## Homework Assignment 2

## Due in class, Wednesday Feb 21st

## SDS 384-11 Theoretical Statistics

- 1. Show that Markov's inequality is tight.
  - (a) Give an example of a non-negative random variable X and a value k > 1 such that  $P(X \ge kE[X]) = 1/k$ .
  - (b) Give an example of a random variable X (with E[X] > 0) and a value k > 1 such that  $P[X \ge kE[X]] > 1/k$ .
- 2. Consider a r.v. X such that for all  $\lambda \in \Re$

$$E[e^{\lambda X}] \le e^{\frac{\lambda^2 \sigma^2}{2} + \lambda \mu} \tag{1}$$

Prove that:

- (a)  $E[X] = \mu$ .
- (b)  $var(X) < \sigma^2$ .
- (c) If the smallest value of  $\sigma$  satisfying the above equation is chosen. Is it true that  $var(X) = \sigma^2$ ? Prove or disprove.
- 3. Remember Hoeffding's Lemma? We proved it with a weaker constant in class using a symmetrization type argument. Now we will prove the original version. Let X be a bounded r.v. in [a,b] such that  $E[X] = \mu$ . Let  $f(\lambda) = \log E[e^{\lambda(X-\mu)}]$ . Show that  $f''(\lambda) \leq (b-a)^2/8$ . Now use the fundamental theorem of calculus to write  $f(\lambda)$  in terms of  $f''(\lambda)$  and finish the argument.
- 4. Bernstein's inequality for bounded i.i.d sequences of random variables  $\{X_i\}$  with  $|X_i| \leq M$  gives:  $P(|\sum_i (X_i E[X_i])| \geq t) \leq 2 \exp\left(\frac{-t^2/2}{\sum_i \operatorname{var}(X_i) + Mt/3}\right)$ . Consider n i.i.d.  $X_i \sim \operatorname{Bernoulli}(p_n)$  r.v's. We will consider two cases to study concentration of  $\bar{X}_n$  around  $p_2 n$ .
  - (a) (Dense case) Let  $np_n/\log n \to \infty$ . Can you apply Hoeffding's bound and Bernstein's inequality to establish concentration of  $\bar{X}_n$ , i.e.  $P(\bar{X}_n \in [p_n(1-\epsilon_n), p_n(1+\epsilon_n)]) = O(1/n)$ ? Do you prefer one bound over another? Why?
  - (b) (Sparse case) Repeat your argument for the case  $np_n = c \log n$  where c is some constant not depending on n.