

Three-dimensional data of wirecut surface scans under the confocal microscope (110 character maximum, inc. spaces)

Yuhang Lin^{1,2,*}, Heike Hofmann^{2,3}, and Curtis Mosher, Alicia Carriquiry, Eden Amin, Jeff Salyards^{1,3}

¹Iowa State University, Department of Statistics, Ames,

²Center for Statistics and Applications in Forensic Evidence (CSAFE), Iowa State University, Ames,

³University of Nebraska-Lincoln, Department of Statistics, Lincoln,

*corresponding author(s): Yuhang Lin (yhlin@iastate.edu)

ABSTRACT

Update later: max of 170 words: describe the study, the assay(s) performed, resulting data, and reuse potential

Wire cut data is important in forensic investigations but lacks a systematic way of analyzing the data. We created a data set of 120 scans of aluminum wire cut in $\times 3p$ format, using 5 wire cutters and 3 locations along the 4 blades, with 2 replicates for each combination. A systematic pipeline with multiple analysis plots was developed to analyze the data and draw conclusions based on numerical measures.

(maximum 170 words) This is a manuscript template for Data Descriptor submissions to *Scientific Data* (<http://www.nature.com/scientificdata>). The abstract must be no longer than 170 words, and should succinctly describe the study, the assay(s) performed, the resulting data, and the reuse potential, but should not make any claims regarding new scientific findings. No references are allowed in this section.

1 Background & Summary

Wire cut data is a type of forensic tool mark data used to identify the source of a wire cutter based on the striations left on the surface. There have been cases where the evidence and testimony on wire cut evidence played a crucial role in the criminal investigation and conviction of a defendant.

However, there is a lack of a standardized method to analyze it except for visual comparison. Although the Association of Firearm and Toolmark Examiners (AFTE) developed the Theory of Identification [incorrect citation format¹](#), which outlines the process of comparing tool marks, it is still subjective and relies on the examiner's experience, making results hard to reproduce and validate. Several reports, such as those from the National Research Council² and the President's Council of Advisors on Science and Technology³, pointed out the abovementioned issues and called for more objective and reproducible methods to analyze tool marks. Early research by⁴ and⁵ has focused on collecting and distributing datasets for this purpose and providing a foundation for future advancements in tool mark analysis. Studies such as those by [\(author?\)⁶](#) and [\(author?\)⁷](#) have demonstrated the efficacy of using numerical methods to improve accuracy and consistency in tool mark analysis. [\(author?\)⁸](#) and [\(author?\)⁹](#) have explored methods for quantifying the similarity between representative signals, but alignment remains a major hurdle.

In this study, we would like to follow the same path and provide a data set of wire cut scans, and also discuss a systematic pipeline to analyze the data and draw conclusions based on numerical measures. Here, we provide a data set containing multiple files, as described in Table 1.

[breakable table, cannot cross-reference inside table](#)

For the reproducibility of all our data and alignment results, we introduce in detail in Section 2.1 [\(cross-reference not working without section number, number-sections: false not working\)](#) how we cut the wire and collect the 120 scans with 5 tools on 3 locations, in Section 2.2 how we extract profiles from the scans, in Section 2.3 how we filter signals from the profiles, in Section 2.4 how we align signals from different scans and optimize the alignment with the cross-correlation function (CCF) values. In Section 3, we discuss where our data is held. Then, in Section 4, a technical validation was conducted to further compare signals from different sources also match our assumption, together with visual aids for drawing conclusions. Finally, in Section 5, we provide available codes for creating the data set and conducting technical validation, as discussed in Section 2 and Section 4. Section 6 discusses where these codes can be found online. We hope this pipeline developed using this data set

Table 1. Structure of available data and files.

	Description	Section
Raw data		
x3p-files	folder containing 120 topographic 3d scans of aluminum wire cuts in x3p format	subsection 2.1
meta.csv	meta information for each cut with tool, blade, and location information (csv format)	subsection 2.1
Manual derivatives		
profiles	folder of files with manually extracted profiles (csv format)	subsection 2.2
Computational derivatives		
signals	folder of signals, processed from corresponding profile (csv format)	subsection 2.3
CCF values	CCF values of all pairwise aligned signals in 1 CSV	subsection 2.4
Image files		
aligned-signals	pictures of pairwise aligned signals from same sources in PNGs	subsection 2.4

38 can be further generalized and applied to real crime scenes to help investigators draw conclusions based on real wire cut data.

39 **2 Methods**

40 In this study, aluminum wire was used to create an optimal scenario where the most amount of information could be transferred
41 from the tool to the substrate, despite the wire in some real cases being made of lead. The physical property of aluminium
42 wire make it an excellent candidate for keeping marks while being relatively easy to bend and non-toxic.

43 **2.1 Cutting wires**

44 The aluminum wire used was 16 Gauge/1.5 mm, anodized. In order to cut the wire, 4-inch pieces were unspooled and cut
45 using Kaiweets wire cutters, model KWS-105, as shown in Figure 1(a), for 1 blade location, either inner, middle, or outer,
46 which gives us 1 replicate. Each piece was then cut into half to create 2-inch pieces for each side, AB and CD, with a sharpie
47 line marking the cut ends, giving us 4 samples. Here, we are showing AB sides only in Figure 1(b) (**remove the tent figure**),
48 and the CD sides are similar from the other side of the cut, with the back of A being C and the back of B being D. Both AB and
49 CD sides form tent structures on the tips of the wire, and we can separate each side of the tent into 2 pieces along the bending
50 position, resulting in 8 scans. We repeated this process for all 3 locations along the blade and 5 wire cutters, with 2 replicates
51 for each tool-edge-location combination, resulting in 120 scans. Each piece was labeled with the naming conventions, T(ool)
52 1/2/3/4/5 (Edge) A/B/C/D W(ire) - L(ocation) I(mmer)/M(iddle)/O(uter) - R(epetition) 1/2, with T1AW-LI-R1 being the piece
53 cut by tool 1 on the A edge at the inner location for the first repetition. Then, we can use the standard scanning protocols for
54 the confocal microscope, shown in Figure 1(c) (**need an extra pic of the very tip**), to scan the wire tip surfaces. The scanned
55 surfaces are saved in a resolution of $0.645\mu\text{m} \times 0.645\mu\text{m}$ per square pixel in an $\times 3\text{p}$ file format.

56 **XXX figure 1 - generally, zoom into these images - we do not want to have a hand in the image, nor a view of the**
57 **crafting aluminum :)** - what are the exact rules on visuals in Scientific Data ? **XXX hard to put the full requirements here, see**
58 **<https://www.nature.com/sdata/publish/submission-guidelines#figures>**

59 **put into quarto layout with (a) (b) (c) on the top left, no tent, add blade C & D (Do later after decide on using qmd or**
60 **Overleaf)**

61 **2.2 Extract profiles**

62 Numerical comparisons between 2 replicates cannot be done directly on the $\times 3\text{p}$ files. We need to extract representative
63 functions from the scans first. A representative function with the most information is considered as a signal for one scan,
64 which can be used for comparison later. To obtain this function, we first need a profile of the scan, which is a sequence of
65 values along a user-drawn line on the surface. The profile should capture most features of the scan, and be orthogonal to the
66 striation marks of the scan, which are formed by ups and downs of grooves. So, we draw the line across the wide region of
67 the scan to maximize the feature captured, as shown in dark blue in Figure 2(a). We can then investigate the values under this
68 profile line. The profile function is along the line is plotted in Figure 2(b).

69 **2.3 Filtered signals**

70 With the profile extracted, we can then obtain the signal. Two Gaussian filters are applied to these resulting profiles. In
71 particular, we first used a large low-pass filter with bandwidths of 400 microns to remove large trend, as it can overwhelm the
72 signals, and then used a small high-pass filter of 40 microns to average across noise and remove spikes, as shown in Figure
73 2(c). ¹⁰ (**add reference: W. S. Cleveland, E. Grosse and W. M. Shyu (1992) Local regression models. Chapter 8 of Statistical**
74 **Models in S eds J.M. Chambers and T.J. Hastie, Wadsworth & Brooks/Cole.**) Finally, the extreme tail values are removed.

75 **2.4 Align signals**

76 Signals extracted from different scans can be put together for comparison, and we maximize the cross-correlation function
77 (CCF) values between the signals to numerically find the best alignment. For example, we compare T1AW-LI-R1 to T1AW-
78 LI-R2, T1CW-LI-R1 to T1CW-LI-R2, and so on. That is comparing each row in Figure 3. We know that signals from two
79 replicates with the same tool-edge-location combination should yield similar signals as in the first and second column of
80 Figure 4, which will give alignments of massive overlapping and high CCF values close to 1. The alignments and values we
81 got in the rightmost column of Figure 4 fulfill our expectations.

82 **3 Data Records**

83 The complete data set is available on the ISU DataShare repository at <https://iastate.figshare.com/>, which is public and open
84 access for every interested researcher. The data set consists of 120 scans in the $\times 3\text{p}$ file format with the naming convention
85 as described before. (**Explain the $\times 3\text{p}$ header info?**)

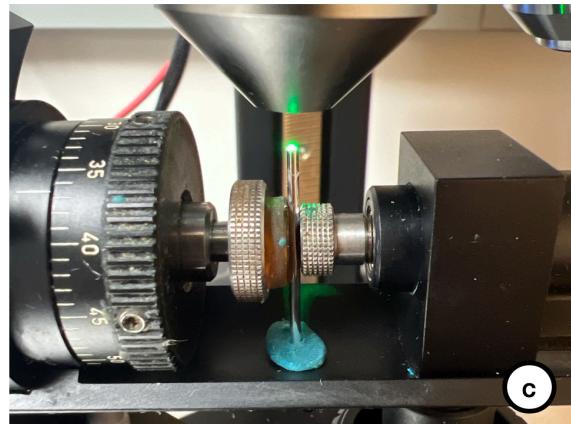
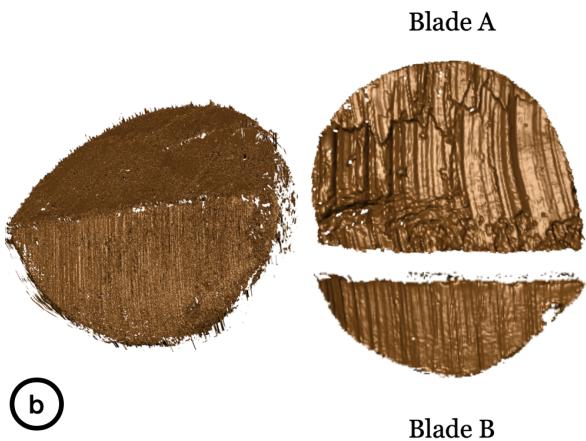


Figure 1. (a) A Kaiweets wire cutter of model KWS-105 was used to cut the wire. (b) A tent structure created by blade AB. After separating 2 tent structures by the connecting position, we obtained 2 samples - 1 sample from blade A and B. (c) A confocal microscope was used to scan the wire surfaces.

86 4 Technical Validation

87 a picture of alignment with ccf from different sources to show if different source, our evaluation returns small ccf, which
88 matches what we thought.

89 For the data collection process, two team members did the cutting and labeling together, then one person did the scanning
90 and named according to the naming convention above. The scanning was done in a specific order to ensure consistency across
91 all scans. The data was saved in a consistent format to ensure they could be easily accessed and analyzed. A third person then
92 checked the data to ensure that the data was consistent in naming and accurate.

93 again - the website is not the right place for the validation - instead, move parts from the website here.

94 For the validation of the scans and their processing we investigate the correlation scores of pair-wise aligned signals. For
95 signals from scans of wires cut with a different tool, we would expect a low correlation score. Large scores between signals
96 are indicative of being made by the same tool. Show boxplots and roc curve.

97 For validation of all other tools and locations of scan replicates, see the detailed [report](#).

98 5 Usage Notes

99 The R package `x3ptools` (available from CRAN) supports working with files in x3p format.

100 Sample scripts in R for processing scans from x3p format to their signal are available from ... [github](#).

101 Further analysis can be conducted with the GitHub R package `wire` and the GitHub R shiny app `wireShiny` ([citation?](#)). We already conduct between-replicate comparisons in the technical validation section, and we can also conduct
102 across-replicate comparisons to establish error rates threshold and produce other analysis plots.

103 Suppose we put the CCF values in a tilemap with different tool, location and edge combinations. In that case, we expect

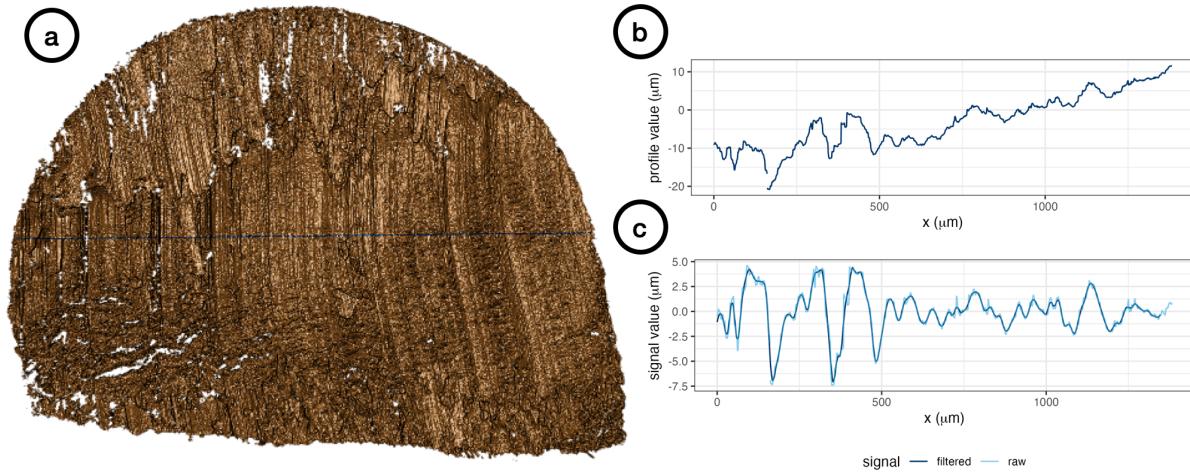


Figure 2. (a) A profile line in dark blue was drawn across the striations of the scan. (b) The profile function extracted along the profile line in (a). (c) The raw signal in light blue is obtained by using the low-pass filter on the profile function in (b) and the filtered signal is obtained by using the high-pass filter on the raw signal.

105 only the diagonal to have high CCF values, close to 1 and marked as orange in the tilemap, as the diagonal represents the same
 106 source, and the rest of the matrix to have low CCF values, close to 0 and marked as gray. In Figure 5, the behavior is consistent
 107 with our expectation overall, except for some rare cases with tool 5 edge D, which is caused by ????? We also put the resulting
 108 CCFs in the boxplot, as in Figure 6. We can see that the CCF values for the same sources are close to 1, while the CCF values
 109 for different sources are much lower than expected. This difference can be used to establish a threshold for CCF and help us
 110 draw conclusions about the similarity between wire cut scans numerically, which can be used in real crime scenes. The density
 111 plot in Figure 7 shows the distribution of the CCF values with the same sources and different sources. The overlapping points
 112 between the tails of these two distributions can be a rough threshold. Furthermore, the receiver operating characteristic (ROC)
 113 curve in Figure 8 shows the sensitivity / true positive rate against the false positive rate (FPR) (1 - specificity). The curve is
 114 very close to the upper left corner, which is excellent for classification and drawing conclusion. It gives us a true threshold of
 115 0.589 to control the FPR to be less than 0.05 with false negative rate (FNR) to be 0, (false positive rate (FPR) / false discovery
 116 rate (FDR) -> define the H0 or call it false identification rate (FIR)???, and 0.658 to control the FPR to be less than 0.01, with
 117 FNR to be 0.02.

118 6 Code availability

119 [table of code-manual?](#)

120 no, we can't use the website as a place for more detailed procedures. This paper is the detailed procedure.no more
 121 README, scanning procedures in another HTML

122 We put together the cutting and the standard scanning procedures mentioned in Section 2.1 cross-ref not working with
 123 more pictures for each step into a [README of the GitHub repository heike/Wirecuts](#) (High-res pics needed in the
 124 README).

125 The data set can be easily accessed with the CRAN R package `x3ptools`. Further analysis can be conducted with the
 126 GitHub R package `wire` and the GitHub R shiny app `wireShiny` ([citation](#)).

127 References

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146 Acknowledgements

147 This work was partially funded by the Center for Statistics and Applications in Forensic Evidence (CSAFE) through Cooperative
148 Agreement 70NANB20H019 between NIST and Iowa State University, which includes activities carried out at Carnegie
149 Mellon University, Duke University, University of California Irvine, University of Virginia, West Virginia University, University
150 of Pennsylvania, Swarthmore College and the University of Nebraska-Lincoln.

151 Author contributions statement

152 Let's follow the Elsevier definitions: <https://www.elsevier.com/researcher/author/policies-and-guidelines/credit-author-statement>
153 Y.L.: Methodology, Software, Validation, Data Curation, Writing - Draft; H.H.: Conceptualization, Methodology, Validation,
154 Writing - Review & Editing; C.M.: Lab supervision; E.A.: Physical Specimen, Scanning; J.S: Forensic advice; A.C.: Funding
155 acquisition.
156 All authors reviewed the manuscript.

157 Competing interests

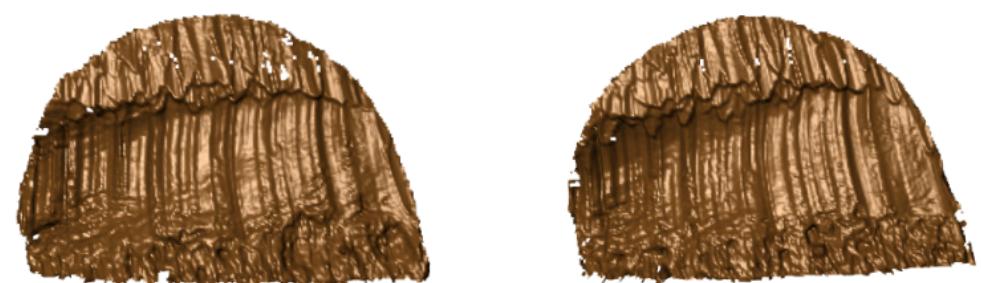
158 (mandatory statement)

159 H.H. is a technical advisor to AFTE (Association of Firearms and Toolmarks Examiners), fellow of the ASA (American
160 Statistical Association), and committee member of the ASA Forensic Science Committee. H.H. has testified as court witness
161 on behalf of judge April Neubauer, NY State Supreme Court Criminal Term in New York City.

Edge A



Edge C



Edge B



Edge D



Figure 3. Scans from different sides of tool 1 at the inner location.

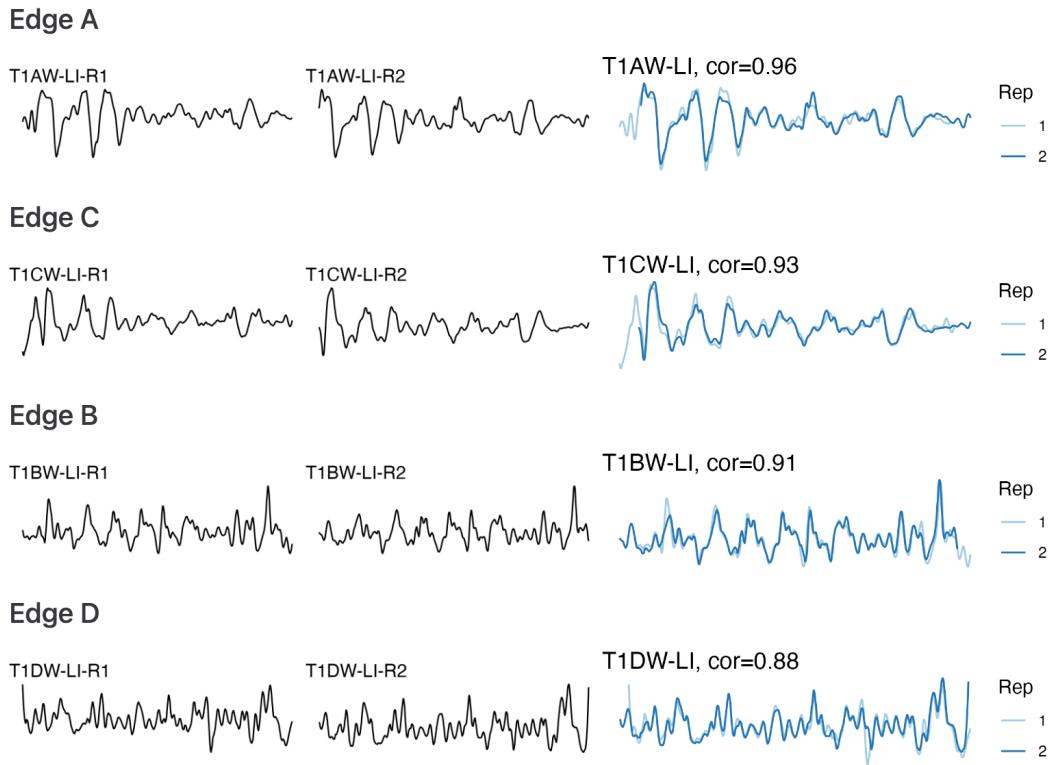


Figure 4. The first and second columns show the signals extracted from Figure 3, and the third column shows the alignments and CCF values between pairs of signals.

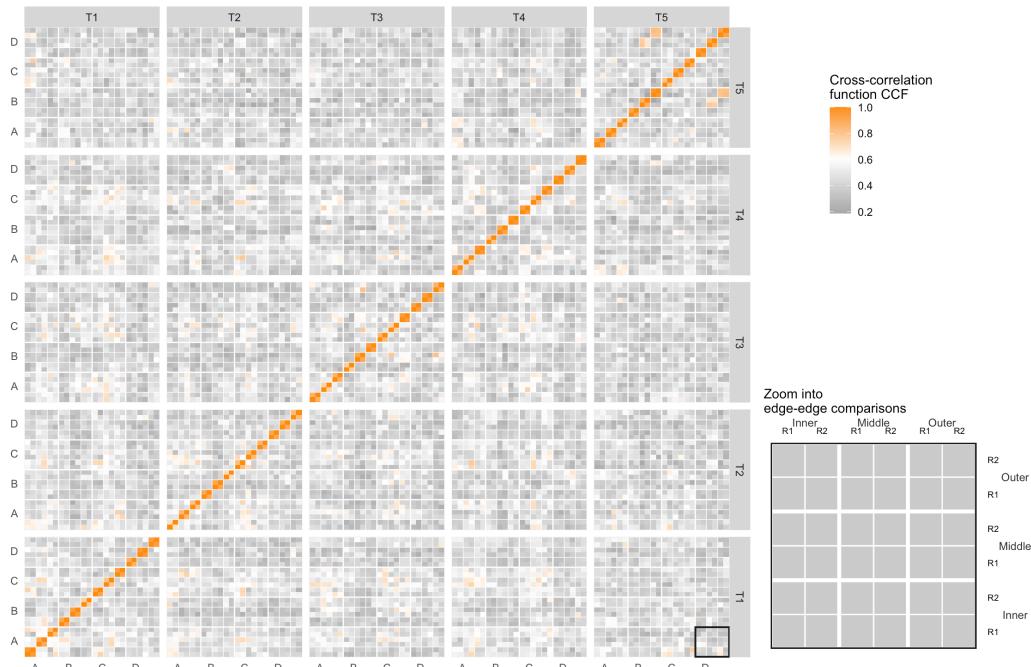


Figure 5. The tilemap shows signals from the same source have CCFs close to 1.

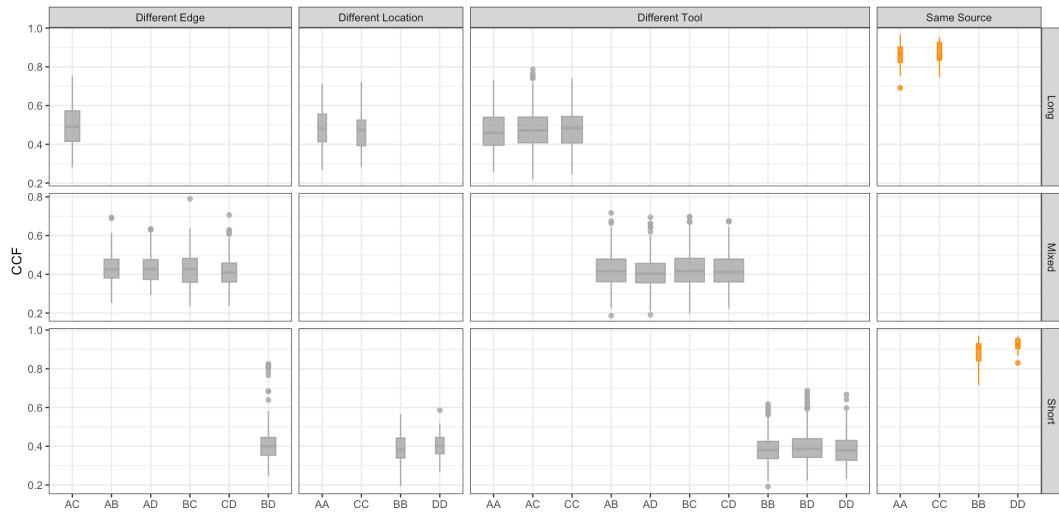


Figure 6. The boxplot shows that signals from the same sources have higher CCFs than those from different sources.

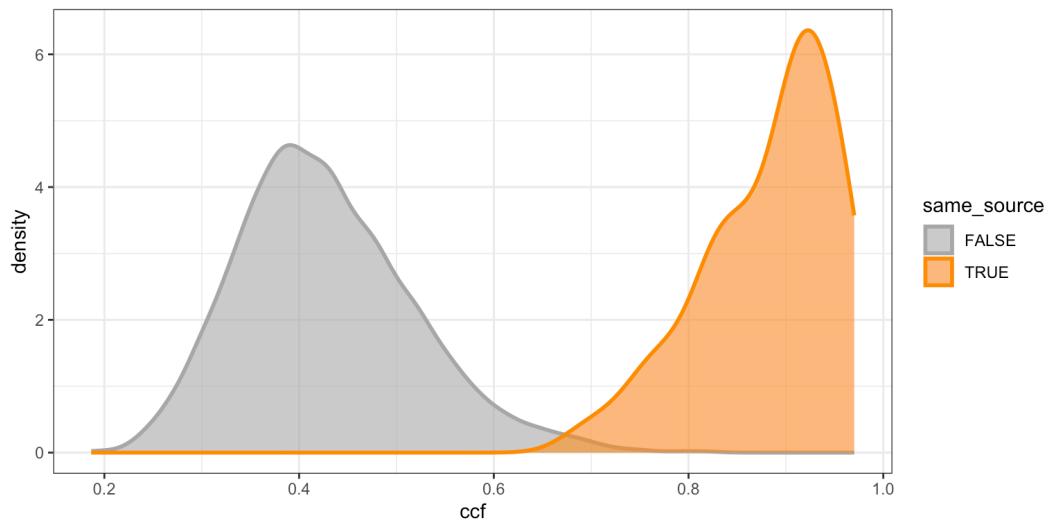


Figure 7. The density plot shows tails of distributions overlap, which can be used as a rough threshold for drawing conclusions.

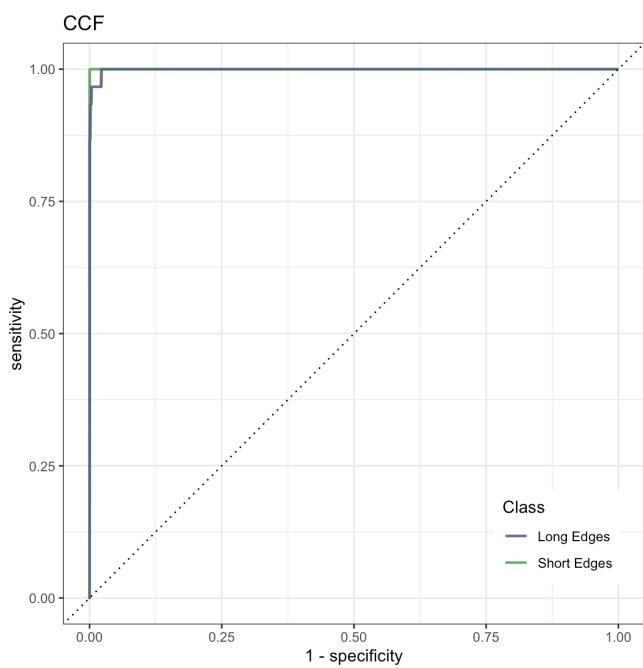


Figure 8. The ROC curve is bending very close to the upper left corner, which means excellent in classification and drawing conclusions.