Leyes y desigualdades notables

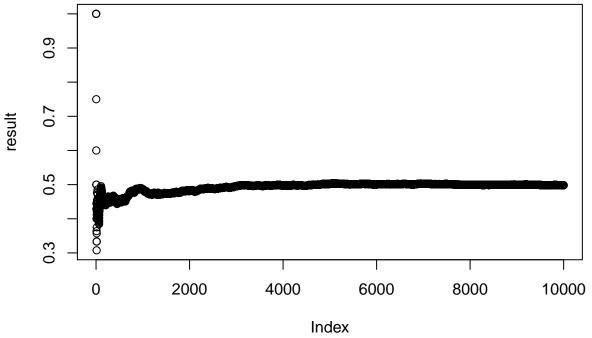
Alexander A. Ramírez M. (alexanderramirez.me) 26 de Agosto, 2016

1. Frecuencia relativa

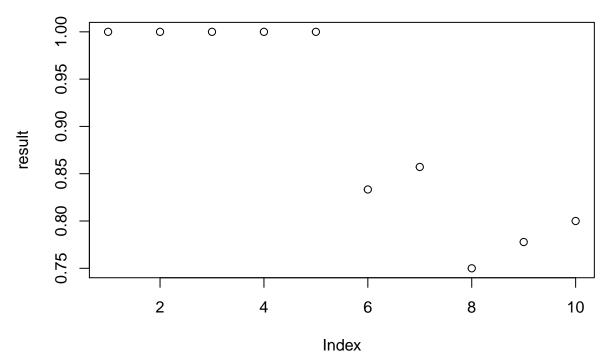
La frecuencia relativa es la cantidad de veces que un evento E ocurre a partir de n resultados mutuamente excluyentes (no ocurren resultados distintos en cada intento del experimento) de un experimento. La denotamos como:

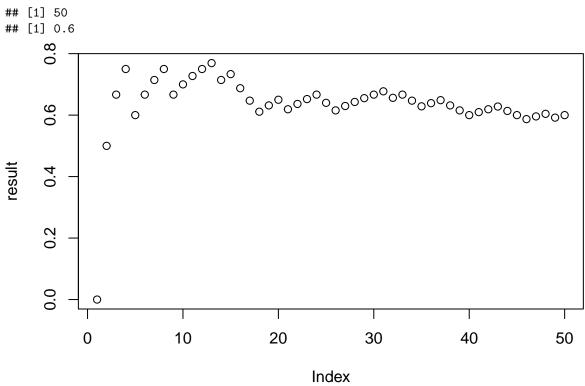
$$P(E) = \frac{1}{n} \sum_{i=1}^{n} \mathbb{1}_{E}(x_i)$$

```
relative_freq<-function(n,vars) {</pre>
  result<-c()
  heads < -0
  for (i in 1:n) {
    trial<-rbinom(vars,1,.5)</pre>
    if (trial==0) {
      heads<-heads+1
    result <- append (result, heads / i)
  return(result)
}
n<-10000
vars<-1
result <- relative_freq(n, vars)
result[n]
## [1] 0.4983
plot(result)
```

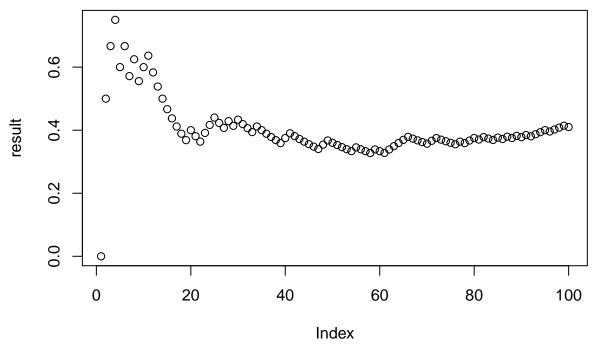


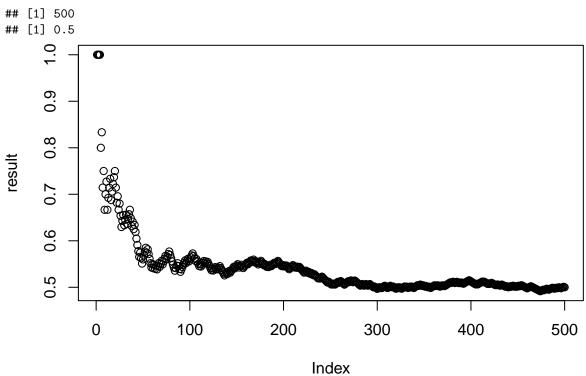
```
mean(result)
## [1] 0.4927526
sd(result)
## [1] 0.01703201
vars<-1
for (n in c(10,50,100,500,1000,5000,10000)) {
    result<-relative_freq(n,vars)
    print(n)
    print(result[n])
    #write("n->",n," mean->", result[n])
    plot(result)
}
## [1] 10
## [1] 0.8
```



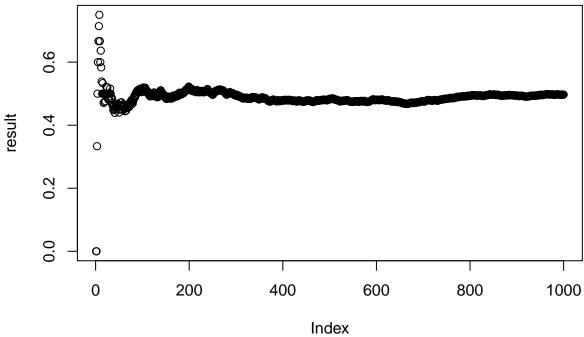


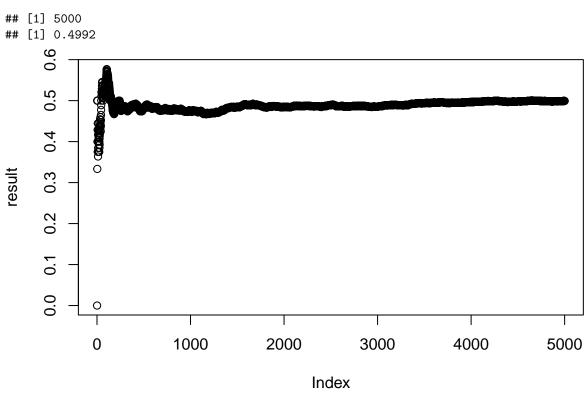
[1] 100 ## [1] 0.41



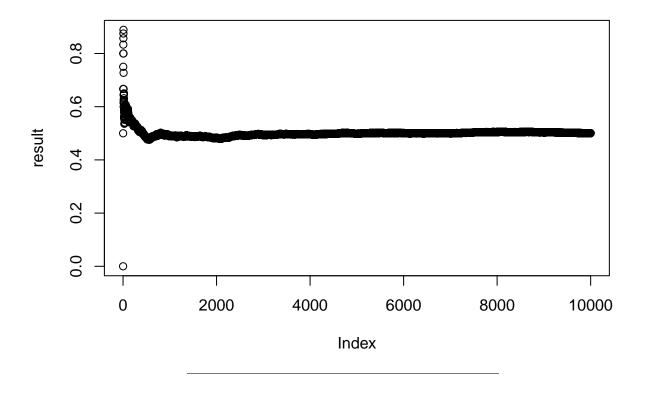


[1] 1000 ## [1] 0.497





[1] 10000 ## [1] 0.5005



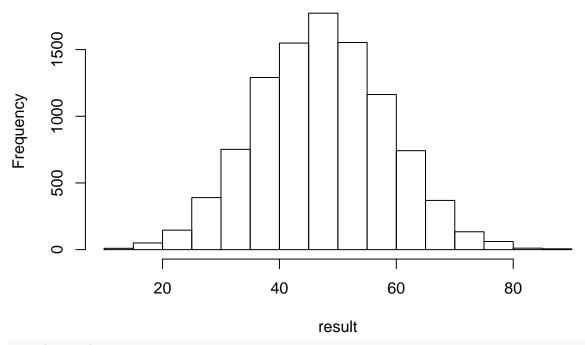
2. Suma de variables aleatorias Bernoulli

3. Media de variables aleatorias Bernoulli

```
meanof_bernoulli<-function(n,vars) {
   result<-c()
   for (i in 1:n) {
      result<-append(result,mean(rbinom(vars,1,.5))*sqrt(n))
   }
   return(result)
}

n<-10000
vars<-20
result<-meanof_bernoulli(n,vars)
hist(result)</pre>
```

Histogram of result



mean(result)

[1] 49.847

sd(result)

[1] 11.28359

El promedio de n medias:

mean(result)

[1] 49.847

sd(result)

[1] 11.28359

4. Desigualdad de Markov

Sea X una variable aleatoria no negativa y a>0,

$$P(X \ge a) \le \frac{E(X)}{a}$$

o también

$$P(\mid X - \mu \mid \ge k\sigma) \le \frac{1}{k^2}$$

5. Desigualdad de Tchebysheff

Sea X una variable aleatoria con media μ y varianza $\sigma^2 < \infty$. Entonces, para k > 0,

$$P(\mid X - \mu \mid < k\sigma) \ge 1 - \frac{1}{k^2}$$

o también

$$P(\mid X - \mu \mid \ge k\sigma) \le \frac{1}{k^2}$$

Otra notación es la siguiente:

$$P(\mid X - E(X) \mid \ge k) \le \frac{\sigma^2}{k^2}$$

6. Ley de los grandes números

6.1. Ley débil

Sean X_1, X_2, X_3, \ldots una sucesión infinita de variables aleatorias independientes y distribuidas idénticamente con $E(Y_i) = \mu$ y $Var(Y_i) = \sigma^2 < \infty$, entonces el promedio

$$\bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i$$

converge en probabilidad a μ .

Es decir, sea $\epsilon > 0$ entonces

$$\lim_{n \to \infty} P(|\bar{X}_n - \mu| < \epsilon) = 1$$

Donde podemos decir que

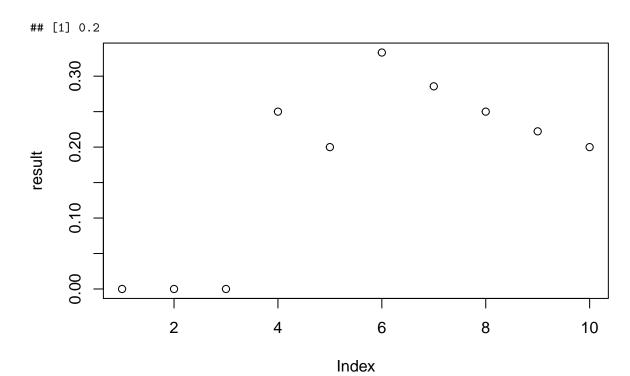
$$P(\mid \bar{X}_n - \mu \mid < \epsilon) = \frac{1}{n} \sum_{i=1}^n \mathbb{1}_{(-\epsilon + \mu, \epsilon + \mu)}(\bar{X}_n)$$

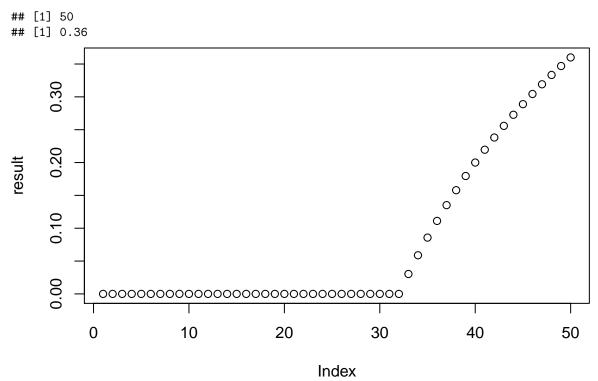
```
weak_law<-function(n,vars,epsilon) {
    result<-c()
    heads<-0
    f<-0
    for (i in 1:n) {
        trial<-rbinom(vars,1,.5)
        if (trial==0) {
            heads<-heads+1
        }
        if (abs((heads/i)-0.5)<epsilon) {
            f<-f+1</pre>
```

```
result<-append(result,f/i)</pre>
  return(result)
}
n<-10000
vars<-1
epsilon<-0.01
result <-weak_law(n, vars, epsilon)
result[n]
## [1] 0.5061
plot(result)
     0.5
     0.4
     0.3
     0.2
     0.1
     0.0
                         2000
                                                                                  10000
             0
                                        4000
                                                      6000
                                                                    8000
                                               Index
mean(result)
## [1] 0.20673
sd(result)
## [1] 0.1610966
vars<-1
epsilon < -0.05
for (n in c(10,50,100,500,1000,5000,10000)) {
  result <-weak_law(n, vars, 0.05)
  print(n)
  print(result[n])
  \#write("n->",n," mean->", result[n])
```

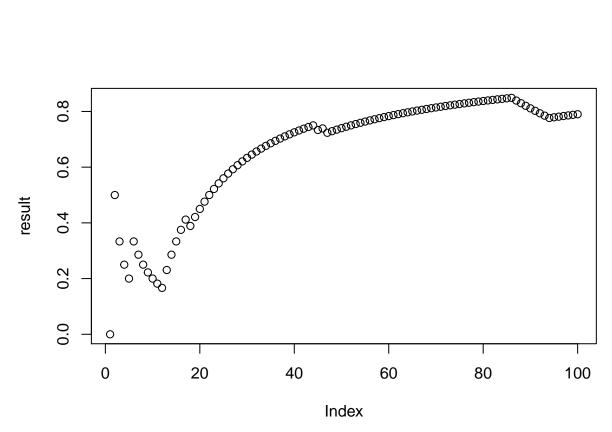
[1] 10

plot(result)

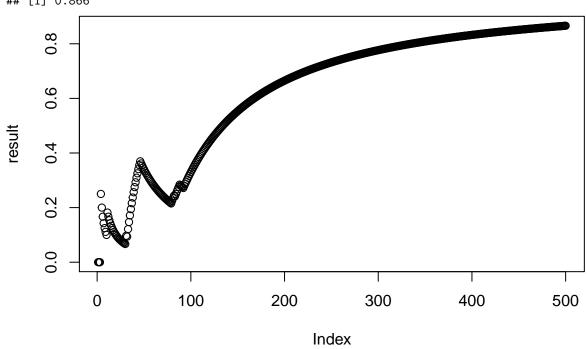




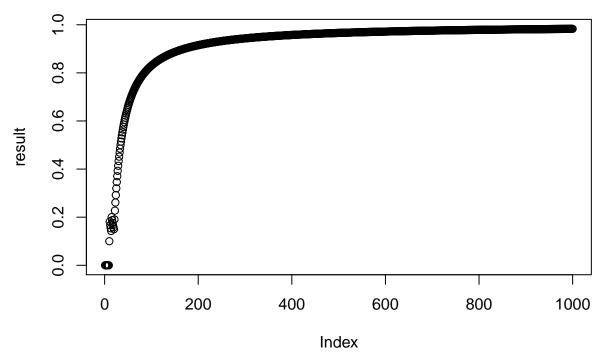
[1] 100 ## [1] 0.79

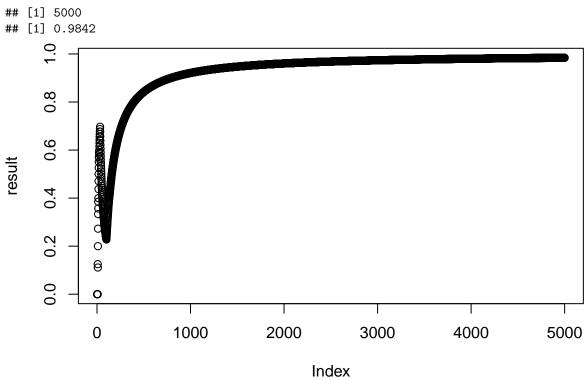




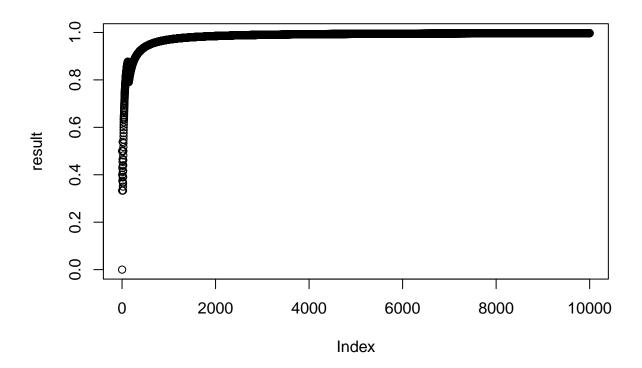


[1] 1000 ## [1] 0.983





[1] 10000 ## [1] 0.997



6.2. Ley fuerte

7. Teorema del límite central

Sean Y_1, Y_2, Y_n variables aleatorias independientes y distribuidas idénticamente con $E(Y_i) = \mu$ y $Var(Y_i) = \sigma^2 < \infty$. Definamos

$$U_n = \frac{\sum Y_i - n\mu}{\sigma\sqrt{n}} = \frac{\bar{Y} - \mu}{\sigma/\sqrt{n}}, \ \bar{Y} = \frac{1}{n} \sum_{i=1}^n Y_i$$

Entonces la función de distribución de U_n converge hacia la función de distribución normal estándar cuando $n \to \infty$. Esto es,

$$\lim_{n\to\infty}P(U_n\leq u)=\int_{-\infty}^u\frac{1}{\sqrt{2\pi}}e^{-\frac{t^2}{2}}dt, \text{ para toda u}$$