Random Variables

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2024-02-22

Introduction

Key Points

- Random variables are numeric outcomes resulting from random processes.
- Statistical inference offers a framework for quantifying uncertainty due to randomness.

Model a random variable with sample() function.

```
set.seed(1)
# define random variable x to be 1 if blue, 0 otherwise
beads <- rep(c("red", "blue"), times = c(2, 3))
x <- ifelse(sample(beads, 1) == "blue", 1, 0)
print(x)</pre>
```

Sampling Models

Key Points

[1] 0

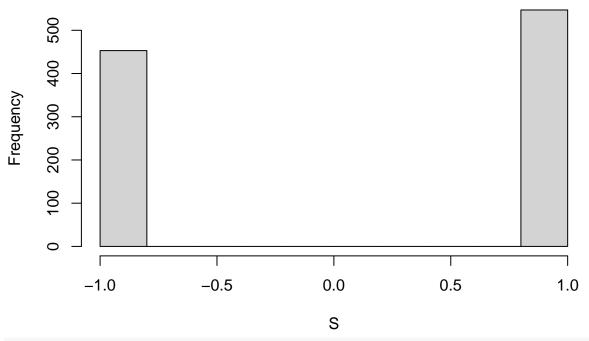
- A sampling model models the random behavior of a process as the sampling of draws from an urn.
- The probability distribution of a random variable is the probability of the observed value falling in any given interval.
- We can define a CDF F(a)=Pr(S<=a) to answer questions related to the probability of S being in any interval.
- The average of many draws of a random variable is called its expected value.
- The standard deviation of many draws of a random variable is called its standard error.

We use sampling model to run a Monte Carlo simulation and use the results to estimate the probability of an event. In this example, we have a casino with 1000 players of roulette each day. We have 38 pockets (18 red, 18 black and 2 green). The player bet 1\$ and the events are:

- red: the casino losses 1\$
- $\bullet\,$ black and green: the casino wins 1\$

The model of this random variable is:

Histogram of S



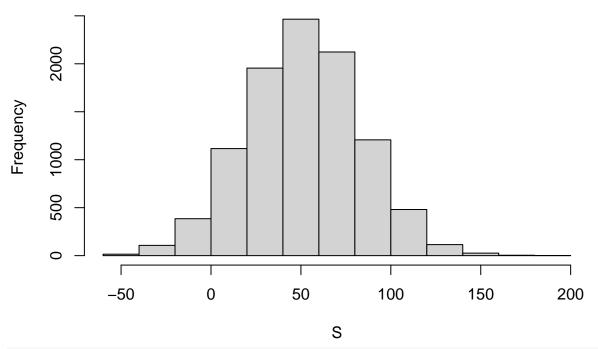
```
sum(S) # total gains
## [1] 94
mean(S) # mean gain
```

[1] 0.094

We use the sampling model to run a Monte Carlo simulation

```
n <- 1000  # number of roulette players
B <- 10000  # number of Monte Carlo experiments
S <- replicate(B, {
    X <- sample(c(-1,1), n, replace = TRUE, prob = c(9/19, 10/19))  # simulate 1000 roulette spins
    sum(X)  # determine total profit
})
hist(S) # distribution of results</pre>
```

Histogram of S



mean(S) # mean results

[1] 52.5392

mean(S < 0) # probability of the casino losing money</pre>

[1] 0.0442