## Math 500, Problem Set 1

## March 3, 2010

- 1. Prove that for every  $\epsilon > 0$  there is a finite  $l_0 = l_0(\epsilon)$  and an infinite sequece of bits  $a_1, a_2, \ldots$  with  $a_i \in \{0, 1\}$ , such that for every  $l > l_0$  and every  $i \ge 1$ , the two binary vectors  $u = (a_i, a_{i+1}, \ldots, a_{i+l-1})$  and  $v = (a_{i+l}, a_{i+l+1}, \ldots, a_{i+2l-1})$  differ in at least  $(\frac{1}{2} \epsilon)l$  coordinates.
- 2. Let G = (V, E) be a simple graph, and suppose each  $v \in V$  is associated with a set S(v) of colors of size at least 10*d*, where  $d \ge 1$ . Suppose, in addition, that for each  $v \in V$  and  $c \in S(v)$  there are at most *d* neighbors *u* of *v* such that *c* lies in S(u). Prove there is a proper coloring of *G* assigning to each vertex *v* a color from its class S(v).
- 3. Let X be a random variable which takes non-negative values only. Show that

$$\sum_{i=1}^{\infty} (i-1) \mathbf{1}_{A_i} \le X \le \sum_{i=1}^{\infty} i \mathbf{1}_{A_i},$$

where  $A_i = \{i - 1 \le X < i\}$ . Deduce that

$$\sum_{i=1}^{\infty} \mathbb{P}(X \ge i) \le \mathbb{E}[X] \le 1 + \sum_{i=1}^{\infty} \mathbb{P}(X \ge i).$$

4. The interval [0,1] is partitioned into n disjoint sub-intervals with lengths  $p_1, p_2, \ldots, p_n$ , and the *entropy* of this partition is defined to be

$$h = -\sum_{i=1}^{n} p_i \log p_i.$$

Let  $X_1, X_2, \ldots$  be independent random variables having uniform distribution on [0, 1], and let  $Z_m(i)$  be the number of  $X_1, \ldots, X_m$  which like in the *i*th interval of the partition above. Show that

$$R_m = \prod_{i=1}^n p_i^{Z_m(i)}$$

satisfies  $m^{-1}\log(R_m) \to -h$  a.s. as  $m \to \infty$  (that is:

$$\mathbb{P}(\lim_{m \to \infty} \frac{\log(R_m)}{m} = -h) = 1.$$

5. Let  $A_1, A_2, \ldots$  be a sequence of events. Show that

$$\mathbb{P}(A_n \ i.o.) \ge \limsup_{n \to \infty} \mathbb{P}(A_n)$$

6. Suppose  $X_1, X_2, \ldots$  are independent and exponentially distributed with paratmeter one. Show that

$$\mathbb{P}\left(\limsup_{n \to \infty} \frac{X_n}{\log(n)} = 1\right) = 1$$

7. Let X be a random variable with  $\mathbb{E}[X] = 0$  and  $\operatorname{Var}(X) = \sigma^2$ . Prove that for all  $\lambda > 0$ ,

$$\mathbb{P}(X \ge \lambda) \le \frac{\sigma^2}{\sigma^2 + \lambda^2}.$$

(Note: this is a slight improvement on Chebyshev, which gives a bound of the form  $\frac{\sigma^2}{\lambda^2}$ .)

8. Let X be a random variable taking integral nonnegative values, let  $\mathbb{E}[X^2]$  denote the expectation of its square, and let Var(X) denote its variance. Prove

$$\mathbb{P}(X=0) \le \frac{\operatorname{Var}(X)}{\mathbb{E}[X^2]}.$$

9. Show that there is a positive constant c such that the following holds. For any n vectors  $a_1, a_2, \ldots, a_n \in \mathbb{R}^2$  satisfying  $\sum_{i=1}^n ||a_i||^2 = 1$  and  $||a_i|| \leq \frac{1}{10}$ , where  $|| \cdot ||$  denotes the usual Euclidean norm, if  $(\epsilon_1, \ldots, \epsilon_n)$  is a  $\{-1, 1\}$ random vector obtained by choosing each  $\epsilon_i$  randomly and independently with uniform distribution to be either -1 or 1, then

$$\mathbb{P}\left(\left\|\sum_{i=1}^{n}\epsilon_{i}a_{i}\right\|\leq\frac{1}{3}\right)\geq c.$$

- 10. Let  $S_n = \sum_{i=1}^n X_i$  where the  $X_i$  are iid variables that are uniformly  $\pm 1$ .
  - (a) Show that

$$\mathbb{P}\left(\sqrt{\left(\frac{S_n}{\sqrt{n}}\right)^2} \ge \frac{1}{3}\right) \ge c$$

for some absolute constant c. (This is a special case of the previous problem.)

(b) Show that

$$\mathbb{E}[|S_n|] = n2^{1-n} \binom{n-1}{\lfloor \frac{n-1}{2} \rfloor}.$$

Apparently a 1974 Putnam problem.

(c) Derive that

$$\mathbb{E}[|S_n|] \sim (\sqrt{\frac{2}{\pi}} + o(1))\sqrt{n}.$$

11. Let  $X_2, X_3, \ldots$  be independent random variables such that:

$$\mathbb{P}(X_n = n) = \mathbb{P}(X_n = -n) = \frac{1}{2n\log n}, \quad \mathbb{P}(X_n = 0) = 1 - \frac{1}{n\log n}$$

Show that this sequence obeys the weak law but not the strong law, in the sense that  $n^{-1} \sum X_i$  converges to zero in probability but not almost surely.

- 12. Construct a sequence  $\{X_r : r \ge 1\}$  of independent random variables with zero mean such that  $n^{-1} \sum_{r=1}^n X_r \to -\infty$  almost surely, as  $n \to \infty$ .
- 13. Prove that  $\mathbb{P}(X = 0) \leq \frac{\sigma^2}{\mu^2}$  when X has mean  $\mu \neq 0$  and variance  $\sigma^2$  (essentially, we did this in class.) Then show that the stronger inequality below is true:

$$\mathbb{P}(X=0) \le \frac{\sigma^2}{\mu^2 + \sigma^2}.$$