

The KKL inequality ~ 1988

Influence + Bool Fns: $\{0, 1\} \xrightarrow{\uparrow} \{-1, 1\}$.

\nearrow \nwarrow \nearrow \nwarrow easier to work with.

Def: Boolean functions $f: \mathbb{Z}_n \rightarrow \{0, 1\}$

$\underbrace{\{-1, 1\}^n}_{\text{flipped}}$ w_i has k th bit $k \in [n]$

Def: k pivotal for f for w if $f(w) \neq f(w_i)$ \sim flipped

Def: Influence - $I_k(f) := P(k \text{ pivotal for } f)$

Ex: $f(x_1, \dots, x_n) = x_1 \rightarrow I_1(f) = 1, I_k(f) = 0, k \neq 1.$

$$\sim \frac{1}{n} \sum I_k(f) = \frac{1}{n}.$$

Bounds on Influence:

Discrete Poincaré: $f: \mathbb{Z}_n \rightarrow \{-1, 1\}$ then $\overline{\text{Var}(f)} \leq \sum I_i(f)$,

so $I_i(f) \geq \overline{\text{Var}(f)}/n$, some i .

Efron-Stein

... KKL gives logarithmic improvement. \sim

KKL: $\exists c > 0$ s.t. if $f: \mathbb{Z}_n \rightarrow \{0, 1\}$, then $\exists i$ s.t.
 $I_i(f) \geq c \overline{\text{Var}(f)} \log n / n$.

Ex: Bound is sharp \Rightarrow Ben-Or + Linial give an example

--- Part $[n]$ into blocks of length

$1, \dots, n$.

$\log_2(n) - \log_2(\log_2(n))$. Set $f_{n,t}$ to be 1 if
 \exists block containing all 1s and 0 otherwise.

(Varies away from 0 + influence smaller than $c(\log n)/n$, $c < \infty$)

$$I_i(f_{n,t}) \leq c \overline{\text{Var}(f)} \log n / n. \quad \forall n \geq 1.$$

A Bit of Fourier analysis on \mathbb{Z}_n :

$(\mathbb{Z}_n, \mathbb{Z}^{\mathbb{Z}_n}, \mathbb{P})$, $\mathbb{P} = (\frac{1}{2}\delta_0 + \frac{1}{2}\delta_1)^{\otimes n}$ ~ uniform meas.

Given $f, g \in L^2(\mathbb{Z}_n)$ the inner product:

$$\langle f, g \rangle := \underline{\mathbb{E}[fg]} \quad \text{~character of } \mathbb{Z}_n \text{ as a group.}$$

Def: $S \subseteq [n]$, $\underline{\chi_S(x)} := \prod_{i \in S} x_i$. ~ Forms orthonormal basis for $L^2(\mathbb{Z}_n)$.

We view \mathbb{Z}_n as the group \mathbb{Z}_2^n . $S = \{1, 3, 4\} \subseteq [4]$, $x = (1011)$.

$$\chi_S(xy) = \prod_{i \in S} x_i y_i = \prod_{i \in S} x_i \prod_{i \in S} y_i = \chi_S(x)\chi_S(y).$$

$$\Rightarrow \boxed{f = \sum_{S \subseteq [n]} \hat{f}(S) \chi_S,} \quad \hat{f}(S) = \underline{\mathbb{E}[f \chi_S]}. \quad (\text{Walsh coeff})$$

How does this target connect to the original question of influence?

~~Prep:~~ $f: \mathcal{L}_n \rightarrow \{0, 1\}$ thru:

$$I_k(f) = 4 \sum_{S \in k \text{-sets}} \hat{f}(S)^2 \quad \text{and}$$

sum of influences

"in direction of k^{th} bit"

Proof: introduce the discrete derivative: $\nabla_k f(w)$ (flip k^{th} bit)

$$\nabla_k f: \mathcal{L}_n \rightarrow \{1, 0, -1\} \quad \nabla_k f(w) = \frac{f(w) - f(wk)}{2}$$

$$\text{easy to see that } \nabla_k f(w) = \sum_{S \in k \text{-sets}} \hat{f}(S) [\chi_S(w) - \chi_S(wk)] = \sum_{S \in k \text{-sets}} 2\hat{f}(S)\chi_S(w)$$

$$\Rightarrow \nabla_k f(S) = \begin{cases} 2\hat{f}(S) & S \in k \\ 0 & \text{else} \end{cases}$$

$$I_k(f) = \|\nabla_k f\|_1 \text{ is clear. Also } \|\nabla_k f\|_1 = \|\nabla_k f\|_2 \quad \checkmark$$

\Rightarrow proportion by parallel.

$$\Rightarrow \|\nabla_k f\|_2^2 = \sum_S (\nabla_k f(S))^2 = \sum_{S \in k \text{-sets}} (2\hat{f}(S))^2 = \sum_S \hat{f}(S)^2 + \sum_S \hat{f}(S) \chi_S(wk)$$

$$\nabla_k f(w) = f(w) - f(wk) \\ = \sum_S \hat{f}(S) \chi_S(w)$$

The idea of the proof of KKL $\underline{4 \sum \hat{f}(S)^2}$.

The idea of the proof of KKL ($\frac{4\sum_i f(S)^2}{M}$). S, T, M

Say the influences are all small enough (or obviously done)

Then since $I_k(f) = \|\nabla_k f\|_2^2$, $\nabla_k f$ has small support.

$\Rightarrow \nabla_k f$ is more spread out (concentration on high freqs).

Let's try something... Take $1 \leq M \leq n$:

$$\sum_{S \in \mathcal{B} \setminus M} \hat{f}(S)^2 \leq 4 \sum_{0 < |S| \leq M} |S| \hat{f}(S)^2$$

$$\underline{\text{Var } f} = \sum_k \sum_{0 < |S| \leq M} \hat{\nabla}_k f(S)^2$$

$$\leq \sum_k \|\hat{\nabla}_k f\|_2^2 \quad (\text{Uncertainty principle} \dots \text{support on higher freqs.})$$

$$= I(f) \quad (\text{Parallel})$$

$$\|\hat{\nabla}_k f\|_2^2 I(f) = 4 \sum_S |S| \hat{f}(S)^2$$

$$\sum_k \sum_{0 < |S| \leq M} \hat{\nabla}_k f(S)^2 = \sum_{0 < |S| \leq M} \sum_k \hat{\nabla}_k f(S)^2 = \sum_{0 < |S| \leq M} |S| \hat{f}(S)^2$$

$$= \sum_{0 < |S| \leq M} |S| \hat{f}(S)^2$$

$$\hat{f}(\phi)^2 = E[f]$$

$$\text{We have } \sum_{|S| > 0} \hat{f}(S)^2 = \underline{\text{Var } f}.$$

If frac of Fourier mass of f conc. below M ,
 $I(f)$ larger than $\text{Var } f$, one influence is big.

If most of mass on larger freqs, $I(f)$ large using:

$$I(f) = 4 \sum_S |S| \hat{f}(S)^2.$$

... How do we understand the spectral info?

One More Tool we need before the proof of KKL:

Hypercontractivity: start w/ a rough function ... want to smooth it

out:
similar to heat flow.
For $f \in L^2(\mathbb{R}^n)$, $\|K_t * f\|_2 \leq \|f\|_2$, $1 < t < 2$. $\|T_p f\|_p \leq \|f\|_p$.

K_t is the heat kernel: $K_t(x) = (4\pi t)^{-n/2} \exp\left(-\frac{|x|^2}{4t}\right)$
with Gaussian meas $R^n(A) = (2\pi)^{-n/2} \int_A \exp\left(-\frac{1}{2}|x|^2\right) d\lambda^n(x)$.

Now in the case of the hypercube

Def^c: $w \in \mathbb{Z}_n$. Then ω_p is the dist of x ab. $x_i = w_i$ w/ prob p and x_i is randomized w/ prob $1-p$.

Def: $f: \mathbb{Z}_n \rightarrow \mathbb{R}$, $T_p f(w) = \mathbb{E}_{x \sim \omega_p} [f(x)]$.

Ex: $T_0 f = \mathbb{E}[f]$. $T_1 f = f$.

Easy to see T_p

Theorem - [Benami-Gross-Beckner]: $f: \mathbb{Z}_n \rightarrow \mathbb{R}$, $p \in [0, 1]$.

$$\|T_p f\|_2 \leq \|f\|_{1+p^2}.$$

"Hypercontractivity" inequality since $\|T_p f\|_2 \leq \|f\|_2$ a contraction,
but we even have $\|T_p f\|_2 \leq \|f\|_{1+p^2}$, $(1+p^2 \leq 2)$.

$$\begin{aligned} T_p \chi_S(w) &= T_p \left(\prod_{i \in S} w_i \right) \\ &= \prod_{i \in S} \mathbb{E}_{x \sim \omega_p} [w_i] \\ &= \prod_{i \in S} p w_i + \frac{1}{2} (-1 + 1) \\ p \prod_{i \in S} w_i &= p^{|S|} \chi_S(w). \end{aligned}$$

Now we see its usefulness in the proof:

$$n^{-3/4} \|\nabla_{\mathbb{R}^n} f\|_2 \geq c \text{Var}(f) \frac{\log n}{n}, \quad c > 0.$$

Case 1: $\exists k \in [n]$ s.t. $I_k(f) \geq n^{-3/4} \text{Var}(f)$. So clearly $I_k(f) \geq c \text{Var}(f) \frac{\log n}{n}$.

Case 2: ("small" influences):

$$\forall k, \quad I_k(f) = \|\nabla_k f\|_2^2 \leq \text{Var}(f) n^{-3/4}.$$

↳ Short spectrum mostly supported on high freqs:

Take $M \geq 1$:

$$\begin{aligned} \sum_{1 \leq |S| \leq M} \hat{f}(S)^2 &\leq \sum_{1 \leq |S| \leq M} |S| \hat{f}(S)^2 \leq 2^{2M} \sum_{1 \leq |S| \leq M} \left(\frac{1}{|S|}\right)^{2/5} |S| \hat{f}(S)^2 \\ &\leq 2^{2M} \sum_{|S|=1} \left(\frac{1}{|S|}\right)^{2/5} |S| \hat{f}(S)^2 \end{aligned}$$

~ Cheren noise level $\rho = \frac{1}{2}$
other more possible.

$$\begin{aligned} f = \sum_S \hat{f}(S) \chi_S &\mapsto \sum_S \rho |\hat{f}(S)| \chi_S. \quad \rho = \frac{1}{4} 2^{2M} \sum_k \|T_{1/2}(\nabla_k f)\|_2^2 \\ &\leq \frac{1}{4} 2^{2M} \sum_k \|\nabla_k f\|_{5/4}^2 \quad (\text{Hypercontractivity}) \\ &\leq 2^{2M} \sum_k I_k(f)^{8/5} \quad \Rightarrow \quad \|\nabla_k f\|_2^2 = I_k(f). \\ &\leq 2^{2M} n \text{Var}(f)^{8/5} n^{-\frac{3}{4} \cdot \frac{8}{5}} \quad (\text{Using } I_k(f) \leq \text{Var}(f) n^{-3/4}). \\ &\leq 2^{2M} n^{-1/5} \text{Var}(f) \quad (\text{Since clearly } \frac{\text{Var}(f)}{\text{Var}(f)} \leq \text{Var}(f)) \end{aligned}$$

~ $\text{Var}(f) \leq 1$ as f bool.

Set $M = \lfloor \frac{1}{20} \log_2 n \rfloor$:

$$\sum_{1 \leq |S| \leq \frac{1}{20} \log_2 n} \hat{f}(S)^2 \leq n^{1/10 - 1/5} \text{Var}(f) = n^{-1/10} \text{Var}(f).$$

... So most of spectrum conc above $\lfloor \log_2 n \rfloor$.

Thus we get:

$$\begin{aligned}
 \liminf_{k \rightarrow \infty} I_k(f) &\geq \frac{\sum_{k=1}^M I_k(f)}{n} = \frac{\sum_{k=1}^M \sqrt{|S| f(S)^2}}{n} \\
 &\geq \sqrt{n} \left[\frac{\sum_{|S| \geq M} |S| \hat{f}(S)^2}{n} \right] \\
 &\geq \frac{M}{n} \left[\sum_{|S| \geq M} \hat{f}(S)^2 \right] \\
 &= \frac{M}{n} \left[\text{Var}(f) - \frac{\sum_{1 \leq k \leq M} \hat{f}(S)^2}{n} \right] \\
 &\geq \frac{M}{n} \text{Var}(f) \left[1 - n^{-1/10} \right] \\
 &\geq c_1 \underbrace{\text{Var}(f)}_{\frac{\log n}{n}},
 \end{aligned}$$

where $c_1 = \frac{1}{20 \log 2} (1 - 2^{-1/10})$. □

Applications of the KKL inequality:

KKL gives a cleaner proof of the Harris-Kesten result ($p_c = 1/2$ on \mathbb{Z}^2 w/ Bernoulli weights)

Ref: <https://arxiv.org/pdf/math/0410359.pdf>

Why do computer scientists care? - implications to crypto.

- First time harmonic analysis has been applied to this kind of results.

Nice explanation of these connections and more: <http://www.math.chalmers.se/~steif/homepage.pdf>

Applications to social choice:

View $f: \mathcal{L}_n \rightarrow \{0,1\}$ as a "voting function"

That is, given the binary choice of n citizens, make sense decisions.

Ex Dictatorship $f(x_1, \dots, x_n) = x_1$ (Only person 1 matters)

- Nice Reference for connections to social choice:

<https://math.mit.edu/~elmos/fall19/survey.pdf>

Another example: $f: \{-1, 1\}^n \rightarrow \{0, 1\}$ monotone "voting" function,

KKL \Rightarrow you can "bribe" $O(\frac{1}{\log n})$ fraction of voters to swing the vote in favor of name $b \in \{0, 1\}$.

Note:

In the original KKL paper, a corollary is given that states there is a set of $\binom{n}{\log n} = O(n)$ variables which dominate f (By repeated application of KKL).