

## Part C Probabilistic Combinatorics 2008

### Remarks on Azuma, concentration of Lipschitz functions, and isoperimetric inequalities

When doing the isoperimetric inequality in lectures, I cheated inadvertently by using a stronger version of the result on concentration of Lipschitz functions than I had previously proved.

Here are a few notes to clarify things for those who are interested.

We proved the following Lemma:

**Lemma 1** *Let  $\alpha > 0$ . If  $|Y| \leq 1$  with probability 1 and  $\mathbb{E}Y = 0$ , then  $\mathbb{E}(e^{\alpha Y}) \leq e^{\alpha^2/2}$ .*

From this we deduced the following form of Azuma's inequality:

**Theorem 1** *Let  $0 = X_0, X_1, \dots, X_m$  be a martingale such that  $|X_{i+1} - X_i| \leq 1$  with probability 1, for  $0 \leq i \leq m - 1$ . Then for any  $t > 0$ ,*

$$\mathbb{P}(X_m > t) \leq \exp\left(-\frac{t^2}{2m}\right).$$

This in turn gave us the concentration result for Lipschitz functions:

**Theorem 2** *Let  $V_1, \dots, V_m$  be independent random variables taking values in sets  $\Omega_1, \dots, \Omega_m$  respectively. Let  $f : \prod_{i=1}^m \Omega_i \rightarrow \mathbb{R}$  be a Lipschitz function. Let  $X = f(V_1, \dots, V_m)$ . Then*

$$\mathbb{P}(|X - \mathbb{E}X| \geq t) \leq 2 \exp\left(-\frac{t^2}{2m}\right).$$

The idea of the proof was to consider the martingale  $(X_i)$  where  $X_i = \mathbb{E}(X|V_1, \dots, V_i)$ . Since the  $i$ th coordinate cannot affect the function  $f$  by more than 1, learning the value of  $V_i$  does not affect the conditional expectation of  $X$  by more than 1. Hence  $|X_{i+1} - X_i| \leq 1$  with probability 1, and we can apply Azuma's inequality.

In fact, more is true. Once we know  $V_1, \dots, V_i$ , the possible range of values of  $X_{i+1}$  is at most 1; that is, given  $V_1, \dots, V_i$ , there exist  $a$  and  $b$  such that  $b - a \leq 1$  and such that  $a \leq X_{i+1} \leq b$  with probability 1.

This observation makes it possible to prove a stronger version of Theorem 2. We use the following generalisation of Lemma 1:

**Lemma 2** *Let  $\alpha > 0$ . If  $a \leq Y \leq b$  with probability 1, and  $\mathbb{E}Y = 0$ , then  $\mathbb{E}(e^{\alpha Y}) \leq e^{\alpha^2(b-a)^2/8}$ .*

Using this one can prove a more general form of Azuma's inequality which leads to the following stronger version of Theorem 2:

**Theorem 3** Let  $V_1, \dots, V_m$  be independent random variables taking values in sets  $\Omega_1, \dots, \Omega_m$  respectively. Let  $f : \prod_{i=1}^m \Omega_i \rightarrow \mathbb{R}$  be a Lipschitz function. Let  $X = f(V_1, \dots, V_m)$ . Then

$$\mathbb{P}(|X - \mathbb{E} X| \geq t) \leq 2 \exp\left(-\frac{2t^2}{m}\right).$$

This is the result I was claiming during the proof of the isoperimetric inequality. Namely, we used Theorem 3 to show:

**Theorem 4** Let  $X = (X_1, X_2, \dots, X_n)$  be a vector with independent entries  $X_i \in \Omega_i$ . Let  $A \subset \Omega = \prod_{i=1}^n \Omega_i$ . Then for any  $t \geq 0$ ,

$$\mathbb{P}(X \in A) \mathbb{P}(d_H(X, A) \geq t) \leq \exp\left(-\frac{t^2}{2n}\right),$$

(where  $d_H$  is the Hamming distance).

If one instead uses the weaker version in Theorem 2, one can get a similar result but with  $\exp\left(-\frac{t^2}{8n}\right)$  on the right instead. (Exercise: prove this using the same method from lectures).

To deduce Theorem 3 from 2, one needs a more general version of Azuma's inequality which needs a slightly more general definition of a martingale than the one we used in the course. The argument is quite straightforward though.

For full details of all of this (and much more beyond), a good place to look is Colin McDiarmid's article "Concentration" in the book *Probabilistic Methods for Algorithmic Discrete Mathematics*, ed. M. Habib et al, Springer 1998.

**Course webpage:** <http://www.stats.ox.ac.uk/~martin/PC.html>