

## Chapter 5

### Limit Theorems

#### 5.1 Markov and Chebychev Inequalities

Markov Inequality is a -typically loose- bound on the value of a nonnegative random variable  $X$  with a known mean  $E(X)$ .

**Markov Inequality:** If a random variable  $X$  can take only nonnegative values, then

$$P(X \geq a) \leq \frac{E(X)}{a}$$

Proof:

**Ex:** Let  $X$  be uniform in  $[5, 10]$ . Compute probabilities that  $X$  exceeds certain values and compare them with the bound given by Markov Inequality.

To be able to bound probabilities for general random variables (not necessarily positive), and to get a tighter bound, we can apply Markov Inequality to  $(X - E(X))^2$  and obtain:

**Chebychev Inequality:** For random variable  $X$  with mean  $E(X)$  and variance  $\sigma^2$ , and any real number  $a > 0$ ,

$$P(|X - E(X)| \geq a) \leq \frac{\sigma^2}{a^2}$$

Proof: Bound  $P((X - E(X))^2 \geq a^2)$  using Markov Inequality.

Note that Chebychev's Inequality uses more information about  $X$  in order to provide a tighter bound about the probabilities related to  $X$ . In addition to the mean (a first-order statistic), it also uses the variance, which is a second-order statistic. You can easily imagine two very different random variables with the same mean: for example, a zero-mean Gaussian with variance 2, and a discrete random variable that takes on the values  $+0.1$ , and  $-0.1$  equally probably. Markov Inequality does not distinguish between these distributions, where as Chebychev Inequality does.

It is possible to get better bounds that use more information- such as the Chernoff bound (but those are beyond the scope of this course.)

**Ex:** Use Chebychev's Inequality to lower-bound the probability that a Gaussian within two standard deviations of its mean.

## 5.2 Probabilistic Convergence

Let us remember the definition of convergence for a deterministic sequence of numbers from basic Calculus. Let  $\{a_n\}$  be a sequence of numbers, indexed by  $n$ .  $\lim_{n \rightarrow \infty} a_n = a$  means, for every  $\epsilon > 0$ , there exists an  $n_o$  such that for all  $n > n_o$ ,  $|a_n - a| < \epsilon$ .

### 5.2.1 Convergence "In Probability"

Now, instead of a sequence of numbers, consider a sequence of random variables  $\{Y_n\}$ , indexed by  $n$ .  $Y_n$  converges in probability to a number  $a$  if for every  $\epsilon > 0$ ,  $\lim_{n \rightarrow \infty} P(|Y_n - a| > \epsilon) = 0$ . The notation is:  $Y_n \xrightarrow{i.p.} a$ .

**Ex:** Suppose for each  $n$ ,  $Y_n$  takes the value  $n$  with probability  $1/n$ , and the value zero with probability  $1 - 1/n$ . Does  $\{Y_n\}$  converge, and if so, to what?