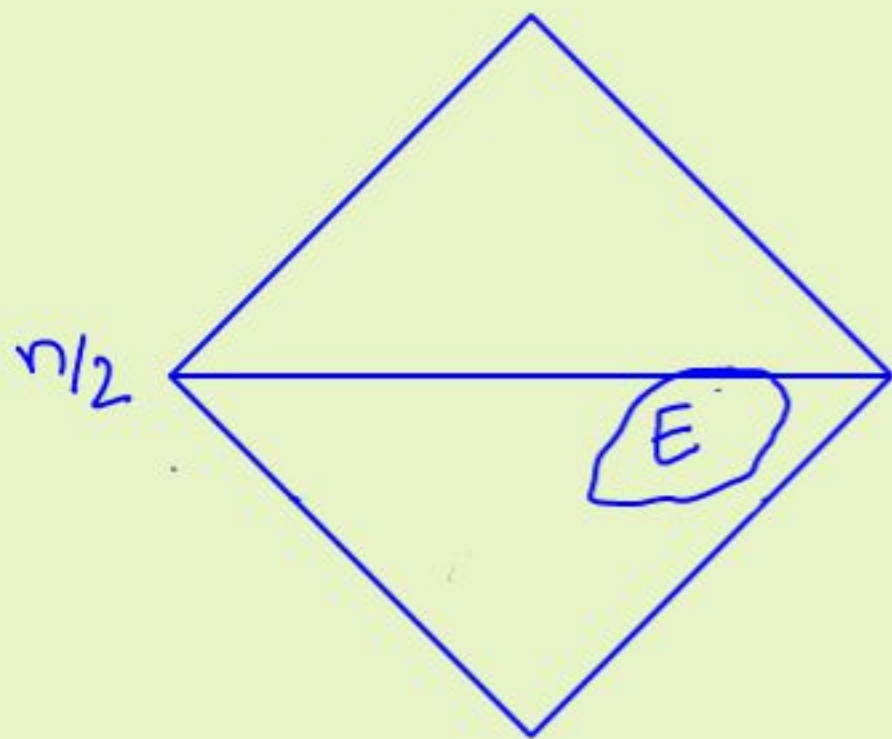
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$$E'' = E \cup E' \Rightarrow |E''| \leq \varepsilon \cdot \binom{n}{n/2} + \binom{n}{n/2 - 2k} + \binom{n}{n/2 - 3k} + \dots$$

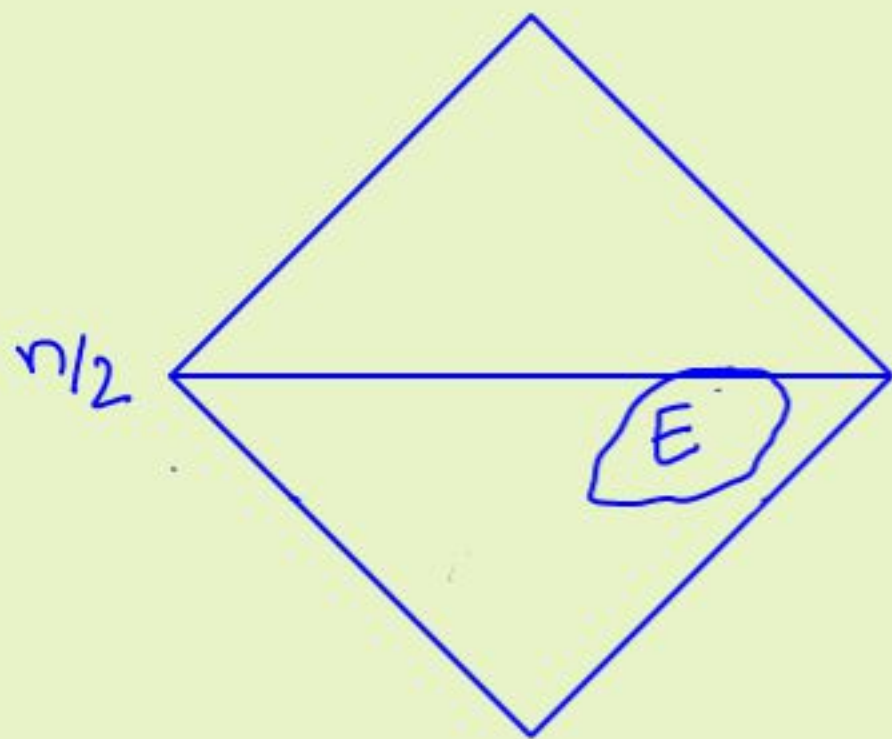
$$\leq 2\varepsilon \cdot \binom{n}{n/2}$$

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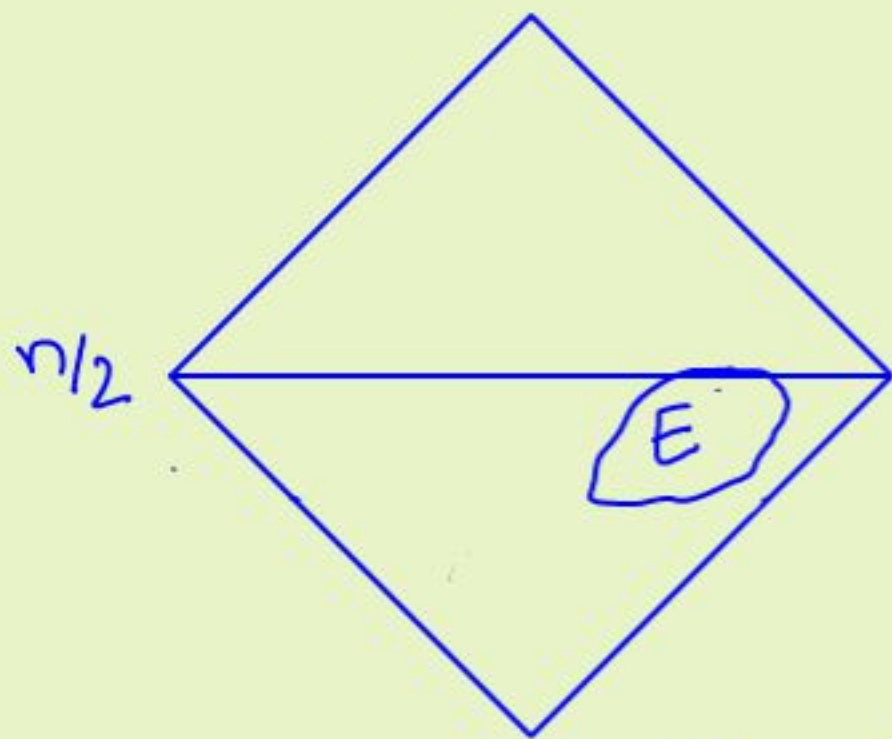


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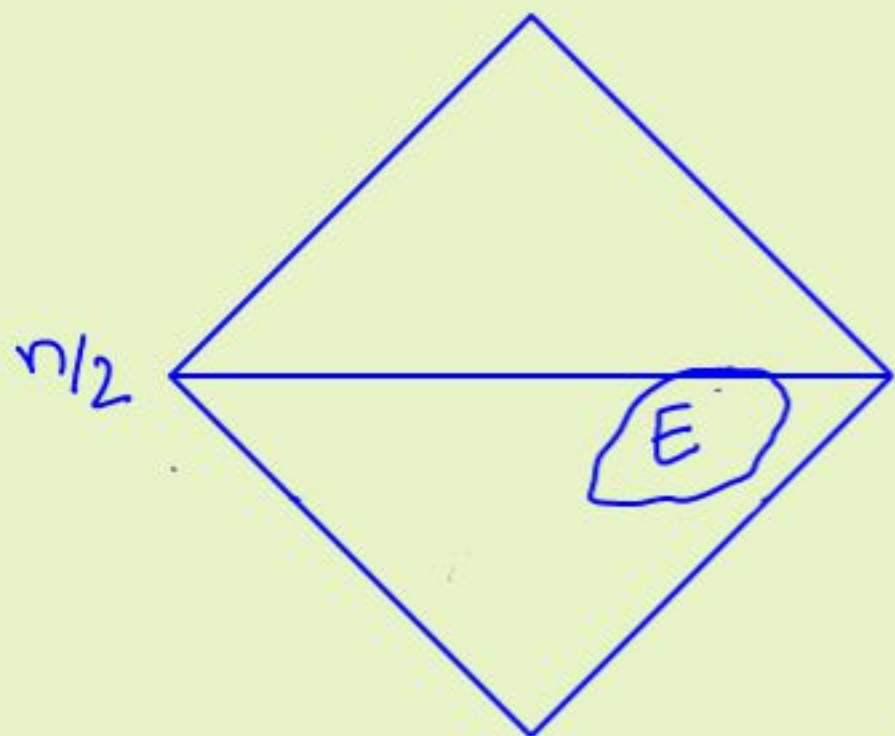
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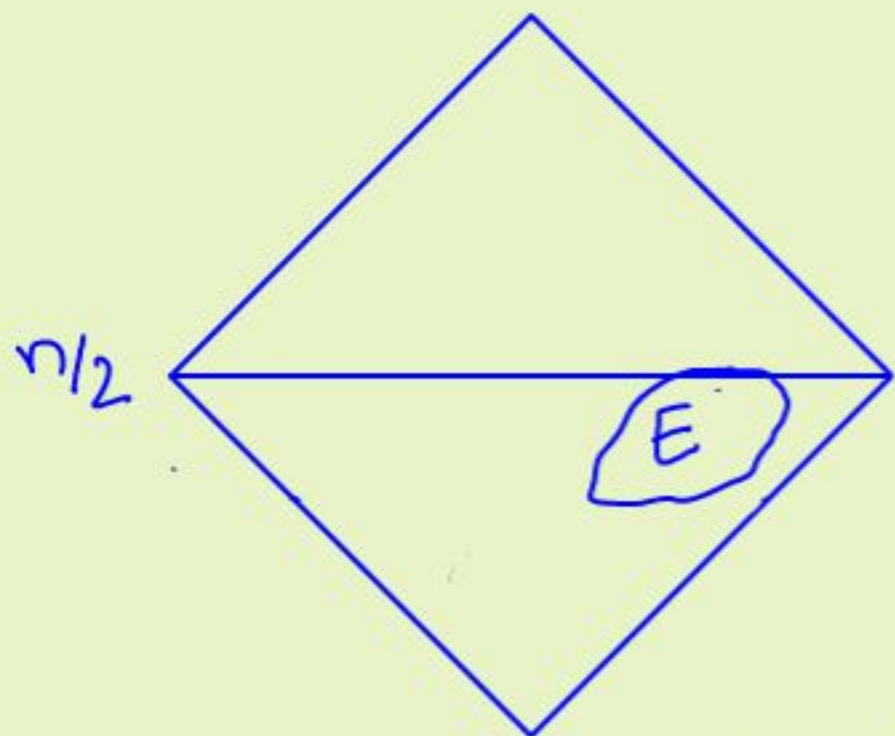
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[Wie-wang (4)]

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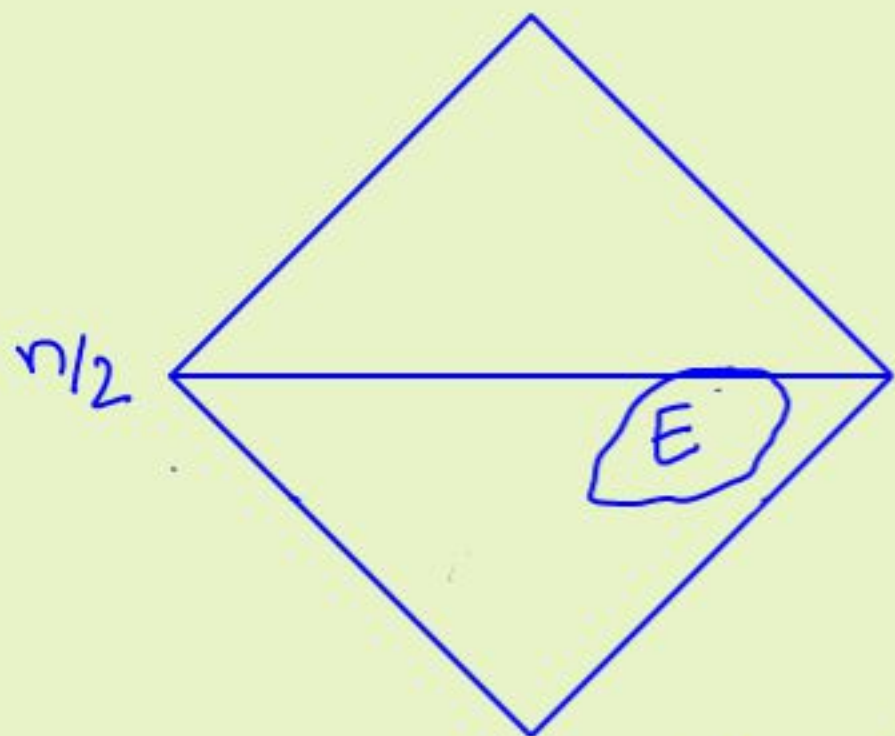
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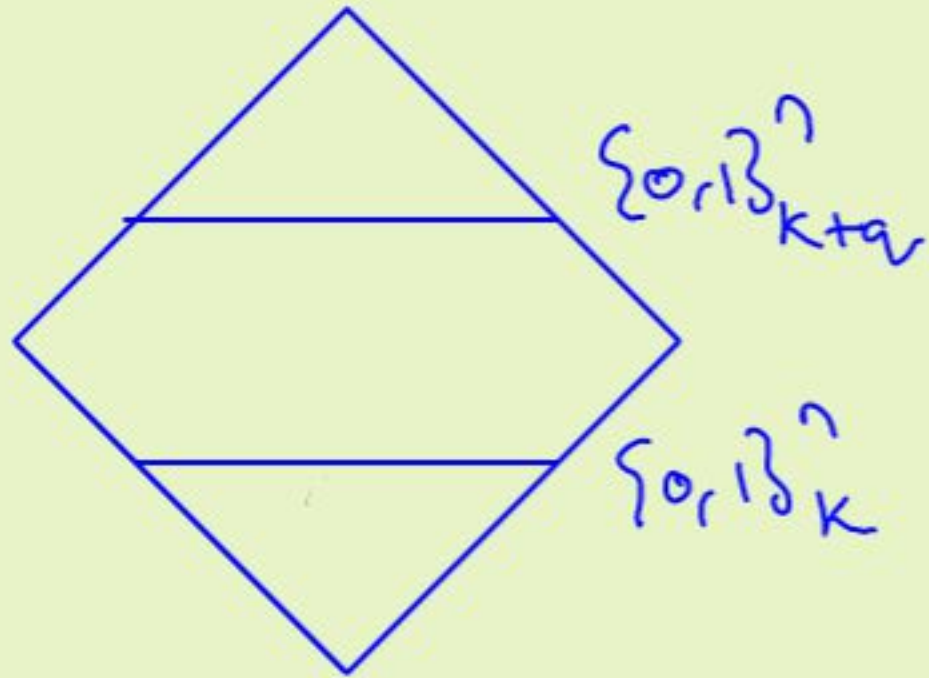
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□



# Main lemma



$$\mathbb{F}[x_1, \dots, x_n]$$

Q: lowest deg.  $R(x_1, \dots, x_n)$  s.t.

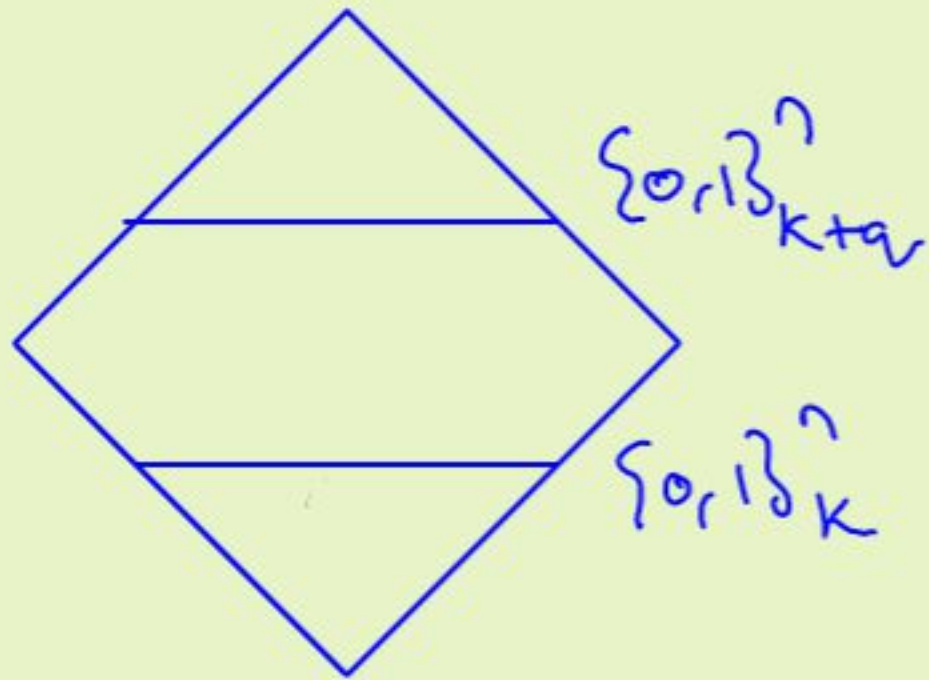
$$(I) \quad P_{|a|=k} [R(a) \neq 0] \leq \varepsilon$$

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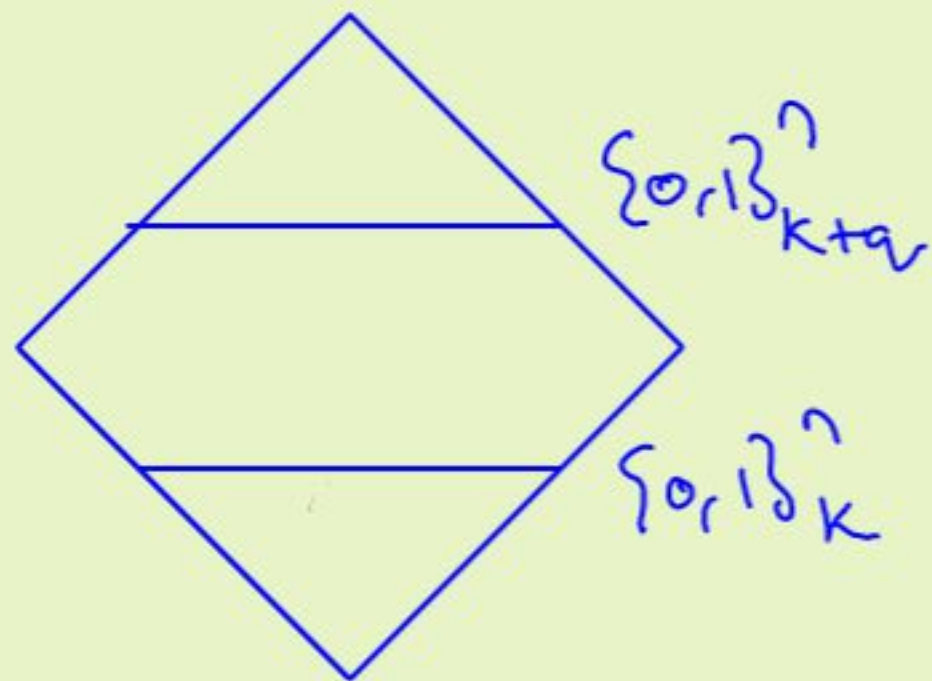
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Others?

Thanks!

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$D_\alpha^n$  - distribution over  $\{0, 1\}^n$ , each bit  $\uparrow$  with prob.  $\alpha$

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Programs

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Smaller  $\epsilon$ ?

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Corollary to Robust Hegedüs:  $\deg(R) = \Omega\left(\frac{1}{\delta} \log \frac{1}{\epsilon}\right)$ .