Examining the Distribution of Experimental Averages of Samples of the Exponential Distribution

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Project Overview

For this project, we generated samples consisting of 40 random number generations (hereby referred to as observations) distributed via the exponential distribution. We then took the mean value among the 40 observations in each sample. It is these 40-observation-means (hereby referred to as averages) that we examined, calculating the mean and variance of the distribution of these averages. We then compared the distribution of these averages to a normal distribution whose mean and variance matched the theoretical mean and variance of our set of averages.

The distribution of the individual observations

Attached in the appendix are plots showing the experimental observations with the theoretical exponential distribution overlayed for comparison. You can see above that the set of observations more closely approximates the theoretical distribution curve as the number of observations increases. This is an expected consequence of the law of large numbers. As the number of observations increases, the more closely we expect the distribution of the observations to match the theoretical distribution.

The distribution of the sample averages of the observations

Function generateBatchesOfObservations() was the function which actually executed the random number generation. Function rexp() was used to generate the observations. The mean() function was used to evaluate the averages of each set of 40 observations. mean() was also used to generate the expiremental mean of the averages; sd() was used to compute the standard deviation of the averages (and therefore the variance as well). Function produceStatisticsReport() output the theoretical and experimental values associated with the generated observation averages. Function

produceSampleDensityHistogramAgainstNormalCurve() created a histogram overlaying the theoretical normal distribution of observation averages over our set of experimental observation averages. In the histogram, our experimental observation averages are displayed as densities, not as frequencies; this allows us to easily use dnorm() to generate the theoretical normal curve necessary for comparison. The code is commented to explain step-by-step how these plots are generated.

The Theoretical mean of the Sample Averages is the population mean, which is 1/lambda for the exponential distribution. As such, with lambda fixed to .2 for our simulations, the population mean was 5, resulting in the theoretical sample average to also have an expected mean of 5.

The Theoretical Population Variance refers to the amount of variation we expect in a random variable. This value is independent of the number of observations which may be sampled from a population. The Theoretical Variance of the Sample Averages is the Population Variance divided by the number of observations n which comprise each Average's sample size. As such, the more observations we have comprising each sample average, the less we expect the distribution of the sample averages to vary; this is due to the law of large numbers taking effect. This differs from the population variance because it refers to the distribution of experimental sample AVERAGES, not the distribution of individual members of the population. Examining the distribution of averages allows us to smooth away outliers in the data population.

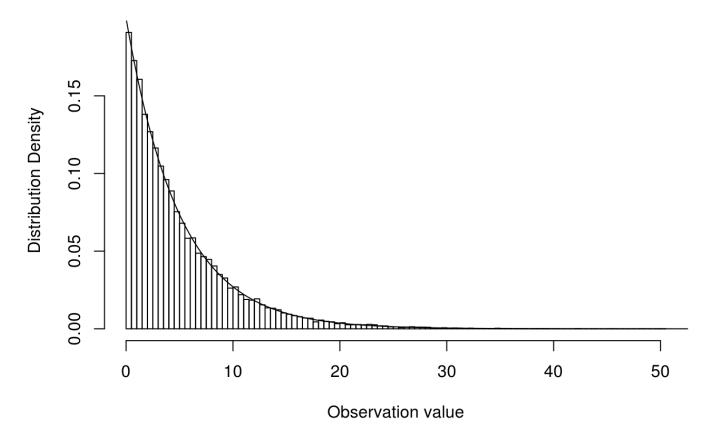
As you can see below, while the experimental mean approximated the population mean for all sample sizes, the variance decreased as the size of the samples increased. Theoretically this is expected as the theoretical variance of a sample mean is the population mean divided by the number of observations per sample, mentioned above. Intuitively this makes sense as the law of large numbers implies that as we take more samples, we expect that the average observation should approach the expected value. This is why the expected value is expected in the first place. The numbers in the table below are compared to their theoretical values in the appendix

```
## 1000 Samples 10000 Samples 100000 Samples
## Mean 4.990898 4.9953912 4.998901
## Variance 0.634041 0.6231779 0.621723
```

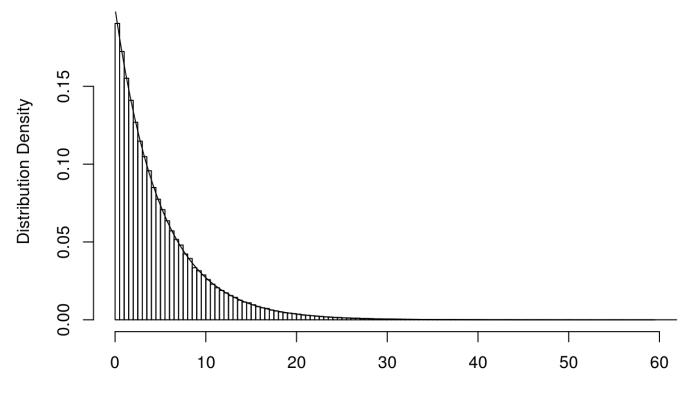
In the histograms shown in the appendix, you can see the distribution of the sample averages more closely approximating a normal distribution as the number of sample averages increases. This is expected as per the Central Limit Theorem and Law of Large Numbers

Appendix

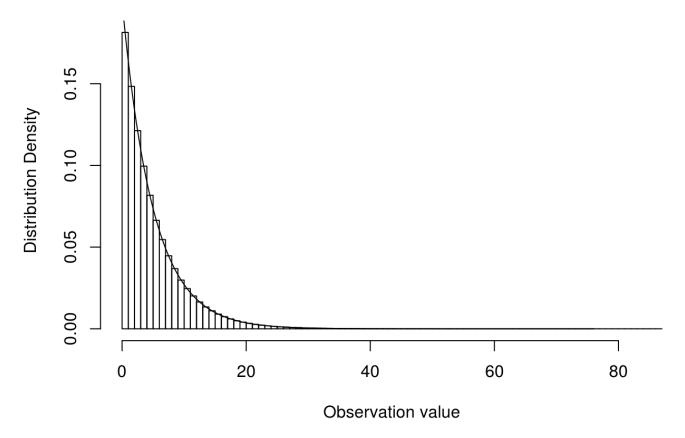
Observation Value Distribution Compared To Normal Distribution for 40000 Observations



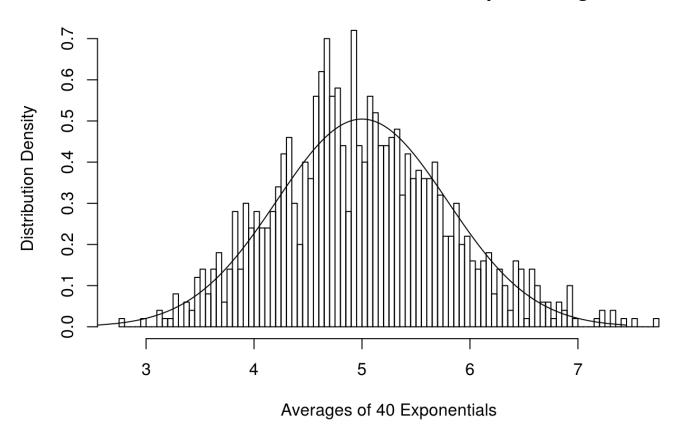
Observation Value Distribution Compared To Normal Distribution for 400000 Observations



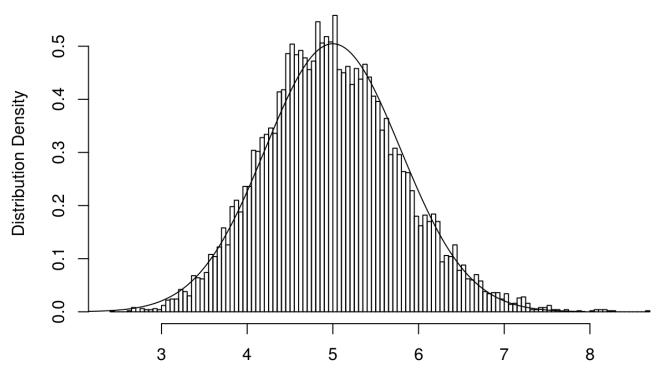
Observation Value Distribution Compared To Normal Distribution for 4000000 Observations



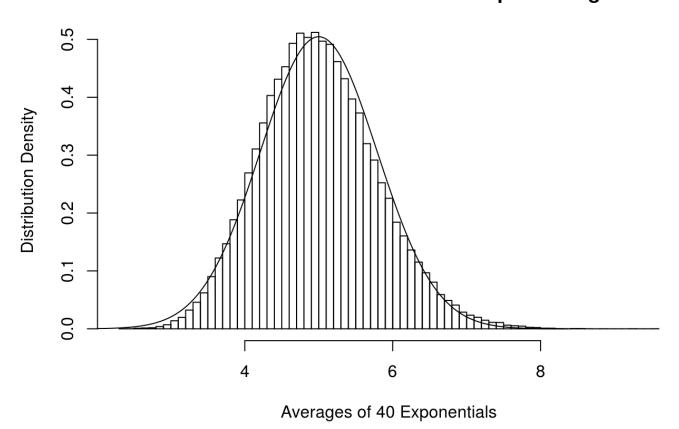
Sample Average Distribution Compared To Normal Distribution for 1000 Sample Averages



Sample Average Distribution Compared To Normal Distribution for 10000 Sample Averages



Sample Average Distribution Compared To Normal Distribution for 100000 Sample Averages



```
[1] "1000 samples of the average of 40 exponentials with lambda=0.2 were take
##
n:"
##
    [2]
    [3] "Population Average: 5"
##
        "Theoretical Mean of Averages of 40 exponentials: 5"
##
##
    [5]
        "Experimental Mean of Averages of 40 exponentials: 4.99089829499993"
##
    [6]
        "Theoretical Variance of Averages of 40 exponentials: 0.625"
##
    [7]
    [8] "Experimental Sample Variance of Averages of 40 exponentials: 0.6340410159
##
73402"
##
    [9]
## [10] "Theoretical Standard Devation of Averages of 40 exponentials: 0.790569415
042095"
## [11] "Experimental Standard Devation of Averages of 40 exponentials: 0.79626692
5078143"
```

```
##
    [1] "10000 samples of the average of 40 exponentials with lambda=0.2 were take
n:"
    [2]
##
    [3]
        "Population Average: 5"
##
        "Theoretical Mean of Averages of 40 exponentials: 5"
       "Experimental Mean of Averages of 40 exponentials: 4.9953911570227"
##
##
    [6]
##
    [7] "Theoretical Variance of Averages of 40 exponentials: 0.625"
    [8] "Experimental Sample Variance of Averages of 40 exponentials: 0.6231778782
7566"
##
    [9]
## [10] "Theoretical Standard Devation of Averages of 40 exponentials: 0.790569415
042095"
## [11] "Experimental Standard Devation of Averages of 40 exponentials: 0.78941616
2917671"
    [1] "100000 samples of the average of 40 exponentials with lambda=0.2 were tak
##
en:"
   [2]
##
##
    [3] "Population Average: 5"
##
    [4] "Theoretical Mean of Averages of 40 exponentials: 5"
        "Experimental Mean of Averages of 40 exponentials: 4.99890111950532"
##
##
   [6]
        "Theoretical Variance of Averages of 40 exponentials: 0.625"
    [8] "Experimental Sample Variance of Averages of 40 exponentials: 0.6217230417
##
42492"
##
    [9]
## [10] "Theoretical Standard Devation of Averages of 40 exponentials: 0.790569415
042095"
## [11] "Experimental Standard Devation of Averages of 40 exponentials: 0.78849416
0880404"
```

Below is the R code used to generate this report

```
exponentialObservationsGenerator <- function(number of batches to construct with =
1000)
{
    lambda <- 0.2;
    observations per batch <- 40;
    number of batches to observe <- number of batches to construct with;
    averages of batches <- NULL;
    sample mean <- NULL;</pre>
    sample standard deviation <- NULL;</pre>
    matrix of observation batches <- NULL;
    vector of all observations <- NULL;</pre>
    absolute filepath to export to <- '/home/parallels/R/coursera/statistical infe
rence/course project/';
    setNumberOfBatchesToObserve <- function(number of batches to set)</pre>
    {
        number of batches to observe <<- number of batches to set;
    }
    initializeBatchData <- function()</pre>
    {
        sample mean <<- NULL;</pre>
        sample_standard_deviation <<- NULL;</pre>
        matrix of observation batches <<- NULL;
        averages of batches <<- NULL;
        vector of all observations <- NULL;</pre>
    }
    getNumberOfObservationsPerBatch <- function()</pre>
    {
        observations per batch;
    }
    getNumberOfBatchesToObserve <- function()</pre>
    {
        number_of_batches_to_observe;
    }
    getTotalObservationsCount <- function()</pre>
        total_observations_count <- getNumberOfObservationsPerBatch() * getNumberO</pre>
fBatchesToObserve();
        total observations count;
    }
    getAllObservations <- function()</pre>
        vector of all observations
    }
```

```
generateBatchesOfObservations <- function()</pre>
        initializeBatchData();
        total number of observations <- number of batches to observe * observation
s_per_batch;
        vector of all observations <<- rexp(total number of observations, lambda);</pre>
        # We will make each row of the matrix be a batch of observations
        matrix of observation batches <<- matrix(data = vector of all observation
s, nrow = number of batches to observe,
                                                  ncol = observations per batch, by
row = TRUE);
        # Compute the averages of the batches
        averages_of_batches <<- apply(matrix_of_observation_batches, 1, mean);</pre>
        computeSampleMean();
        computeSampleStandardDeviation();
        invisible(matrix of observation batches);
    }
    # Output the theoretical sample average mean/variance and the experimental sam
ple average mean/variance
    produceStatisticsReport <- function()</pre>
        #statistics report file <- getAbsoluteFilePathForFile('statistics report.l
og');
        observations per batch <- getNumberOfObservationsPerBatch();
        number of batches to observe <- getNumberOfBatchesToObserve();</pre>
        lines to write <- c(
            paste0(as.integer(number_of_batches_to_observe), " samples of the aver
age of ", observations per batch ," exponentials with lambda=", lambda, " were tak
en:", collapse=""),
            # Compare Mean of Averages with Theoretical Value
            paste0('Population Average: ', getTheoreticalPopulationMean()),
            pasteO('Theoretical Mean of Averages of ', observations per batch ,' e
xponentials: ', getTheoreticalSampleMean(), collapse=""),
            paste0('Experimental Mean of Averages of ', observations per batch ,'
exponentials: ', getSampleMean(), collapse=""),
            # Compare Variance of Averages with Theoretical Value
            pasteO('Theoretical Variance of Averages of ', observations per batch
,' exponentials: ', getTheoreticalSampleVariance(), collapse=""),
            pasteO('Experimental Sample Variance of Averages of ', observations_pe
r batch ,' exponentials: ', getSampleVariance(), collapse=""),
            # Compare Standard Deviation of Averages with Theoretical Value
            pasteO('Theoretical Standard Devation of Averages of ', observations p
er batch ,' exponentials: ', getTheoreticalSampleStandardDeviation(), collaps
```

```
e=""),
            pasteO('Experimental Standard Devation of Averages of ', observation
s per batch ,' exponentials: ', getSampleStandardDeviation(), collapse="")
        );
        #writeLines(lines to write, statistics report file);
        print(lines to write);
    # Show a histogram of the experimental sample averages with the theoretical no
rmal distribution the sample averages
        should be distributed by overlayed as a curve
    produceSampleDensityHistogramAgainstNormalCurve <- function()</pre>
        #density comparison histogram filepath <- getAbsoluteFilePathForFile('samp
le density comparison histogram.jpg');
        #jpeg(density comparison histogram filepath);
        # Produce the histogram representing the sample data.
        x_axis_label <- paste0("Averages of ", observations_per_batch ," Exponenti</pre>
als", collapse="");
        main_label <- paste0("Sample Average Distribution Compared To\nNormal Dist</pre>
ribution for ", as.integer(number of batches to observe), " Sample Averages");
        # Display the samples as DENSITIES, NOT AS FREQUENCIES to allow for compar
ison to the normal density curve
        hist(x=averages of batches, breaks=100, xlab=x axis label, freq=FALSE, mai
n=main label, ylab="Distribution Density");
        # Produce the points which will comprise the expected normal distributon c
urve
        # Get a vector of probability steps, with one step for every sample averag
е
        probability steps for quantile calculation <- getProbabilitySteps(number o</pre>
f batches to observe);
        theoretical sample mean <- getTheoreticalSampleMean();</pre>
        theoretical sample standard deviation <- getTheoreticalSampleStandardDevia
tion();
        # Compute the theoretical quantile for every sample average
        normal quantiles <- qnorm(probability steps for quantile calculation, mea
n=theoretical sample mean, sd=theoretical sample standard deviation);
        # Compute the theoretical density at every theoretical quanitle
        normal densities <- dnorm(normal quantiles, mean=theoretical sample mean,
sd=theoretical sample standard deviation);
        # Plot the theoretical sample average density distribution, which is appro
ximated as a normal distribution curve
        points(x=normal quantiles, y=normal densities, type="l");
        #dev.off();
        #invisible(density comparison histogram filepath);
    }
    getProbabilitySteps <- function(number of steps)</pre>
```

```
probability steps for quantile calculation < seq((1/number of steps), 1,
(1/number of steps));
        probability steps for quantile calculation;
    }
    drawTheoreticalExponentialDistributionOverSampleDistribution <- function()</pre>
        #exponential_observation_distribution_plot_filepath <- getAbsoluteFilePath
ForFile('exponential observation distribution plot filepath.jpg');
        #ipeg(exponential observation distribution plot filepath);
        total number of observations <- getTotalObservationsCount();</pre>
        probability steps for quantile calculation <- getProbabilitySteps(total nu
mber of observations);
        main label <- paste0("Observation Value Distribution Compared To\nNormal D
istribution for ", as.integer(total number of observations), " Observations");
        hist(x=getAllObservations(), breaks=100, freg=FALSE, main=main label, xla
b="Observation value", ylab="Distribution Density");
        exp quantiles <- qexp(probability steps for quantile calculation, rate=lam
bda);
        exp densities <- dexp(exp quantiles, rate=lambda);</pre>
        points(x=exp quantiles, y=exp densities, type="l");
        #dev.off();
        #invisible(exponential observation distribution plot filepath);
    }
    getAveragesOfBatches <- function()</pre>
    {
        averages_of_batches;
    }
    # The sample mean is the mean of the batch averages
    computeSampleMean <- function()</pre>
    {
        sample mean <<- mean(averages of batches);</pre>
        sample mean;
    }
    getSampleMean <- function()</pre>
    {
        sample mean;
    }
    computeSampleStandardDeviation <- function()</pre>
        sample standard deviation <<- sd(averages of batches);</pre>
        sample standard deviation;
    getSampleStandardDeviation <- function()</pre>
```

```
sample standard deviation;
    # Var(sample mean) = standard deviation(sample mean)^2
    getSampleVariance <- function()</pre>
    {
        sample variance <- (sample standard deviation^2);</pre>
        sample variance;
    }
    getTheoreticalSampleMean <- function()</pre>
    {
        getTheoreticalPopulationMean();
    \# Var(sample mean) = Var(X) / n
    getTheoreticalSampleVariance <- function()</pre>
        theoretical variance of population <- getTheoreticalPopulationVariance();
        theoretical variance of sample mean <- theoretical variance of population
/ observations per batch;
        theoretical_variance_of_sample_mean;
    # standard deviation(sample mean) = sqrt(Var(sample mean))
    getTheoreticalSampleStandardDeviation <- function()</pre>
    {
        theoretical variance of sample mean <- getTheoreticalSampleVariance();
        theoretical standard deviation of sample mean = sqrt(theoretical varianc
e of sample mean);
        theoretical_standard_deviation_of_sample_mean;
    }
    # For exponential distribution, E[X] = 1 / lambda = mu
    getTheoreticalPopulationMean <- function()</pre>
    {
        theoretical_mean <- (1.0 / lambda);</pre>
        theoretical mean
    # For exponential distribution, Var(X) = 1 / lambda^2
    getTheoreticalPopulationVariance <- function()</pre>
    {
        theoretical variance <- (1.0 / (lambda^2));
        theoretical variance;
    # For exponential distribution, sd(X) = 1 / lambda
    getTheoreticalPopulationStandardDeviation <- function()</pre>
        theoretical standard deviation <- (1.0 / lambda);
        theoretical_standard_deviation
    # Functions for outputting the data and the density comparison histogram
```

```
getAbsoluteFilePathForFile <- function(file name)</pre>
    {
        absolute file path <- pasteO(absolute filepath to export to, as.integer(nu
mber_of_batches_to_observe), "_samples_", file_name, collapse = '');
        absolute file path;
    }
    list(generateBatchesOfObservations = generateBatchesOfObservations,
         produceStatisticsReport = produceStatisticsReport,
         produceSampleDensityHistogramAgainstNormalCurve = produceSampleDensityHis
togramAgainstNormalCurve,
         drawTheoreticalExponentialDistributionOverSampleDistribution = drawTheore
ticalExponentialDistributionOverSampleDistribution,
         getSampleMean = getSampleMean,
         getSampleVariance = getSampleVariance
     );
}
```