## High-Dimensional Measures and Geometry Lecture Notes from April 29, 2010

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Last time we have shown that if  $\mu(X) \leq 1/2$  then we have the following inequality

$$\mu(X(1)) \ge \left(1 + \frac{\lambda_1(\Delta)}{2d}\right)\mu(X)$$

which means that X grows at a minimum rate  $1 + \lambda_1(\Delta)/2d$ . Now by using some iteration we obtain

$$\mu(X(r+1)) \ge \left(1 + \frac{\lambda_1(\Delta)}{2d}\right)^{r+1} \mu(X)$$

as long as  $\mu(X_r) \leq 1/2$ . Consequently

$$\mu(X) \le \left(1 + \frac{\lambda_1(\Delta)}{2d}\right)^{-r} \mu(X(r)) \le \frac{1}{2} \left(1 + \frac{\lambda_1(\Delta)}{2d}\right)^{-r}.$$

Now we wish to use this result for measure concentration.

Note: If  $X = V \setminus Y(r)$  for some  $Y \subseteq V$  then

$$X = \{v \in V : d(v, Y) > r\}$$

We claim that  $X(r) \cap Y = \emptyset$  This is because  $v \in X(r)$  iff  $d(v, X) \leq r$ . But if  $v \in Y$  then  $d(v, X) \geq d(X, Y) > r$ . So  $v \notin Y$ .

Now by going to complements, if  $Y \subseteq V$  with  $\mu(Y) \geq 1/2$  and  $X = V \setminus Y(r)$  then  $X(r) \subseteq V \setminus Y$  and  $\mu(X(r)) \leq 1/2$ . Thus

$$\mu(Y) \ge \frac{1}{2} \Rightarrow \mu(X) = \mu(\{v \in V : d(v, Y) > r\}) \le \frac{1}{2} \left(1 + \frac{\lambda_1(\Delta)}{2d}\right)^{-r}$$

Recall that for  $I_n$ , if we define

$$B = \{ v \in I_n : \sum_{i=1}^n v_i \le n/2 \}$$

then we know

$$\mu(I_n \setminus B(\alpha n)) = \mu(\{x \in I_n : f(x) - \frac{n}{2} \ge \alpha n\}) \le e^{-2n\alpha^2} e^{Cn\alpha^3}.$$

We compare this result with  $r = \alpha n$ ,

$$\mu(\{v \in V : d(v, B) > r\}) \le \frac{1}{2} \left(1 + \frac{1}{n}\right)^{-r} \le \frac{1}{2} e^{-r/n} = \frac{1}{2} e^{-\alpha}.$$

Clearly the estimation above is better. However we can improve this estimation.

**0.0.1 Theorem.** Let G = (V, E) be a connected graph,  $X \subseteq V$  be a set and  $deg(v) \leq d$  for all v in V. Then for every  $r \geq 1$ , if  $\mu(X(r)) \leq 1/2$ , we have

$$\mu(X(r)) \ge \left(1 + \frac{\lambda_1(\Delta)}{2d}r^2\right)\mu(X).$$

*Proof.* We apply the preceding edge-count estimate to X = A,  $B = V \setminus X(r)$ , then  $E \setminus (E(A) \cup E(B))$  "connects" between X and  $V \setminus X(r)$ . Thus

$$|X(r) \setminus X| \ge \frac{1}{d} (E - E(X) - E(V \setminus X(r))) \ge \frac{\lambda_1(\triangle)r^2}{d} \frac{|X| |V \setminus X(r)|}{|X| + |V \setminus X(r)|}$$

However,

$$\frac{|V \setminus X(r)|}{|X| + |V \setminus X(r)|} = \left(1 + \frac{|X|}{|V \setminus X(r)|}\right)^{-1} = \left(1 + \frac{\frac{|X|}{|V|}}{\frac{|V \setminus X(r)|}{|V|}}\right)^{-1} = \left(1 + \frac{\mu(X)}{\mu(V \setminus X(r))}\right)^{-1}$$

now considering  $\mu(X) \leq 1/2$  and  $\mu(V \setminus X(r)) = 1 - \mu(X(r)) \geq 1/2$  we obtain that the latter expression on the above is greater than or equal 1/2. Hence we have

$$|X(r) \setminus X| \geq \frac{\lambda_1(\triangle)r^2}{d} \frac{|X|}{2}$$

equivalently,

$$|X(r)| \ge \frac{\lambda_1(\triangle)r^2}{d} \frac{|X|}{2} + |X| = \left(1 + \frac{\lambda_1(\triangle)}{2d}r^2\right)|X|.$$

Thus

$$\mu(X(r)) \ge \left(1 + \frac{\lambda_1(\Delta)}{2d}r^2\right)\mu(X).$$

 $\square$ 

As before, we conclude that if  $Y\subseteq V$  with  $\mu(Y)\geq 1/2$  and  $X=V\setminus Y(r),$  then  $\mu(X(r))\leq 1/2$  and

$$\mu(\{v \in V : d(v, Y) > r\}) \le \frac{1}{2} \left(1 + \frac{\lambda_1(\Delta)}{2d} r^2\right)^{-1}$$

This estimation doesn't seem to be as good as what we had, at least for large r. But, if  $\lambda_1 r/d$  is small then

$$\left(1+\frac{\lambda_1}{d}\right)^r \cong 1+\frac{\lambda_1 r}{2d} < 1+\frac{\lambda_1 r^2}{2d}.$$

So this new bound is better. We iterate the bound to get good estimates when r is large.

**0.0.2 Corollary.** Let G = (V, E) be a connected graph such that  $deg(v) \leq d$  for all v in V. Let  $A \subseteq V$  be a set with  $\mu(A) > 1/2$ . Then for any  $t \in \mathbb{N}$ 

$$\mu(\{v \in V : d(v, A) > t\}) \le \frac{1}{2} \left(1 + \frac{\lambda_1(\triangle)r^2}{2d}\right)^{-\lfloor t/r \rfloor}$$

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*Proof.* Take  $B = V \setminus A(t)$ . Again  $B(t) \subseteq V \setminus A$  and thus  $\mu(B(t)) \leq 1/2$ . Take  $s = \lfloor t/r \rfloor$  and construct a sequence

$$X_0 = B, \ X_1 = X_0(r), \ X_2 = X_1(r), ..., X_k = X_{k-1}(r) \text{ for } k \le s.$$

Then  $X_s = B(rs) \subseteq B(t)$ . So  $\mu(X_k) \le 1/2$  for k = 0, 1, 2, ..., s. Now using our refined estimate we have

$$\frac{1}{2} \ge \mu(X_s) \ge \left(1 + \frac{\lambda_1(\Delta)r^2}{2d}\right)^s \mu(B)$$

or

$\mu(B) \le \frac{1}{2}$	$\left(1+\frac{\lambda_1}{\lambda_1}\right)$	$\frac{\Delta}{2d}$	) .

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