STAT 206A: Polynomials of Random Variables

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Lecture 0

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Consider finite probability spaces  $\Omega_1, \ldots, \Omega_n$ , with measures  $\mu_1, \ldots, \mu_n$ . Let  $\alpha_i$  be size of the smallest atom of  $(\Omega_i, \mu_i)$ , and set  $\alpha = \min_i \alpha_i$ . Let  $f \in L^2(\prod_i \mu_i)$  be a real function. Let  $\Delta_i f = \sum_{S:i \in S} \hat{f}(S)U_S$ .

Theorem 1 (Generalizaion of Talagrand, 1994) There exists some universal constant C such that

 $var(f) \le C \log(1/\alpha) \sum_{i \le n} \frac{||\Delta_i f||_2^2}{\log(||\Delta_i f||_2/||\Delta_i f||_1)}.$ 

Corollary 2 (Kahn, Kalai and Linial, 1988) Consider  $f: \{0,1\}^n \to \{0,1\}$ , where  $\{0,1\}^n$  is endowed with the uniform measure, then there exists a constant C>0 such that

 $\max_{i} I_i(f) \ge Cvar(f) \frac{\log n}{n} .$ 

**Proof:** [of Corollary 2] Recall that  $||\Delta_i f||_2^2 = I_i(f)$ , and that  $x/\log(1/x)$  is increasing on (0,1). By the identity  $\Delta_i f = f - E[f \mid X_j, \quad j \neq i]$ , it is easy to check that that  $||\Delta_i f||_1 = I_i(f)$ . So by Theorem 1 we get

$$Cvar(f) \le n \frac{\max_{i} I_i(f)}{\log(\max_{i} I_i(f))},$$

and since  $y/\log(1/y) \ge x$  implies  $y \ge Kx/\log(1/x)$  for some constant K for all  $x \in (0, 1/2)$ , we get the result.  $\square$ 

**Remark 3** Similarly we can prove that for all  $p \in (0,1)$  there exists a constant  $C_p$  such that if  $f: \{0,1\}^n \to \{0,1\}$ , where  $\{0,1\}^n$  is endowed with the Bin(n,p) measure, then

$$\max_{i} I_i(f) \ge C_p var(f) \frac{\log n}{n}.$$

**Proof:** [of Theorem 1] For a real function g from our space, denote

$$M^2(g) = \sum_{S: i \in S} \frac{\hat{g}(S)^2}{|S|}.$$

So

$$var(f) = \sum_{S \neq \emptyset} \hat{f}(S)^2 = \sum_{i \le n} M^2(\Delta_i f),$$

and hence it suffices to prove that for any function g with  $\mathbf{E}g = 0$ ,

$$M^2(g) \le K \log(1/\alpha) \frac{||g||_2^2}{\log(||g||_2/||g||_1)}$$
 (1)

To prove (1) we use hypercontractivity. The following proposition is proved in the end of this note.

**Proposition 4** Let  $q \in (1,2)$  and  $\Theta \in (0,1)$  satisfies

$$\Theta^2 \le \frac{\alpha^2}{3} (q - 1) \,,$$

then for all functions g we have,

$$||T_{\Theta}g||_2 \le ||g||_q$$

where  $T_{\Theta}$  is the Bonami-Beckner operator.

Recall that

$$T_{\Theta}g = \sum_{S} \Theta^{|S|} \hat{g}(S) U_{S} \,,$$

and apply the previous with q=3/2, and  $\Theta^2=\frac{\alpha^2}{6}$ . This gives that for any integer k>0,

$$\Theta^{2k} \sum_{|S|=k} \hat{g}(S)^2 \le \sum_{S} \Theta^{2|S|} \hat{g}(S)^2 = ||T_{\Theta}g||_2^2 \le ||g||_{3/2}^2,$$

hence

$$\sum_{|S|=k} \hat{g}(S)^2 \le \left(\frac{6}{\alpha^2}\right)^k ||g||_{3/2}^2.$$

Fix an integer m > 0, and sum the previous for all  $k \leq m$  to get

$$\sum_{|S| \le m} \frac{\hat{g}(S)^2}{|S|} \le \sum_{k \le m} \frac{\left(\frac{6}{\alpha^2}\right)^k}{k} ||g||_{3/2}^2 \le \frac{2\left(\frac{6}{\alpha^2}\right)^m}{m} ||g||_{3/2}^2,$$

where the last inequality comes from the fact that the ratio between two consecutive summands in the sum is greater than 2. We now have

$$M^{2}(g) = \sum_{|S| \le m} \frac{\hat{g}(S)^{2}}{|S|} + \sum_{|S| > m} \frac{\hat{g}(S)^{2}}{|S|} \le \frac{2\left(\frac{6}{\alpha^{2}}\right)^{m}}{m} ||g||_{3/2}^{2} + \frac{||g||_{2}^{2}}{m}$$
$$\le \frac{2}{m} \left[ \left(\frac{6}{\alpha^{2}}\right)^{m} ||g||_{3/2}^{2} + ||g||_{2}^{2} \right]. \tag{2}$$

We now choose optimal m. Choose largest m such that  $\left(\frac{6}{\alpha^2}\right)^m ||g||_{3/2}^2 \leq ||g||_2^2$ , hence

$$\left(\frac{6}{\alpha^2}\right)^{m+1}||g||_{3/2}^2 \ge ||g||_2^2 \implies m+1 \ge \frac{2\log\left(||g||_2/||g||_{3/2}\right)}{\log(6/\alpha^2)}.$$

Plugging this back into (2) gives

$$M^2(g) \le C \frac{\log(6/\alpha^2)||g||_2^2}{\log(||g||_2/||g||_{3/2})}.$$

An application of Cauchy-Schwartz gives

$$||g||_{3/2}^3 \le ||g||_1 ||g||_2^2$$

hence

$$\left(\frac{||g||_{3/2}}{||g||_2}\right)^3 \le \frac{||g||_1}{||g||_2},$$

which concludes the proof of (1) and so we are done.  $\square$ 

Let  $A \subset \{0,1\}^n$  be a monotone increasing set. Let  $\mu_p$  be the Bin(n,p) measure on  $\{0,1\}^n$ . Note that since A is increasing,  $\mu_p(A)$  is an increasing function in p. Moreover, it is a polynomial and in particular it is infinitely differentiable.

## Lemma 5 (Russo's Lemma)

$$\frac{\partial \mu_p(A)}{\partial p} = \frac{\sum_{i \le n} I_i^{(p)}(A)}{p(1-p)}.$$

**Proof:** Let  $\varphi(p_1, p_2, \ldots, p_n) : [0, 1]^n \to [0, 1]$  be a function returning the measure of A in the space  $L^2(\prod_i \mu_i)$  where  $\mu_i$  is a measure on the two point space  $\{0, 1\}$  which gives 1 weight  $p_i$  and gives 0 weight  $1 - p_i$ . The clearly  $\mu_p(A) = \varphi(p, \ldots, p)$ , so by the chain rule

$$\frac{\partial \mu_p(A)}{\partial p} = \sum_{i < n} \frac{\partial \varphi}{\partial p_i}(p, \dots, p) = \sum_{i < n} \frac{I_i^{(p)}(A)}{p(1-p)},$$

where the last equality is due to the easy fact

$$\frac{\partial \varphi}{\partial p_i}(p,\ldots,p) = \frac{I_i^{(p)}(A)}{p(1-p)}.$$

A graph property P on n vertices is a set of graphs on n vertices which is invariant under vertex permutations. The following theorem states that any graph property which is monotone experiences a 'sharp threshold'.

**Theorem 6 (Friedgut and Kalai, 1996)** Let P be a monotone increasing graph property on n vertices. If  $p \in (0,1)$  is such that  $\mu_p(P) > \epsilon$ , then

$$\mu_q(P) > 1 - \epsilon \,,$$

for  $q = p + c_1 \frac{\log(\frac{1}{2\epsilon})}{\log n}$ , where  $c_1 > 0$  is a universal constant.

**Proof:** Invariance under vertex permutation gives that all influences of the indicator function of A are equal (note the edges of graph are the variables of the function). Hence by Theorem 1 and Remark 3 we have that

$$\sum_{i} I_i(A) \ge C\mu_p(A)(1 - \mu_p(A)) \log n.$$

For any r > p such that  $\mu_r(A) \le 1/2$ , by Lemma 5 and the previous line we have that

$$\frac{\partial \mu_r(A)}{\partial r} \ge C\mu_r(A)\log n\,,$$

where we consider p to be fixed (and hence so is 1/p). Last equation can be written as

$$\frac{\partial \log(\mu_r(A))}{\partial r} \ge C \log n \,,$$

and so if we take  $q' = p + \frac{\log(\frac{1}{2\epsilon})}{C\log n}$  we get by the fundamental theorem of calculus that

$$\log(\mu_{q'}(A)) \ge \log(\mu_p(A)) + \int_p^{q'} C \log n \ge \log(\epsilon) + \log(\frac{1}{2\epsilon}) = \log(1/2).$$

And so  $\mu_{q'}(A) \ge 1/2$ . Similarly, if we take  $q = q' + \frac{\log(\frac{1}{2\epsilon})}{C \log n}$  we get that  $\mu_q(A) \ge 1 - \epsilon$ .

**Proof:** [of Proposition 4] We have learned that the hypercontractive constant for the space  $L^2(\prod_i \mu_i)$  is

$$\Theta(q) = \left(\frac{(1-\alpha)^{2-2/q} - \alpha^{2-2/q}}{(1-\alpha)\alpha^{1-2/q} - \alpha(1-\alpha)^{1-2/q}}\right)^{1/2},$$

for all  $q \in (1,2)$ . Thus in order to prove the claim, we just need to lower bound  $\Theta(q)$ . Let

$$f(x) = x^{2-2/q}, \quad g(x) = -(1-x)x^{1-2/q},$$

and by Lagrange's theorem we have

$$\Theta(q)^2 = \frac{f'(\xi_1)}{g'(\xi_2)},$$

for some  $\xi_1, \xi_2 \in (\alpha, 1 - \alpha)$ . By computing, one can check that f' and g' are decreasing, and hence

$$\begin{split} \Theta(q)^2 & \geq \frac{f'(1-\alpha)}{g'(\alpha)} = \frac{(2-2/q)(1-\alpha)^{1-2/q}}{\alpha^{1-2/q} + (2/q-1)\alpha^{-2/q}(1-\alpha)} \\ & = \frac{2(q-1)}{q} \Big(\frac{1-\alpha}{\alpha}\Big)^{-2/q} \Big[\frac{1-\alpha}{\alpha + (2/q-1)(1-\alpha)}\Big] \geq \frac{(q-1)\alpha^2}{3} \,. \end{split}$$