

DEVELOPING AN EXCLUSION BOUNDARY (EB) FOR THE DASHER-XLS

STAT 938

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Outline

1. Introduction
2. Description Problem and Data
3. Methodology
4. Results
5. Conclusions

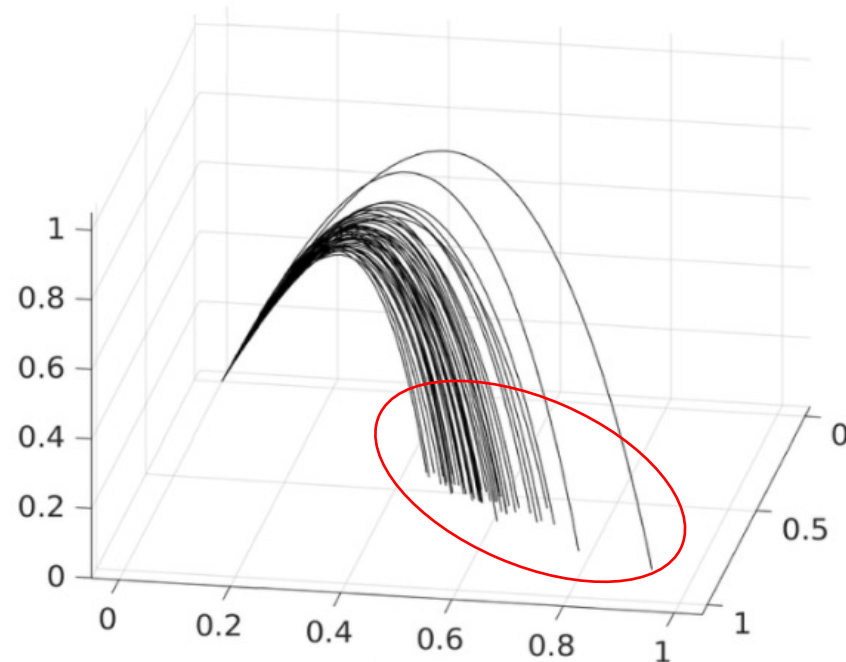
1. Introduction



2. Description of Problem

- **Problem**

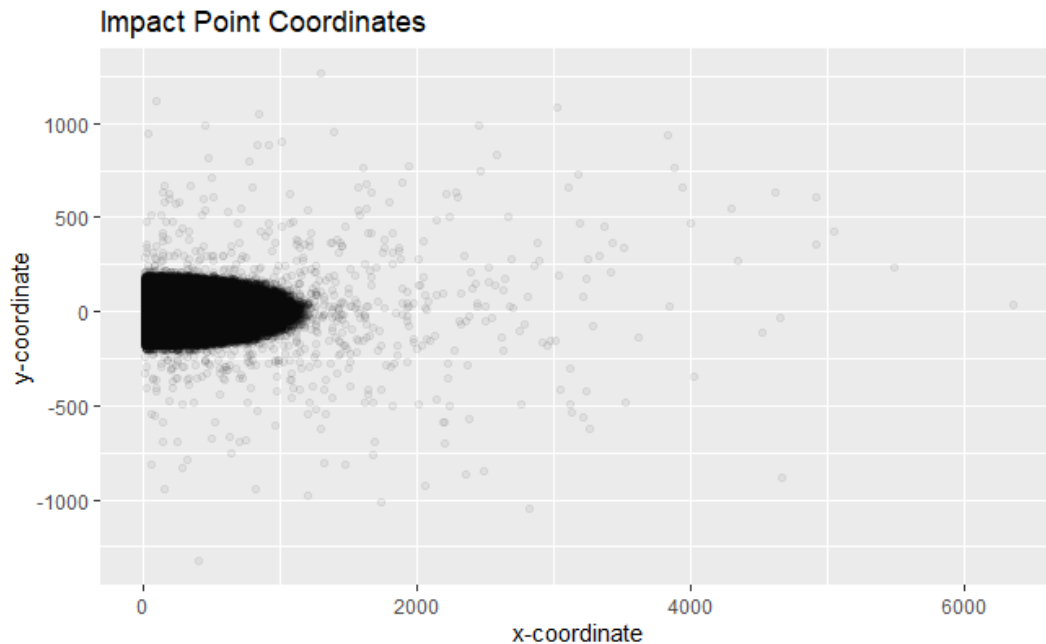
Define an **exclusion boundary** with a confidence level of $\gamma\%$ and a probability of crossing θ based on the impact point data



2. Description of Data

- **Data**

The impact point coordinates (x,y) for multiple Dasher-XLS launches

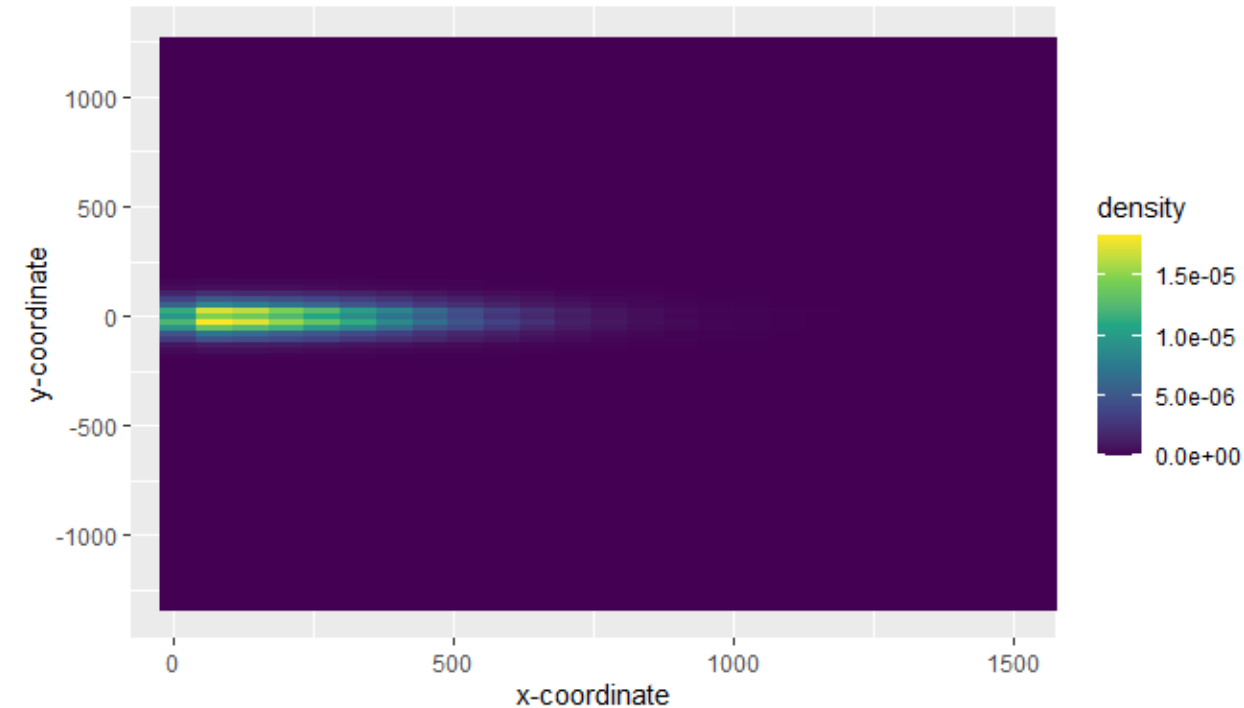


	X	Y
Min	5	-1325
Max	6355	1265
Median	215	5
Mean	251.7	-0.01

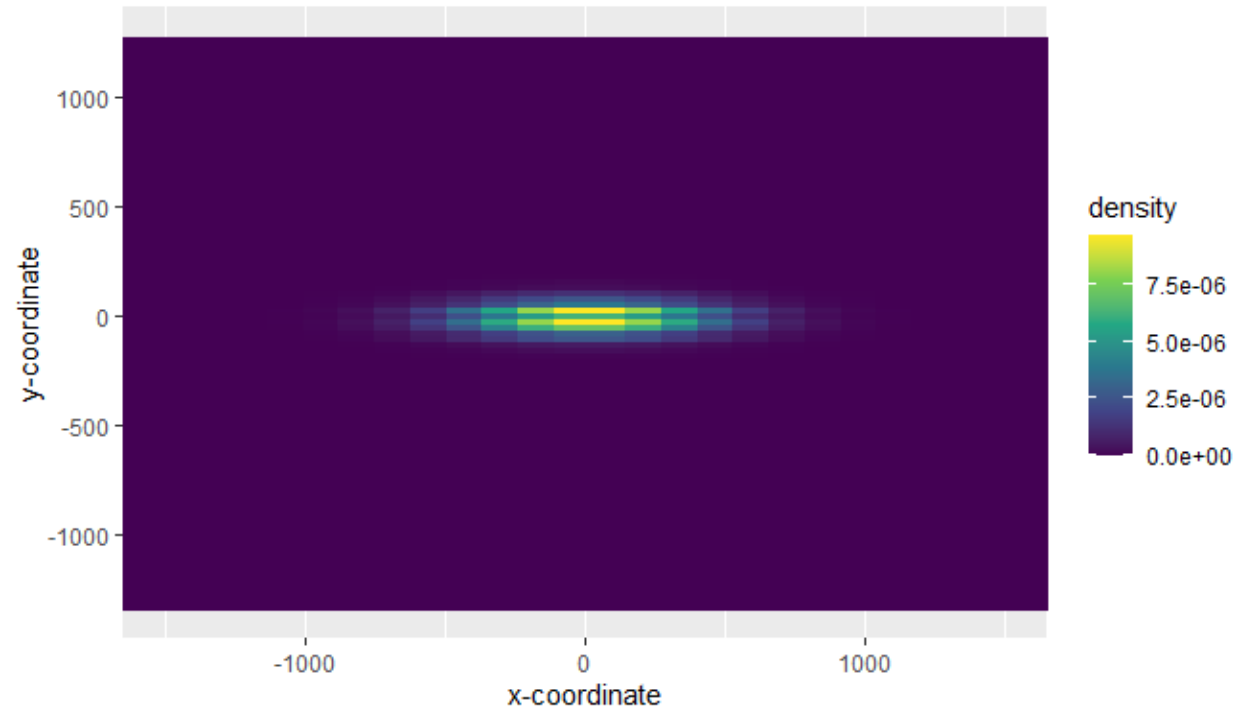


2. Description of Data

Density of Impact Points



Density of Impact Points and Duplicates



When duplicated the impact points, the distribution resembles gaussian

→ Fit a Gaussian model to data



3. Methodology

Step 1 :

Create a Gaussian Mixture Model by analyzing the data and its mirrored version

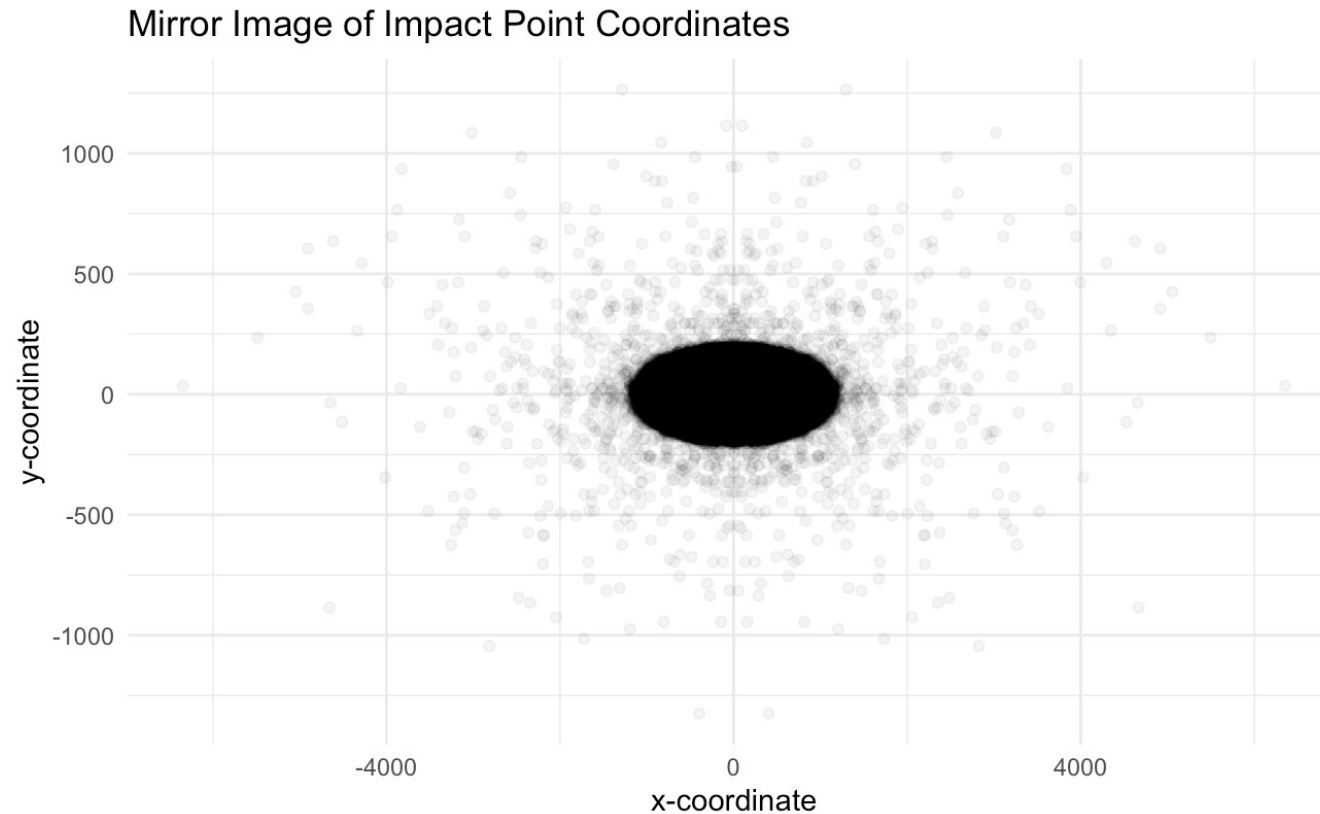
Step 2 :

Find an ellipse that leaves out $100 \cdot \theta\%$ of data points using the model's parameters

Step 3 :

Bootstrap B ellipses, each excluding $100 \cdot \theta\%$ of data points
Estimate the exclusion boundary with $\gamma\%$ confidence and θ probability

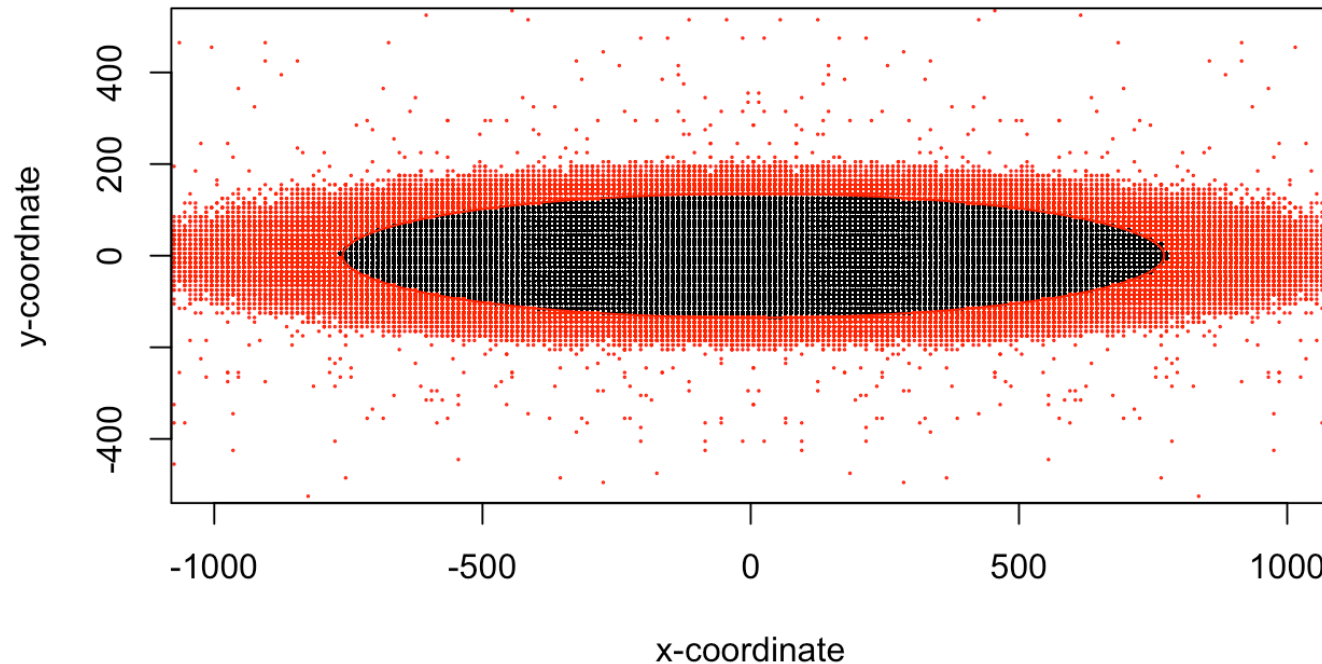
Step 1. Create a Gaussian Mixture Model by analyzing the data and its mirrored version



Density estimates are produced for each data point using a Gaussian mixture model



Step 2. Find an ellipse that leaves out $100 \cdot \theta\%$ of data points using the model's parameters



- **Black points** : non-excluded data points
- **Red points** : excluded data points that comprise 5% of the total data points
- **Red ellipse boundary** : ellipse that excludes 5% of the data points



Step 3. Estimate the exclusion boundary with $\gamma\%$ confidence and θ probability

- A. Resample the data using **bootstrap** (with replacement) based on the GMM model.
- B. Extract **mean** and **covariance** matrix parameters for each resampled data set.
- C. Calculate the **trace** of each covariance matrix in the resampled data.
- D. Determine the ellipse size using the **$\gamma\%$ quantile of the eigenvalues**.
- E. Calculate the **ellipse radius** based on the exclusion probability θ .
- F. Create ellipses representing the exclusion boundary with a confidence interval of γ and the exclusion probability θ .



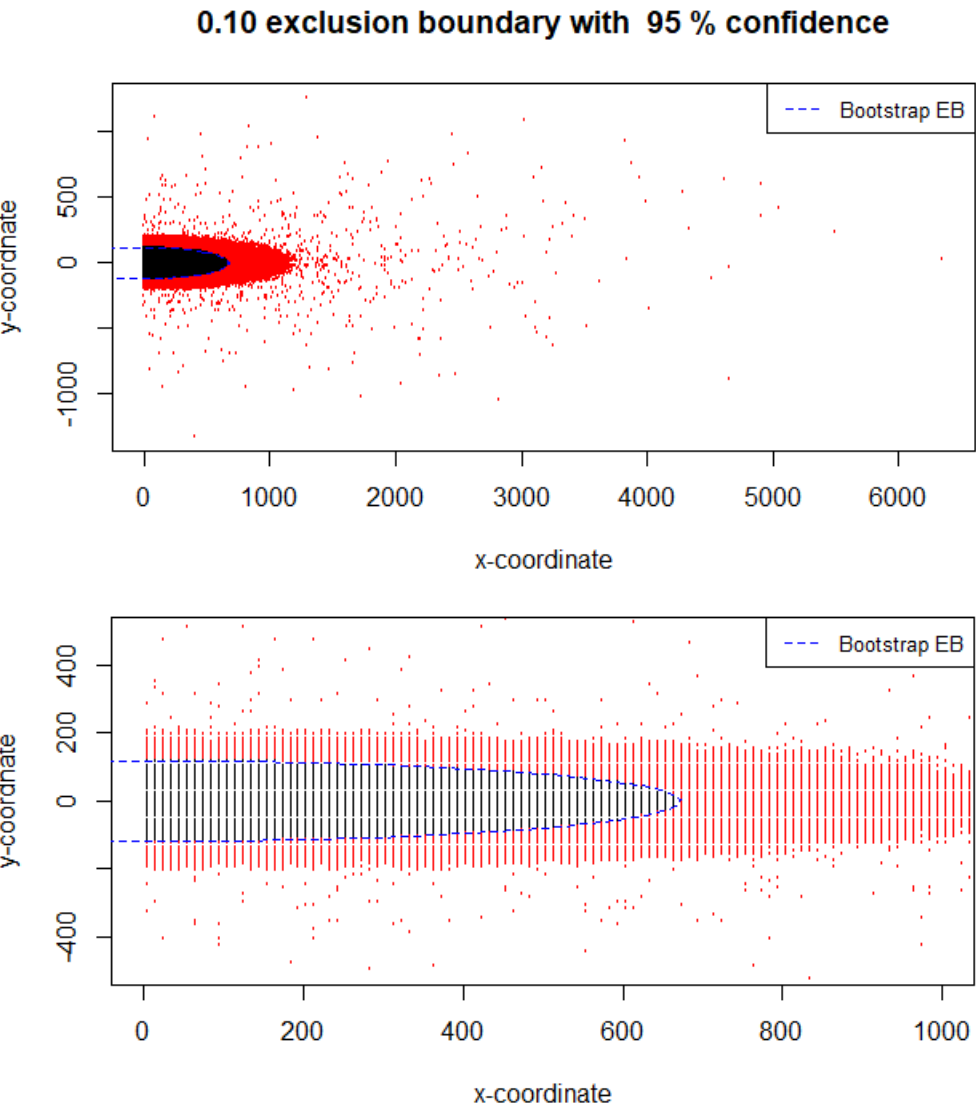
4. Results

Simulation Steps:

- A. Specify exclusion probability θ and confidence level $\gamma\%$
- B. Using the methodology from the previous section, construct an exclusion boundary that has probability θ of crossing with $\gamma\%$ confidence
- C. Take 100 bootstrap samples of size N from the data, and compute the percentage of points in each sample that lies outside of the exclusion boundary
- D. Count the number of bootstrap samples which excludes at most $100 \cdot \theta \cdot N$ data points to the pre-determined confidence level γ
- E. If our method is exact, there should be $(1 - \gamma\%) \cdot 100$ bootstrap samples that have more than $100 \cdot \theta \cdot N$ data points that lie outside of the exclusion boundary

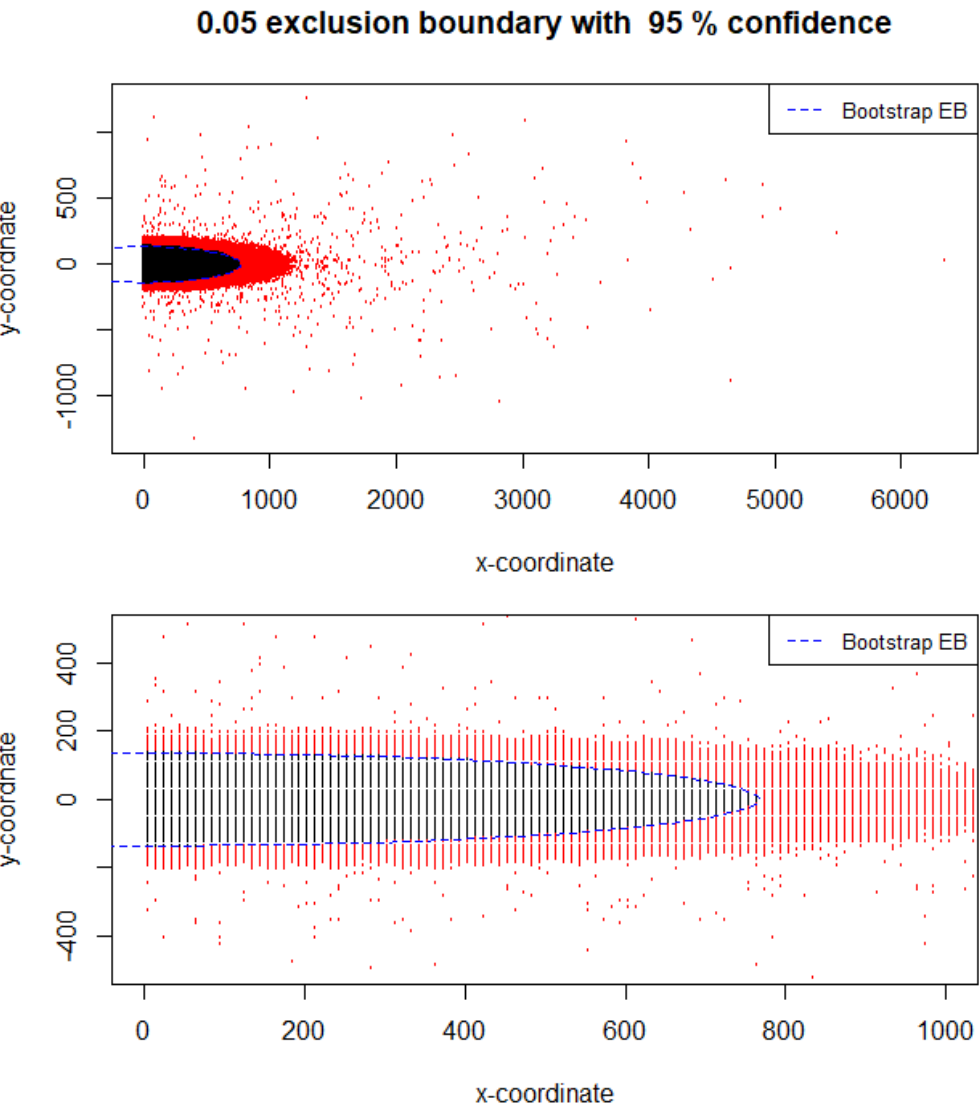


4. Results



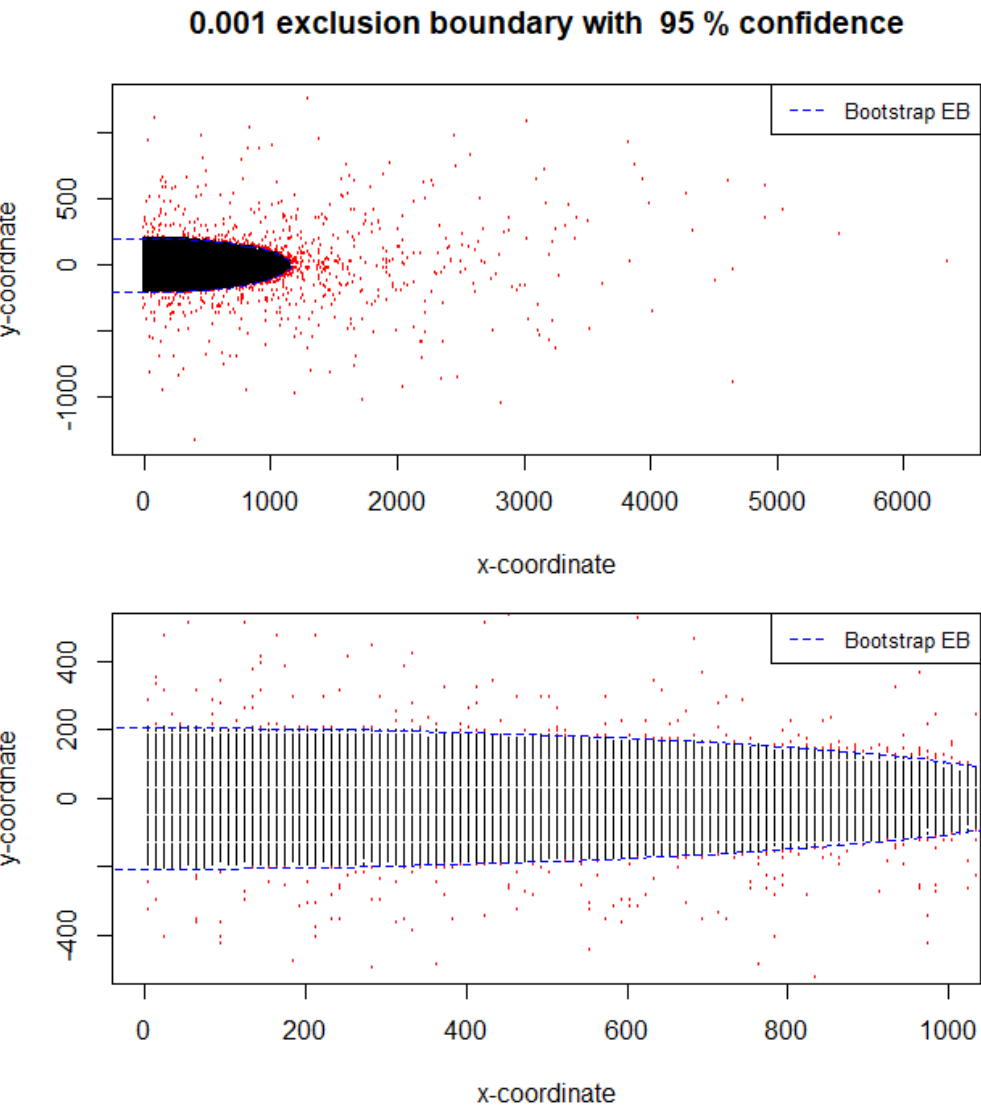
N	Number of Unsatisfactory Bootstrap Samples	Mean Exclusion (10%)	Standard Deviation of Exclusion
10,000	27 / 100	9.8035%	0.2998%
100,000	3 / 100	9.8283%	0.0842%
1,000,000	0 / 100	9.8232%	0.0317%

4. Results



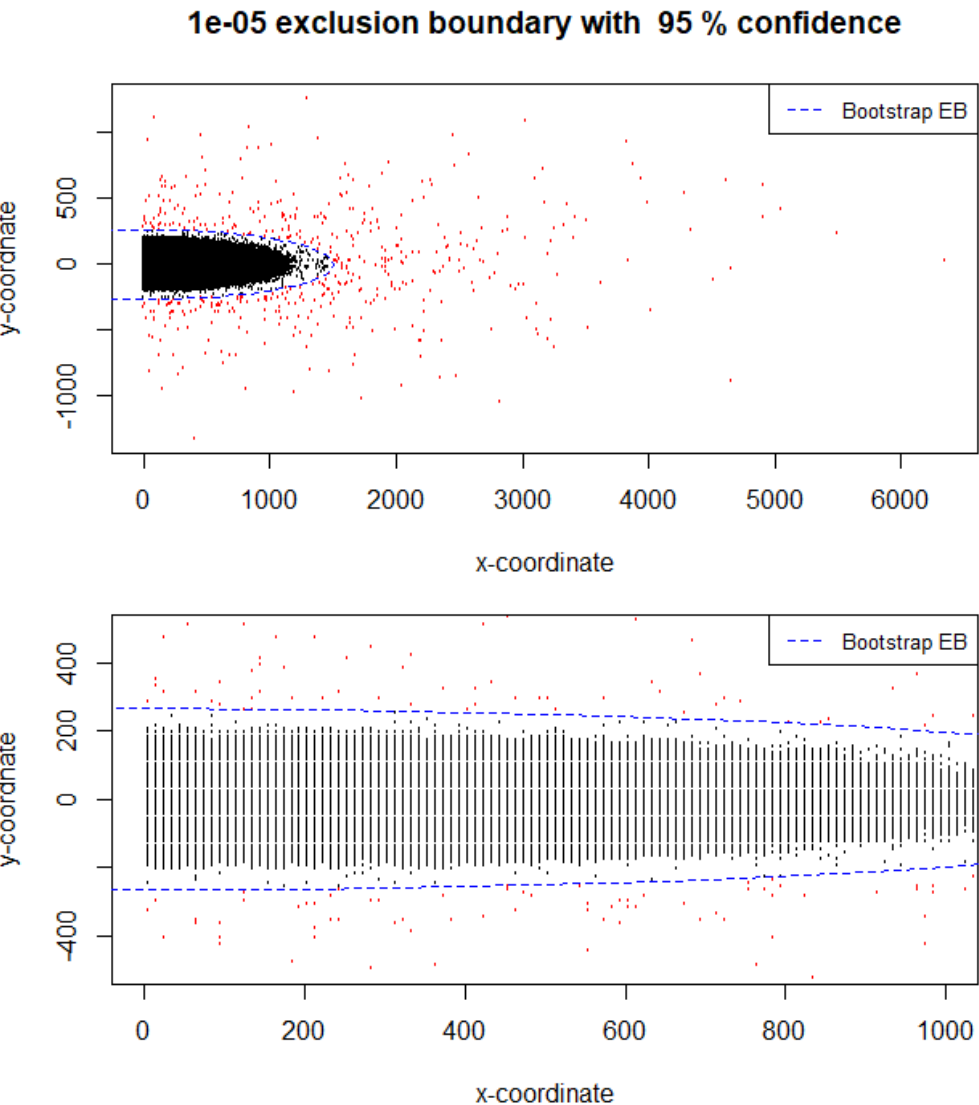
N	Number of Unsatisfactory Bootstrap Samples	Mean Exclusion (5%)	Standard Deviation of Exclusion
10,000	34 / 100	4.9098%	0.2167%
100,000	8 / 100	4.9199%	0.0620%
1,000,000	0 / 100	4.9278%	0.0200%

4. Results



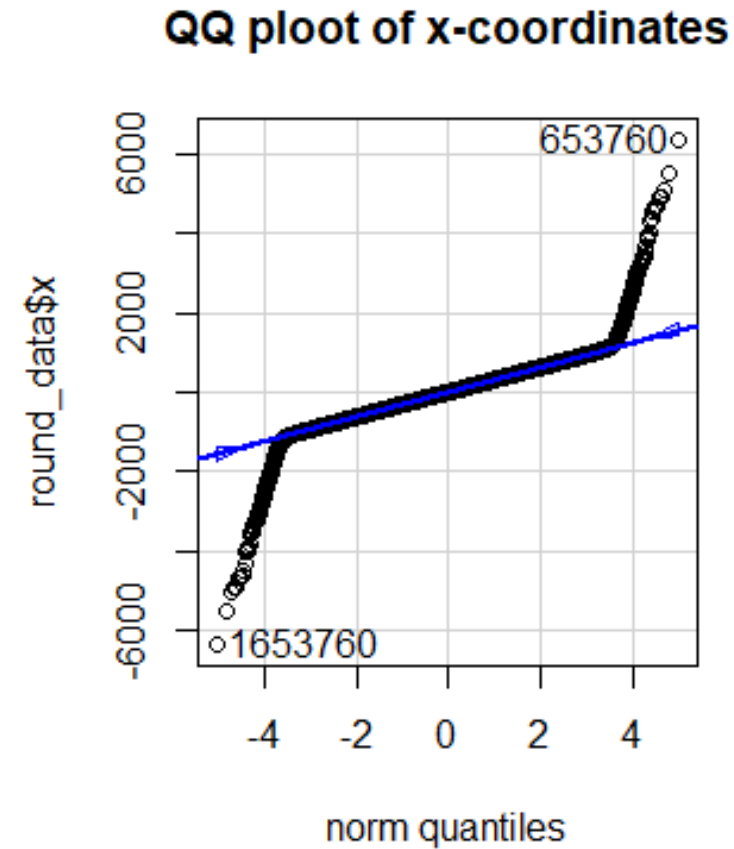
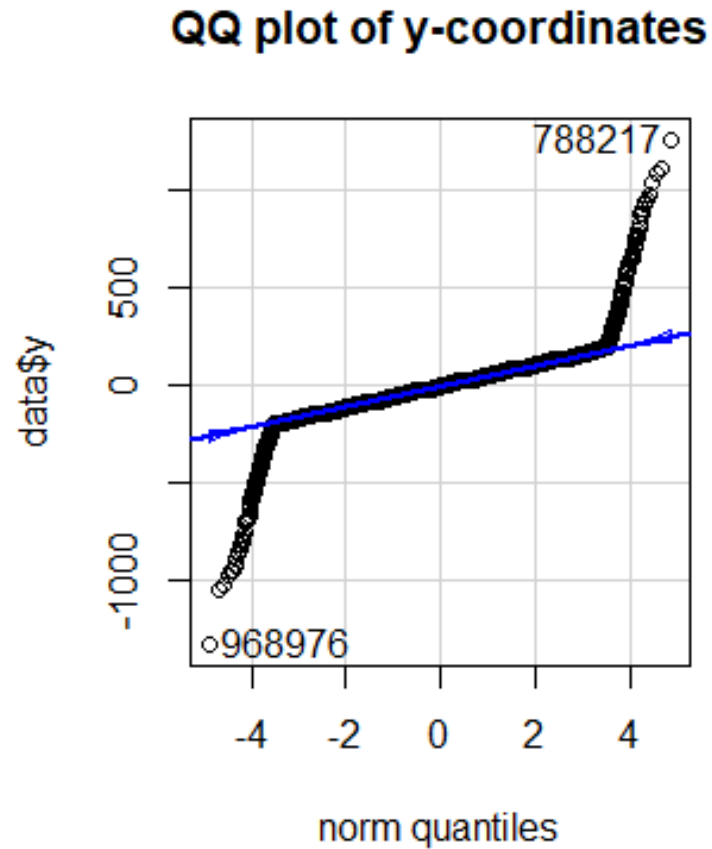
N	Number of Unsatisfactory Bootstrap Samples	Mean Exclusion (0.1%)	Standard Deviation of Exclusion
10,000	26 / 100	0.0856%	0.0297%
100,000	7 / 100	0.0876%	0.0001%
1,000,000	0 / 100	0.0884%	0.0000%

4. Results - Concern

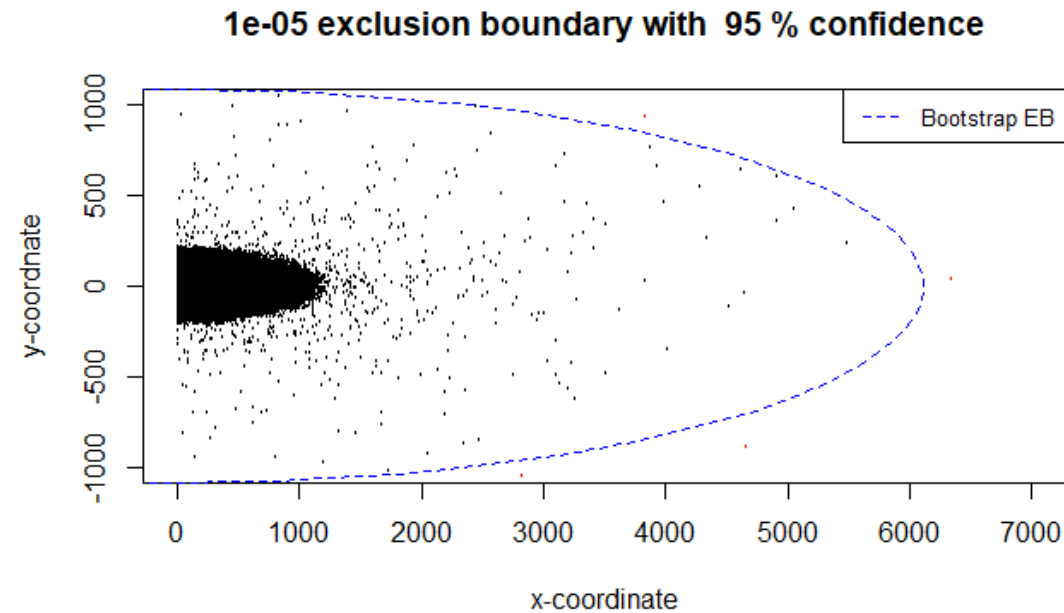


N	Number of Unsatisfactory Bootstrap Samples	Mean Exclusion (0.001%)	Standard Deviation of Exclusion
10,000	100 / 100	0.0453%	0.0172%
100,000	100 / 100	0.0462%	0.0001%
1,000,000	100 / 100	0.0463%	0.0000%

4. Results - Adjustment for Extremely Small Theta



4. Results - Concern



N	Number of Unsatisfactory Bootstrap Samples	Mean Exclusion % (0.001% Expected)	Standard Deviation of Exclusion %
10,000	8 / 100	0.0009%	0.0032%
100,000	21 / 100	0.0009%	0.0010%
1,000,000	23 / 100	0.0008%	0.0003%



5. Conclusion

- Our method of identifying an exclusion boundary utilizes the distribution of the data by transforming it to a more familiar distribution and fitting the well-known Gaussian Mixture Model.
- The method can provide an exclusion boundary that has θ probability of being crossed with $\gamma\%$ confidence.
- In the case of extremely small θ , the method performs poorly due to the presence of many extreme observations in the data and relatively small data size. We recommend collecting a sufficiently large sample when θ of interest is extremely small in order to identify an exclusion boundary with high levels of confidence.

5. Conclusion

- Our method uses familiar statistical techniques to create an exclusion boundary.
- The boundary has a specific probability of being crossed and a confidence level.
- When dealing with extremely small probabilities, the method may not perform well due to extreme data points and a small dataset.
- To ensure accurate results with high confidence, gather a larger sample when working with very small probabilities.



THANK YOU!

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