Random Walk With Drift

Le'Sean Roberts

2 December 2015

Use a biased coin to simulate a random walk of 30 steps on the line. If the coin comes up heads (H), take one step to the right, if it comes up tails (T), take one step left. After 30 steps, note the final position. Take the probability P(H) = 0.75. Plot a sample path of the random walk for 30 time steps. Make a histogram of the final position for 1 million such random walks. Compute the sample mean and the sample variance.

Brownian Motion can be recognised from a proposed Drunk's trek. GEOMETRIC DISPLAYS MAY TAKE SOME TIME TO EXHIBIT

```
n<-30 # 30 steps in drunkard's walk.
prob<-0.75# probability of heads via binomial simulation.
size<-1e6
u<-runif(n)
u<prob</pre>
```

```
## [1] FALSE TRUE FALSE TRUE TRUE FALSE TRUE TRUE TRUE TRUE TRUE
## [12] TRUE TRUE TRUE TRUE FALSE TRUE TRUE TRUE TRUE TRUE
## [23] TRUE FALSE TRUE TRUE TRUE TRUE TRUE FALSE
```

```
staggerstep<-function(size,n,prob) sum(runif(n)<prob) #action catering movement staggers. sample<-replicate(size,staggerstep(size, n,prob)) # 1e6 trials for drunk. hist(sample,100,col="green") # E(x)=n*p; Var(x)=n*p*(1-p) 30*0.75 # analytic formula for binomial expectation.
```

```
## [1] 22.5
```

mean(sample) # resultis consistent with analytic formula of binomial expectation.

```
## [1] 22.49858
```

30*0.75*0.25 # analytic formula for binomial variance.

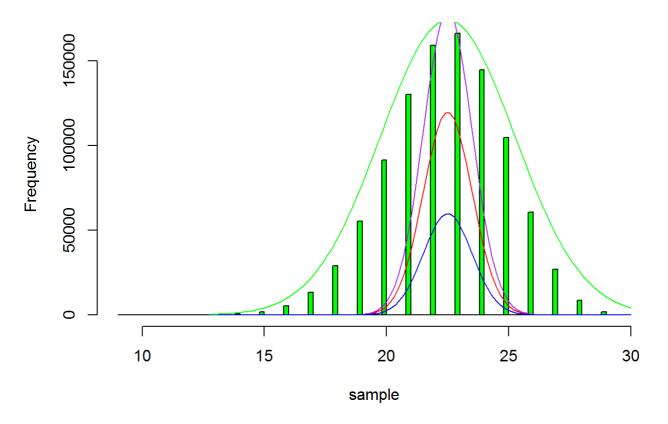
```
## [1] 5.625
```

mean(sample)*(1-prob) # result consistent with analytic formula for binomial variance.

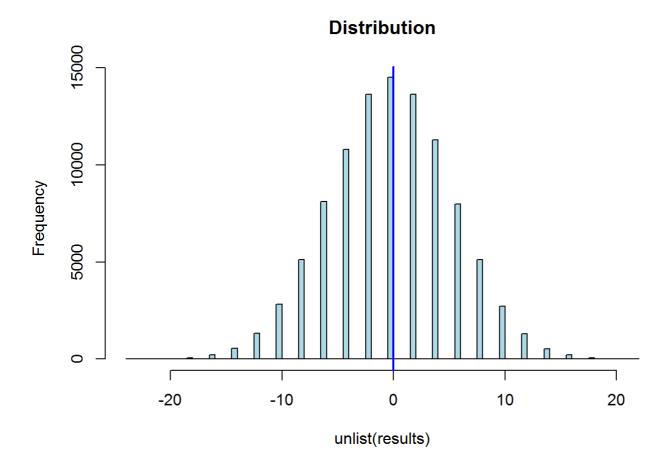
```
## [1] 5.624646
```

```
curve(300000*dnorm(x,22.5,1),12.75,30,col="red",add=T)
curve(450000*dnorm(x,22.5,1),12.75,30,col="purple",add=T)
curve(150000*dnorm(x,22.5,1),12.75,30,col="blue",add=T)
curve(1.2e6*dnorm(x,22.5,2.75),12.75,30,col="green",add=T)
```

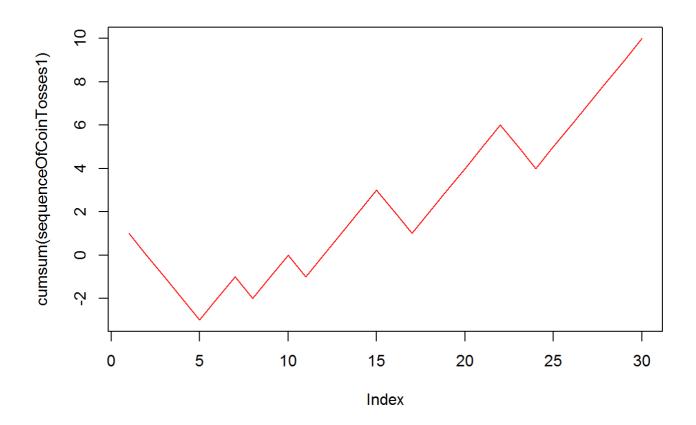
Histogram of sample



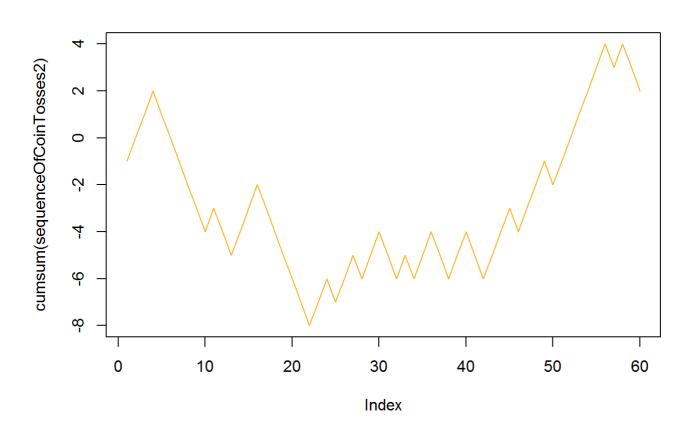
```
# OBSERVING NOW A MORE DIRECT APPROACH WITH FAIRNESS FOR GENERALITY.
# Empty list creation to store results
results <- list()
for(i in 1:100000) {
    coinTosses <- cumsum(sample(c(-1,1), 30, replace = TRUE))
    results[[i]] <- coinTosses[length(coinTosses)]
}
# Unlist the list and create a histogram.
hist(unlist(results), main = "Distribution",col = "lightblue", breaks = 100)
# Vertical line at 0, of breadth 2 to exhibit the expectation of the distribution
abline(v = 0, col = "blue", lwd = 2)</pre>
```



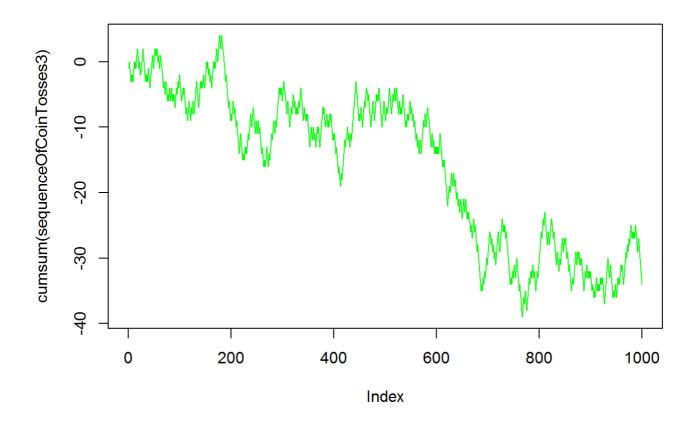
```
sequenceOfCoinTosses1 <- sample(c(-1,1), 30, replace = TRUE)
sequenceOfCoinTosses2 <- sample(c(-1,1), 60, replace = TRUE)
sequenceOfCoinTosses3 <- sample(c(-1,1), 1000, replace = TRUE)
plot(cumsum(sequenceOfCoinTosses1), type = 'l',col="red")</pre>
```



plot(cumsum(sequenceOfCoinTosses2), type = 'l',col="orange")

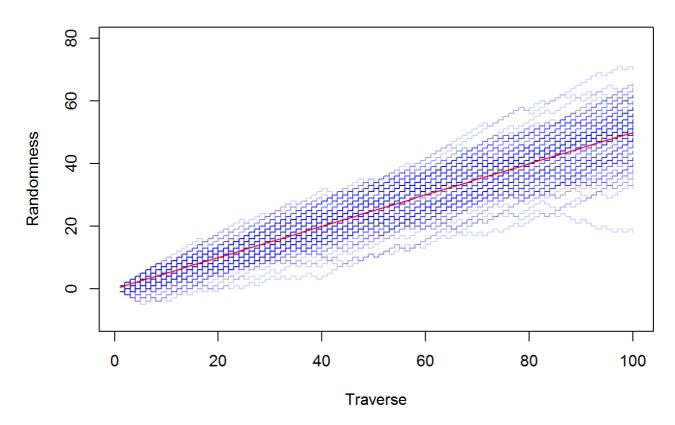


plot(cumsum(sequenceOfCoinTosses3), type = '1', col="green")



10/30/21, 5:05 AM Random Walk With Drift

100 Sample Paths of Random Walk (Probability 0.75)



CONCLUSION: A sample size of 1e6 simulations of the like random variables implies the Central Limit Theorem. Such is known with higher number experiments of binomial processes. As well from observation of simulations with "Sequence of Coin Tosses" functions with fairness (being binomial), the higher the mount of trials, leads to more "randomness" in successive geometric figures.