

**ECE 299: STATISTICAL LEARNING THEORY**

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HOMEWORK 1

Assigned February 2, 2011; due February 16, 2011

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- (1) Let  $X_1, \dots, X_n$  be  $n$  independent Bernoulli( $\theta$ ) random variables. Prove the following *multiplicative Chernoff bound*:

$$\mathbb{P}\left(\frac{1}{n} \sum_{i=1}^n X_i \geq (1 + \gamma)\mu\right) \leq e^{-\gamma^2 n \mu / 3}$$

for any  $0 \leq \gamma \leq 1$  and any  $\theta \leq \mu \leq 1$ . You may need the fact that

$$\sum_{k \geq (1+\gamma)\mu n} \binom{n}{k} \mu^k (1-\mu)^{n-k} \leq e^{-\gamma^2 n \mu / 3}.$$

- (2) Let  $X_1, \dots, X_n$  be  $n \geq 2$  real-valued random variables (not necessarily independent). Suppose that there exists some constant  $\sigma > 0$ , such that for all  $s > 0$  and all  $i = 1, \dots, n$

$$\mathbb{E}[e^{sX_i}] \leq e^{s^2 \sigma^2 / 2}$$

Prove the following:

(a)

$$\mathbb{E}\left[\max_{1 \leq i \leq n} X_i\right] \leq \sigma \sqrt{2 \log n}.$$

**Hint.** Try to bound  $e^{s \mathbb{E}[\max_{1 \leq i \leq n} X_i]}$ , exploit convexity of the function  $\phi(x) = e^{sx}$ .

(b) Assuming that  $X_i \geq 0$  for all  $i = 1, \dots, n$ ,

$$\mathbb{P}\left(\max_{1 \leq i \leq n} X_i \geq \sqrt{2\sigma^2 n}\right) \leq \sqrt{\frac{\log n}{n}}.$$

- (3) Let  $X^n = (X_1, \dots, X_n)$  be an  $n$ -tuple of independent random variables taking values in some space  $\mathcal{X}$ . The *Hamming distance* between any  $n$ -tuples  $x^n, y^n \in \mathcal{X}^n$  is defined as the number of coordinates in which  $x^n$  and  $y^n$  differ:

$$d(x^n, y^n) \triangleq \sum_{i=1}^n \mathbf{1}_{\{x_i \neq y_i\}}.$$

If  $B$  is an arbitrary (measurable) subset of  $\mathcal{X}^n$ , then the Hamming distance between an  $n$ -tuple  $x^n \in \mathcal{X}^n$  and  $B$  is defined as

$$d(x^n, B) \triangleq \min_{y^n \in B} d(x^n, y^n).$$

Use McDiarmid's inequality to prove the following fact: if

$$\varepsilon \geq \sqrt{\frac{1}{2n} \log \frac{1}{\mathbb{P}(B)}},$$

then

$$\mathbb{P}(d(X^n, B) \geq n\varepsilon) \leq \exp\left(-2n\left(\varepsilon - \sqrt{\frac{1}{2n} \log \frac{1}{\mathbb{P}(B)}}\right)^2\right),$$

where  $\mathbb{P}(B) = \mathbb{P}(X^n \in B)$  is the probability of the set  $B \in \mathcal{X}^n$  under the joint distribution of  $X^n$ .

**Hint.** You may find the following fact handy:  $d(x^n, B) \leq 0$  if and only if  $x^n \in B$ .