

## Proof of the Bounded Differences Inequality

Peter Bartlett. March 9, 2006.

**Theorem** Suppose that  $X_1, \dots, X_n \in \mathcal{X}$  are independent, and  $f : \mathcal{X}^n \rightarrow \mathbb{R}$ . Let  $c_1, \dots, c_n$  satisfy

$$\sup_{x_1, \dots, x_n, x'_i} |f(x_1, \dots, x_{i-1}, x_i, x_{i+1}, \dots, x_n) - f(x_1, \dots, x_{i-1}, x'_i, x_{i+1}, \dots, x_n)| \leq c_i,$$

for  $i = 1, \dots, n$ . Then

$$P(f - \mathbb{E}f \geq t) \leq \exp\left(\frac{-2t^2}{\sum_{i=1}^n c_i^2}\right).$$

*Proof:* We define

$$V_i = \mathbb{E}[f|X_1, \dots, X_i] - \mathbb{E}[f|X_1, \dots, X_{i-1}].$$

These  $V_i$ s will play the same role as that played by the terms of the sum in the proof of Hoeffding's inequality. In particular, since the sum telescopes, we have

$$f - \mathbb{E}f = \sum_{i=1}^n V_i.$$

Using this, and the Chernoff bounding technique, we see that

$$\begin{aligned} & P(f - \mathbb{E}f \geq t) \\ &= P\left(\sum_{i=1}^n V_i \geq t\right) \\ &\leq \inf_{s>0} \exp(-st) \mathbb{E} \exp\left(s \sum_{i=1}^n V_i\right) \\ &= \inf_{s>0} \exp(-st) \mathbb{E} \left(\prod_{i=1}^n e^{sV_i}\right). \end{aligned}$$

So we need to bound the moment generating function of this sum of (dependent) random variables. Now, we have seen that  $\mathbb{E}(V_i|X_1, \dots, X_{i-1}) = 0$ . If we define

$$\begin{aligned} L_i &= \inf_x \mathbb{E}(f|X_1, \dots, X_{i-1}, x) - \mathbb{E}(f|X_1, \dots, X_{i-1}), \\ U_i &= \sup_x \mathbb{E}(f|X_1, \dots, X_{i-1}, x) - \mathbb{E}(f|X_1, \dots, X_{i-1}), \end{aligned}$$

then we have

$$\begin{aligned} V_i - L_i &= \mathbb{E}(f|X_1, \dots, X_i) - \mathbb{E}(f|X_1, \dots, X_{i-1}) \\ &\quad - \inf_x \mathbb{E}(f|X_1, \dots, X_{i-1}, x) + \mathbb{E}(f|X_1, \dots, X_{i-1}) \\ &= \mathbb{E}(f|X_1, \dots, X_i) - \inf_x \mathbb{E}(f|X_1, \dots, X_{i-1}, x), \end{aligned}$$

and so  $L_i \leq V_i$  almost surely. Similarly,  $V_i \leq U_i$  a.s. Furthermore, from the independence of the  $X_i$ ,

$$\begin{aligned} U_i - L_i &= \sup_x \mathbb{E}(f|X_1, \dots, X_{i-1}, x) - \inf_x \mathbb{E}(f|X_1, \dots, X_{i-1}, x) \\ &= \sup_x \int f(X_1, \dots, X_{i-1}, x, x_{i+1}, \dots, x_n) dP(x_{i+1}, \dots, x_n) \\ &\quad - \inf_x \int f(X_1, \dots, X_{i-1}, x, x_{i+1}, \dots, x_n) dP(x_{i+1}, \dots, x_n) \\ &= \sup_{x,y} \int (f(X_1, \dots, X_{i-1}, x, x_{i+1}, \dots, x_n) \\ &\quad - f(X_1, \dots, X_{i-1}, y, x_{i+1}, \dots, x_n)) dP(x_{i+1}, \dots, x_n) \\ &\leq c_i, \end{aligned}$$

from the bounded differences assumption. Notice that the independence allows us to write the difference of the conditional expectations in terms an integral of a difference, and hence appeal to the bounded differences property. Thus, we may apply the Hoeffding lemma to the bounded, zero-mean random

variables  $V_i$  conditioned on  $X_1, \dots, X_{i-1}$ , as follows.

$$\begin{aligned}
\mathbb{E} \left( \prod_{i=1}^n e^{sV_i} \right) &= \mathbb{E} \mathbb{E} \left( \prod_{i=1}^{n-1} e^{sV_i} e^{sV_n} \mid X_1, \dots, X_{n-1} \right) \\
&= \mathbb{E} \left( \prod_{i=1}^{n-1} e^{sV_i} \mathbb{E} (e^{sV_n} \mid X_1, \dots, X_{n-1}) \right) \\
&\leq \exp(s^2 c_n^2 / 8) \mathbb{E} \left( \prod_{i=1}^{n-1} e^{sV_i} \right) \\
&= \exp(s^2 c_n^2 / 8) \mathbb{E} \mathbb{E} \left( \prod_{i=1}^{n-1} e^{sV_i} \mid X_1, \dots, X_{n-2} \right) \\
&= \exp(s^2 c_n^2 / 8) \mathbb{E} \left( \prod_{i=1}^{n-2} e^{sV_i} \mathbb{E} (e^{sV_{n-1}} \mid X_1, \dots, X_{n-2}) \right) \\
&\leq \exp(s^2 (c_{n-1}^2 + c_n^2) / 8) \mathbb{E} \left( \prod_{i=1}^{n-2} e^{sV_i} \right) \\
&\vdots \\
&\leq \exp \left( s^2 \sum_{i=1}^n c_i^2 / 8 \right).
\end{aligned}$$

Substituting, we have

$$\begin{aligned}
P(f - \mathbb{E}f \geq t) &\leq \inf_{s>0} \exp \left( -st + \frac{s^2 \sum_{i=1}^n c_i^2}{8} \right) \\
&= \exp \left( \frac{-2t^2}{\sum_{i=1}^n c_i^2} \right),
\end{aligned}$$

where we selected the optimizing value  $s = 4t / \sum_i c_i^2$ .