1) Law of Large Numbers (LLN), Convergence in Probability and Distribution

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Markov Inequality

Let X be a nonnegative R.V. and t > 0.

$$E(X) = \sum_{x} xf(x) = \sum_{x < t} xf(x) + \sum_{x \ge t} xf(x)$$

$$E(X) \ge \sum_{x \ge t} xf(x) \ge \sum_{x \ge t} tf(x) = tPr(X \ge t)$$

$$-tPr(X \ge t) \ge -E(X)$$

$$Pr(X \ge t) \le \frac{E(X)}{t}$$

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Chebyshev Inequality

Let
$$Y = [X - E(X)]^2$$
, then $E(Y) = Var(X)$

$$Pr(X \ge t) \le \frac{E(X)}{t}$$

$$Pr(|X - E(X)| \ge t)$$

$$= Pr(Y \ge t^2) \le \frac{Var(X)}{t^2}$$

$$Pr(|X - E(X)| \ge t) \le \frac{Var(X)}{t^2}$$

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Properties of the Sample Mean

$$\bar{X}_n = \frac{1}{n}(X_1 + \ldots + X_n)$$

$$E(\bar{X}_n) = \frac{1}{n}E(X_1 + ... + X_n) = \frac{1}{n}n\mu = \mu$$

$$Var(\bar{X}_n) = \frac{1}{n^2} Var(\sum_{i=1}^n X_i) = \frac{1}{n^2} n\sigma^2 = \frac{\sigma^2}{n^2}$$

$$Se(\bar{X}_n) = \frac{\sigma}{\sqrt{n}}$$

Weak Law of Large Numbers (WLLN)

$$Pr(|X - E(X)| \ge t) \le \frac{Var(X)}{t^2}$$
 $Pr(|\bar{X}_n - \mu| < \epsilon) \ge 1 - \frac{\sigma^2}{n\epsilon^2}$
 $\lim_{n \to \infty} Pr(|\bar{X}_n - \mu| < \epsilon) = 1$

Convergence in Probability: $\bar{X}_n \stackrel{p}{\to} \mu$

 μ is the probability limit (plim) of $ar{X}_n$

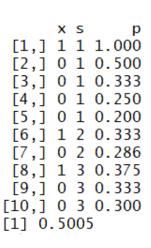
Fair Coin Simulation

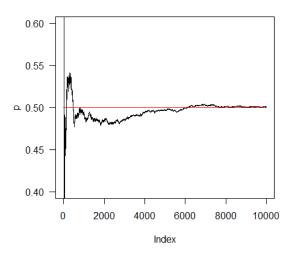
```
set.seed(7)
n < -10000
x <- sample(0:1, n, repl=T)
s < -cumsum(x); p < -s/(1:n)
plot(p, ylim=c(.4, .6), type="l")
lines(c(0,n), c(.5,.5), col = "red")
round(cbind(x,s,p), 3)[1:10,]; p[n]
```

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LLN: Fair Coin



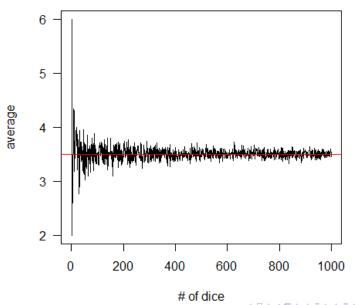


Dice Roll Simulation

```
die < -1.6
roll <- function(n) {</pre>
mean(sample(die, size = n, replace = TRUE))
plot(sapply(1:1000, roll), type = "l",
xlab = \# of dice , ylab = "average")
abline(h = 3.5, col = "red")
```

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Expected Value of Rolling Dice



Convergence in Distribution

A sequence of R.Vs. $\{b_n\}$ converge in distribution to a R.V. b if

$$\lim_{n\to\infty} F_n = F$$

at every continuity point of F, where F_n is the distribution of b_n , F is the distribution of b.

$$b_n \stackrel{d}{\rightarrow} b$$

F is the limit distribution of $\{b_n\}$

Continuous Mapping and/or Slutsky's Theorem

$$b_n \xrightarrow{p} b \Rightarrow b_n \xrightarrow{d} b$$

$$b_n \xrightarrow{p} b \Rightarrow g(b_n) \xrightarrow{p} g(b)$$

$$a_n \xrightarrow{d} a \Rightarrow g(a_n) \xrightarrow{d} g(a)$$

$$1) a_n + b_n \xrightarrow{d} a + b$$

$$2) a_n b_n \xrightarrow{d} ab$$

$$3) \frac{a_n}{b_n} \xrightarrow{d} \frac{a}{b} \text{ if } b \neq 0$$