

Distributed Server Project

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Overview

The purpose of this project was to learn how to use CORBA to set up a distributed system across servers and measure the difference in transfer time from one server to another. The tools we used for this experiment were provided but included a Java implemented server and client. CORBA stands for “Common Object Request Broker Architecture”, CORBA provides “vendor-independent architecture and infrastructure that computer applications use to work together over networks”.^[1] The Java program we were provided requires us to first setup a ORBD Daemon which the Java program will then use to establish a connection between the server and the client.

Experiment Conditions

Clipper is a server located at Shippensburg University. Clipper is an Ubuntu 18.04.4 server with Linux Kernel 4.15.0-88-generic, it has an Intel Xeon 2680 v3 and has 20 CPU cores and 24 GB's of ram, which have been assigned to the virtual machine running Clipper. The ORBD daemon was started on CLipper and our team used ports 1054 and 1055. Java 8 was used to build and run the server and the client.

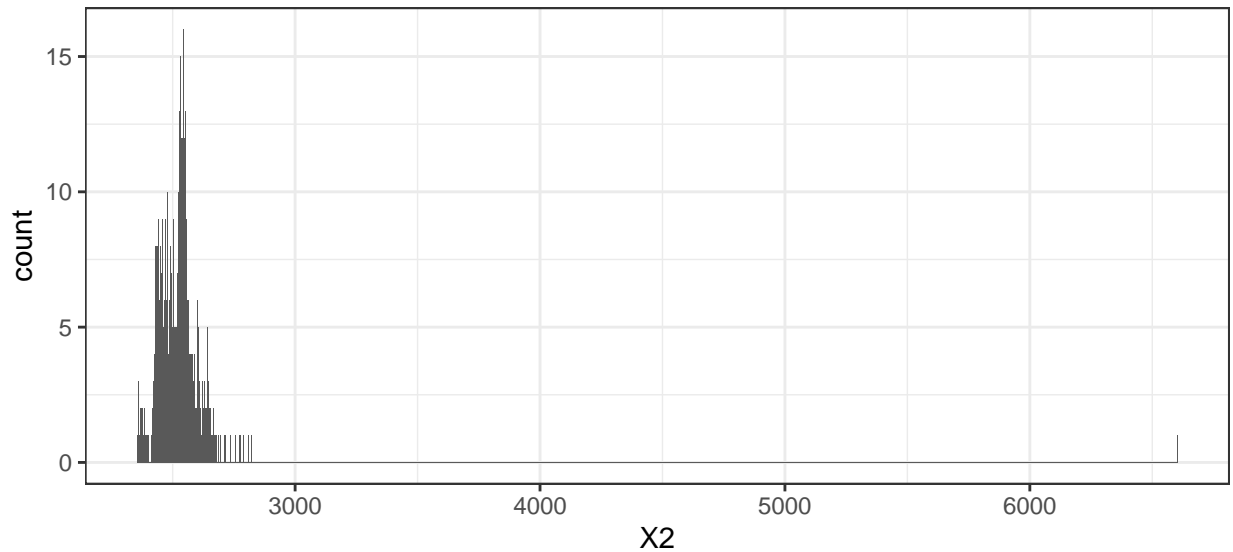
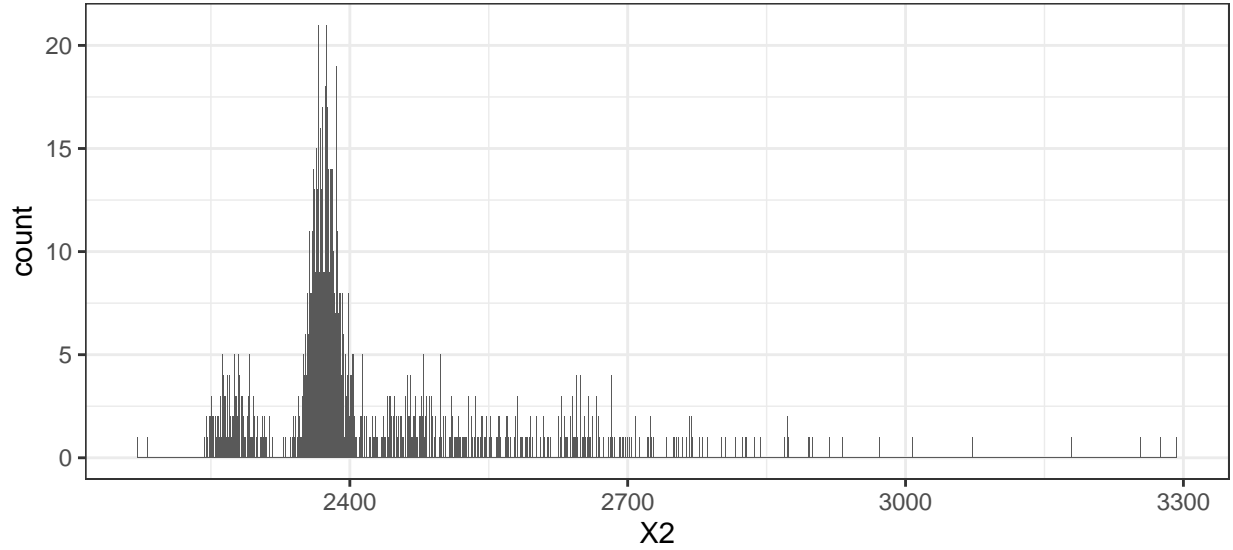
The German server is a remote virtual server hosted in Germany and uses an Intel Xeon e5-2660 v4 but is assigned one virtual core and 1 GB of ram. The server runs Ubuntu 18.04.5 with Linux Kernel 4.15.0-122-generic x86_64. The ORBD daemon was started on CLipper and our team used ports 1054 and 1055. Java 8 was used to build and run the server and the client.

The Spanish server is also a remote virtual server hosted in Spain instead of Germany. It uses an Intel Xeon E5-2683 v3 but is also assigned one virtual core and 1 GB of ram. The server runs Ubuntu 18.04.5 with Linux Kernel 4.15.0-122-generic x86_64. The ORBD daemon was started on CLipper and our team used ports 1054 and 1055. Java 8 was used to build and run the server and the client.

First Experiment

We initially set up our experiment using a bash script that would start and stop the ORBD Daemon and the Java Server, then run the client and measure the time the client took to complete the file transfer. In our experiment we used a text document full of random characters that was 1 Mb in size, we stored that on our server in the appropriate folder in the Java Server build path and then transferred that file to our client. We ran this bash script 1000 times on the German server and the Spanish server and stored the results in a csv.

First Results



Second Experiment

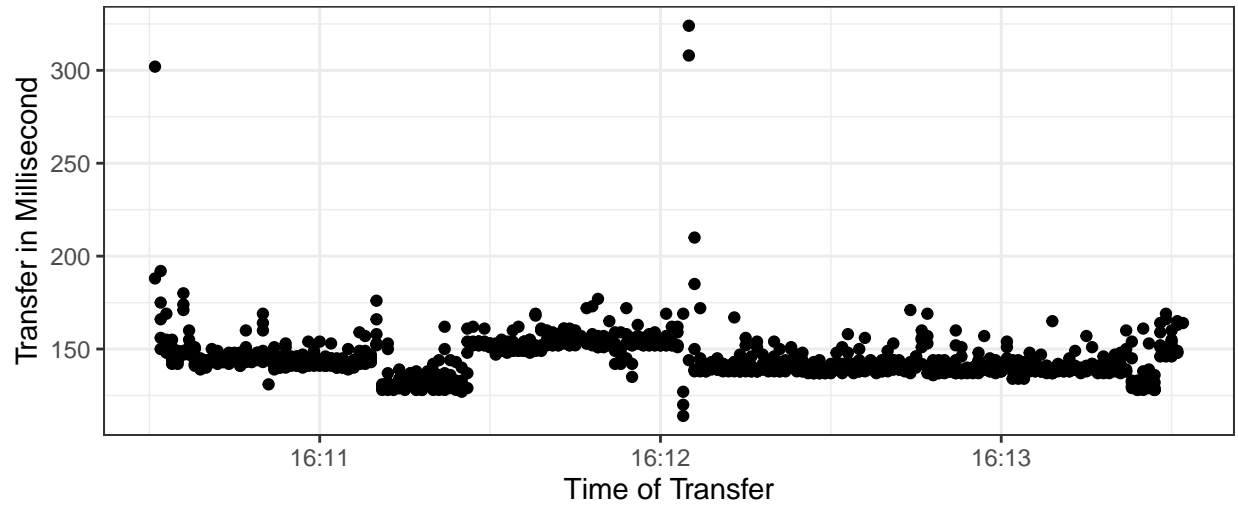
After some consideration we became concerned that the differences in compile time on the each server may have influenced our results and rendered them void.

We set up a second experiment where we transferred in the same manner a 2 MB text file, but instead of measuring total time to compile and transfer we only measured the time that it took to transfer using a Java method to measure the start and stop time of the transfer file method. We nested this method in a loop and ran it 1000 times on Clipper, the German and the Spanish server. We ran these tests 3 times each for each server and stored the results in corresponding csv files. The Clipper test was ran as a baseline to see best case scenarios for the transfer. The file size chose was intended to be large enough to be significant but not too large to be overwhelming for the remote servers. We ran the tests 3 times on each server to ensure we could have a large enough sample size to make significant analysis.

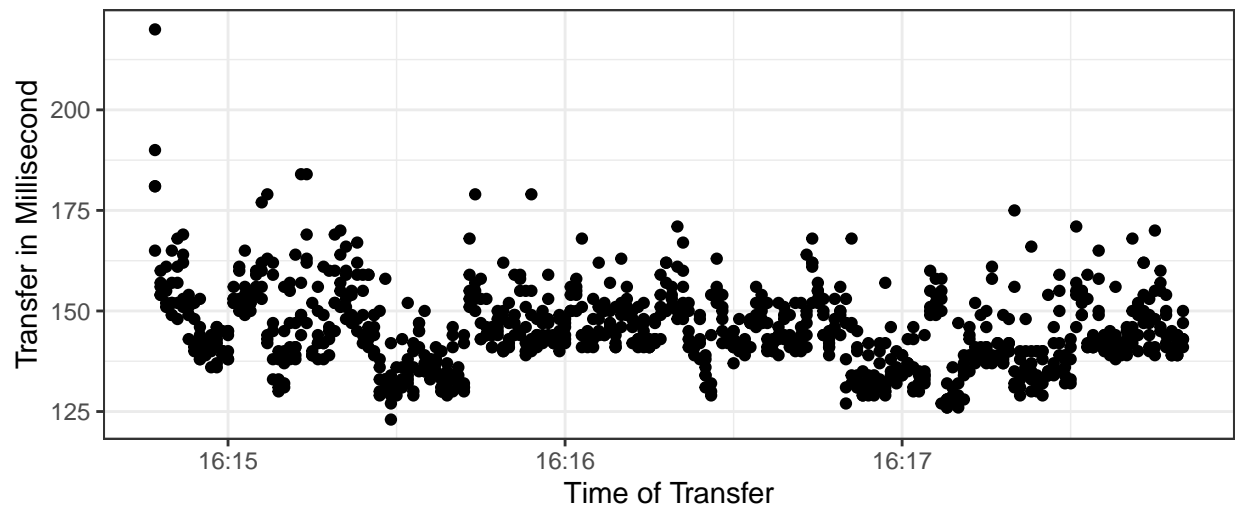
Clipper Server Results

We started analysis by looking at a point plot of the data over time.

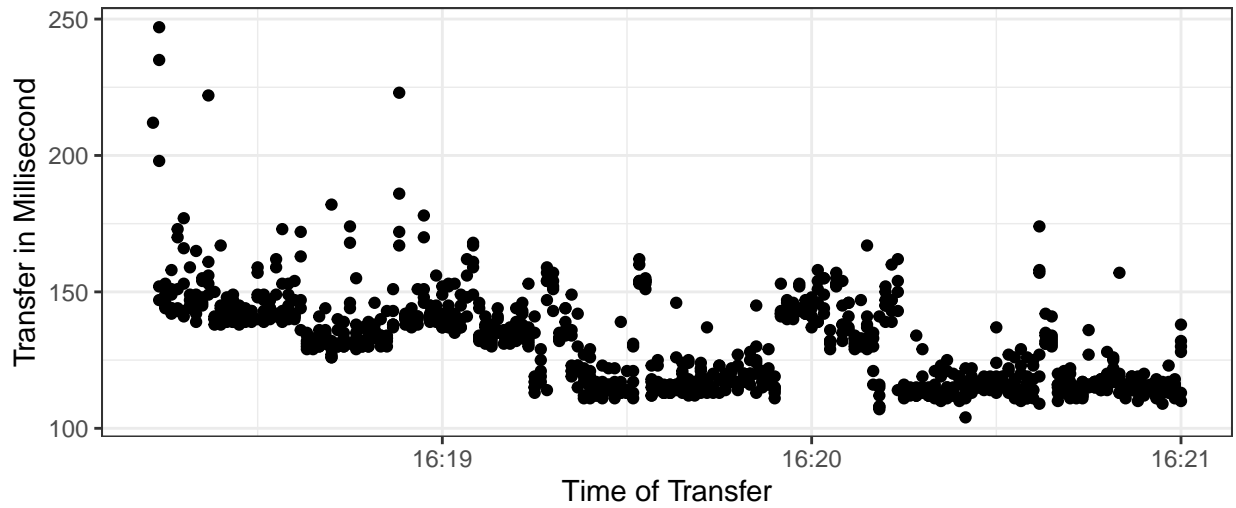
Clipper Results 1



Clipper Results 2

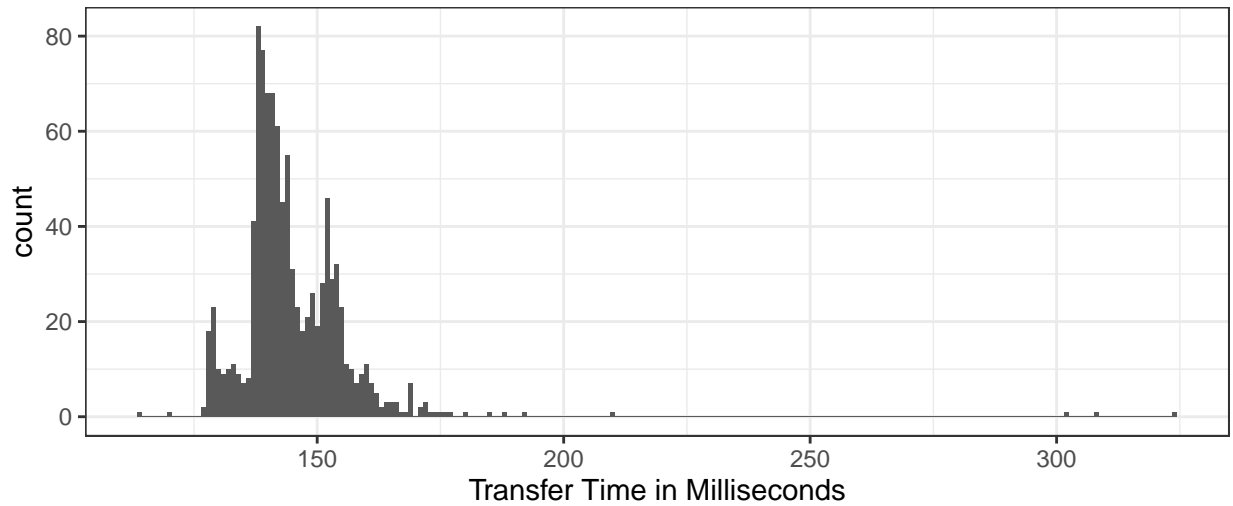


Clipper Results 3

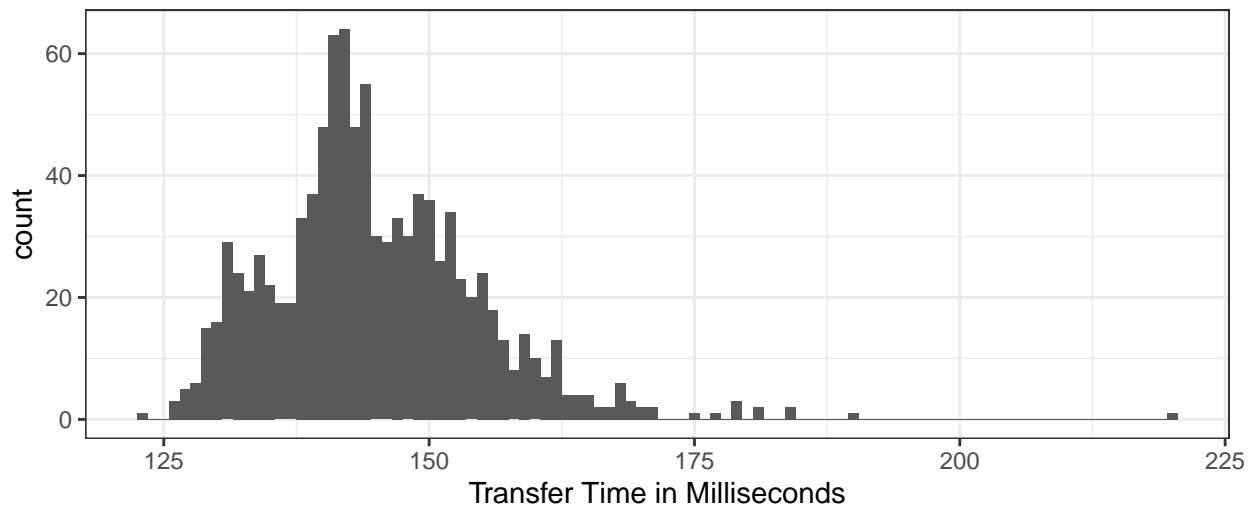


The time series data seems to show a mostly flat line across the first two tests with a few outliers except for the third plot which seems to stepped into two separate lines. To get a better look at these and to see if the data was normally distributed we made histograms using the data.

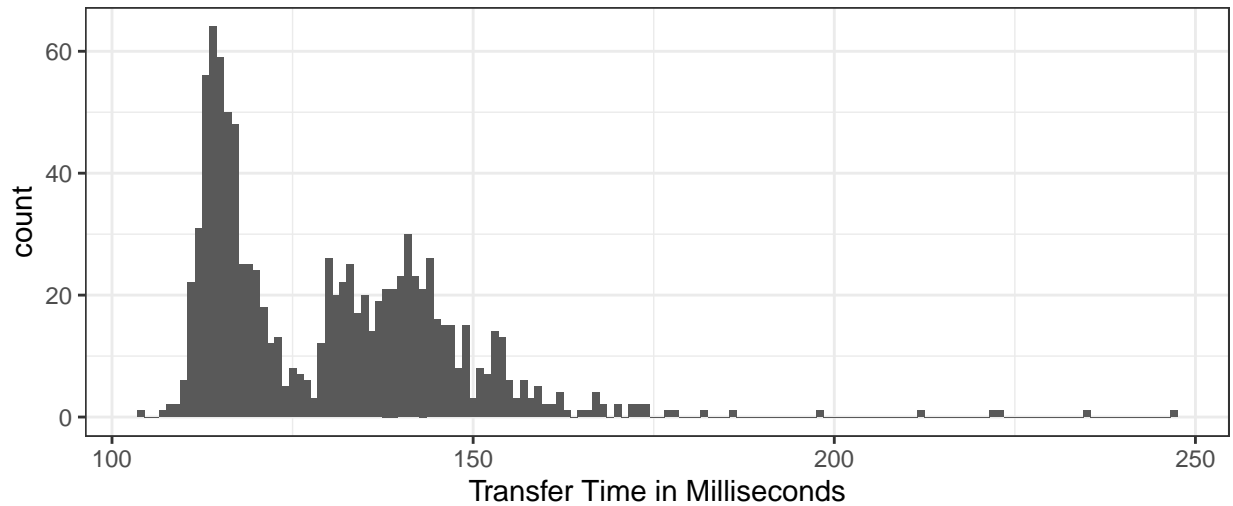
Clipper Transfer Time Histogram 1



Clipper Transfer Time Histogram 2

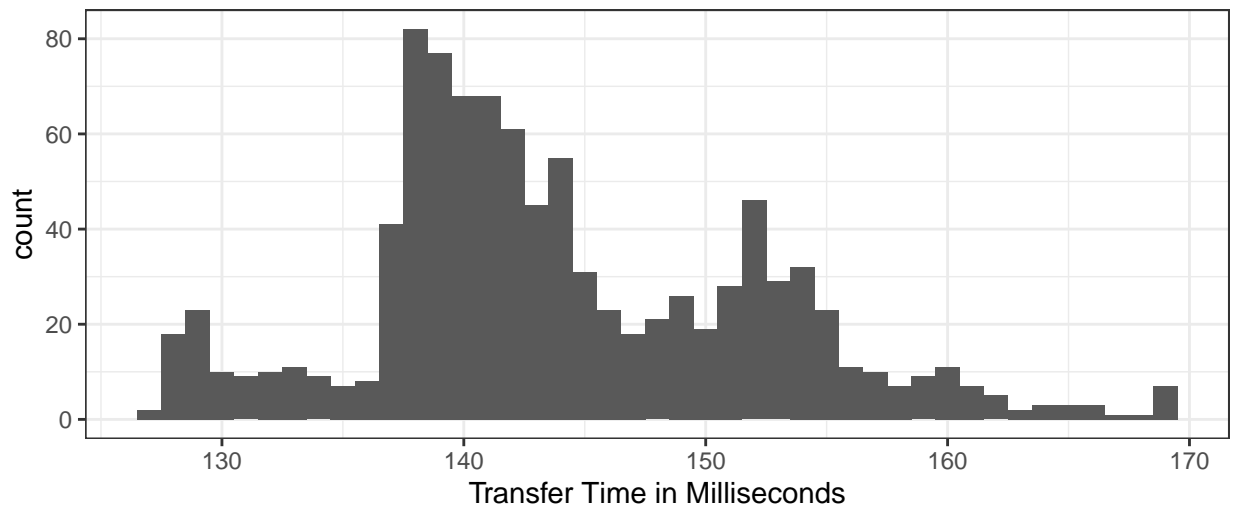


Clipper Transfer Time Histogram 3

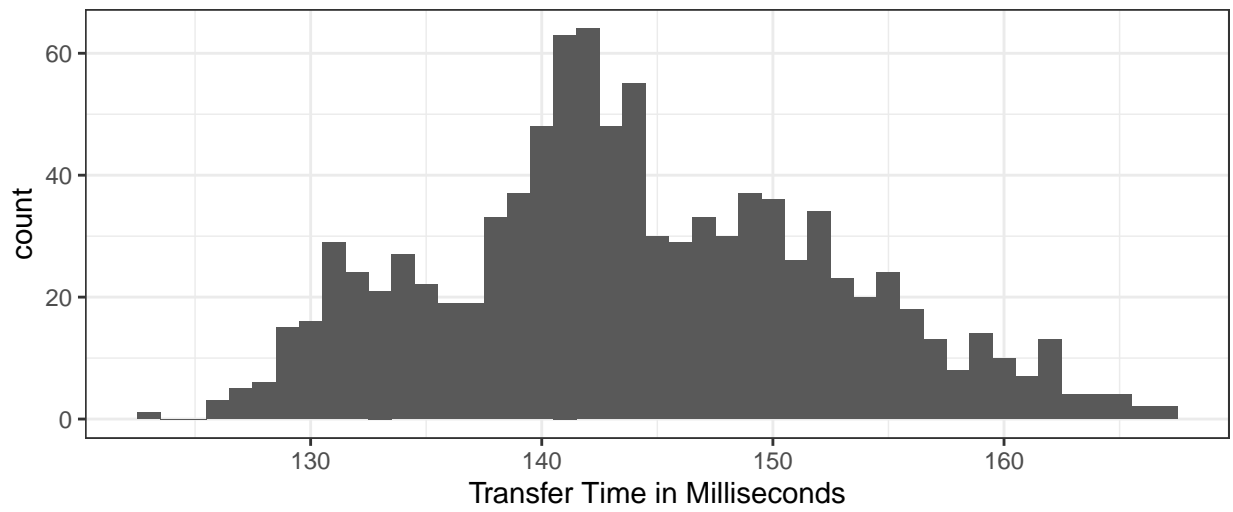


It became evident that we did have a few outliers and likely a bimodal distribution in the third set of data. To account for the outliers we used the boxplot function contained in the ggstatsplot package to remove outliers and clean up the data. The boxplot function can be used to identify outliers “which lie beyond the extremes of the whiskers”. [2] After removing the outliers, the following were our resulting histograms.

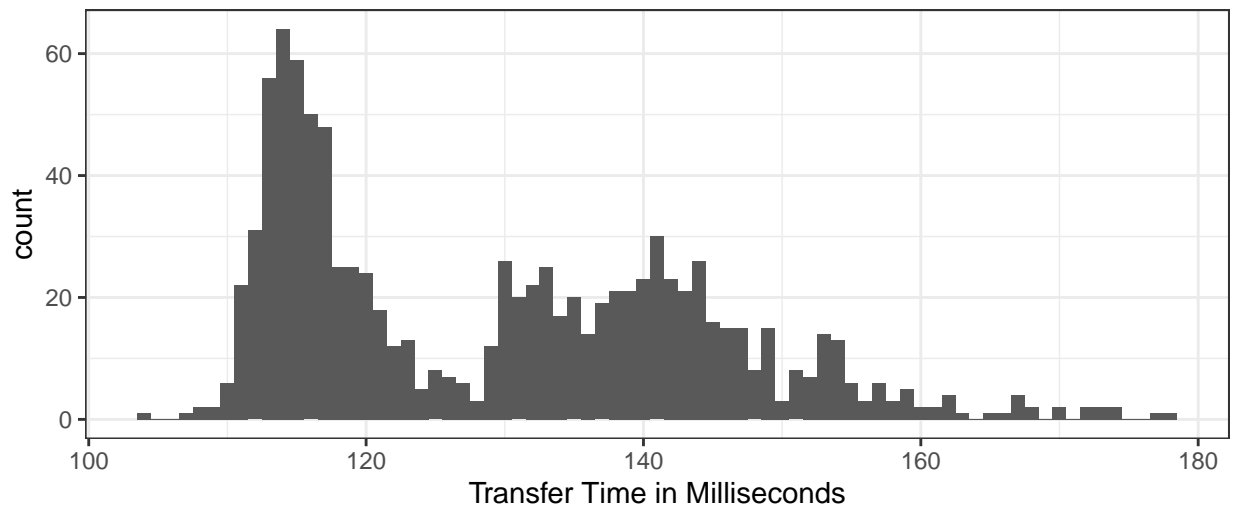
Clipper Transfer Time Histogram 1



Clipper Transfer Time Histogram 2

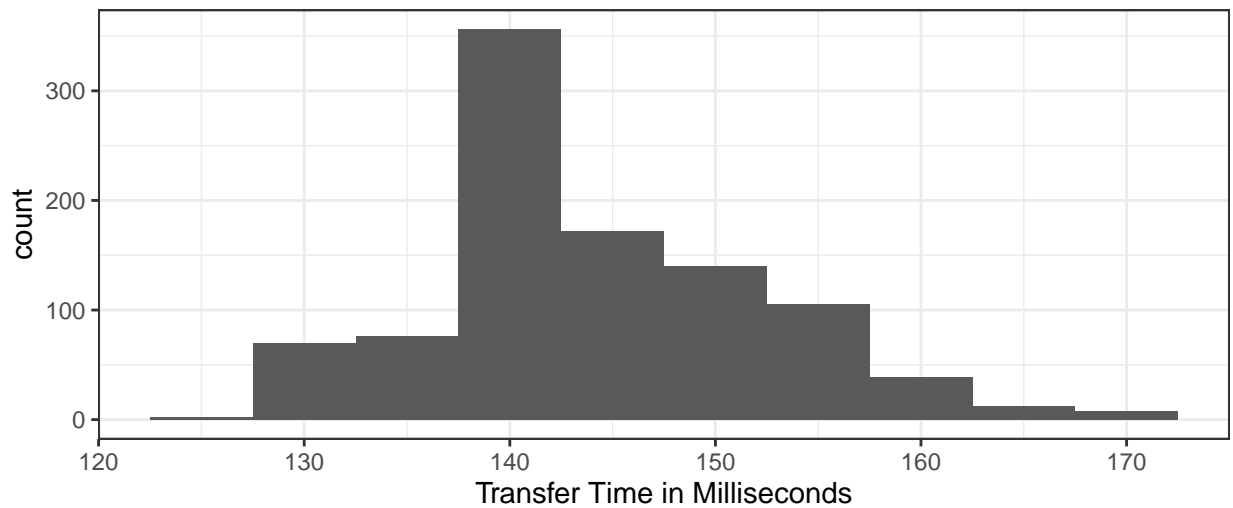


Clipper Transfer Time Histogram 3

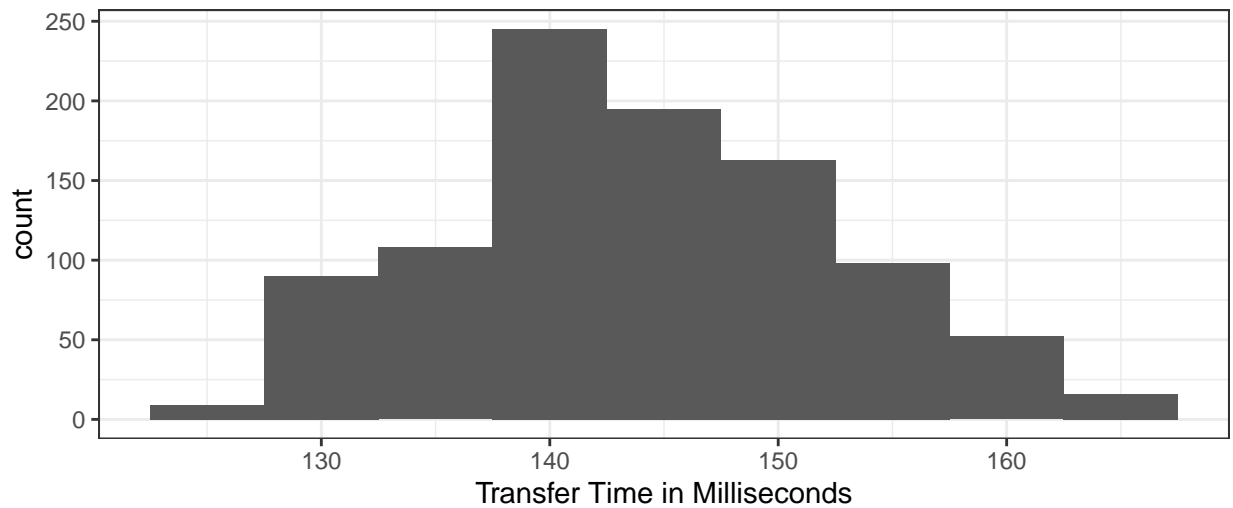


As can be seen in the resulting histograms the second test on Clipper seems to approach a normal distribution. The first and third test both seem to be bimodal at first. Upon further examination we found that if we changed the bin width for the histograms we can see that with an adjusted bin width of 5, the first test on Clipper also seems to be normally distributed. The third test is certainly bimodal.

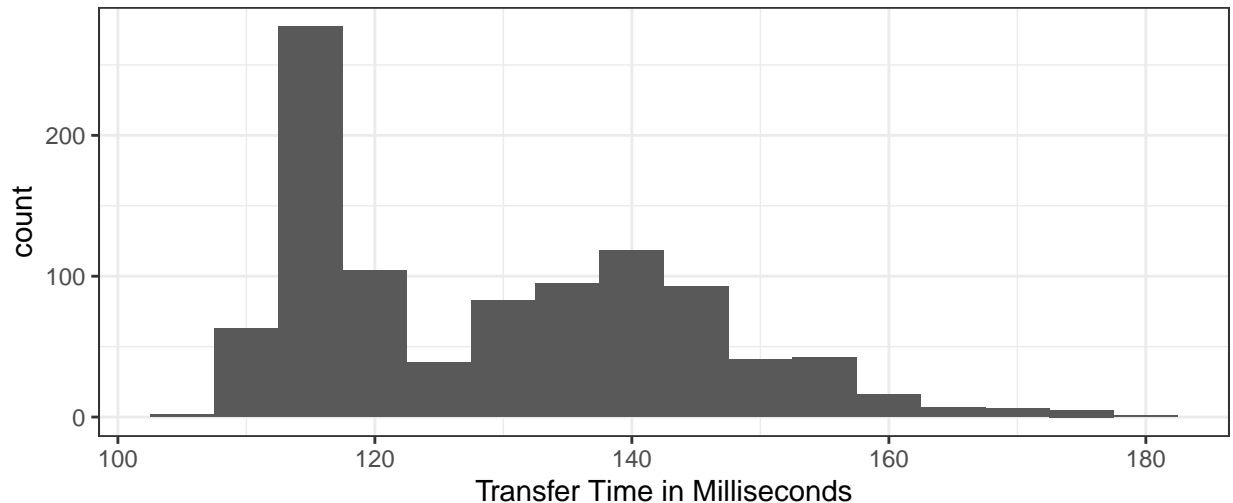
Clipper Transfer Time Histogram 1



Clipper Transfer Time Histogram 2



Clipper Transfer Time Histogram 3



After we removed the outliers and confirmed that the first two data sets were normally distributed we needed ensure that the first and second test on clipper accurately represent the same underlying distribution. We chose a t-test to compare the first and the second dataset, because a t-test can be used to determine if there is a significant difference between the means of two groups. First we'll assume a null hypothesis which is that Test 1 = Test 2 , Test 2 = Test 3, and Test 1 = Test 3, that is the data from three separate tests are = and there is no statistical difference between them. Our Alternate Hypothesis will be that Test 1 != Test 2 , Test 2 != Test 3, and Test 1 != Test 3, or that is there is a statistically significant difference between the three sets of data.

```
##
## Two Sample t-test
##
## data: javaResults01$X2 and javaResults02$X2
## t = -0.24977, df = 1954, p-value = 0.8028
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.8342817 0.6457822
```



```

## sample estimates:
## mean of x mean of y
## 143.9969 144.0912

##
## Two Sample t-test
##
## data: javaResults02$X2 and javaResults03$X2
## t = 26.673, df = 1966, p-value < 2.2e-16
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 13.56879 15.72246
## sample estimates:
## mean of x mean of y
## 144.0912 129.4456

##
## Two Sample t-test
##
## data: javaResults01$X2 and javaResults03$X2
## t = 26.8, df = 1970, p-value < 2.2e-16
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 13.48655 15.61620
## sample estimates:
## mean of x mean of y
## 143.9969 129.4456

```

As can be seen in the results the mean of the first test was 144.973 and the mean of the second test was 144.859. The p-value was .8258 indicating the underlying distributions are similar and we can not reject a null hypothesis in this instance. Our second t-test has an extremely low p-value (2.2e-16) and shows a sample mean of 144.859 for Test 1 and 130.115 for Test 2. This low p-value indicates that we should reject the null hypothesis. There is a statistically significant difference between our results in Test 2 and our results in Test 3. Likewise when comparing Test 1 and Test 3 we get similar results with a p-value of 2.2e-16. With all that in mind we can conclude that test's 1 and 2 belong to the same underlying distribution and will be ok to use when comparing the performance of other servers. As test 3 was bimodal and did not fit with Test 1 and Test 2 we will not use Test 3 as a baseline or include it in further analysis.

```

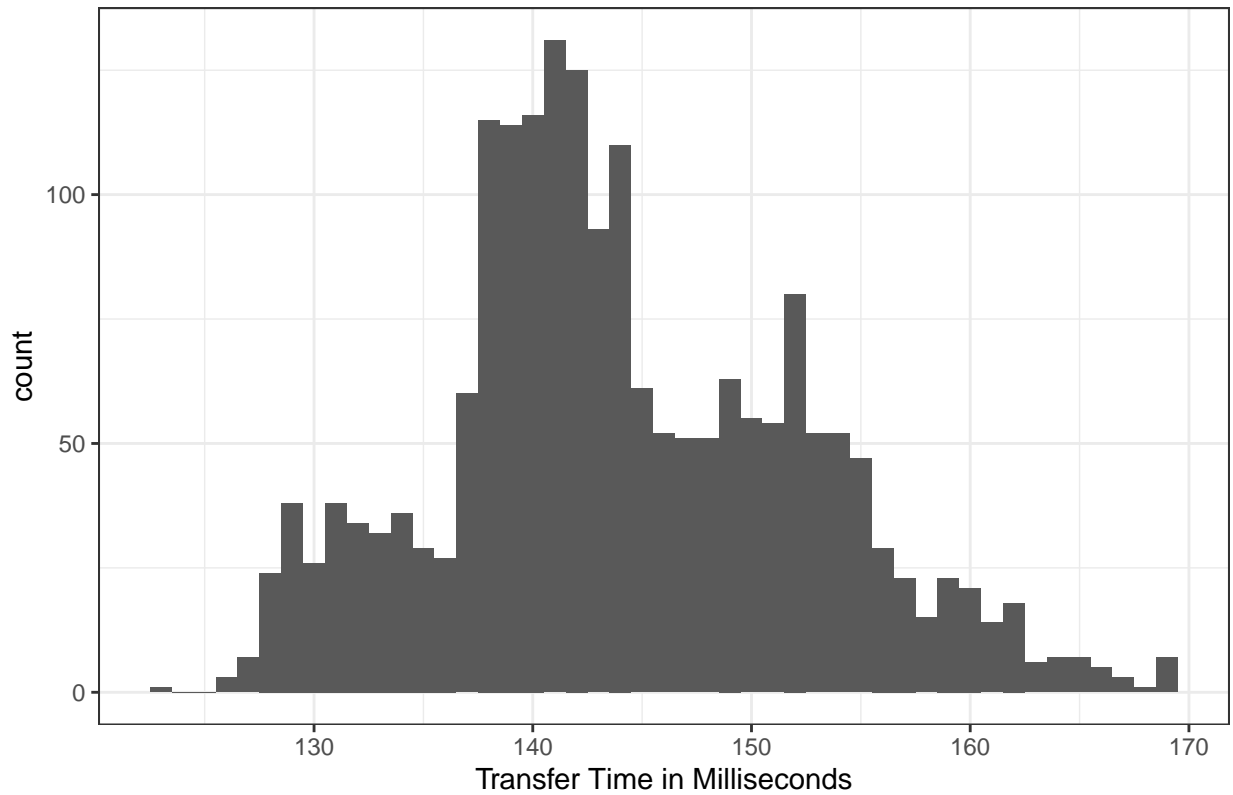
## 0% 25% 50% 75% 100%
## 123 139 143 150 169

## Max = 169
## Min = 123
## Mean = 144.04
## SD = 8.34

```

So our final results for best case scenario transfer time is a mean time of 144.92, with a max time of 324 milliseconds and a min time 144 milliseconds. Our standard deviation is 11.58 and 50% of all our values fall between 139 and 151 milliseconds.

Clipper Final Transfer Times

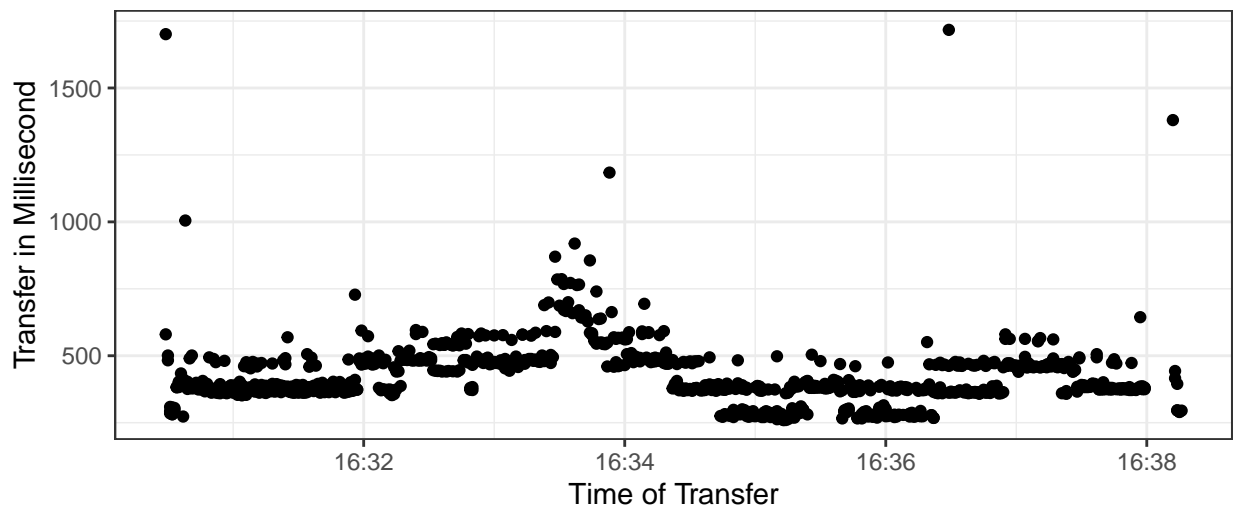


As can be seen the combined dataset approaches a normal distribution.

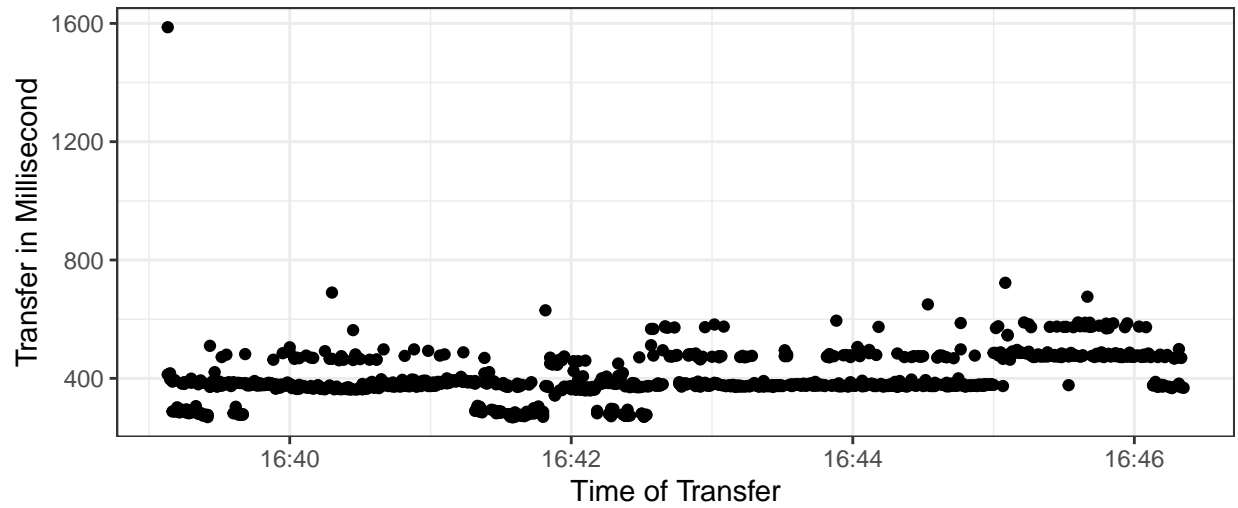
German Server Results

Having established a baseline we ran the same tests on the German server

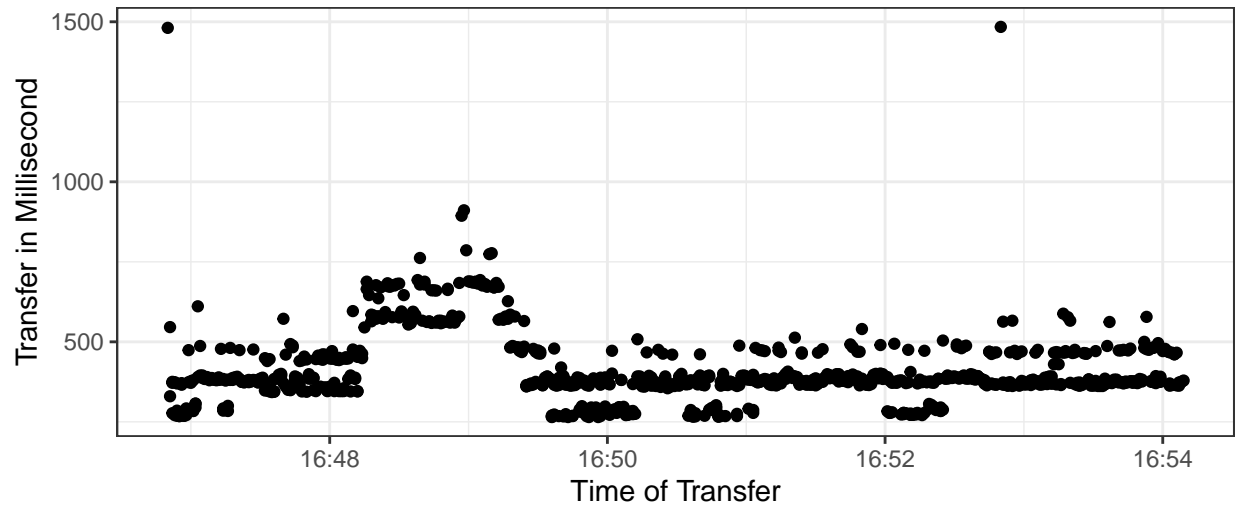
Germany Results 1



Germany Results 2

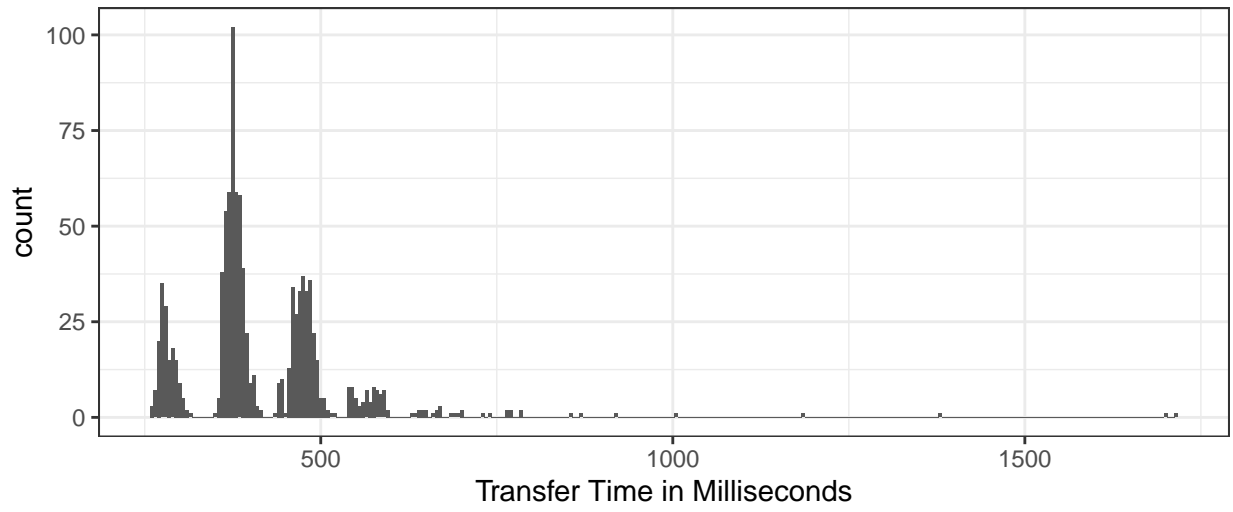


Germany Results 3

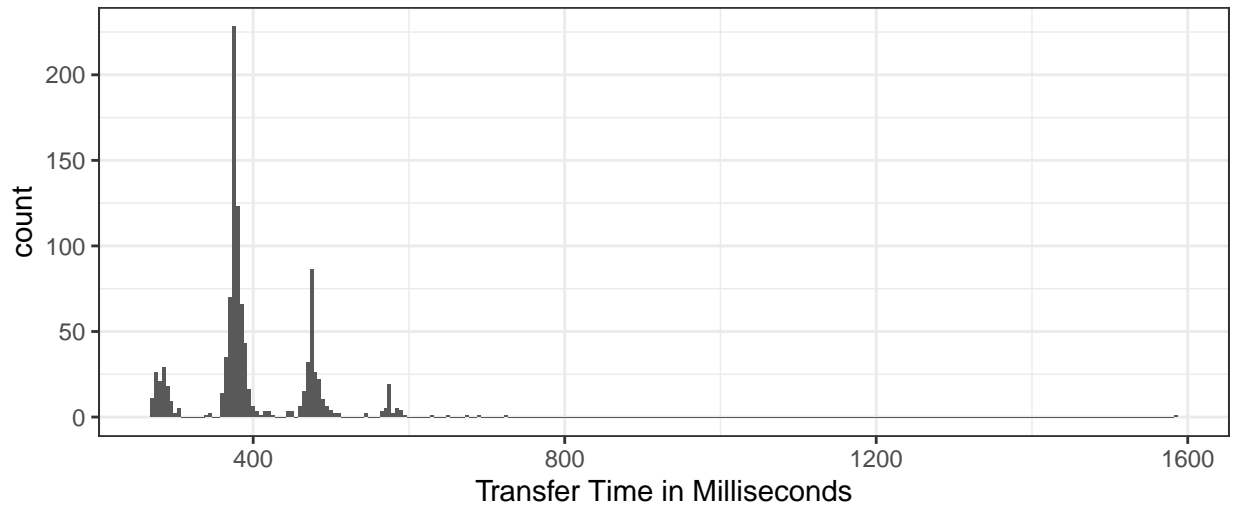


As can be seen the time series data looks similar to our previous test. There seems to be a spike in activity that happens around a minute in for all three tests but all three seem to follow a similar pattern.

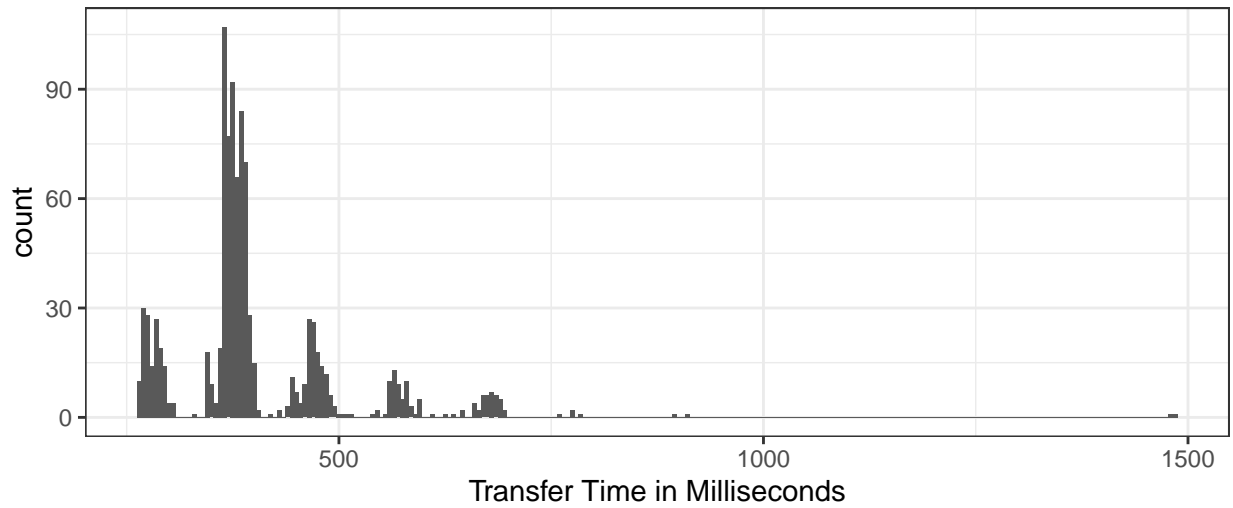
German Transfer Time Histogram 1



German Transfer Time Histogram 2

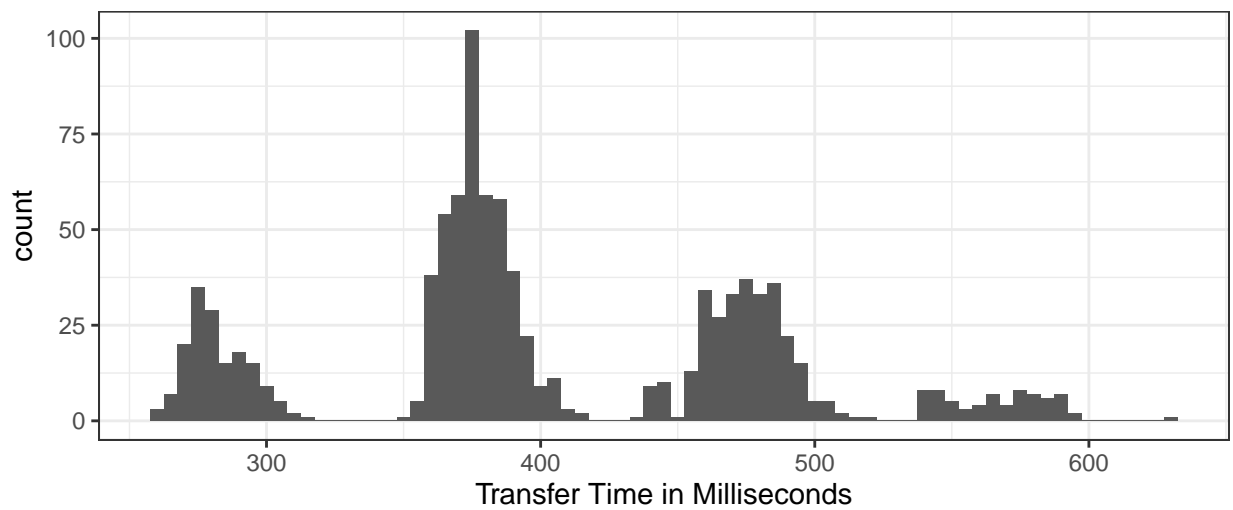


German Transfer Time Histogram 3

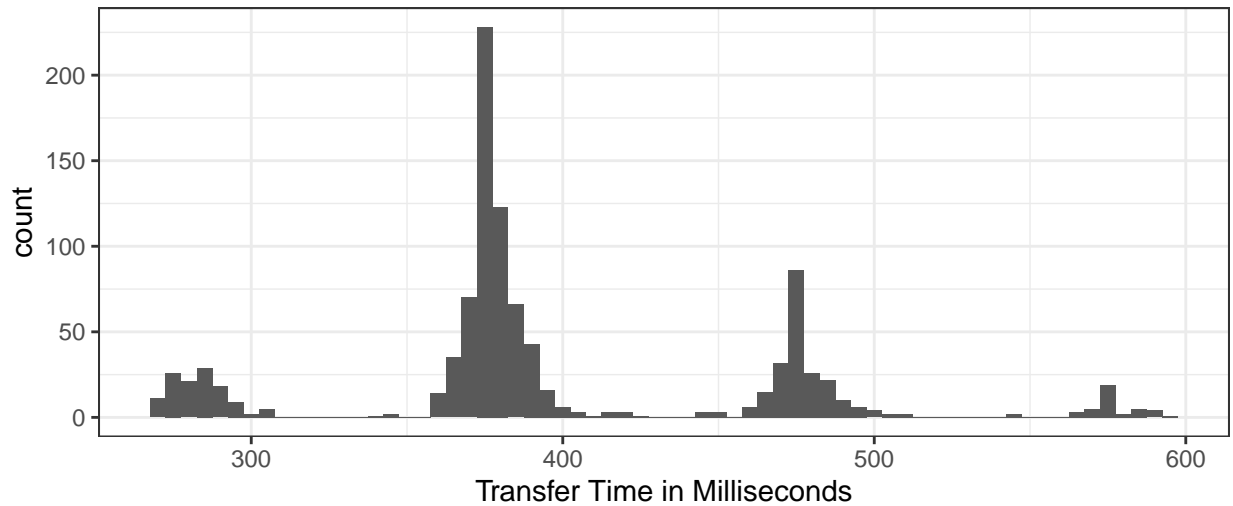


When making our histograms there is an very apparent difference from our first results on Clipper. All three tests run on the German server seem to group around similar transfer time intervals.

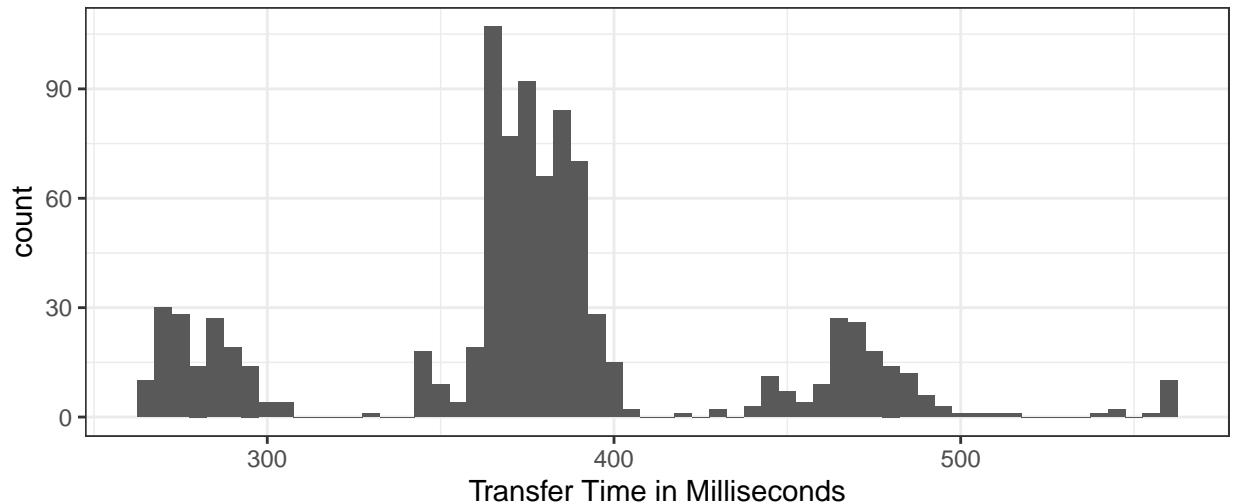
Germany Transfer Time Histogram 1



Germany Transfer Time Histogram 2



Germany Transfer Time Histogram 3



After removing outliers as we did before and grouping these into larger bins (width of 5) the grouping is even more apparent. There seems to be groupings around 275 milliseconds, 375 milliseconds, and 475 milliseconds on tests 1 and 2. While this does not seem to be exactly normally distributed, it is a consistent distribution. Test 3 seems to be normally distributed after removing the outliers. Running the t-tests as before seems to indicate all these tests belong to the same underlying distribution.

```
##
## Two Sample t-test
##
## data: javaResultsGermany01$X2 and javaResultsGermany02$X2
## t = 2.2767, df = 1968, p-value = 0.02291
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
##  1.047186 14.063054
## sample estimates:
## mean of x mean of y
## 403.5994 396.0443
```

```
##
## Two Sample t-test
##
## data: javaResultsGermany02$X2 and javaResultsGermany03$X2
## t = 6.0896, df = 1895, p-value = 1.367e-09
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 12.15257 23.69897
## sample estimates:
## mean of x mean of y
## 396.0443 378.1185

##
## Two Sample t-test
##
## data: javaResultsGermany01$X2 and javaResultsGermany03$X2
## t = 7.8206, df = 1877, p-value = 8.698e-15
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 19.09087 31.87091
## sample estimates:
## mean of x mean of y
## 403.5994 378.1185
```

As we can see from our two sample t-test's the means for Test 1 = 403.6, Test 2 = 396.04, and Test 3 = 378.12 and all three tests fall below a significant level of .05. This low significance level indicates that all three are not part of the same underlying distribution.

```
## 0% 25% 50% 75% 100%
## 260 367 379 454 628

## Max = 628

## Min = 260

## Mean = 392.98

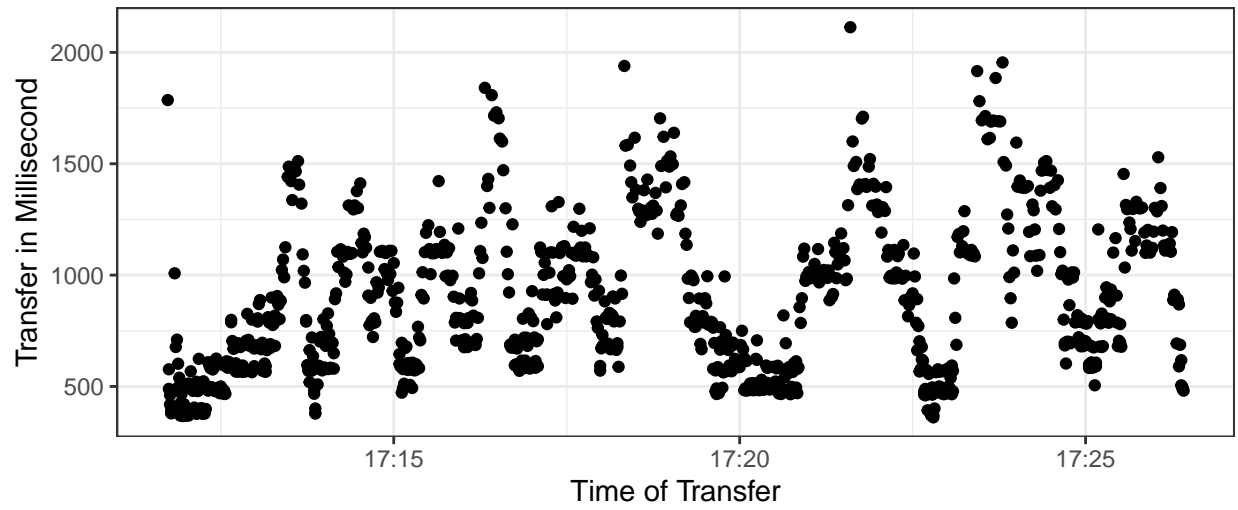
## SD = 70.35
```

Once combined we can see 50% of all transfer time values fall between 367 and 454 milliseconds and our mean transfer time is 392.98 milliseconds. Our fastest time was 260 milliseconds and max transfer time was 628, the standard deviation was 70.35.

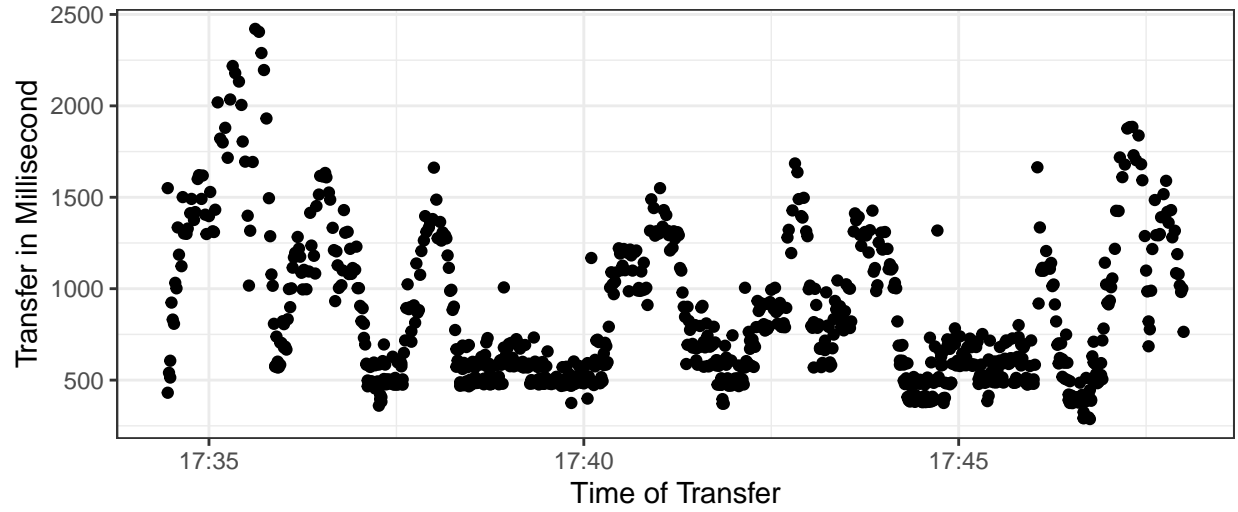
Spanish Server Results

After we ran our tests on the German server we ran our last tests on the server in Spain. As before we plotted the time series data and also plotted the histograms.

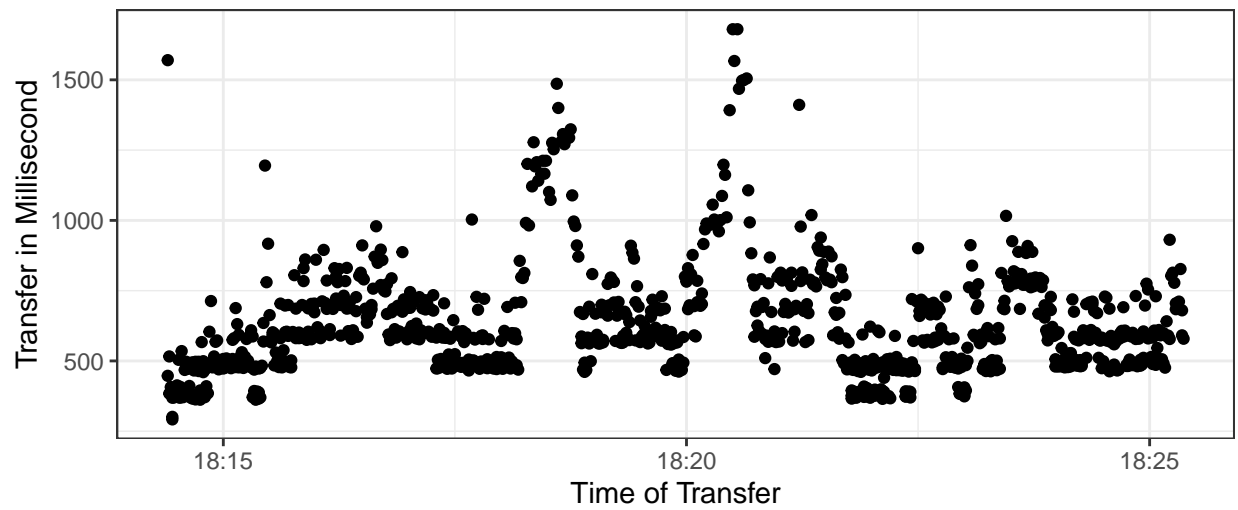
Spain Results 1



Spain Results 2

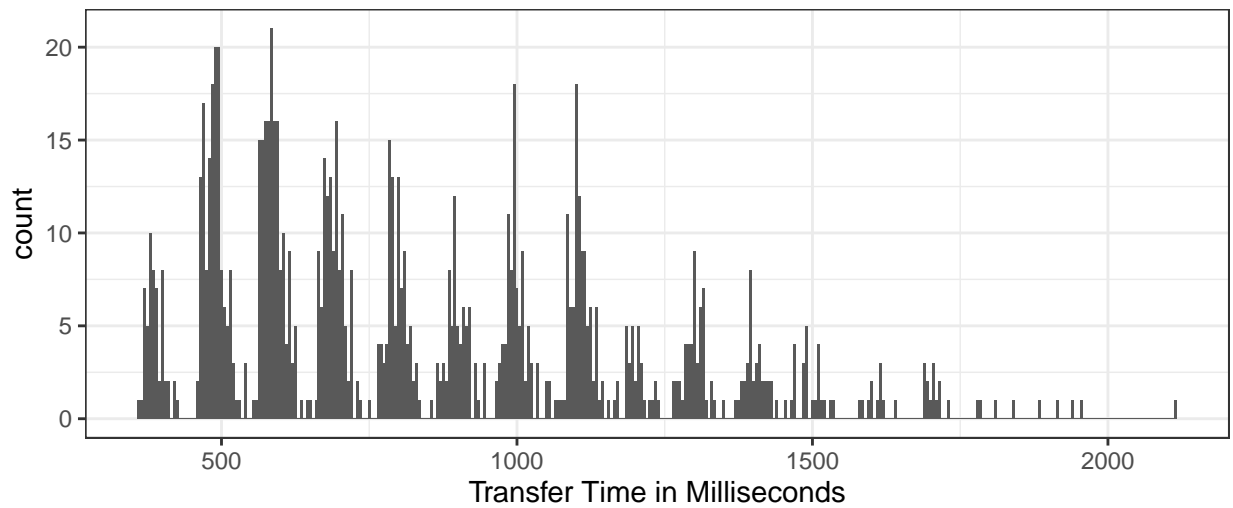


Spain Results 3

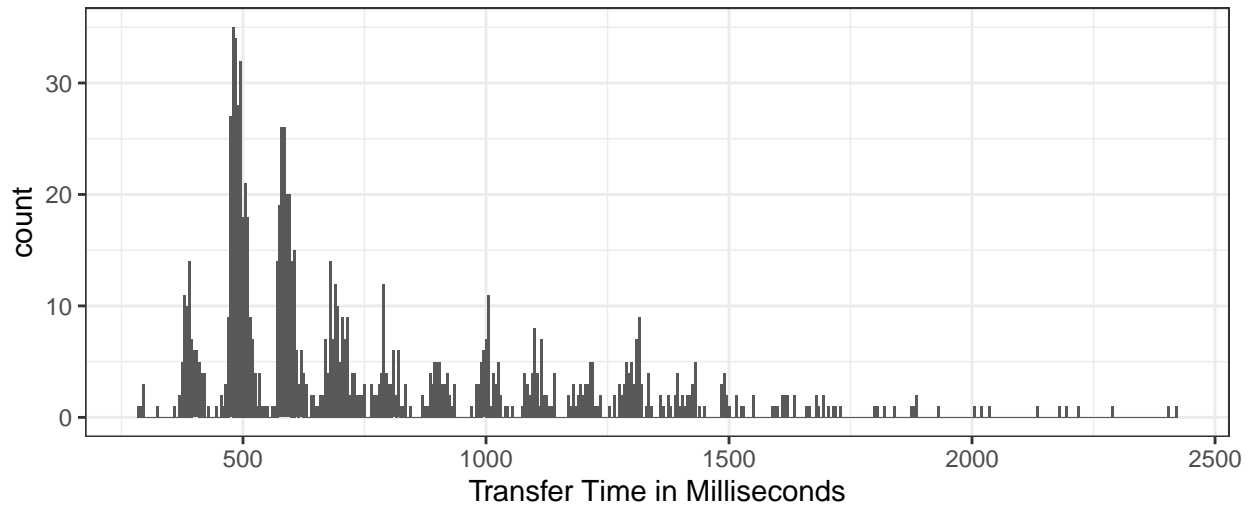


The time series data for the Spain server seems to show quite a bit more variation than either Clipper or the German server. Data does not seem to group appropriately around a time, the data seems to be distributed in a non-linear pattern.

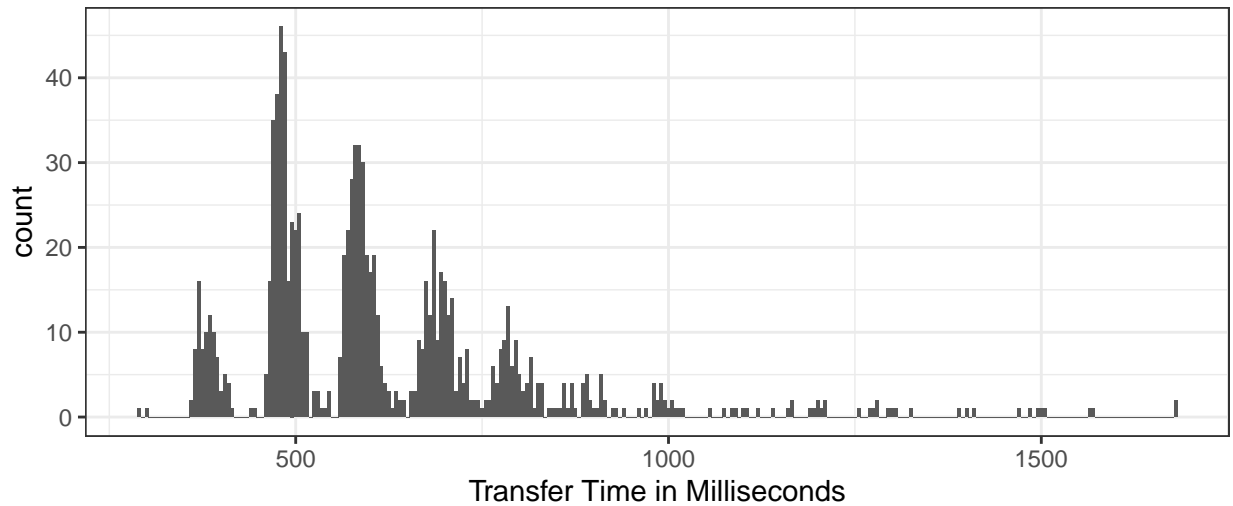
Spain Transfer Time Histogram 1



Spain Transfer Time Histogram 2

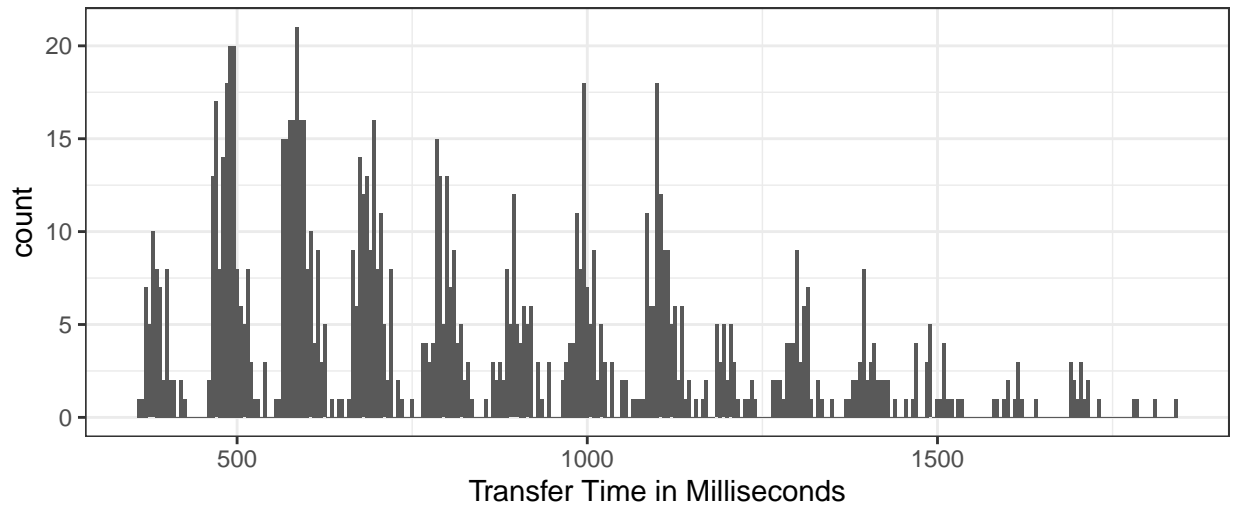


Spain Transfer Time Histogram 3

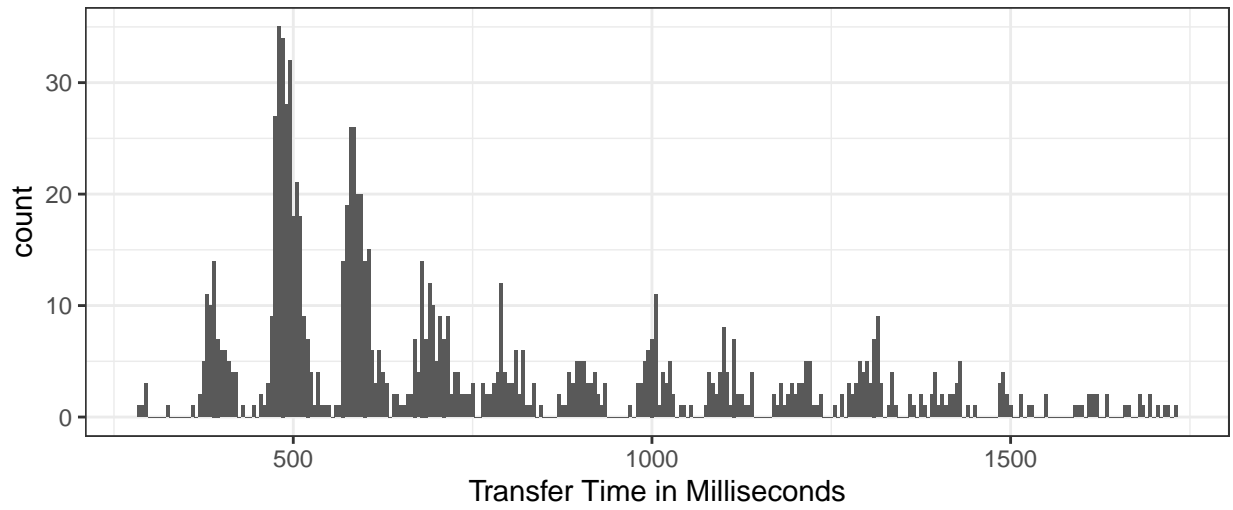


After looking at the histograms it would appear that the data is grouping together in a similar way to the German server, but the range seems like it may be greater. To check this we'll first remove the outliers as we have for all data sets so far.

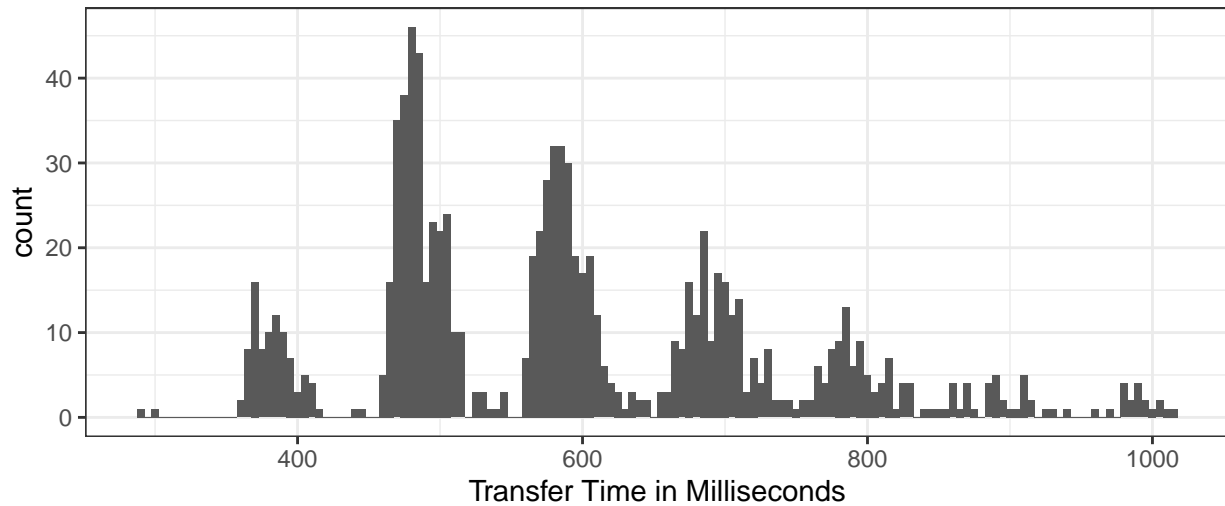
Spain Transfer Time Histogram 1



Spain Transfer Time Histogram 2



Spain Transfer Time Histogram 3



After removing outliers and regraphing out histogram it does appear that our overall x-axis has a greater range than what we saw with the German data.

```
range(javaResultsSpain01$X2)
```

```
## [1] 361 2113
```

```
range(javaResultsSpain02$X2)
```

```
## [1] 287 2421
```

```
range(javaResultsSpain03$X2)
```

```
## [1] 292 1680
```

```
range(javaResultsGermany01$X2)
```

```
## [1] 260 628
```

```
range(javaResultsGermany02$X2)
```

```
## [1] 268 595
```

```
range(javaResultsGermany03$X2)
```

```
## [1] 265 562
```

After looking it is very obvious, the Spanish server results seem to have a wider range of transfer times than the German server results

```
##
```

```
## Two Sample t-test
```

```
##
```

```
## data: javaResultsSpain01$X2 and javaResultsSpain02$X2
```

```
## t = 4.1195, df = 1998, p-value = 3.952e-05
```

```
## alternative hypothesis: true difference in means is not equal to 0
```

```
## 95 percent confidence interval:
```

```
## 34.28086 96.57914
```

```
## sample estimates:
```

```
## mean of x mean of y
```

```
## 844.347 778.917
```

```
##
## Two Sample t-test
##
## data: javaResultsSpain02$X2 and javaResultsSpain03$X2
## t = 11.833, df = 1998, p-value < 2.2e-16
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 131.1529 183.2651
## sample estimates:
## mean of x mean of y
## 778.917 621.708

##
## Two Sample t-test
##
## data: javaResultsSpain01$X2 and javaResultsSpain03$X2
## t = 17.998, df = 1998, p-value < 2.2e-16
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 198.3789 246.8991
## sample estimates:
## mean of x mean of y
## 844.347 621.708
```

As with the German results all three t-tests indicate that we will reject the null hypothesis, these tests do not belong to the same underlying distribution.

```
##      0%      25%      50%      75%     100%
## 287.00 499.00 622.50 902.25 2421.00

## Max = 2421
## Min = 287
## Mean = 748.32
## Range =
## [1] 287 2421
```

Looking at all results across all Spain data sets they seem to differ quite a bit from the German server results. The mean time is 748.32 milliseconds between all three sets of data, which is significantly more than the German server which was approximately 392 milliseconds. The max transfer of 2421 milliseconds is the highest transfer time we've encountered for any of our tests. Our range also shows quite a bit of variance from a min transfer time of 287 to 2421 milliseconds.

When we plot the final data we can see the data does not seem to approach a normal distribution but is very right skewed. This could be for a number of reasons (bottlenecking of either the CPU or network at different times of our tests), but ultimately this data seems to all fall within the same distribution.

Conclusions

With our data finally cleaned, examined and grouped we can start to look at the initial question of this project, is there a difference between transfer times between the German and Spanish server.

To test this we compared the German server tests with the Spanish server tests using t-tests in a piecewise manner.

```
##
```

```

## Two Sample t-test
##
## data: javaResultsSpain01$X2 and javaResultsGermany01$X2
## t = 39.663, df = 1974, p-value < 2.2e-16
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 418.9545 462.5407
## sample estimates:
## mean of x mean of y
## 844.3470 403.5994

##
## Two Sample t-test
##
## data: javaResultsSpain01$X2 and javaResultsGermany02$X2
## t = 40.982, df = 1992, p-value < 2.2e-16
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 426.8499 469.7556
## sample estimates:
## mean of x mean of y
## 844.3470 396.0443

##
## Two Sample t-test
##
## data: javaResultsSpain01$X2 and javaResultsGermany03$X2
## t = 40.851, df = 1901, p-value < 2.2e-16
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 443.8454 488.6116
## sample estimates:
## mean of x mean of y
## 844.3470 378.1185

##
## Two Sample t-test
##
## data: javaResultsSpain02$X2 and javaResultsGermany01$X2
## t = 30.897, df = 1974, p-value < 2.2e-16
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 351.4947 399.1405
## sample estimates:
## mean of x mean of y
## 778.9170 403.5994

##
## Two Sample t-test
##
## data: javaResultsSpain02$X2 and javaResultsGermany02$X2
## t = 31.984, df = 1992, p-value < 2.2e-16
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 359.3964 406.3490
## sample estimates:

```

```

## mean of x mean of y
## 778.9170 396.0443

##
## Two Sample t-test
##
## data: javaResultsSpain03$X2 and javaResultsGermany03$X2
## t = 35.775, df = 1901, p-value < 2.2e-16
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 230.2358 256.9432
## sample estimates:
## mean of x mean of y
## 621.7080 378.1185

##
## Two Sample t-test
##
## data: javaResultsSpain03$X2 and javaResultsGermany01$X2
## t = 32.204, df = 1974, p-value < 2.2e-16
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 204.8263 231.3909
## sample estimates:
## mean of x mean of y
## 621.7080 403.5994

##
## Two Sample t-test
##
## data: javaResultsSpain03$X2 and javaResultsGermany02$X2
## t = 34.235, df = 1992, p-value < 2.2e-16
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 212.7364 238.5911
## sample estimates:
## mean of x mean of y
## 621.7080 396.0443

##
## Two Sample t-test
##
## data: javaResultsSpain03$X2 and javaResultsGermany03$X2
## t = 35.775, df = 1901, p-value < 2.2e-16
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 230.2358 256.9432
## sample estimates:
## mean of x mean of y
## 621.7080 378.1185

```

As we expected based on our examination of means and range of the data before there is a significant statistical difference between all sets of data.

With this in mind we can see that the Spanish server results as a whole were much slower and had a greater range than the results on the German server. Mean transfer time on the Spanish server was 748, 356 milliseconds slower than the mean transfer time of 392 milliseconds on the German server. Since all tests took

part in similar times of day, on similar servers we can assume those were not factors in the differences of our times. We can thus assume the differences between the transfer times may be because of differences in network infrastructure from country to country or the path our server chose to route the transfer. Geographically Spain is closer to Shippensburg so we should see faster transfer times from that server. The tests we ran are likely inadequate in measuring that true difference though. In the future it could be beneficial to vary the test types to get a wider range of data to see if that difference would hold up in multiple contexts.

[1] “CORBA FAQ.” [Online]. Available: <https://corba.org/faq.htm>.

[2] “graphics.” [Online]. Available: <https://www.rdocumentation.org/packages/graphics/versions/3.6.2/topics/boxplot>.

[3] https://github.com/mjangle/lsa_project_3