

# Estimating Population Size

In any war, it is always of value to one side to have good intelligence on the weapons resources of the other side. During the Second World War, for example, Allied military planners eagerly searched for ways to accurately estimate the Axis production of tanks, aircrafts and numerous other weapons platforms. In the specific case of German tanks, a very clever way to do that was based on using either the stamped serial numbers or gearbox markings on captured Mark I and Mark V tanks, respectively. This type of problem has far wider applications.

We can experimentally see how well the formula that was used works in practice with a Monte Carlo Simulation. That is, the program first randomly picks a value for N (integer between 100 and 1000) with a value for the sample size as a percentage of N.

The program then generates, randomly, n different integers in the interval 1 to N, the maximum of those integers is then used in the estimation formula to estimate a value for N. This estimate can be compared to the actual value of N to determine how well the formula has performed. We investigate samples that are 2%, 5%, 10% and 20% of the population.

## Function to implement Monte Carlo simulation using formula

```
runSim <- function(nSim=1000, sample.size.perc=10/100) {  
  
  #set up an array to store parameter estimates  
  estArray <- array(0, dim=c(nSim,3))  
  
  for (i in 1:nSim) {  
  
    N <- sample(100:1000,1) # select a population size  
  
    n <- round(sample.size.perc*N,0) # the sample size  
  
    sample2 <- sample(N, n) # select a sample from the population  
  
    est <- (n+1)/n*max(sample2) - 1 # apply formula  
  
    est <- round(est,0) # estimate of pop size  
  
    estArray[i,1] <- (est - N)/ N *100 # percentage error is captured  
    estArray[i,2] <- N  
    estArray[i,3] <- sample.size.perc  
  
  }  
  
  list(estArray=estArray )  
}
```

## Sample size 2% of population

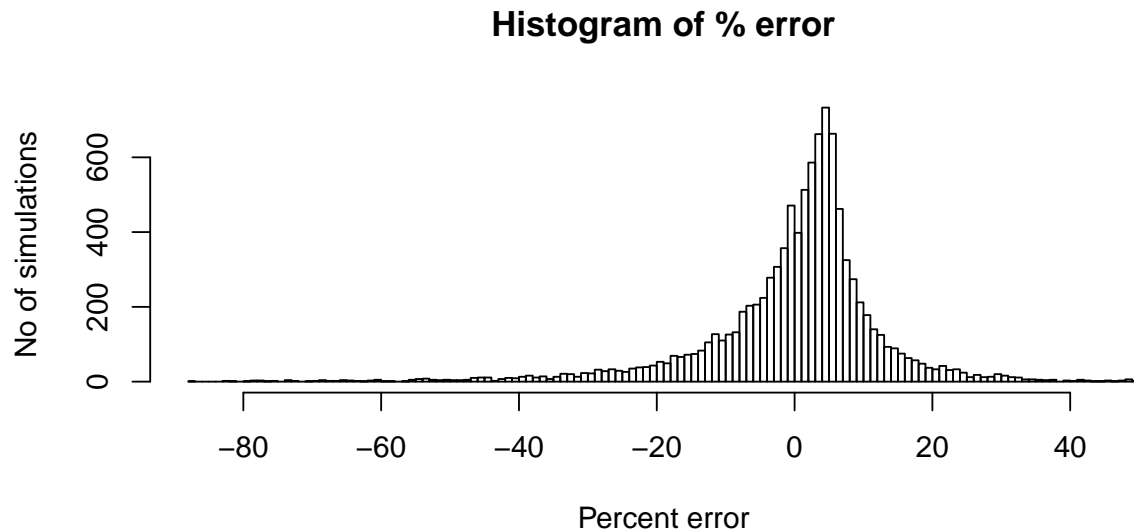
```
x<- runSim(nSim=10000, sample.size.perc=2/100) # run simulation  
apply(x$estArray, 2, mean, na.rm=TRUE)[1] # mean % error
```

```
[1] 0.01385591
```

```
quantile( x$estArray[,1] , prob=c(0.025, 0.5, 0.975)) # median and 95% confidence intervals for % error

      2.5%      50%      97.5%
-33.333333  2.290867  22.222222

hist(x$estArray[,1], breaks=100, main="Histogram of % error" , xlab="Percent error", ylab="No of simulations")
```



### Sample size 5% of population

```
x<- runSim(nSim=10000, sample.size.perc=5/100) # run simulation
apply(x$estArray, 2, mean, na.rm=TRUE)[1]      # mean % error

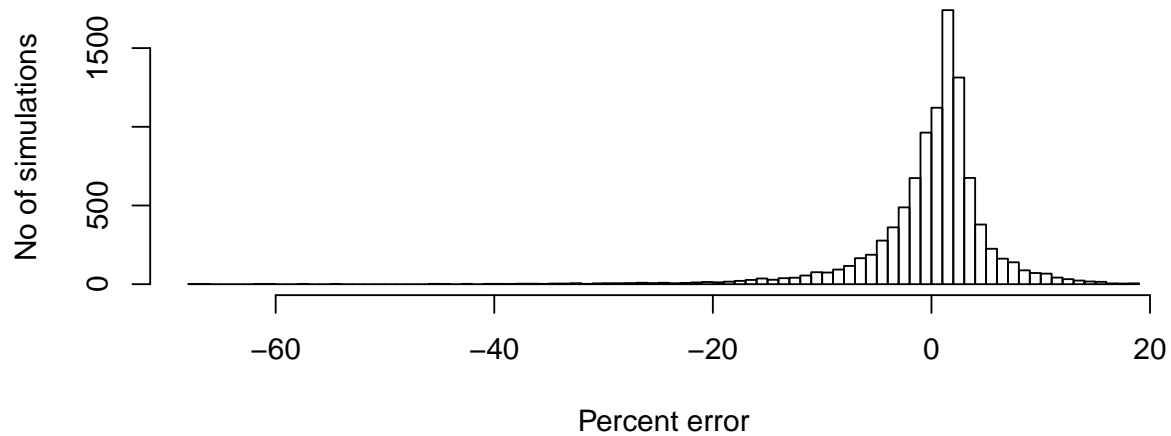
[1] -0.0187242

quantile( x$estArray[,1] , prob=c(0.025, 0.5, 0.975)) # median and 95% confidence intervals for % error

      2.5%      50%      97.5%
-14.525259  1.003344  9.524408

hist(x$estArray[,1], breaks=100, main="Histogram of % error" , xlab="Percent error", ylab="No of simulations")
```

## Histogram of % error



## Sample size 10% of population

```
x<- runSim(nSim=10000, sample.size.perc=10/100) # run simulation
apply(x$estArray, 2, mean, na.rm=TRUE)[1]        # mean % error
```

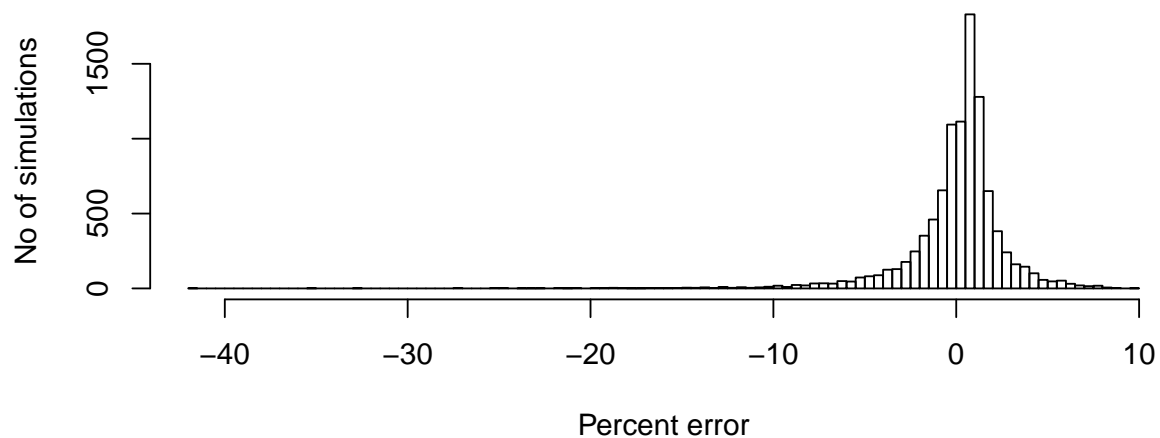
```
[1] 0.04915648
```

```
quantile( x$estArray[,1] , prob=c(0.025, 0.5, 0.975)) # median and 95% confidence intervals for % error
```

```
      2.5%      50%      97.5%
-6.8185347  0.5107252  4.5005653
```

```
hist(x$estArray[,1], breaks=100, main="Histogram of % error" , xlab="Percent error", ylab="No of simulations")
```

## Histogram of % error



## Sample size 20% of population

```
x<- runSim(nSim=10000, sample.size.perc=20/100) # run simulation
apply(x$estArray, 2, mean, na.rm=TRUE)[1]        # mean % error

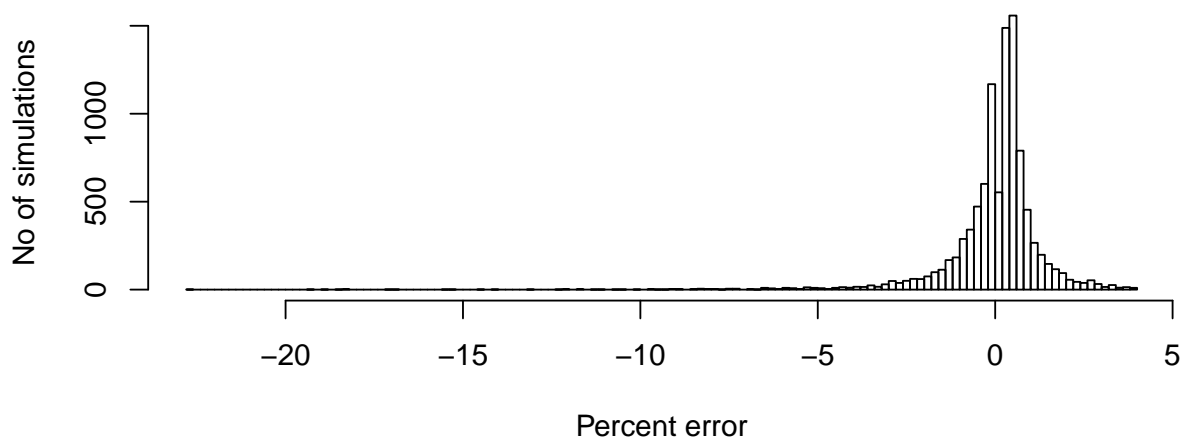
[1] 0.01389191

quantile( x$estArray[,1] , prob=c(0.025, 0.5, 0.975)) # median and 95% confidence intervals for % error

      2.5%      50%      97.5%
-3.1392783  0.2453988  2.1739130

hist(x$estArray[,1], breaks=100, main="Histogram of % error" , xlab="Percent error", ylab="No of simulations")
```

**Histogram of % error**



## Computing Environment

```
sessionInfo()
```

```
R version 3.6.0 (2019-04-26)
Platform: x86_64-w64-mingw32/x64 (64-bit)
Running under: Windows 10 x64 (build 17134)
```

```
Matrix products: default
```

```
locale:
```

```
[1] LC_COLLATE=English_United Kingdom.1252
[2] LC_CTYPE=English_United Kingdom.1252
[3] LC_MONETARY=English_United Kingdom.1252
[4] LC_NUMERIC=C
[5] LC_TIME=English_United Kingdom.1252
```

```
attached base packages:
```

```
[1] stats      graphics  grDevices  utils      datasets
[6] methods    base
```

other attached packages:

[1] knitr\_1.23

loaded via a namespace (and not attached):

[1] compiler\_3.6.0 magrittr\_1.5 tools\_3.6.0  
[4] htmltools\_0.3.6 yaml\_2.2.0 Rcpp\_1.0.1  
[7] stringi\_1.4.3 rmarkdown\_1.13 stringr\_1.4.0  
[10] xfun\_0.7 digest\_0.6.19 evaluate\_0.14

This took 0.87 seconds to execute.