

Generalizations of KKL Theorem & Friedgut's Junta Theorem

Sushant Sachdeva
Simons Institute

Joint work with Madhur Tulsiani

Aug 27th, 2013

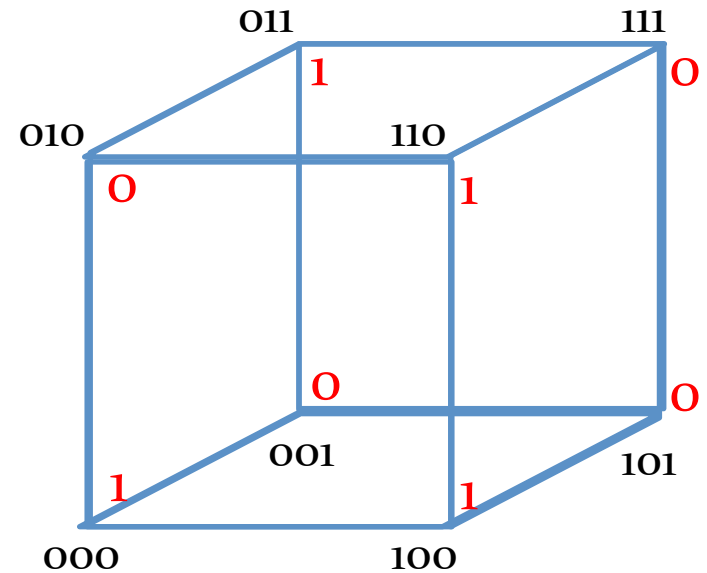
Boolean Functions and Influence

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

Influence (of coordinate i)

$$\text{Inf}_i(f) = \Pr_x[f(x) \neq f(x \oplus e_i)]$$

Flip i^{th} bit

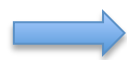


KKL Theorem

Assume f is balanced $\Pr_x[f(x) = 1] = 1/2$

Simplest such function $g(x) = x_j$ $\sum_i \text{Inf}_i(g) = 1$

Easy: For any balanced f $\sum_i \text{Inf}_i(f) \geq 1$



$$\exists i \text{ Inf}_i(f) \geq \frac{1}{n}$$

[Kahn-Kalai-Linial]

For every boolean f , $\exists i \text{ Inf}_i(f) \geq \frac{\log n}{n}$

Friedgut's Junta Theorem

Easy: For any balanced f s.t. $\sum_i \text{Inf}_i(f) = 1 \implies \exists i f(x) = x_i$
or $1 - x_i$

What if $\mathbb{I} = \sum_i \text{Inf}_i(f)$ is larger but still constant?

f depends only on 1 coordinate

[Friedgut]

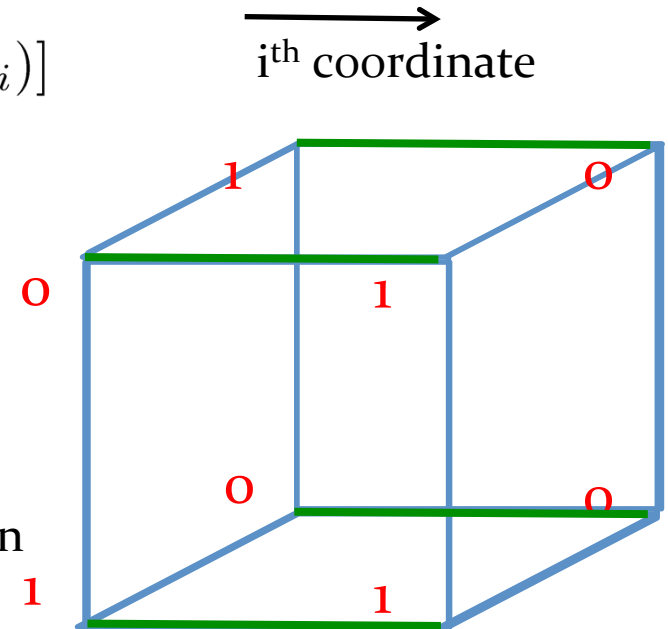
Every boolean f is ϵ -close to a boolean g that depends only on $\exp(\mathbb{I}/\epsilon)$ coordinates

g is a Junta

$$\epsilon\text{-close} \iff \Pr_x[f(x) \neq g(x)] \leq \epsilon$$

Influence

$$\begin{aligned} \text{Inf}_i(f) &= \Pr_x[f(x) \neq f(x \oplus e_i)] \\ &= \mathbf{E}_x \mathbb{1}[\text{Is } f \text{ non-constant on } (x, x \oplus x_i)] \\ &= \mathbf{E}_x \mathbf{Var}_{x_i} f(x) \\ &= \text{fraction of edges cut in } i^{\text{th}} \text{ direction} \\ &= \text{fractional size of boundary in } i^{\text{th}} \text{ direction} \end{aligned}$$



Identify f with the cut

$$\{ x \mid f(x) = 1 \}$$

Generalizations of KKL

For every balanced $f : \text{Domain}^n \rightarrow \{0,1\}$, \exists coordinate i s.t.

$$\text{Inf}_i(f) \geq \frac{\log n}{n}$$

Domain

$\text{Inf}_i(f)$

p-biased $\{0,1\}$

$$\Pr_x[f(x) \neq f(x \oplus e_i)]$$

[Mossel] Finite probability space Ω

$$\mathbf{E}_x \mathbf{Var}_{x_i} f(x)$$

[Bourgain-Kahn-Kalai-Katznelson-Linial]

$[0,1]$

$\mathbf{E}_x \mathbb{1}[f(x|_{i \leftarrow y}) \text{ non-constant for } y \in [0,1]]$

[Keller-Mossel-Sen]

Gaussian

measure of boundary in i^{th} direction

[S-Tulsiani]

Vertices of a Graph

fraction of edges cut in i^{th} direction $\frac{\sqrt{\log n}}{n}$ instead of $\frac{\log n}{n}$

[Cordero Erasquin-Ledoux]

$G^n = n\text{-fold Cartesian product of } G$

depend on the particular domain chosen ₆

Other Generalizations

[O'Donnell-Wimmer]

Certain classes of generalizations
of Cayley graphs
(non-product setting)

[Cordero Erasquin-Ledoux]

A framework that encompasses [KKL],
[S-Tulsiani],[O'Donnell-Wimmer]

[Keller]

Other generalizations of Influence
in the product setting

Generalizations of Junta Thm

Every $f : \text{Domain}^n \rightarrow \{0,1\}$ is ϵ^2 -close to a boolean g depending only on $\exp(\mathbb{I}/\epsilon)$ coordinates

$$\mathbb{I} = \sum_i \text{Inf}_i(f)$$

Domain

$\text{Inf}_i(f)$

[Friedgut]

p-biased $\{0,1\}$

$\Pr_x[f(x) \neq f(x \oplus e_i)]$

[S-Tulsiani]

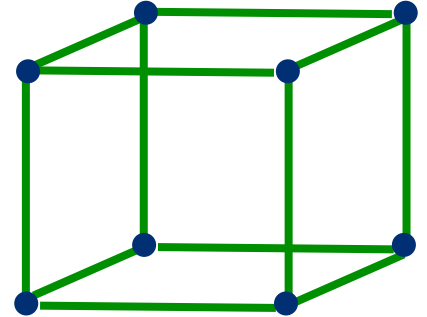
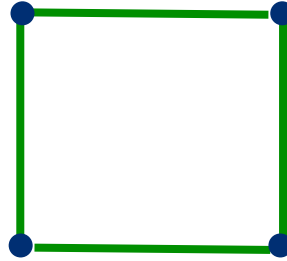
Vertices of a Graph

fraction of edges cut in i^{th} direction

Applications to hardness of approximation results [S-Saket][Khot-Saket]

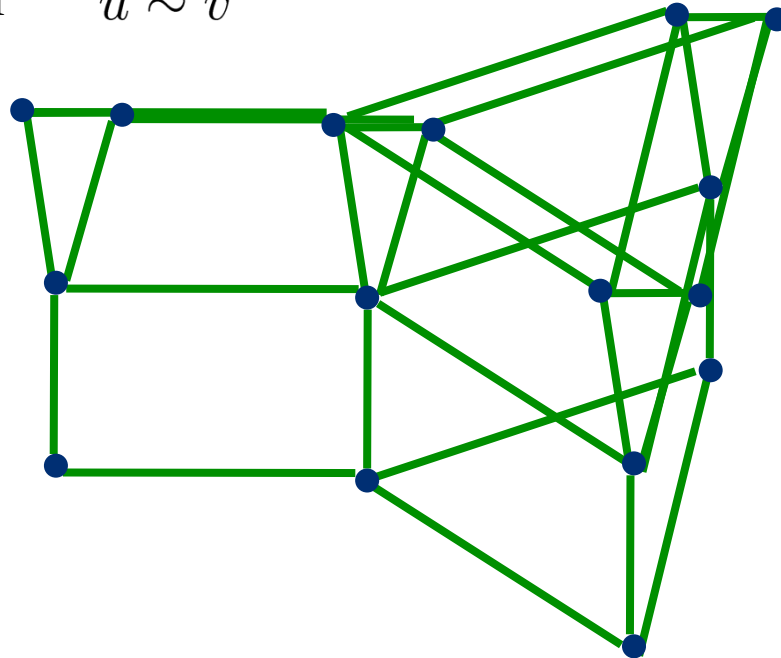
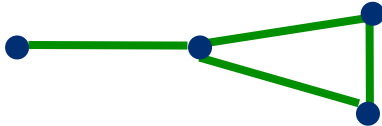
* The constants depend on the particular domain chosen

Cartesian Product of Graphs



$$(x_1, \dots, x_{i-1}, u, x_{i+1}, \dots, x_n) \sim (x_1, \dots, x_{i-1}, v, x_{i+1}, \dots, x_n)$$

whenever $u \sim v$



Log Sobolev Inequality

Given a graph $G(V,E)$, for any $f: V(G) \rightarrow \mathbb{R}$

$$\mathbf{E}_{(x,y) \in E(G)} (f(x) - f(y))^2 \geq \lambda(G) \mathbf{Var}(f)$$

Spectral Gap

Spectral Gap of G

for f is
nt

Log Sobolev Inequality

Given a graph $G(V,E)$, for any $f: V(G) \rightarrow \mathbb{R}$

$$\mathbf{E}_{(x,y) \in E(G)} (f(x) - f(y))^2 \geq \alpha(G) \mathbf{Ent}(f^2)$$

Log-Sobolev

Log Sobolev constant of G

$$\mathbf{Ent}(f^2) = \mathbf{E} f^2 \log f^2 - (\mathbf{E} f^2) \log(\mathbf{E} f^2)$$

Behavior under Cartesian Products

$$\alpha(G^n) = \frac{1}{n} \alpha(G)$$

Results for Cartesian Products

Given graph $G(V,E)$, define

$\text{Inf}_i(f)$ = fraction of edges cut along i^{th} direction

$$\mathbb{I} = \sum_i \text{Inf}_i(f)$$

[S-Tulsiani]

For every $f: V(G^n) \rightarrow \{0,1\}$, $\exists i \text{ Inf}_i(f) \geq \alpha(G) \frac{\log n}{n}$

Also follows from [Cordero Erasquin-Ledoux]

[S-Tulsiani]

Every $f: V(G^n) \rightarrow \{0,1\}$ is ϵ^2 -close to a boolean g that depends only on $\exp(\mathbb{I}/\alpha(G)\epsilon)$ coordinates

Proof Idea for Hypercube

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

Define

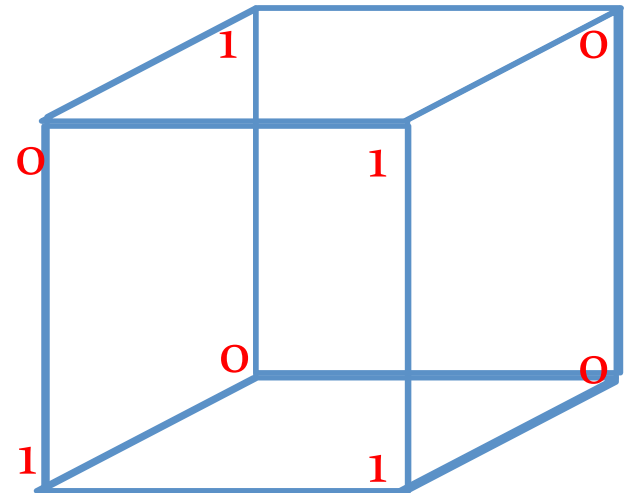
$$g_1(x) = \mathbf{E}_{x_1} f(x)$$

$$g_2(x) = \mathbf{E}_{x_2} g_1(x)$$

$$g_3(x) = \mathbf{E}_{x_3} g_2(x)$$

...

$$g_n(x) = \mathbf{E}_{x_n} g_{n-1}(x)$$



Proof Idea for Hypercube

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

Define

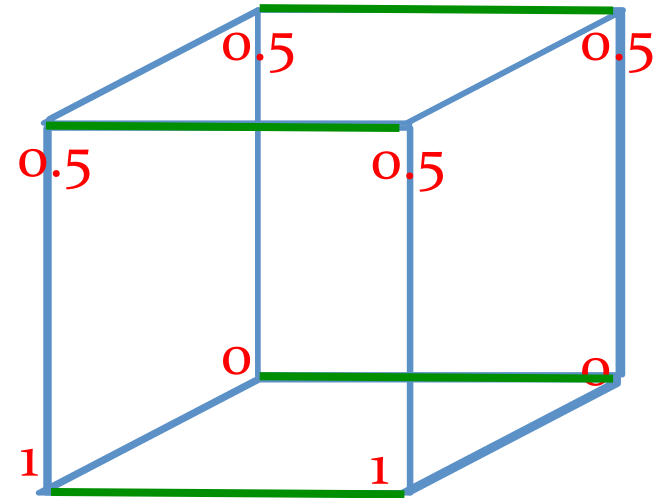
$$g_1(x) = \mathbf{E}_{x_1} f(x)$$

$$g_2(x) = \mathbf{E}_{x_2} g_1(x)$$

$$g_3(x) = \mathbf{E}_{x_3} g_2(x)$$

...

$$g_n(x) = \mathbf{E}_{x_n} g_{n-1}(x)$$



Proof Idea for Hypercube

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

Define

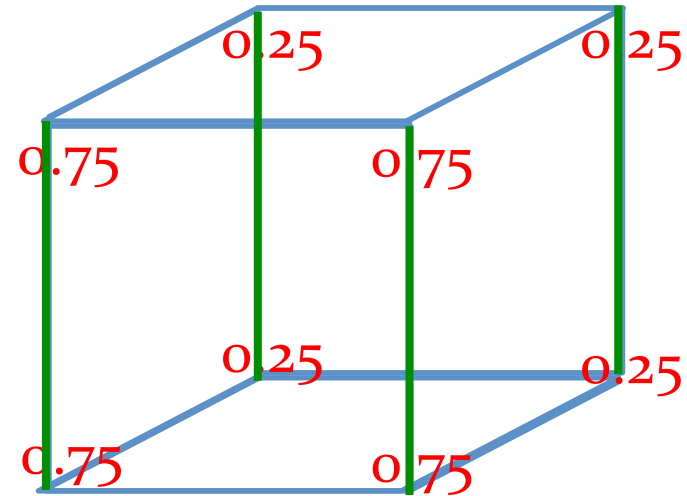
$$g_1(x) = \mathbf{E}_{x_1} f(x)$$

$$g_2(x) = \mathbf{E}_{x_2} g_1(x)$$

$$g_3(x) = \mathbf{E}_{x_3} g_2(x)$$

...

$$g_n(x) = \mathbf{E}_{x_n} g_{n-1}(x)$$



Proof Idea for Hypercube

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

Define

$$g_1(x) = \mathbf{E}_{x_1} f(x) \quad f_1(x) = f(x) - g_1(x)$$

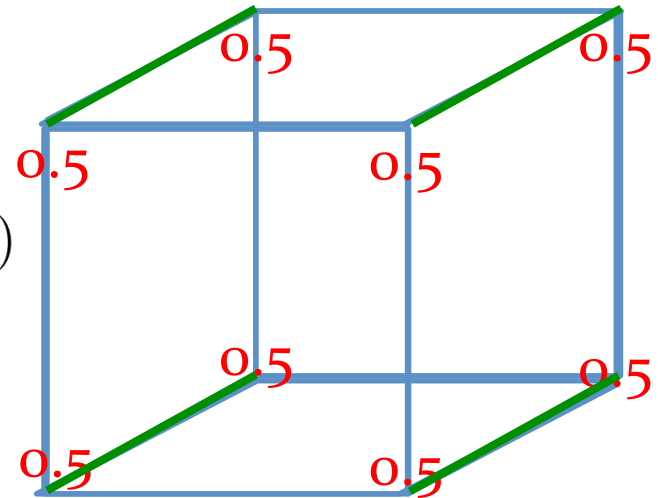
$$g_2(x) = \mathbf{E}_{x_2} g_1(x) \quad f_2(x) = g_1(x) - g_2(x)$$

$$g_3(x) = \mathbf{E}_{x_3} g_2(x)$$

...

...

$$g_n(x) = \mathbf{E}_{x_n} g_{n-1}(x) \quad f_n(x) = g_{n-1}(x) - g_n(x)$$



Proof Idea for Hypercube

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

Define

$$f_j(x) = \mathbf{E}_{x_1, \dots, x_j} f(x) - \mathbf{E}_{x_1, \dots, x_{j-1}} f(x)$$

Apply log-Sobolev inequality to each f_j “Martingale increments”

For a general graph G , define f_j with the measure $\mu(v) = \frac{\deg(v)}{\sum_u \deg(u)}$

Proof Idea for KKL

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

Define $f_j(x) = \mathbf{E}_{x_1, \dots, x_j} f(x) - \mathbf{E}_{x_1, \dots, x_{j-1}} f(x)$

Some nice properties

Zero mean $\mathbf{E}_x f_i(x) = 0$

Orthogonality $\mathbf{E}_x f_i(x) f_j(x) = 0$ for $i \neq j$

$$\sum_i \mathbf{E}_x f_i(x)^2 = \mathbf{E}_x (\sum_i f_i(x))^2 = \mathbf{Var}(f) = 1$$

E-Orthogonality

$$\sum_i \mathbf{E}_{(x,y) \sim E(G^n)} (f_i(x) - f_i(y))^2 = \mathbf{E}_{(x,y) \sim E(G^n)} (f(x) - f(y))^2$$

Norm bounds

$$\mathbf{E}_x f_j(x)^2 \leq \mathbf{Var}_j(f)$$

$$\mathbf{E}_x |f_j(x)| \leq \mathbf{Var}_j(f)$$

Results for Cartesian Products

Given graph $G(V,E)$, define

$\text{Inf}_i(f) =$ fraction of edges cut along i^{th} direction

$$\mathbb{I} = \sum_i \text{Inf}_i(f)$$

[S-Tulsiani]

For every $f: V(G^n) \rightarrow \{0,1\}$, $\exists i \text{ Inf}_i(f) \geq \alpha(G) \frac{\log n}{n}$

Also follows from [Cordero Erasquin-Ledoux]

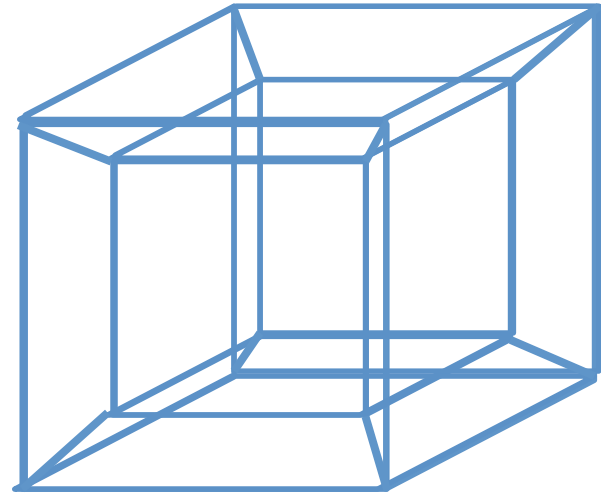
[S-Tulsiani]

Every $f: V(G^n) \rightarrow \{0,1\}$ is ϵ^2 -close to a boolean g that depends only on $\exp(\mathbb{I}/\alpha(G)\epsilon)$ coordinates

Future Directions

- Generalize the theorems to more general product settings
- How about other non-product settings?
- If $\mathbb{R}(G)$ the best constant?

Thanks!



Proof idea for KKL

Let $V = \max_j \mathbf{Var}_j(f)$

Case 1: $V \geq \frac{\log n}{n}$

$$\exists i \text{ Inf}_i(f) \geq \Phi(G)V \geq \alpha(G) \frac{\log n}{n}$$

Case 2: $V \leq \frac{\log n}{n}$

Lemma: $\mathbf{E} f_j^2 \log f_j^2 \geq -\frac{\alpha(G)}{n} \left(\frac{1}{n} + \mathbf{E} f_j^2 \right) \log nV$

$$\mathbf{E}_{(x,y) \in E(G^n)} (f_j(x) - f_j(y))^2 \geq -\frac{\alpha(G)}{n} \left(\frac{1}{n} \log nV + \mathbf{E} f_j^2 \log nV^2 \right)$$

$$\mathbf{E}_{(x,y) \in E(G^n)} (f(x) - f(y))^2 \geq -\frac{\alpha(G)}{n} (\log nV + \log nV^2) = \Omega \left(\frac{\alpha(G)}{n} \log n \right)$$

$$\max_j \text{Inf}_j(f) \geq \mathbf{E}_{(x,y) \in E(G^n)} (f(x) - f(y))^2$$