

Homework Assignment 1

Due in class, Wednesday Feb 7th

SDS 384-11 Theoretical Statistics

- Given densities p_n and q_n with respect to some measure μ , let X be distributed according to the distribution with density p_n . Define the likelihood ratio $L_n(X)$ as $L_n(X) = q_n(X)/p_n(X)$ for $p_n(X) > 0$. $L_n(X) = 1$ if $p_n(X) = q_n(X) = 0$ and $L_n(X) = \infty$ otherwise. Show that the likelihood ratio is a uniformly tight sequence. $E[|L_n(X)|] = E[L_n(X)] = \int_{x:p_n(x)>0} \frac{q_n(x)}{p_n(x)} p_n(x) dx \leq 1$. So for $\epsilon > 0$, take $P(L_n \geq 1/\epsilon) \leq \epsilon$ for all n . So its UT.
- Consider a sequence of iid random variables $\{X_n\}$ such that $X_i \sim \text{Beta}(\theta, 1)$, where $\theta > 0$. Let \bar{X}_n denote the sample mean. The method of moments estimator of θ is $\hat{\theta}_n = \bar{X}_n/(1 - \bar{X}_n)$. Derive the asymptotic distribution of $\sqrt{n}(\hat{\theta}_n - \theta)$. Recall that the expectation of a $\text{beta}(\beta, 1)$ random variable is $\theta/(1 + \theta)$. So $\bar{X}_n \xrightarrow{P} \theta/(1 + \theta)$ and variance $\sigma^2 = \frac{\theta}{(\theta+1)^2(\theta+2)}$. Now

$$\begin{aligned} \sqrt{n}(\hat{\theta}_n - \theta) &= \sqrt{n} \left(\frac{\bar{X}_n}{1 - \bar{X}_n} - \theta \right) \\ &= \sqrt{n}(1 + \theta) \frac{\bar{X}_n - \frac{\theta}{1+\theta}}{1 - \bar{X}_n} \end{aligned}$$

Using CLT we have $\sqrt{n}(\bar{X}_n - \frac{\theta}{1+\theta}) \xrightarrow{d} N(0, \sigma^2)$. $1 - \bar{X}_n \xrightarrow{P} 1/(1 + \theta)$. So $\sqrt{n}(\hat{\theta}_n - \theta) \xrightarrow{d} N(0, \theta(\theta + 1)^2/(\theta + 2))$.

- Derive the following one sided improvement of Chebyshev's inequality for a random variable X with variance σ^2 and any $t > 0$

$$P(X - E[X] \geq t) \leq \frac{\sigma^2}{\sigma^2 + t^2} \quad (1)$$

Take $E[X] = 0$ WLOG.

$$P(X \geq t) = P((X + u)^2 \geq (t + u)^2) \leq \frac{\sigma^2 + u^2}{(t + u)^2}$$

Minimizing the RHS w.r.t $u \geq 0$ gives the answer.

- If $X_n \xrightarrow{d} X \sim \text{Poisson}(\lambda)$, is it necessarily true that $E[g(X_n)] \rightarrow E[g(X)]$?
 (a) $g(x) = 1(x \in (0, 10))$ $g(x)$ is discontinuous at 0 and 10, and $P(X \in \{0, 10\}) \neq 0$. So our theorem does not apply. We will create a counter-example. Let $X_n = X + 1/n$. $E[g(X_n)] = P(X \in (-1/n, 10 - 1/n)) \rightarrow P(X \in [0, 9])$. On the other hand $E[g(X)] = P(X \in [1, 9])$.

- (b) $g(x) = e^{-x^2} e^{-x^2}$ is continuous and bounded, so the Portmanteau theorem proves that the expectation converges.
- (c) $g(x) = \text{sgn}(\cos(x))$ [$\text{sgn}(x) = 1$ if $x > 0$, -1 if $x < 0$ and 0 if $x = 0$.] $g(x)$ is discontinuous at $(\pi/2, 3\pi/2, \dots)$. However since X takes values in integers, we can safely say that $E[g(X_n)] \rightarrow E[g(X)]$
- (d) $g(x) = x$ $g(x)$ is continuous but unbounded. So let's find a counter example. Let $X_n = X$ with probability $1 - 1/n$ and $X_n = n$ with probability $1/n$. So $X_n \xrightarrow{P} X$ and $X_n \xrightarrow{d} X$. But $E[g(X_n)] = \lambda(1 - 1/n) + 1 \rightarrow \lambda + 1$, but $E[g(X)] = \lambda$.
5. Consider n i.i.d random variables $\{X_n\}$ uniformly distributed on the set of n points $\{1/n, 2/n, \dots, 1\}$. Show that $X_n \xrightarrow{d} X$ where $X \sim \text{Uniform}(0, 1)$. Does $X_n \xrightarrow{P} X$?
WLOG let $x \in [i/n, (i+1)/n]$. $P(X_n \leq x) = i/n$ as $n \rightarrow \infty$. So X_n is converging in distribution to a Uniform r.v. Now create a X independently from the sequence.
 $P(|X_n - X| \geq \epsilon) \leq P(X_n + X \geq \epsilon) \leq P(X_n \geq \epsilon/2) + P(X \geq \epsilon/2) \not\rightarrow 0$