

Automatic Matching of Scans from Hamby Sets Friends, False Friends and Clones

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An Investigation ... - the statistical way

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Outline

- ★ Matching bullets - very brief overview of the algorithm: *getting RF scores*
- ★ Applications for quantifying identifications: *using RF scores*
- ★ Scores as a diagnostics tool: *mis(?)using RF scores*

Hamby Sets

- ★ in collaboration with: St Louis PD, DCI Ankeny, Jim Hamby, Paul Murphy, NIST NBTRD
- ★ Hamby Sets 10, 36, 44, 173, 224, 252
- ★ Hamby Clones 159
- ★ Hamby Clone Test Set 224
- ★ each set consists of
 - ★ 20 known bullets (2 from each of ten consecutively manufactured P-85 barrels)
 - ★ 15 questioned bullets

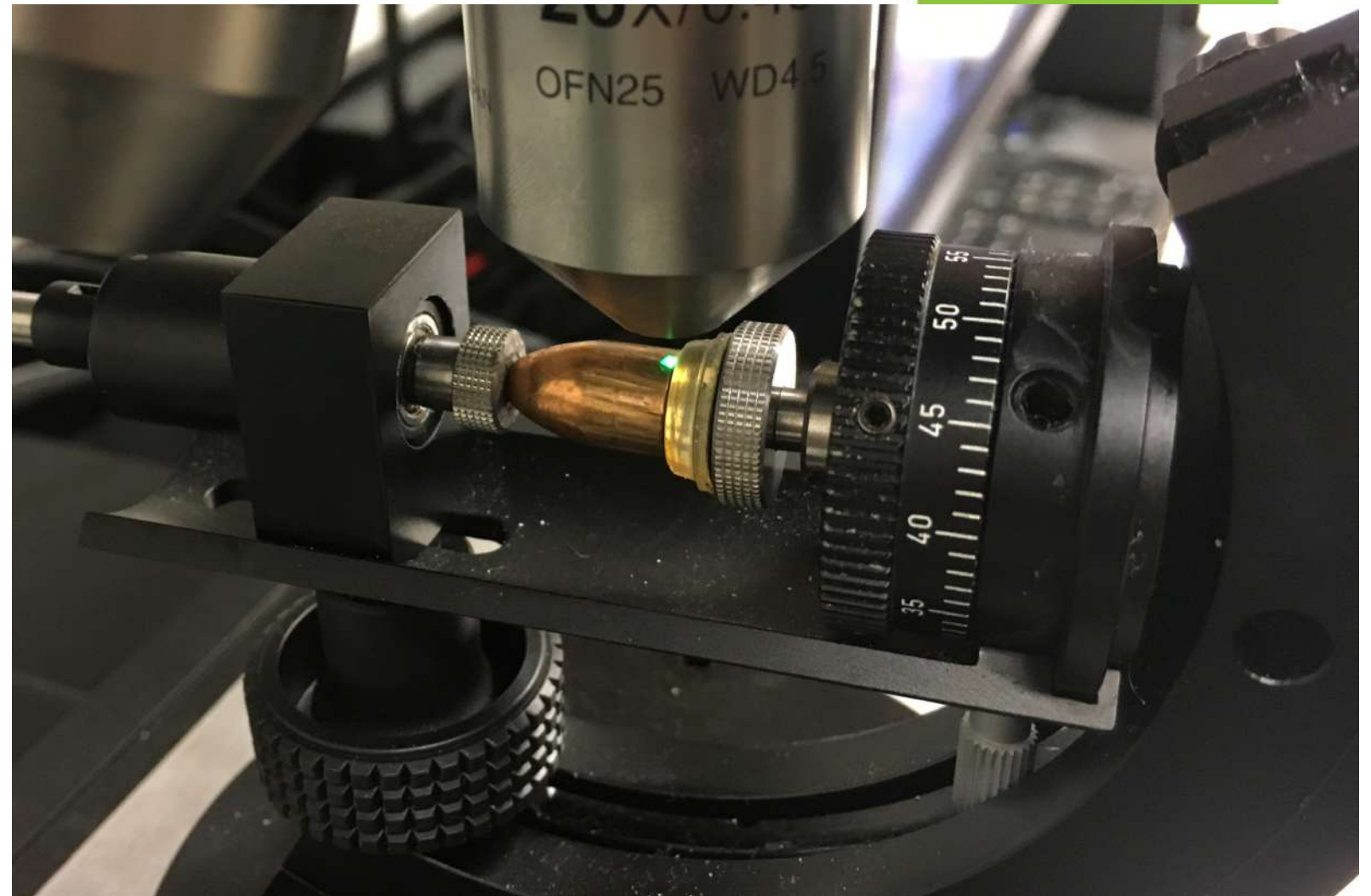
Data: Microscope Facility

Roy J. Carver High Resolution Microscopy Facility

Two Sensofar Confocal
Light Microscopes

Six undergraduates
scanning bullet lands

3d topographic images:
height measurements on
x-y grid



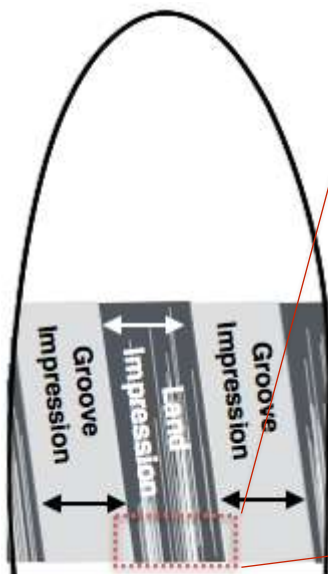
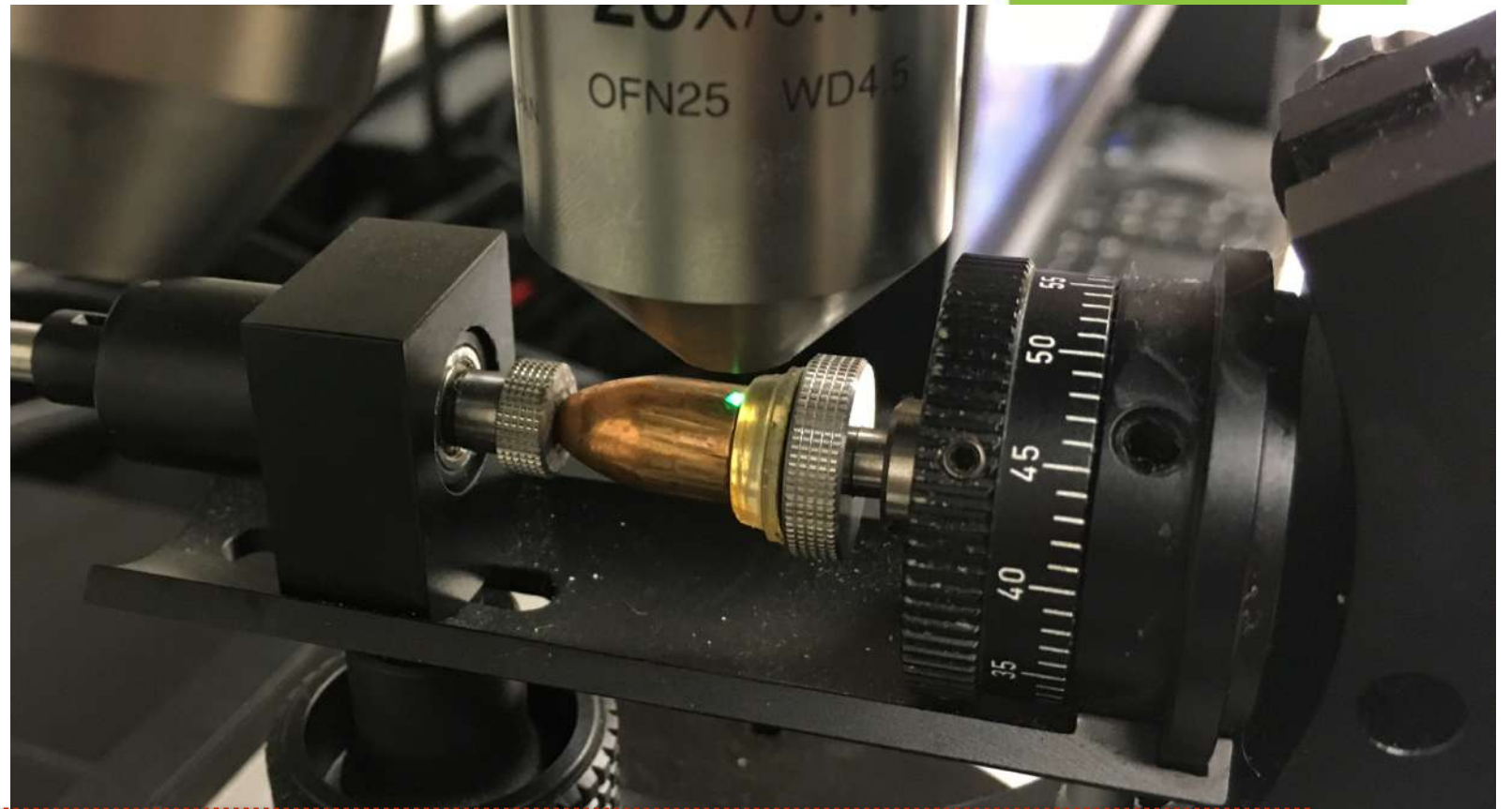
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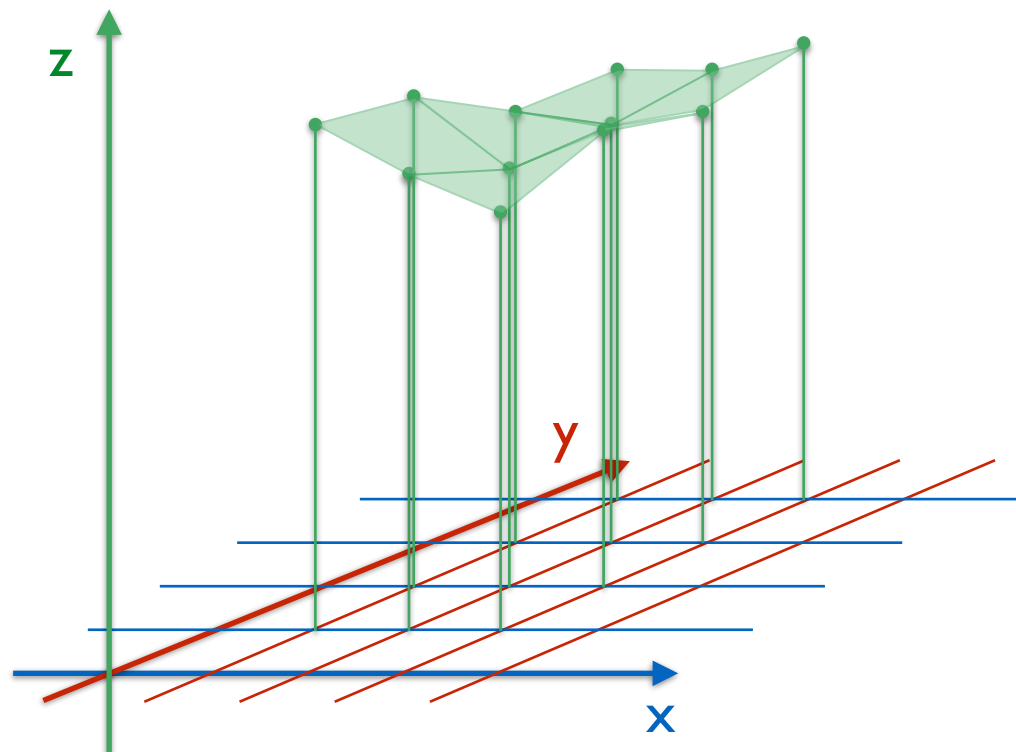
3d topographic images:
height measurements on
x-y grid



Data from CL Microscope

x-y-z files

Data captured on a regular
grid of $0.645 \mu m \times 0.645 \mu m$
Total captured area for each
land $\sim 2.2 mm \times 0.6 mm$



x - y - z file

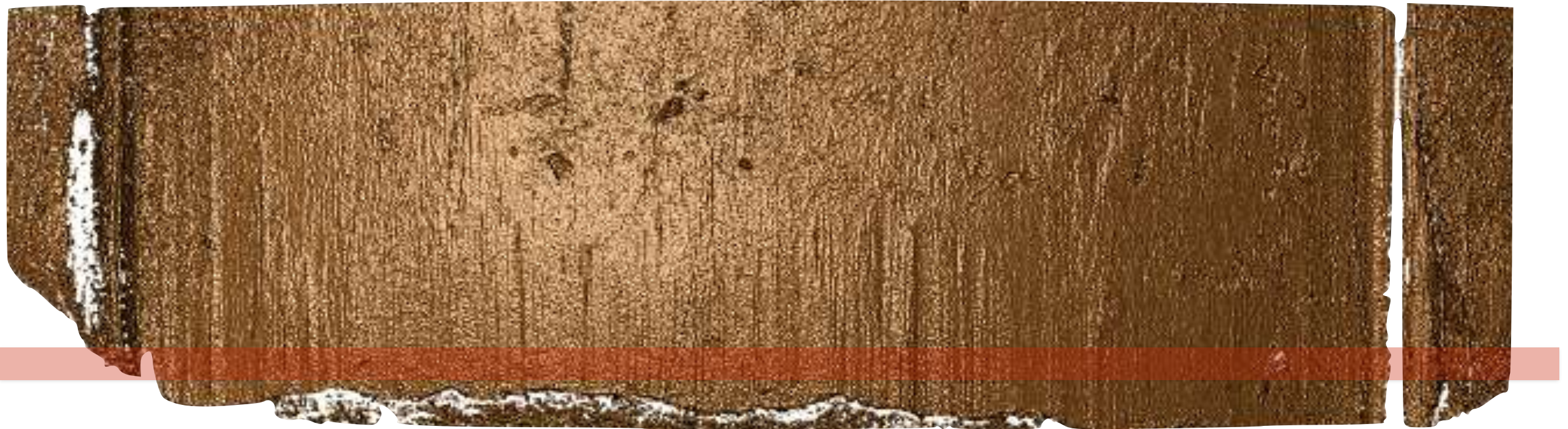
x	y	z
18.705	0.000	-25.221138
19.350	0.000	-25.253155
19.995	0.000	-25.335022
20.640	0.000	-25.418171
21.285	0.000	-25.477917
21.930	0.000	-25.541687
22.575	0.000	-25.673903
23.220	0.000	-25.966341
23.865	0.000	-40.070286
24.510	0.000	-40.407612
25.155	0.000	-40.587063
25.800	0.000	-33.437973
26.445	0.000	-33.691895
27.090	0.000	-39.690674
27.735	0.000	-40.317741

.
. .
. .

Automatic matching score

Step 1: identify region suitable for matching

land from bullet fired from Smith & Wesson



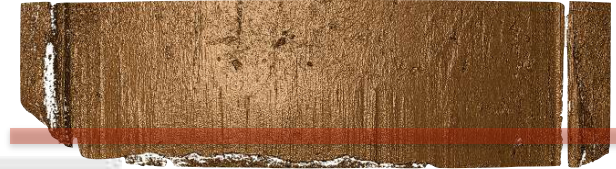
Region close to heel of bullet

Avoid break-off

Automatic matching score

Step 1b: from scan to crosscut

Identify matching region



Automatic matching score

Step 1b: from scan to crosscut

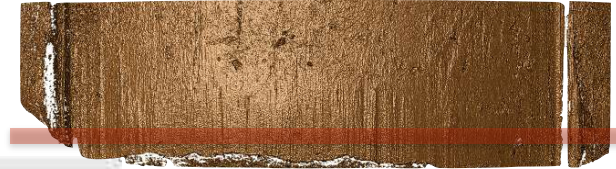
Identify matching region



Automatic matching score

Step 1b: from scan to crosscut

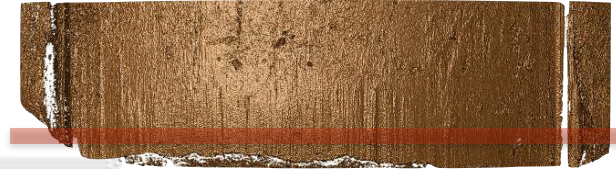
Identify matching region



Automatic matching score

Step 1b: from scan to crosscut

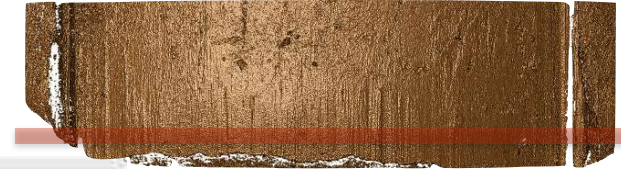
Identify matching region



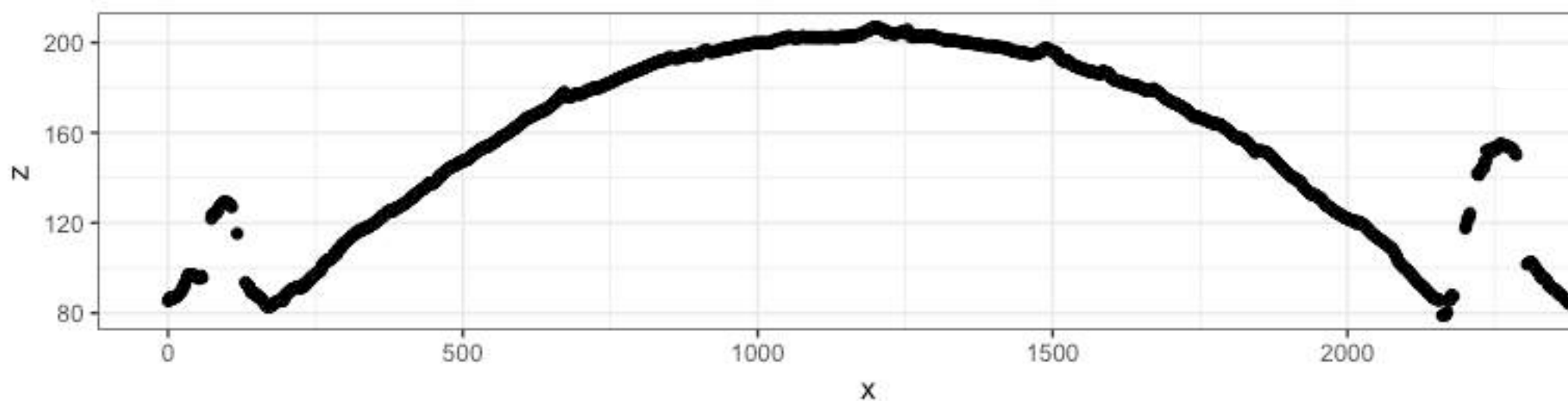
Automatic matching score

Step 1b: from scan to crosscut

Identify matching region



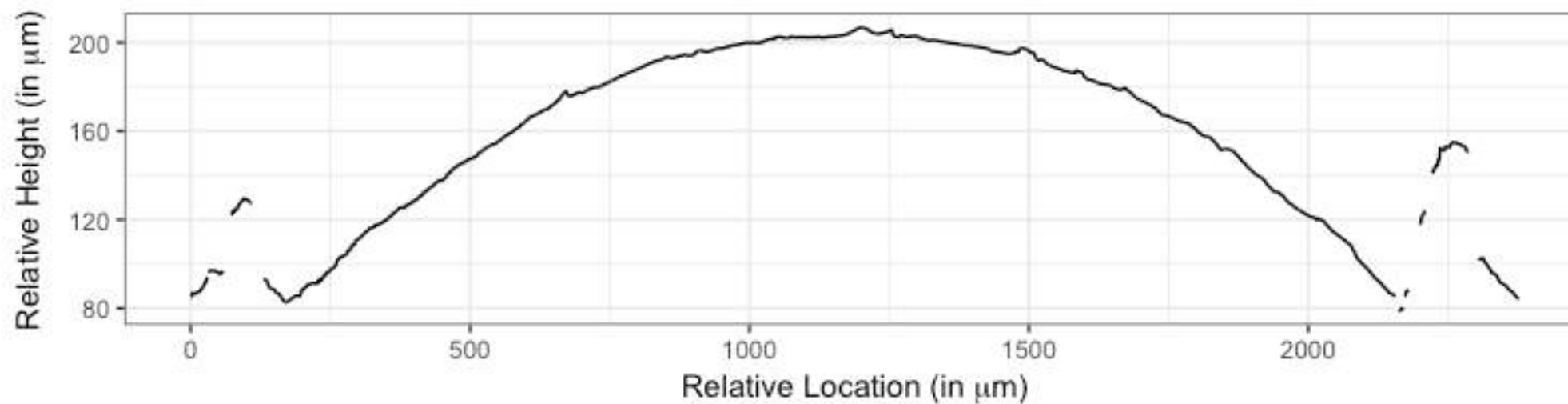
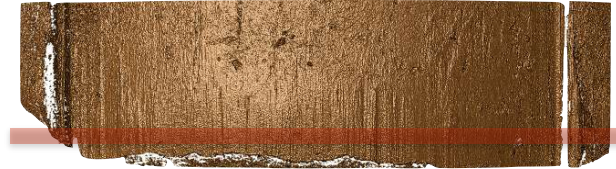
RGL device 3 [Focus]



Automatic matching score

Step 2: Identify groove locations

Identify matching region

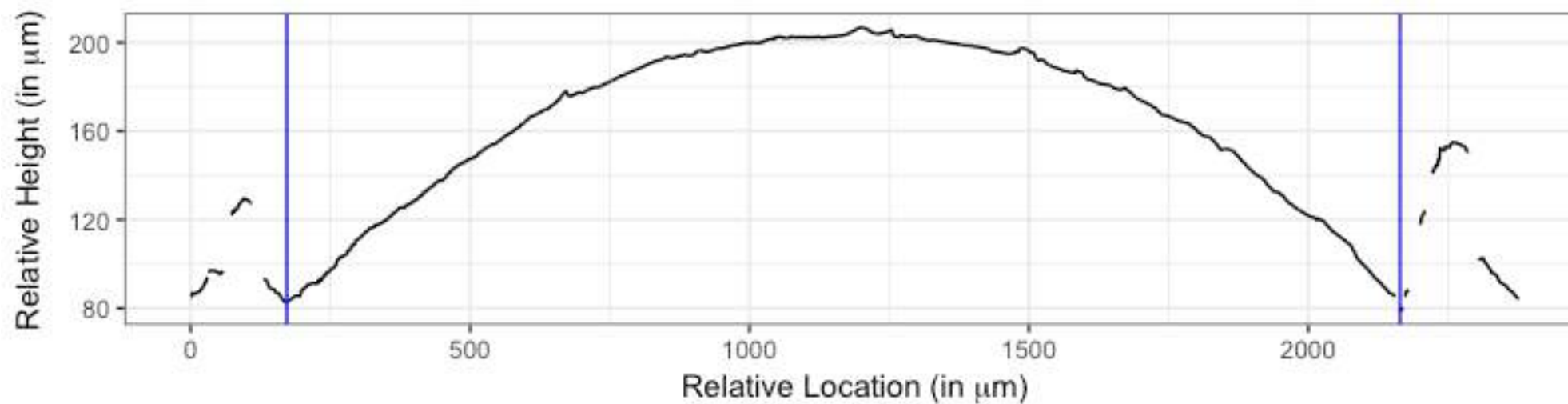
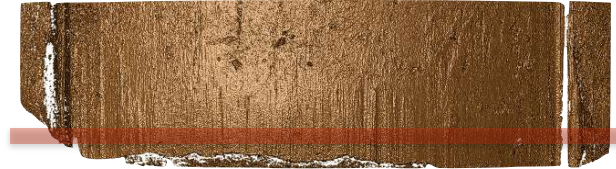


Shoulders (locations outside the grooves) are removed

Automatic matching score

Step 2: Identify groove locations

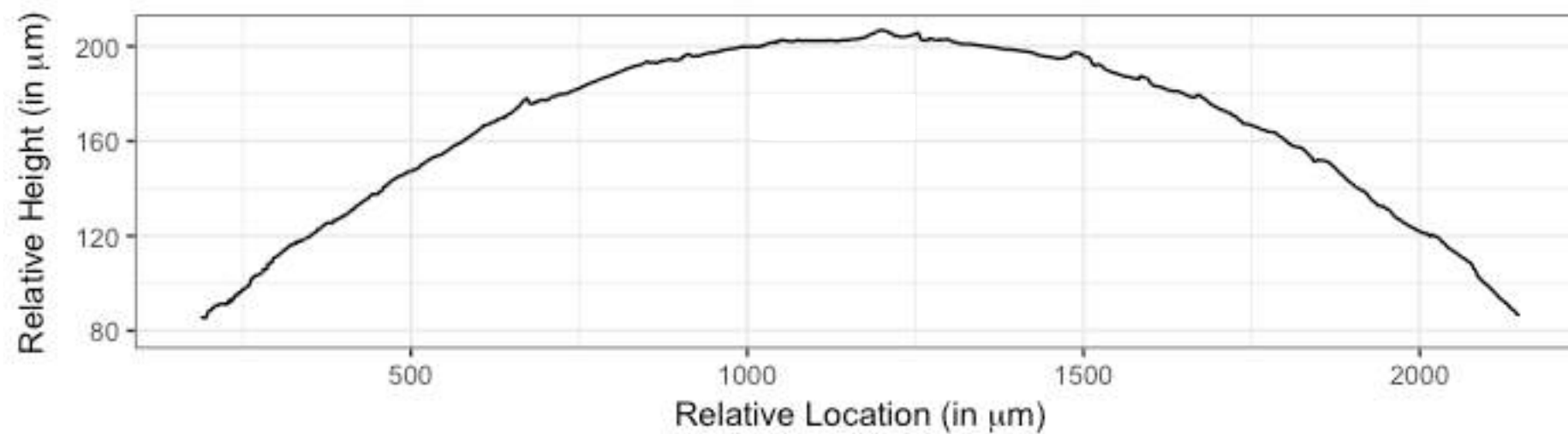
Identify matching region



Shoulders (locations outside the grooves) are removed

Automatic matching score

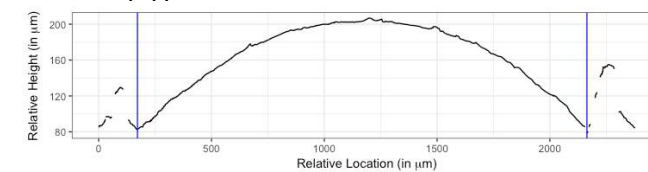
Step 3: Fit curvature



Identify matching region

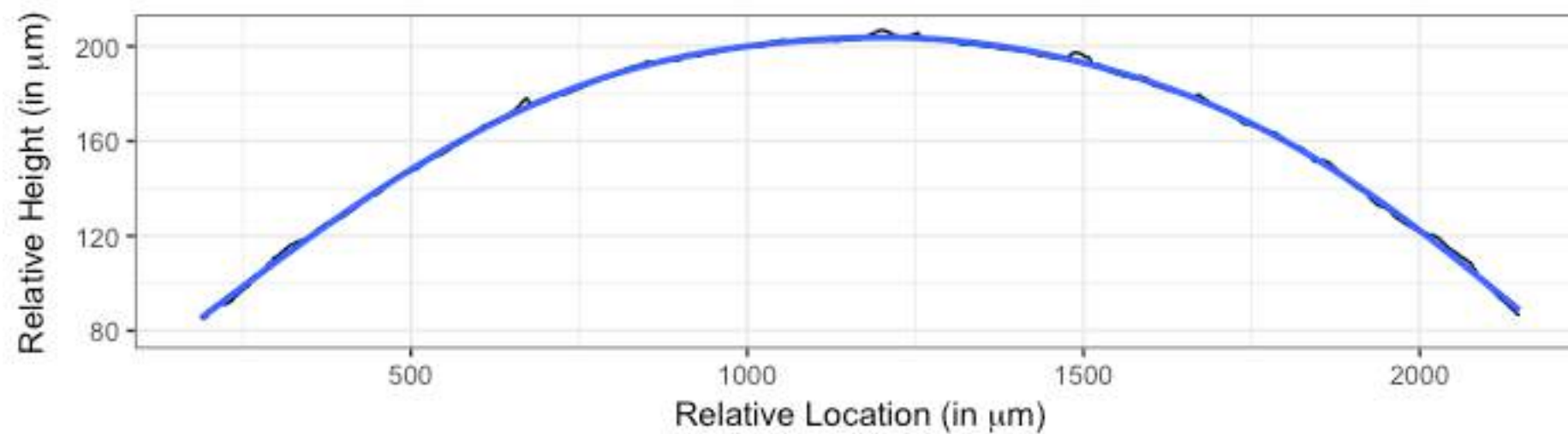


Identify groove locations



Automatic matching score

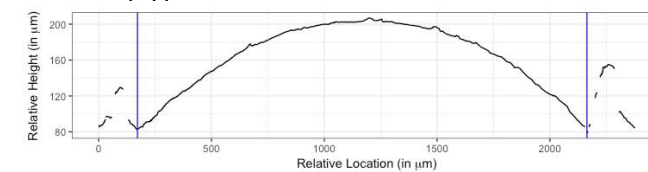
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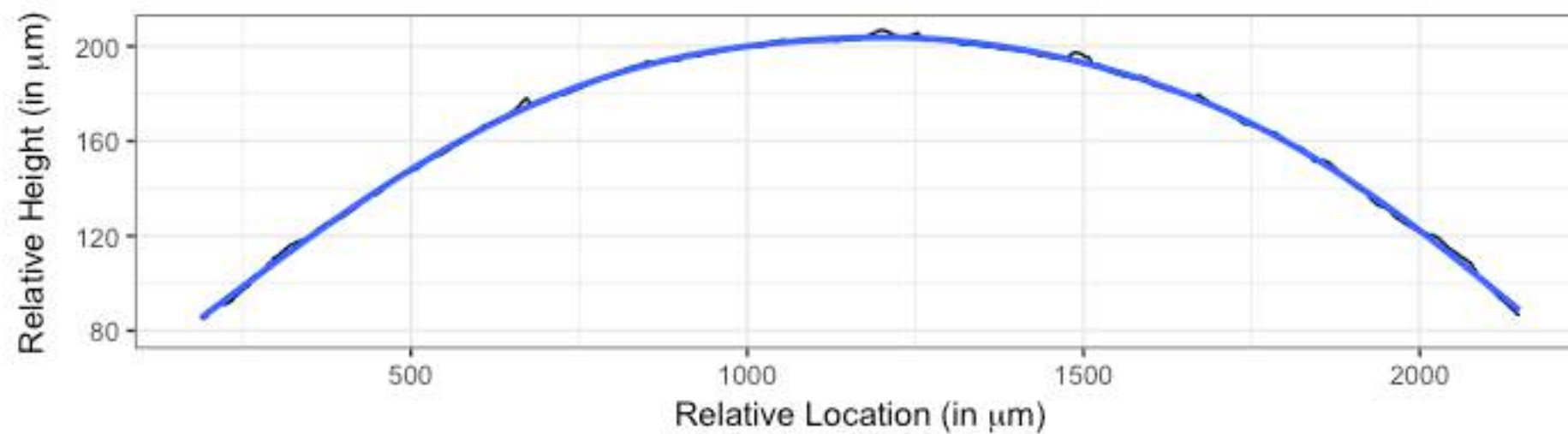


Identify groove locations



Automatic matching score

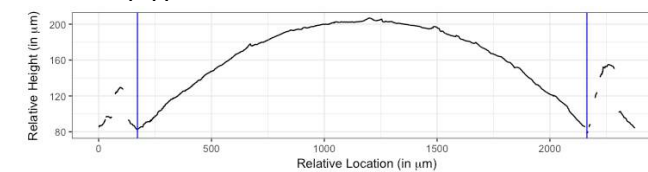
Step 3: Fit curvature & get signature



Identify matching region

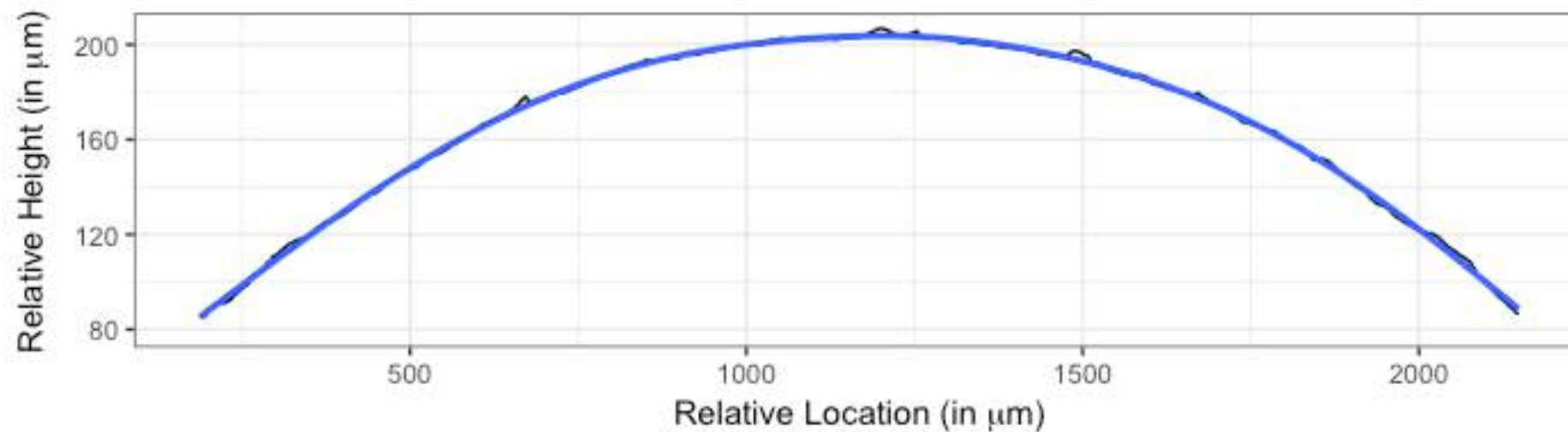


Identify groove locations



Automatic matching score

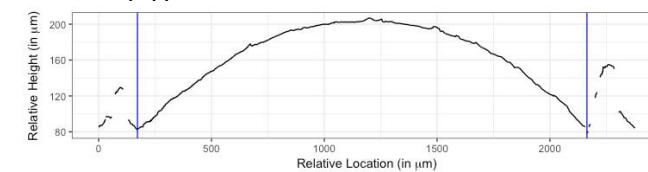
Step 3: Fit curvature & get signature



Identify matching region

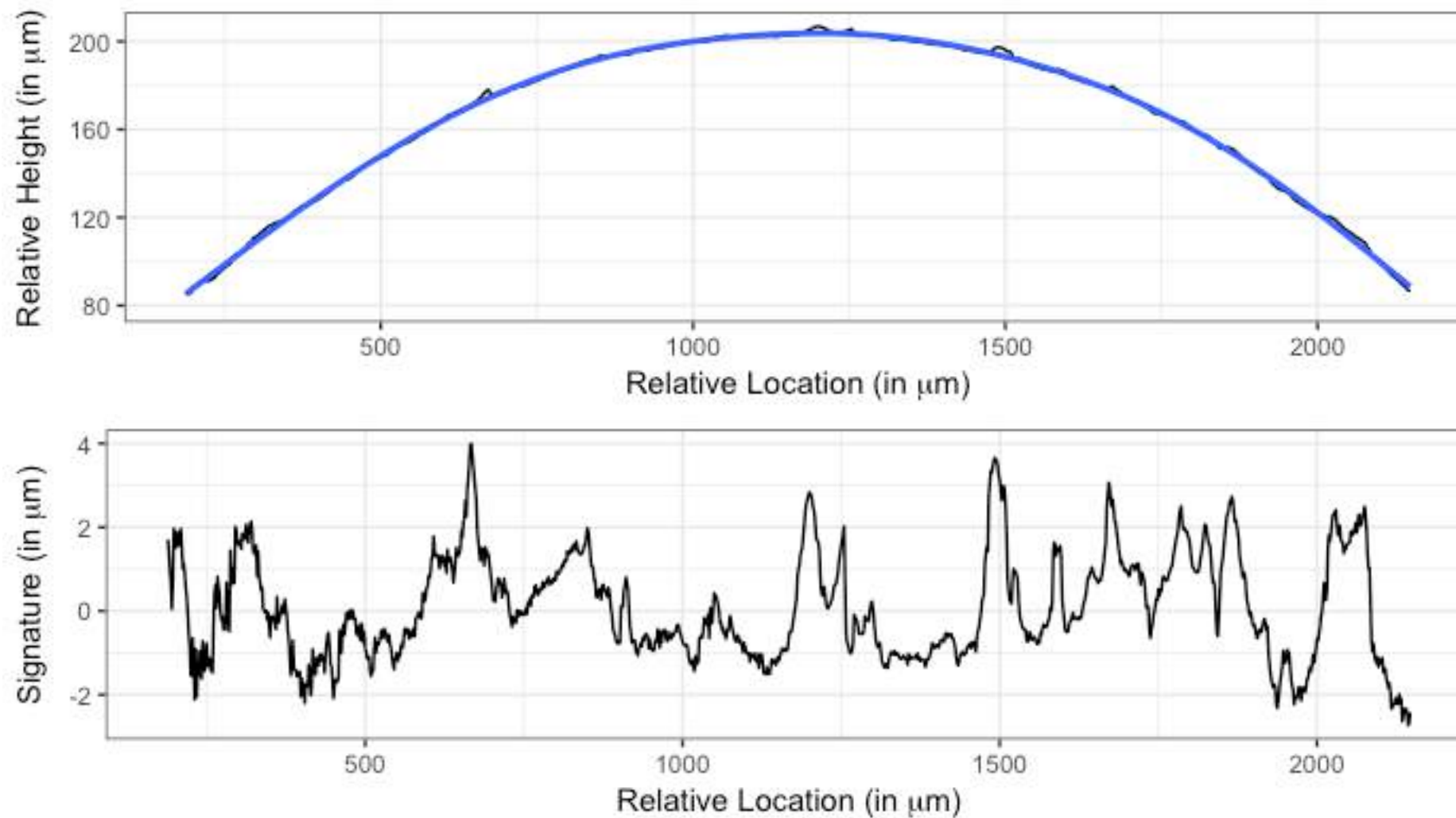


Identify groove locations



Automatic matching score

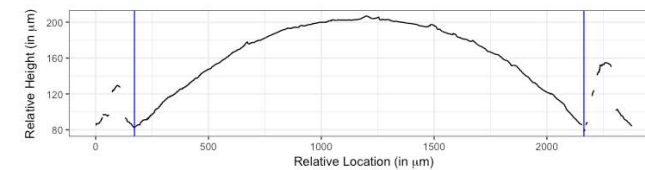
Step 3: Fit curvature & get signature



Identify matching region



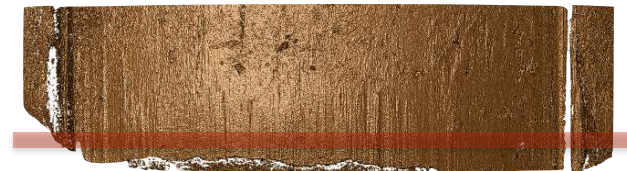
Identify groove locations



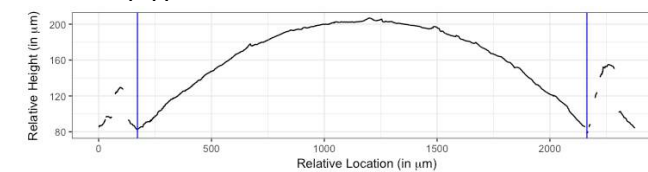
Automatic matching score

Step 4: Align signatures

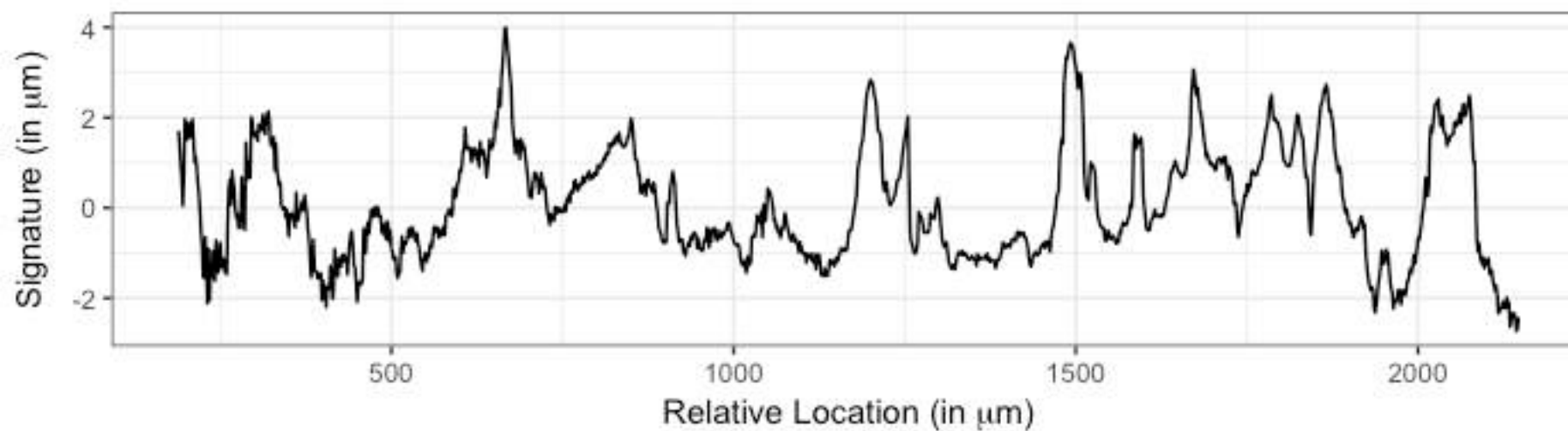
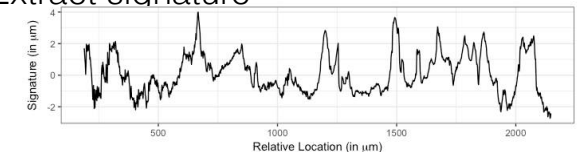
Identify matching region



Identify groove locations

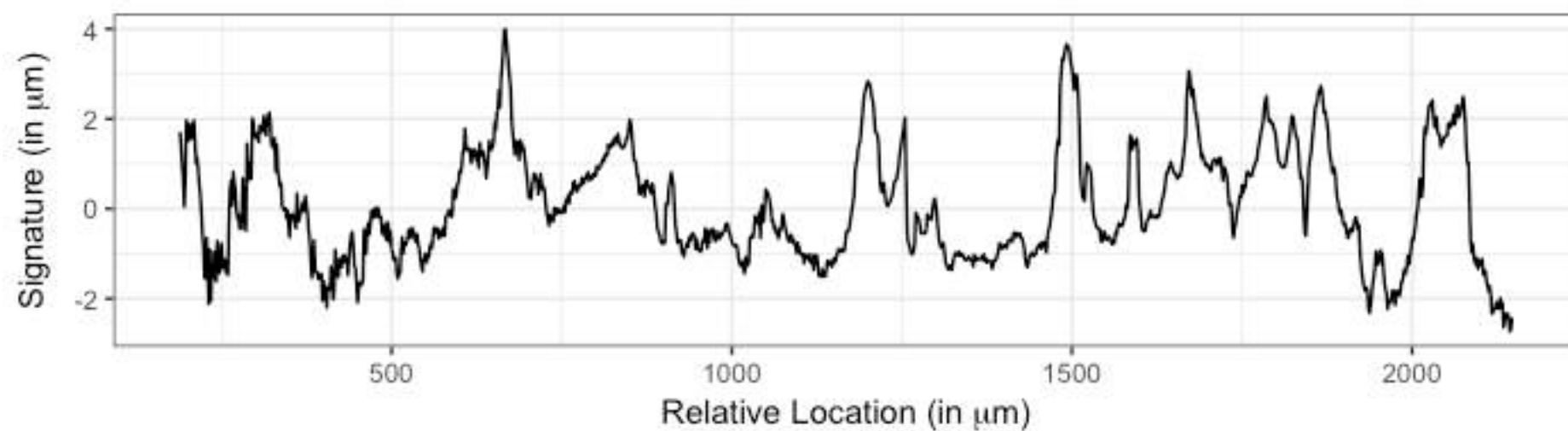
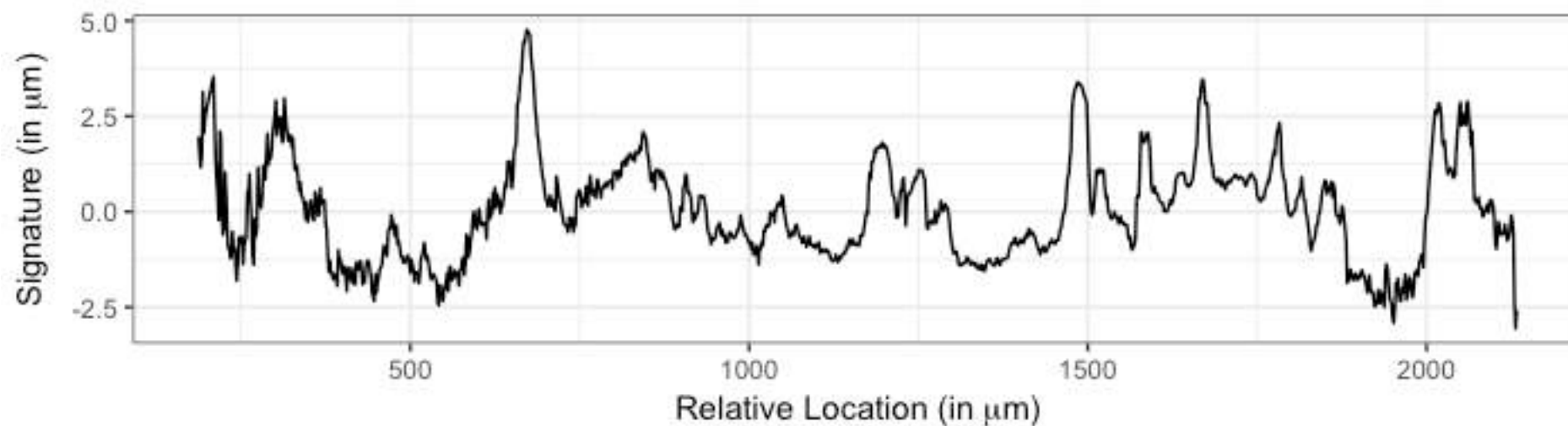


Extract signature



Automatic matching score

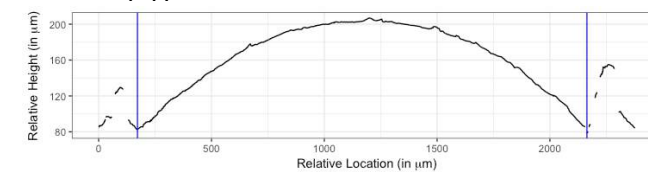
Step 4: Align signatures



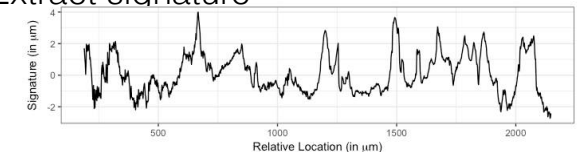
Identify matching region



Identify groove locations

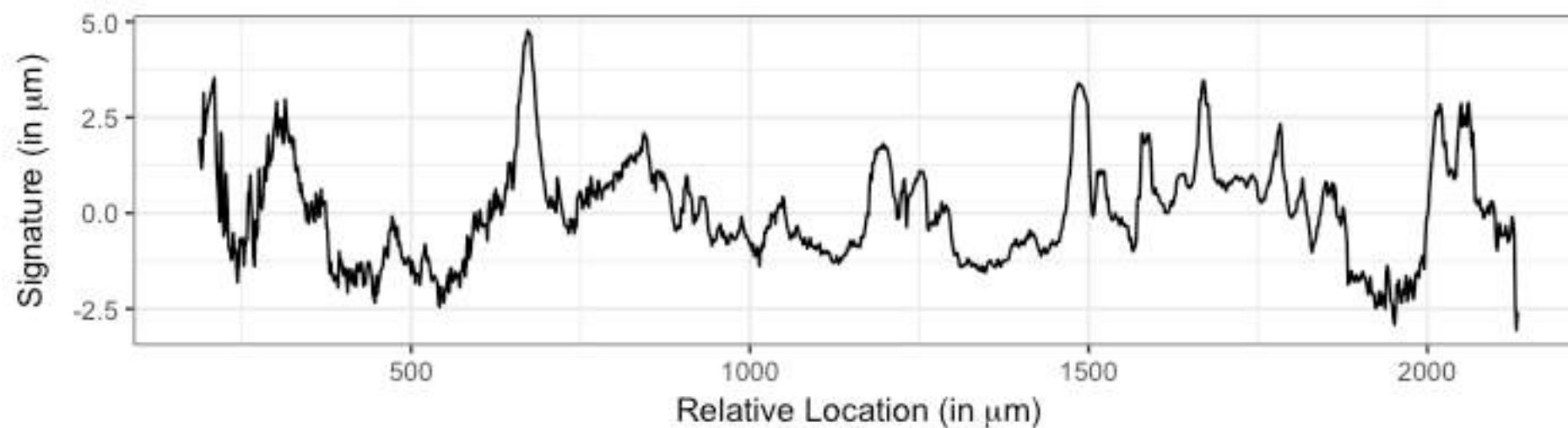


Extract signature

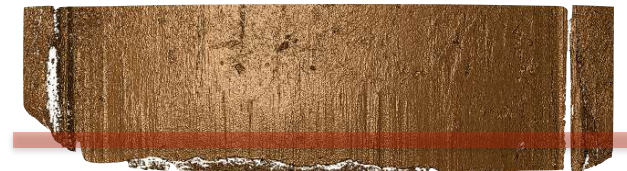


Automatic matching score

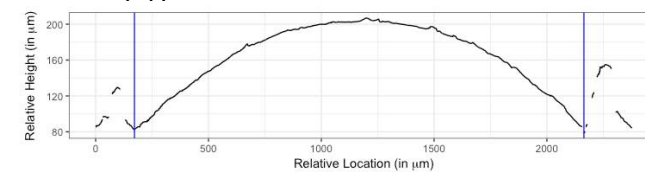
Step 4: Align signatures



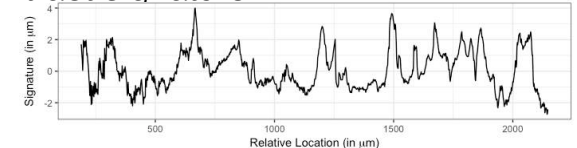
Identify matching region



Identify groove locations



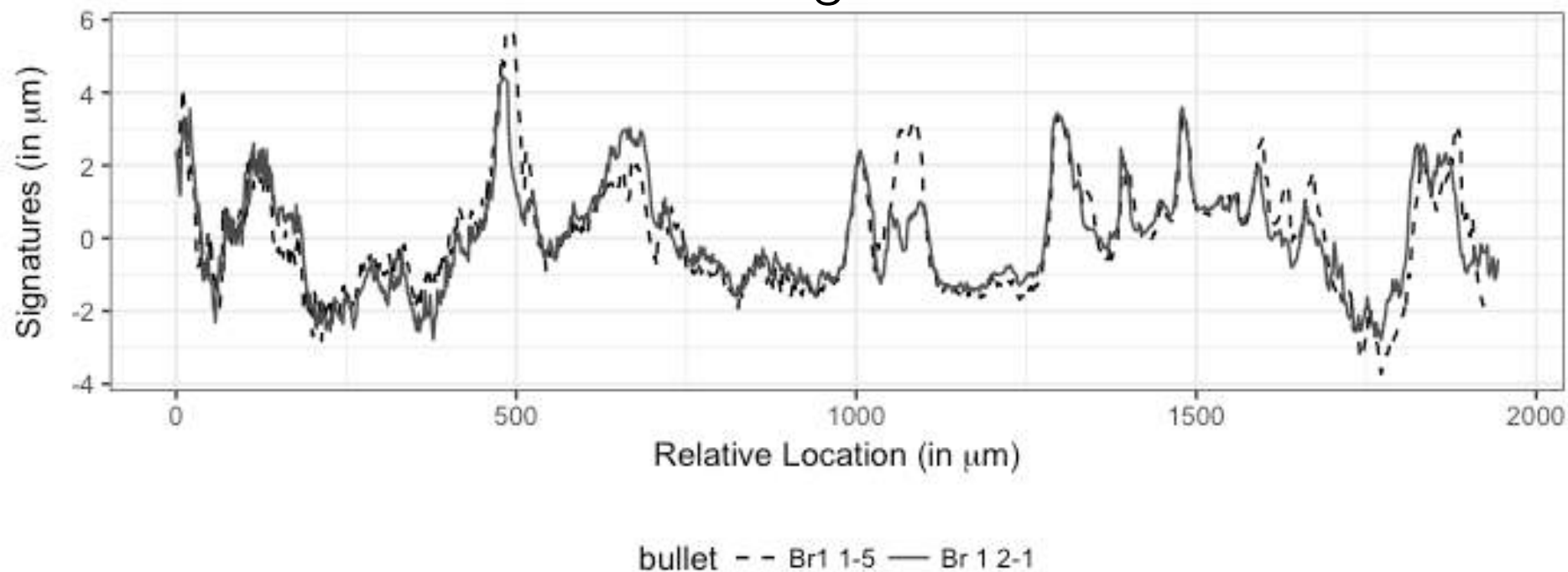
Extract signature



Automatic matching score

Step 4: Align signatures

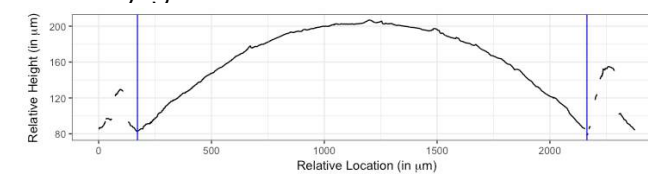
Horizontal shifts to find best alignment



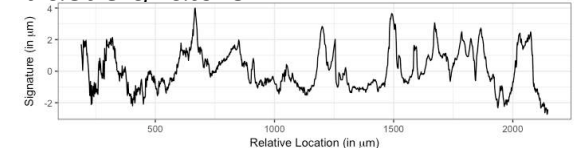
Identify matching region



Identify groove locations



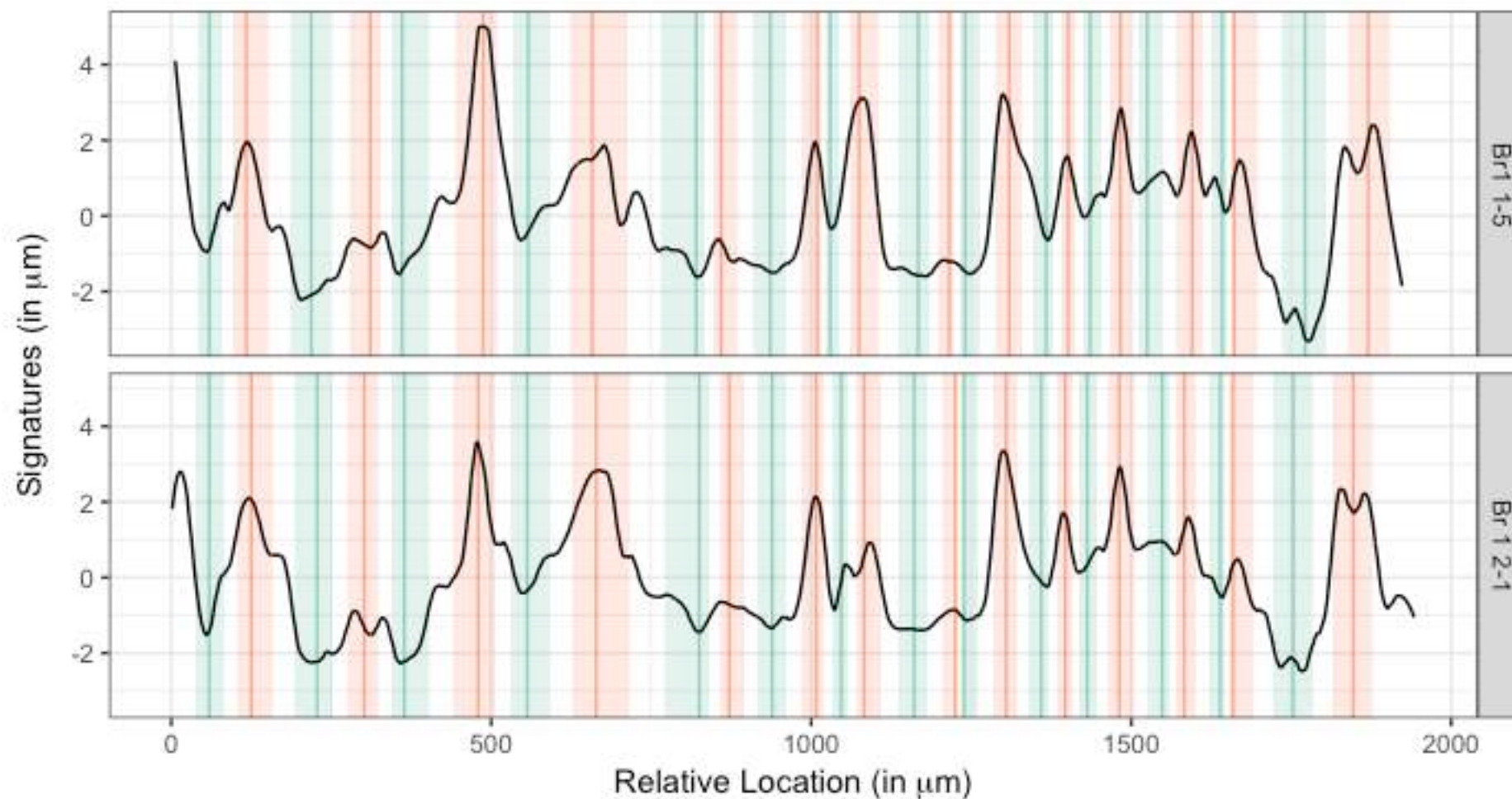
Extract signature



Automatic matching score

Step 5: Identify Peaks & Valleys

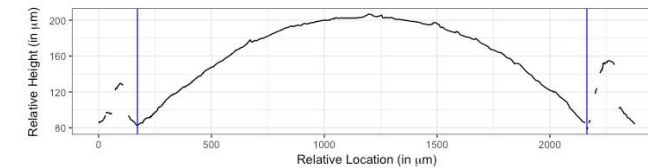
Peaks and valleys in the same locations of two lands are matching striae on the scans



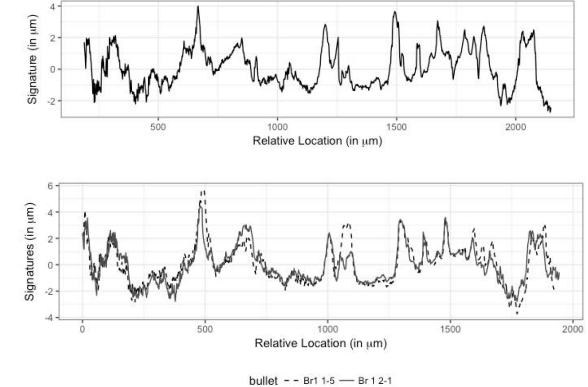
Identify matching region



Identify groove locations



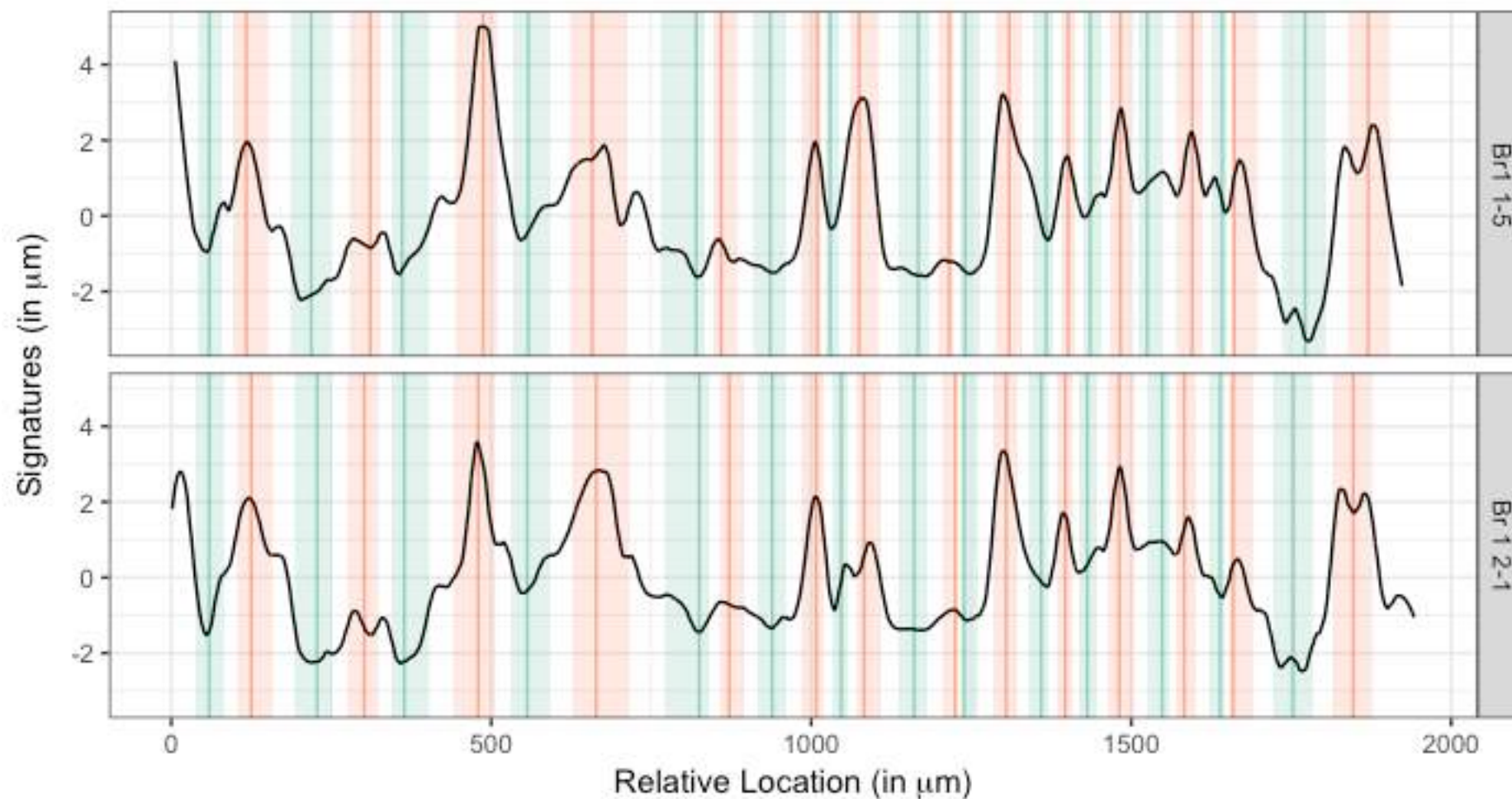
Extract signature



Automatic matching score

Step 6: Extract features

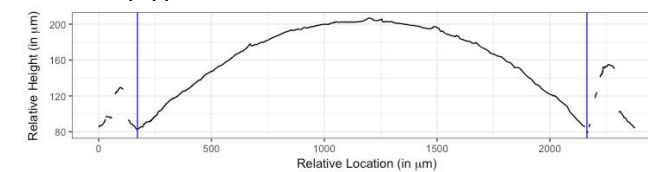
Feature should distinguish between a match and a non-match



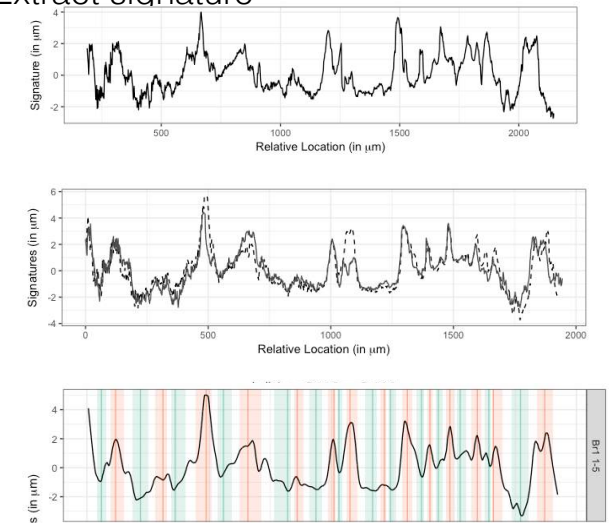
Identify matching region



Identify groove locations



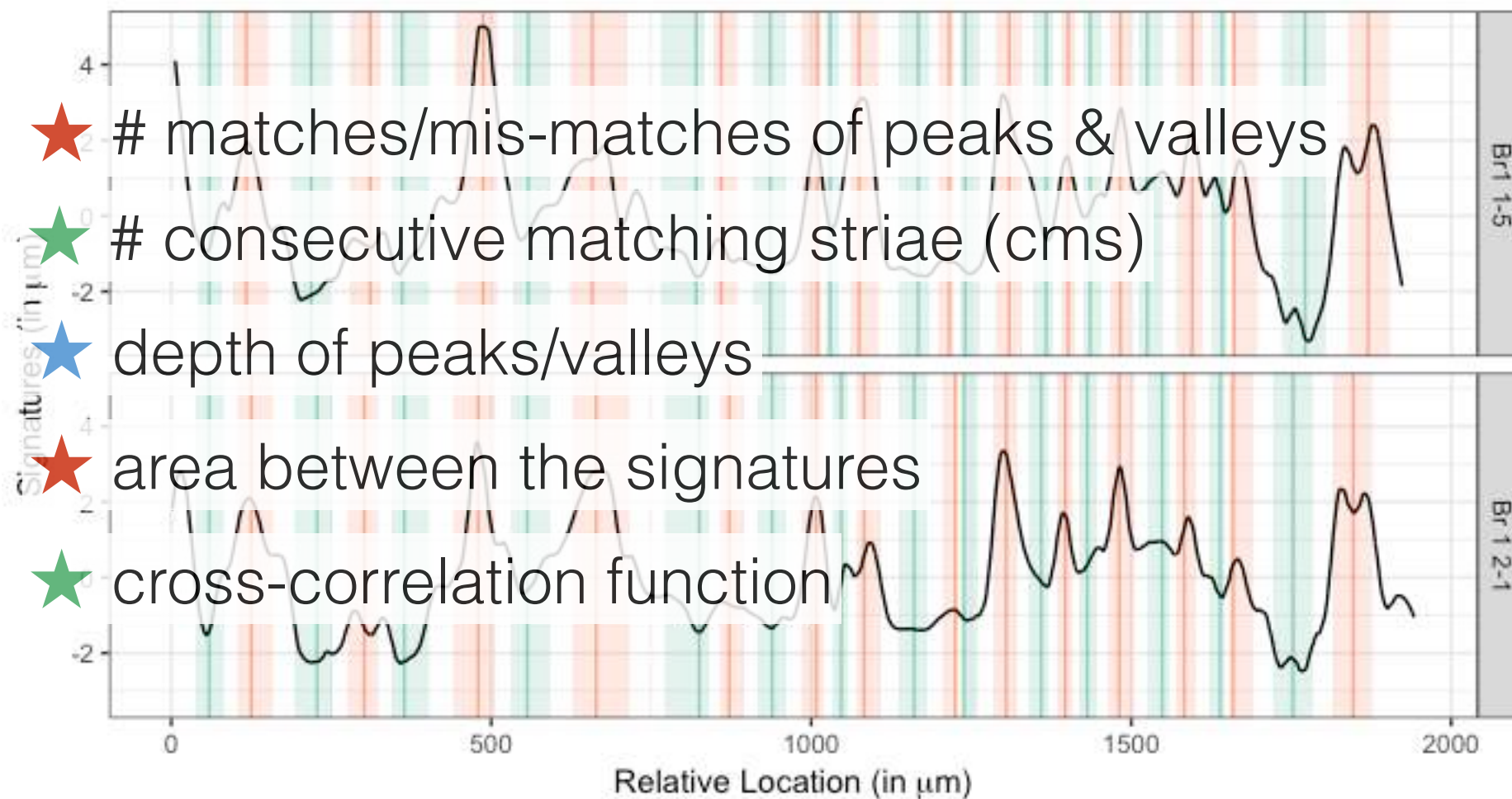
Extract signature



Automatic matching score

Step 6: Extract features

Feature should distinguish between a match and a non-match

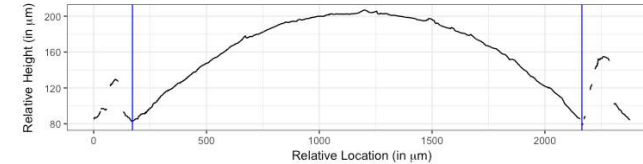


- ★ # matches/mis-matches of peaks & valleys
- ★ # consecutive matching striae (cms)
- ★ depth of peaks/valleys
- ★ area between the signatures
- ★ cross-correlation function

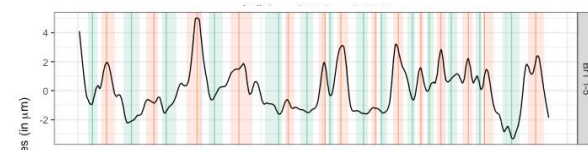
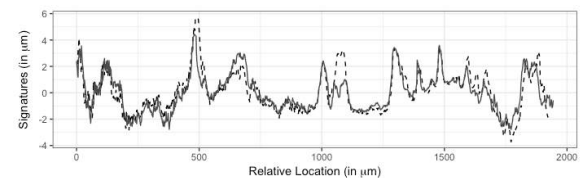
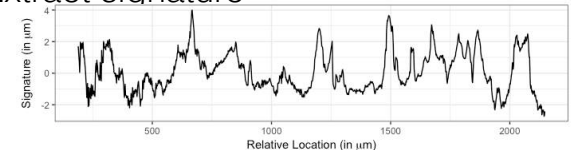
Identify matching region



Identify groove locations



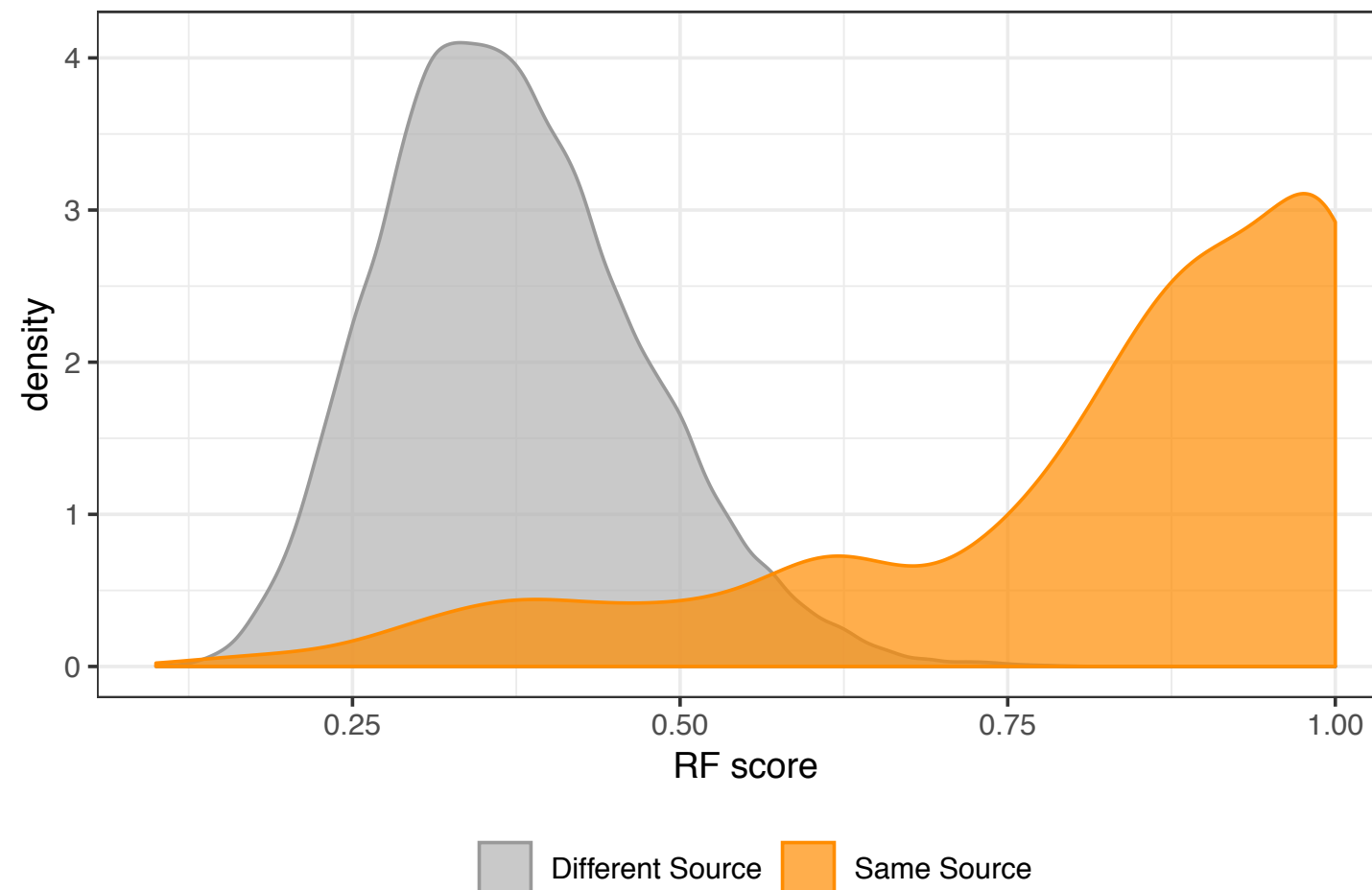
Extract signature



Automatic matching score

Step 7: Assign Score (Random Forest Model)

- ★ Matching score between 0 and 1
- ★ Higher score indicates more similarity between two lands

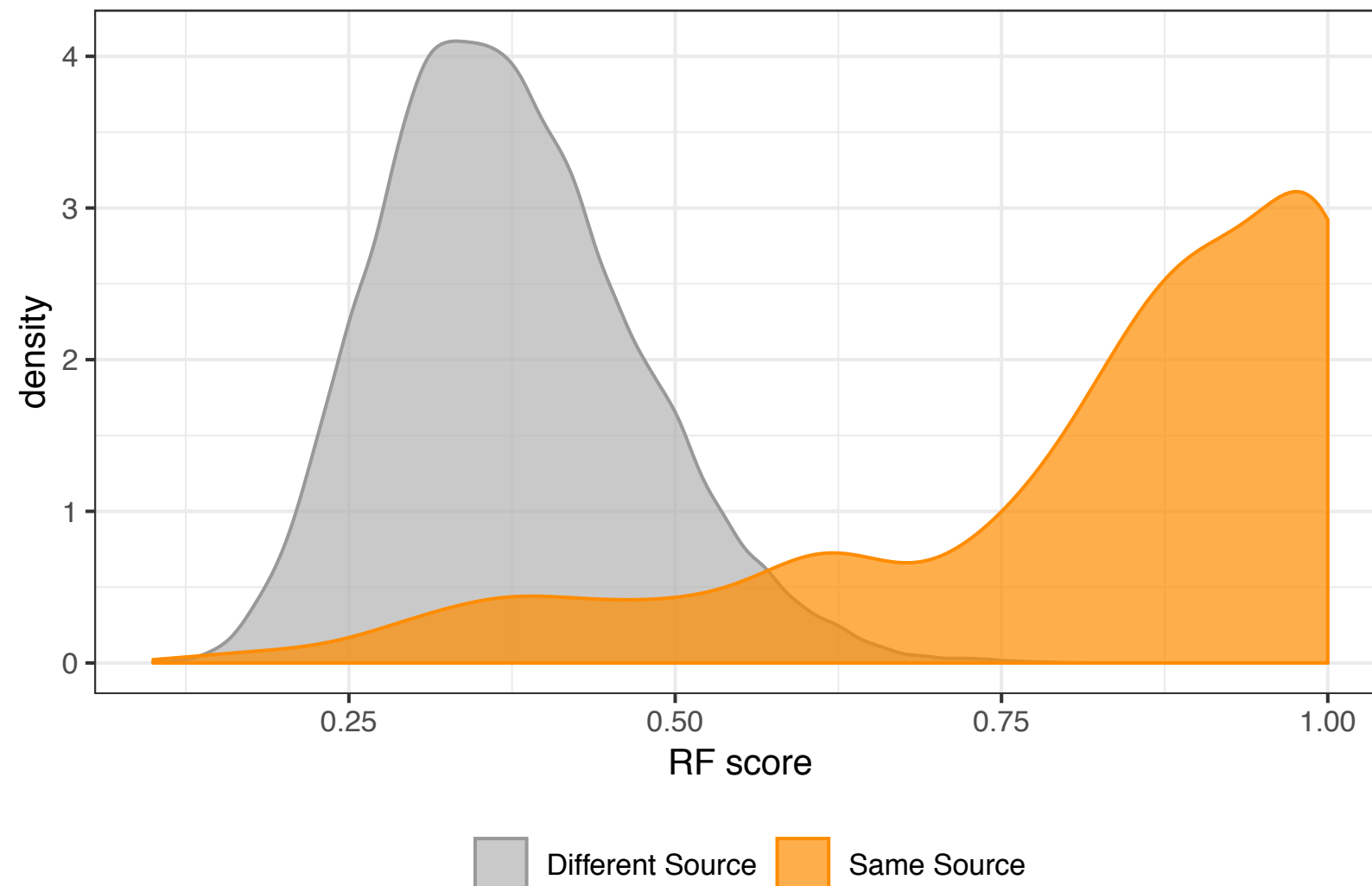


*Automatic matching of
bullet land impressions,*
Annals of Applied
Statistics,
Eric Riemer Hare, Heike
Hofmann, and Alicia
Carriquiry

*Algorithmic approaches to
match degraded land
impressions*
Eric Hare; Heike Hofmann;
Alicia Carriquiry
Law, Probability and Risk,
Volume 16, Issue 4, 1
December 2017, 203–221,
[https://doi.org/10.1093/lpr/
mgx018](https://doi.org/10.1093/lpr/mgx018)

What does it mean?

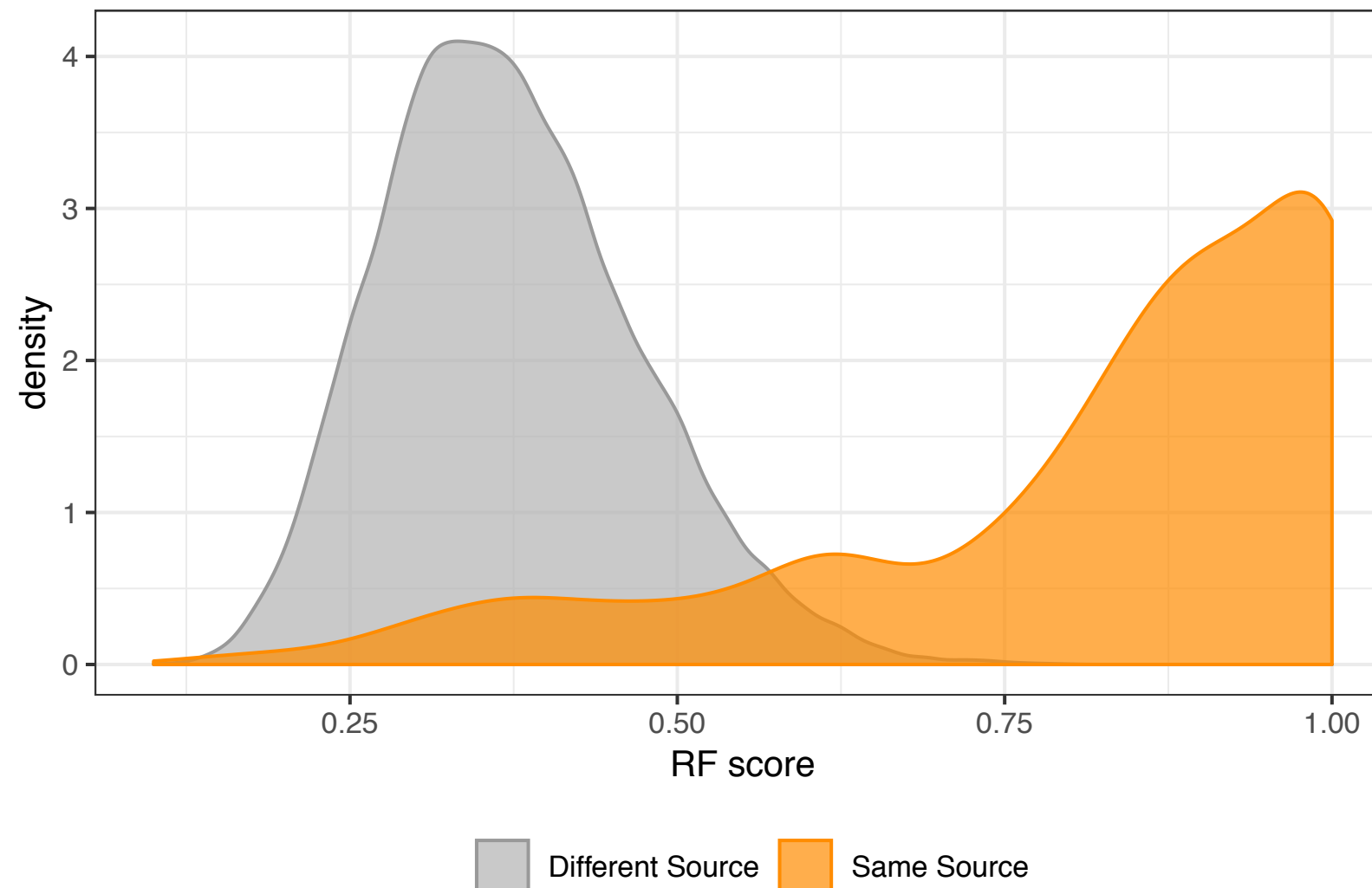
- ★ let's assume that two bullet (land)s are compared, the model spits out a number ...



What does it mean?

★ let's assume that two bullet (land)s are compared, the model spits out a number ...

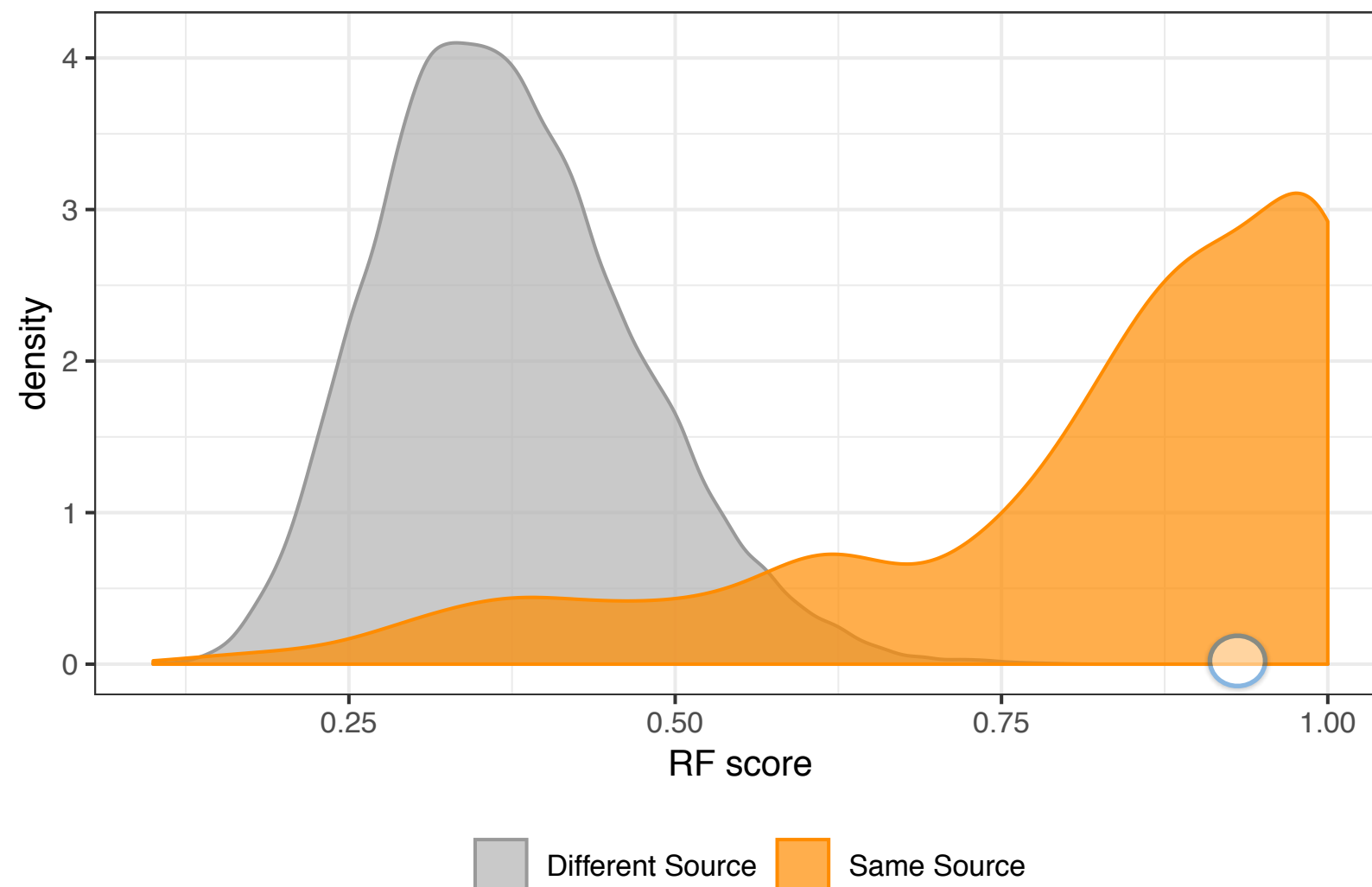
○ 0.91



What does it mean?

★ let's assume that two bullet (land)s are compared, the model spits out a number ...

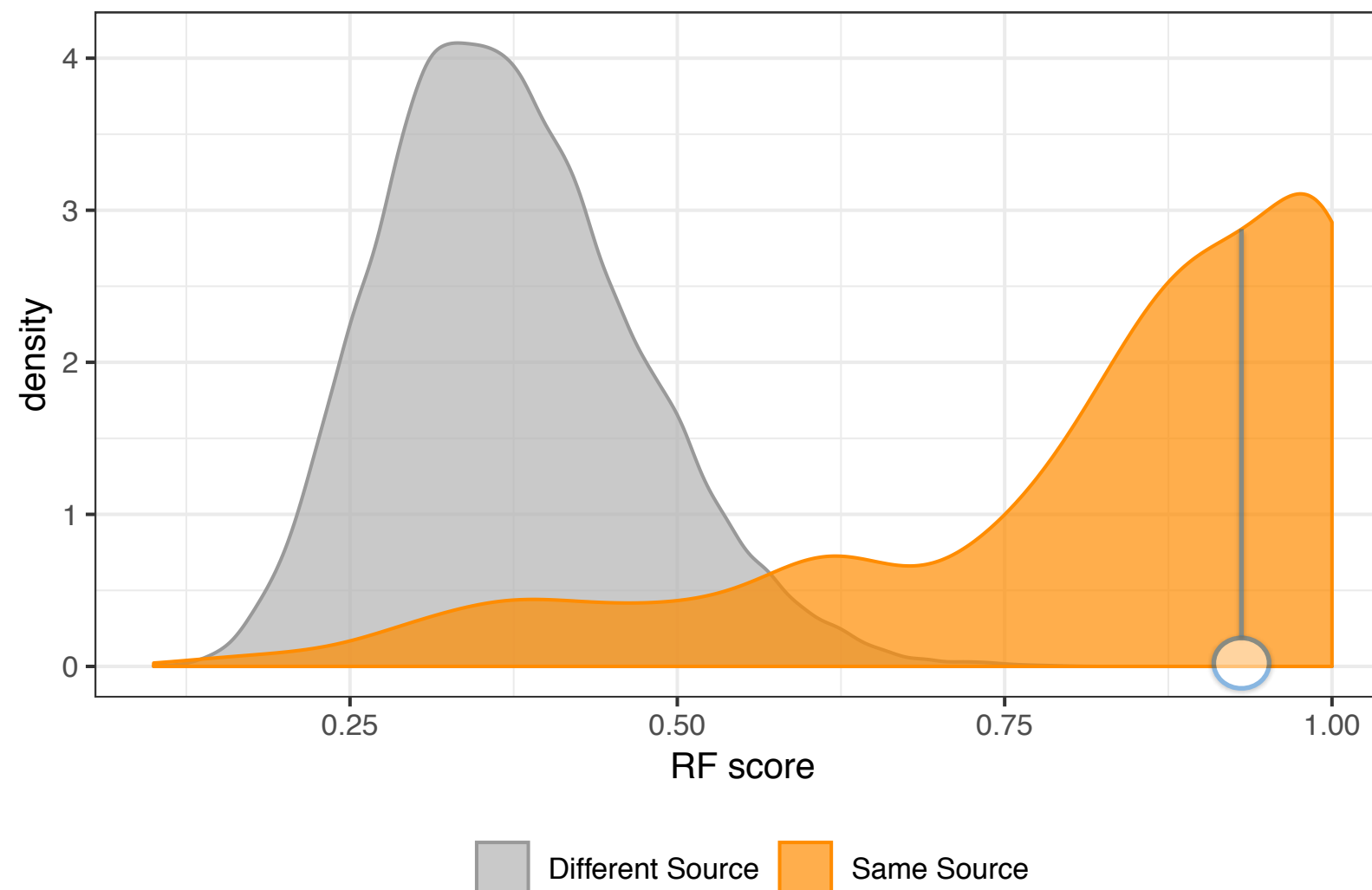
○ 0.91



What does it mean?

★ let's assume that two bullet (land)s are compared, the model spits out a number ...

○ 0.91

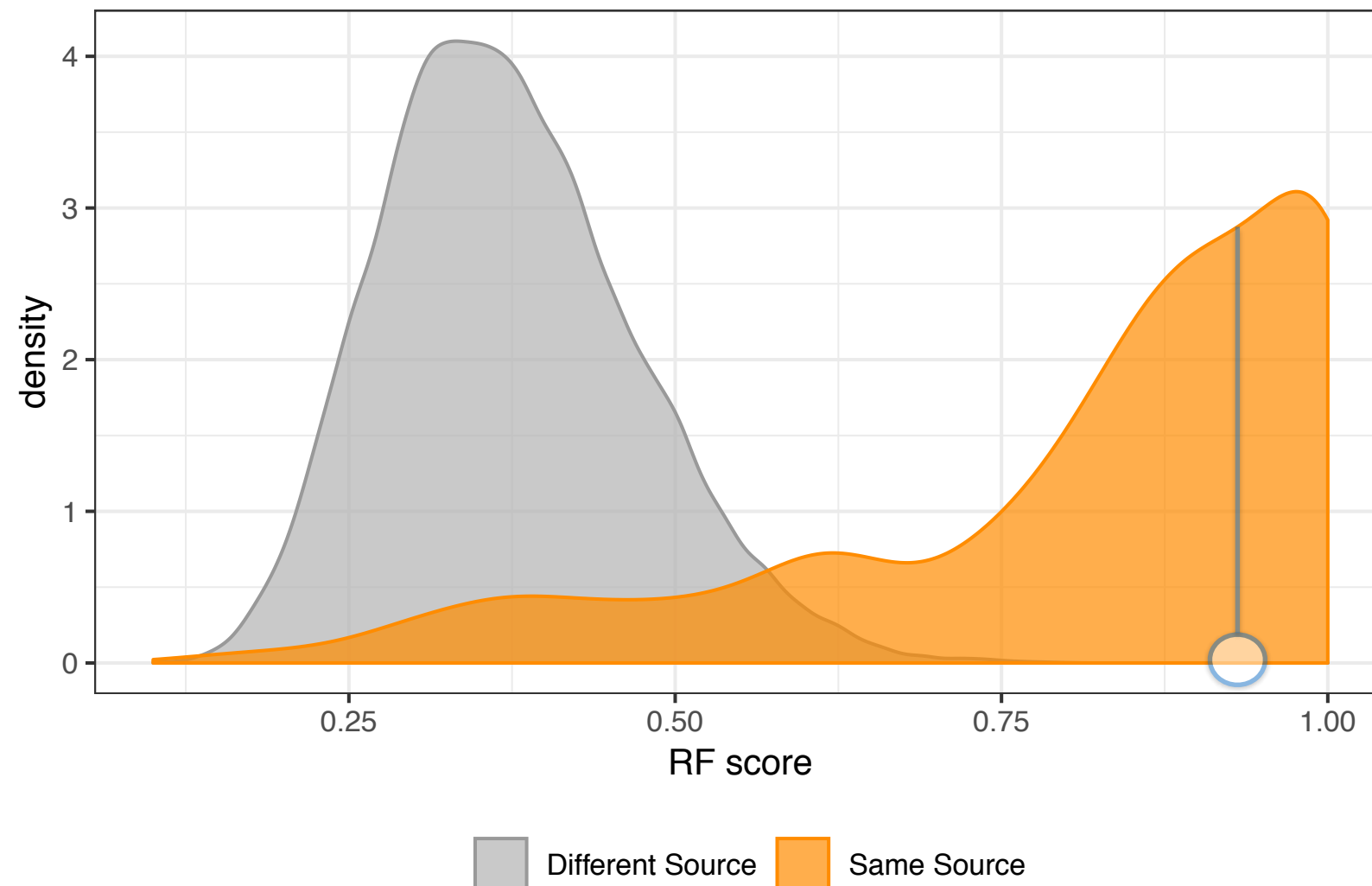


What does it mean?

★ let's assume that two bullet (land)s are compared, the model spits out a number ...

○ 0.91

2.77

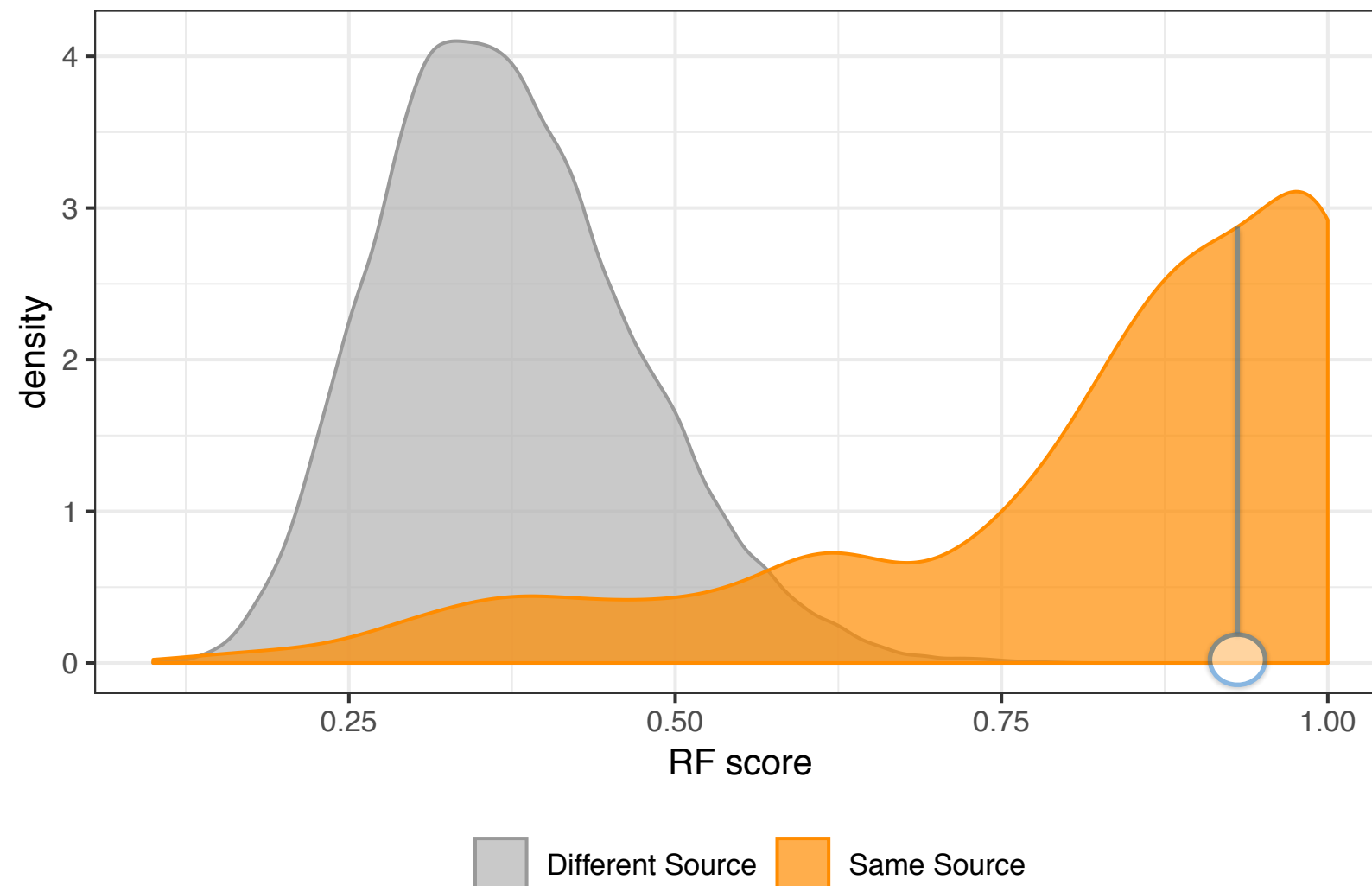


What does it mean?

★ let's assume that two bullet (land)s are compared, the model spits out a number ...

○ 0.91

2.77 1.6×10^{-16}

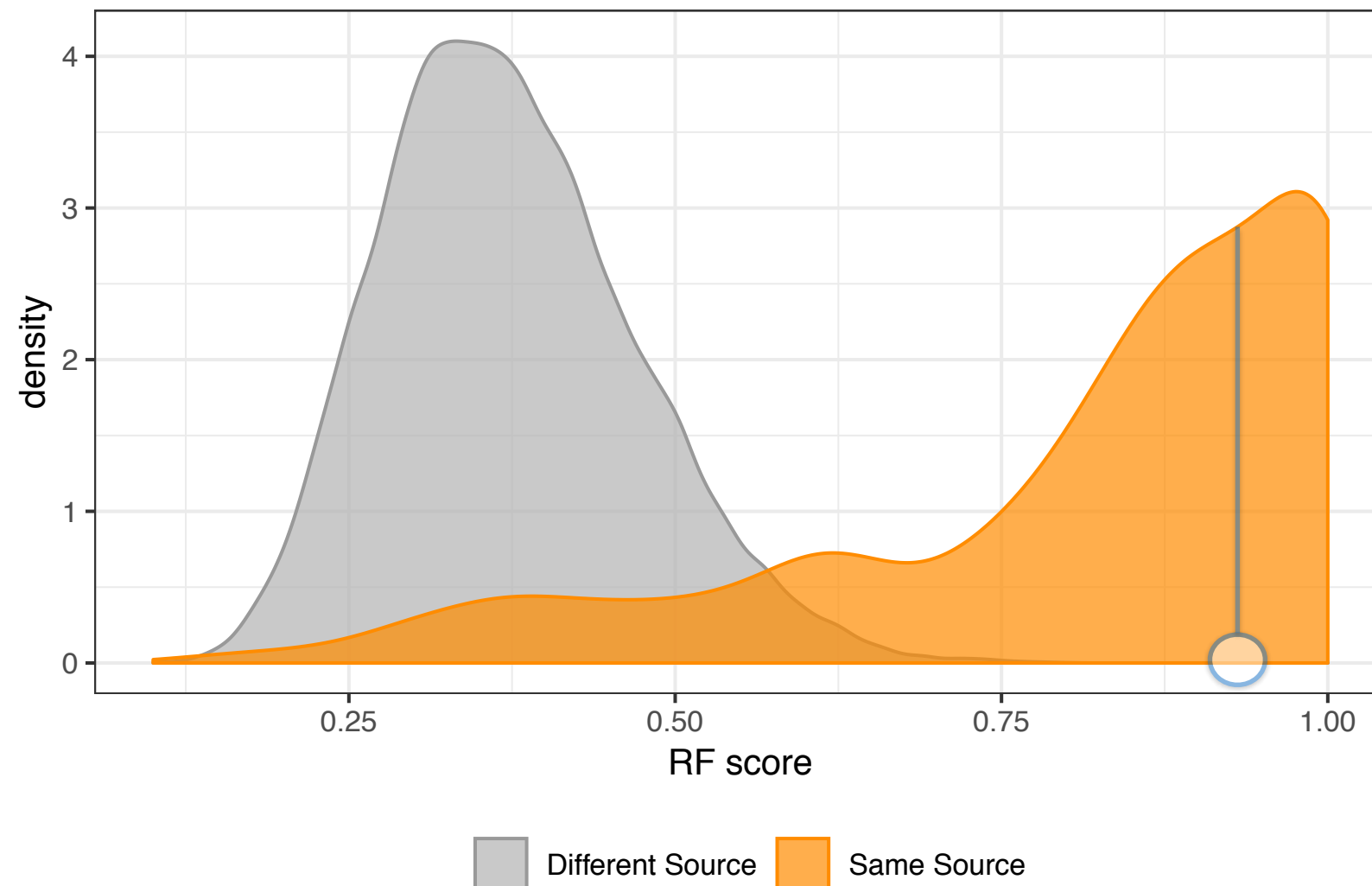


What does it mean?

★ let's assume that two bullet (land)s are compared, the model spits out a number ...

○ 0.91

$$2.77 / 1.6 \times 10^{-16} = 1.7 \times 10^{16}$$

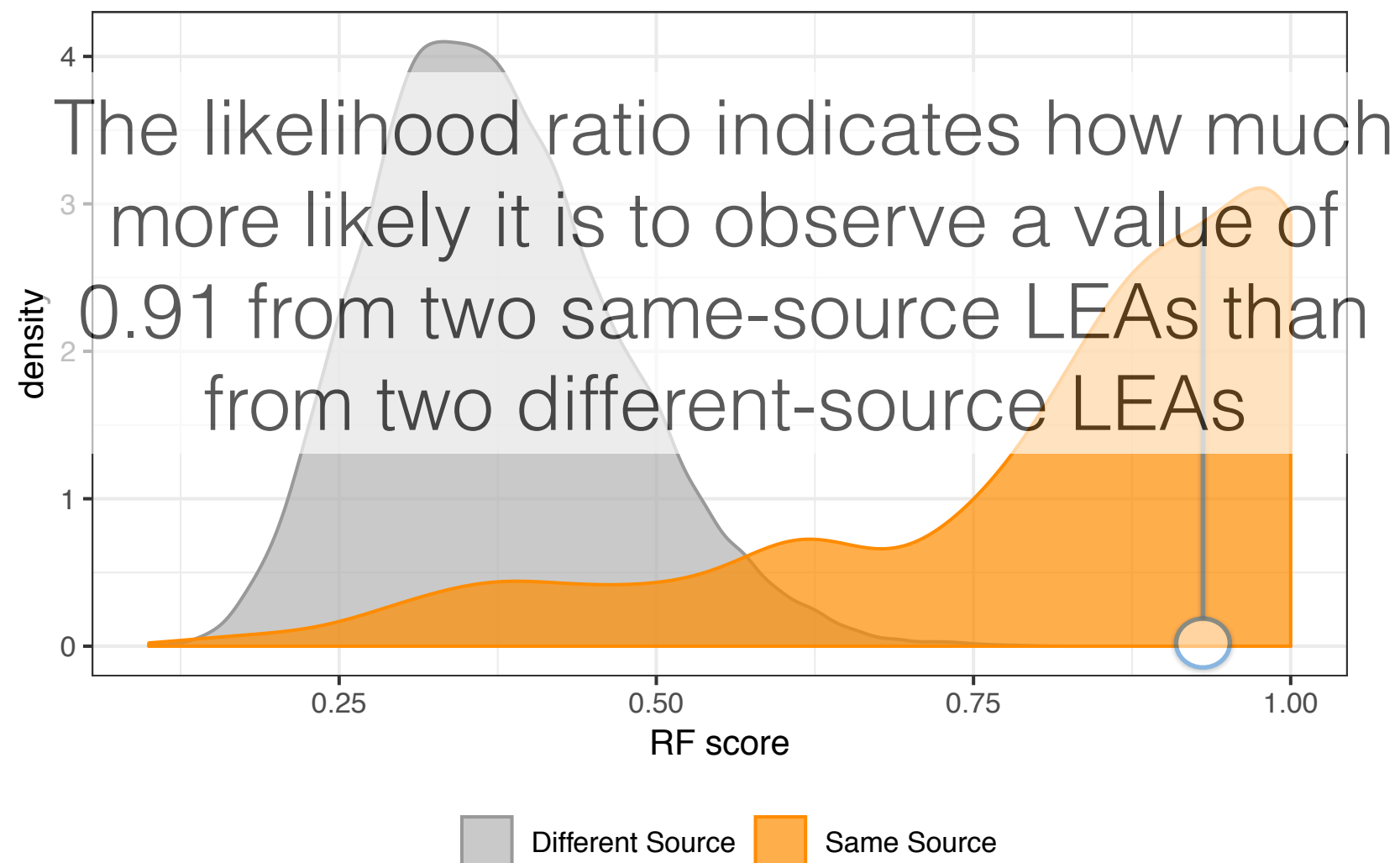


What does it mean?

★ let's assume that two bullet (land)s are compared, the model spits out a number ...

○ 0.91

$$2.77 / 1.6 \times 10^{-16} = 1.7 \times 10^{16}$$

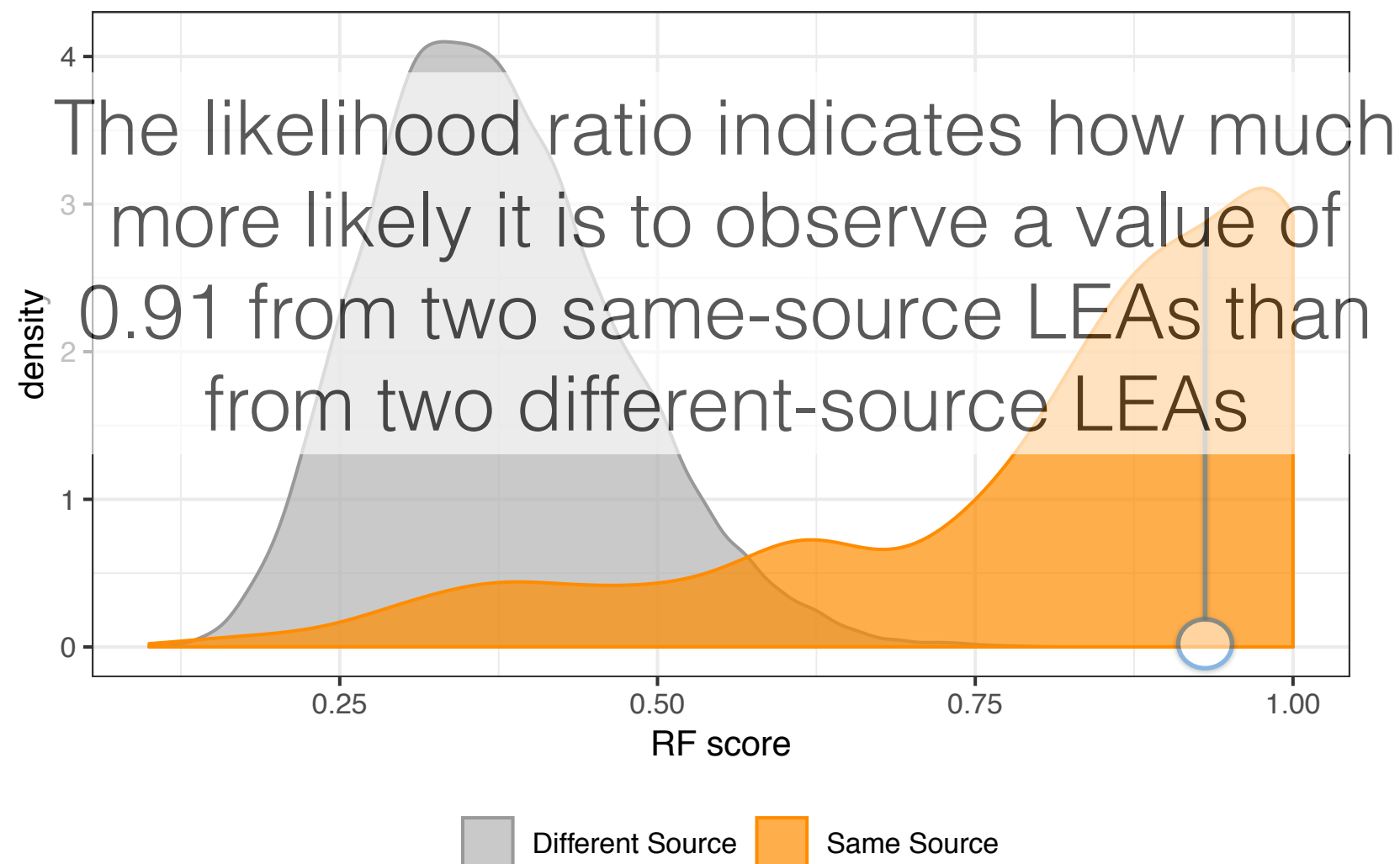


What does it mean?

★ let's assume that two bullet (land)s are compared, the model spits out a number ...

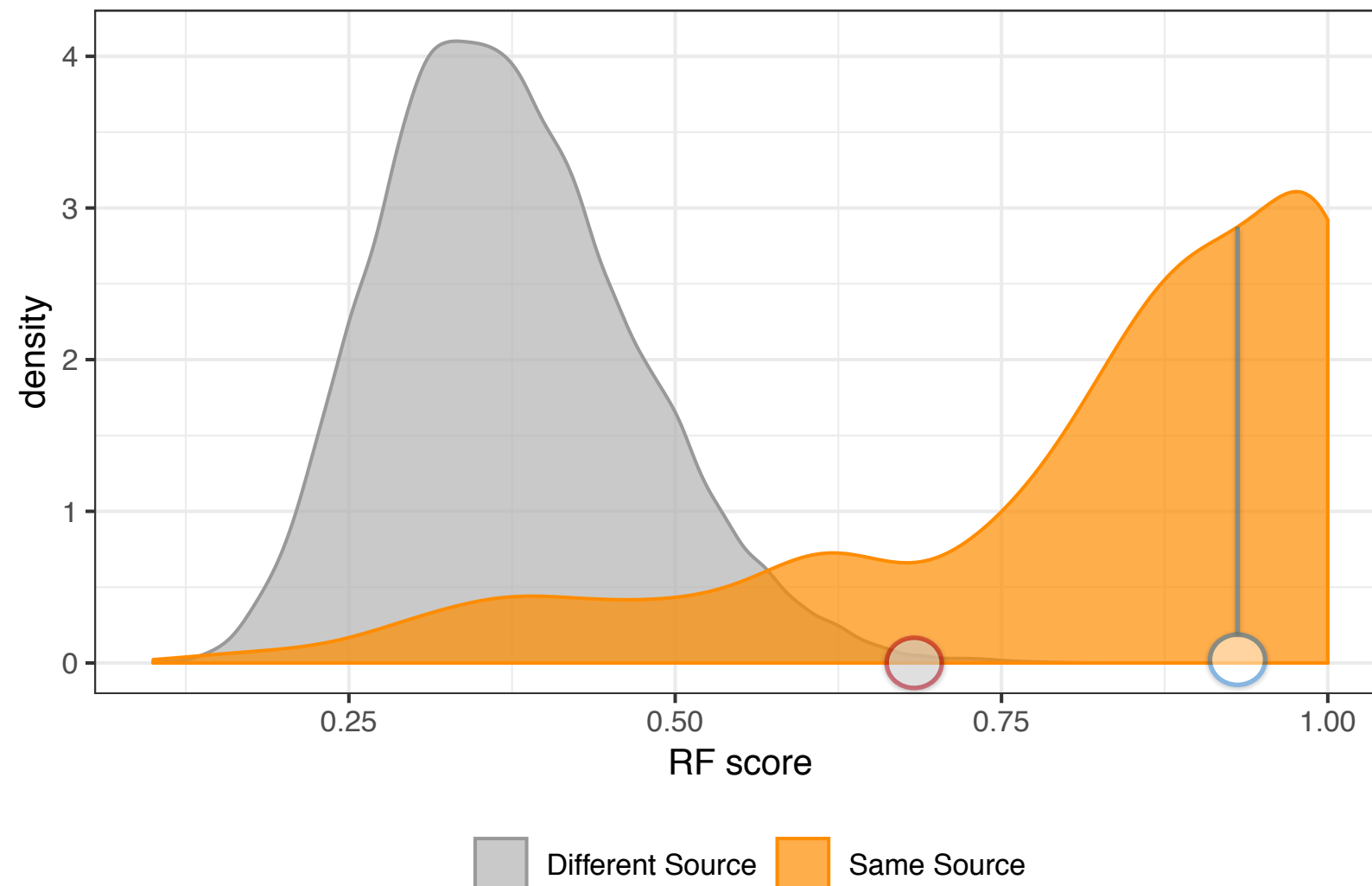
○ 0.91

$2.77 / 1.6 \times 10^{-16} = 1.7 \times 10^{16}$
supports identification



What does it mean?

★ let's assume that two bullet (land)s are compared, the model spits out a number ...



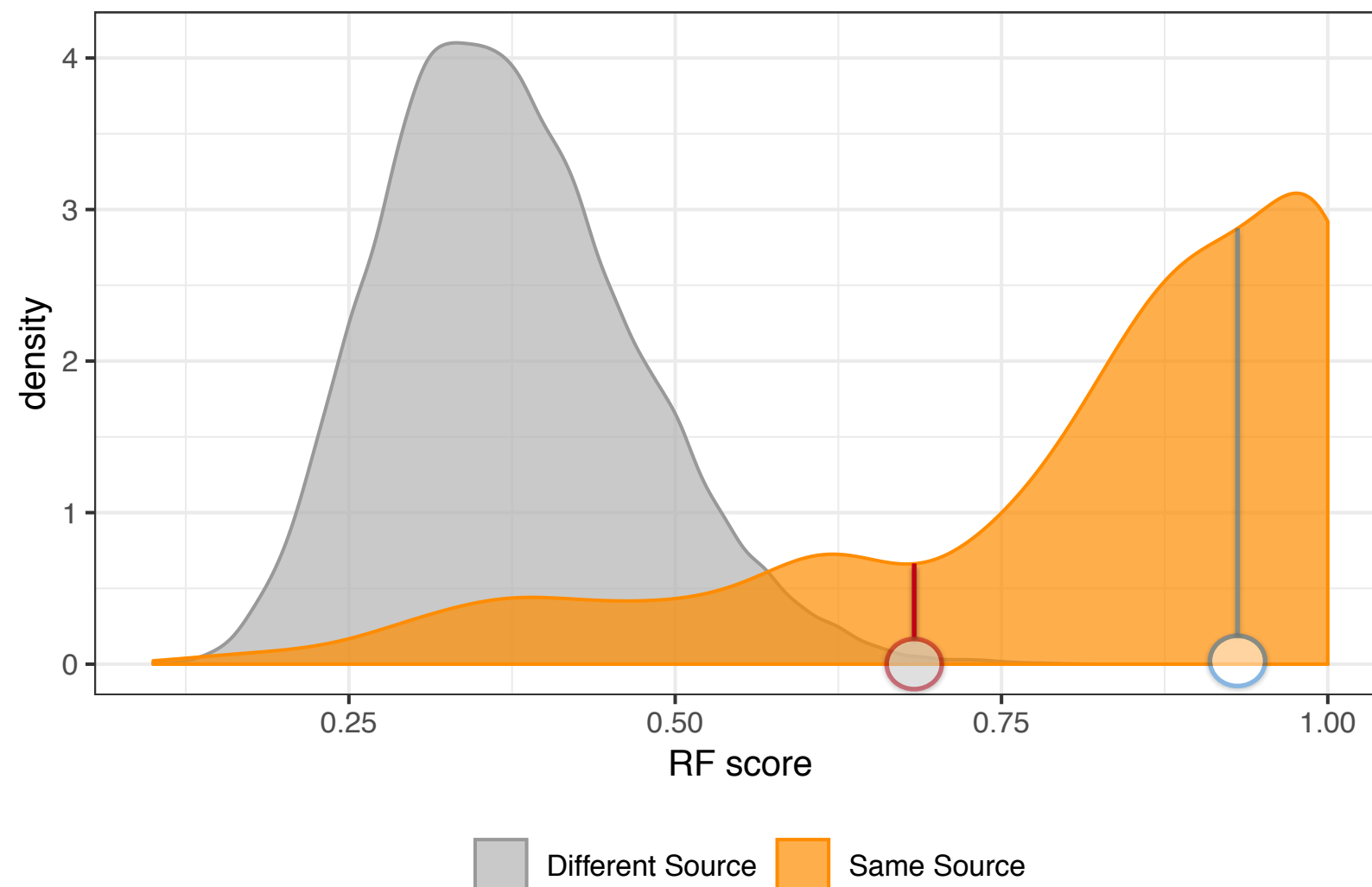
○ 0.91

$2.77 / 1.6 \times 10^{-16} = 1.7 \times 10^{16}$
supports identification

○ 0.69

What does it mean?

★ let's assume that two bullet (land)s are compared, the model spits out a number ...



○ 0.91

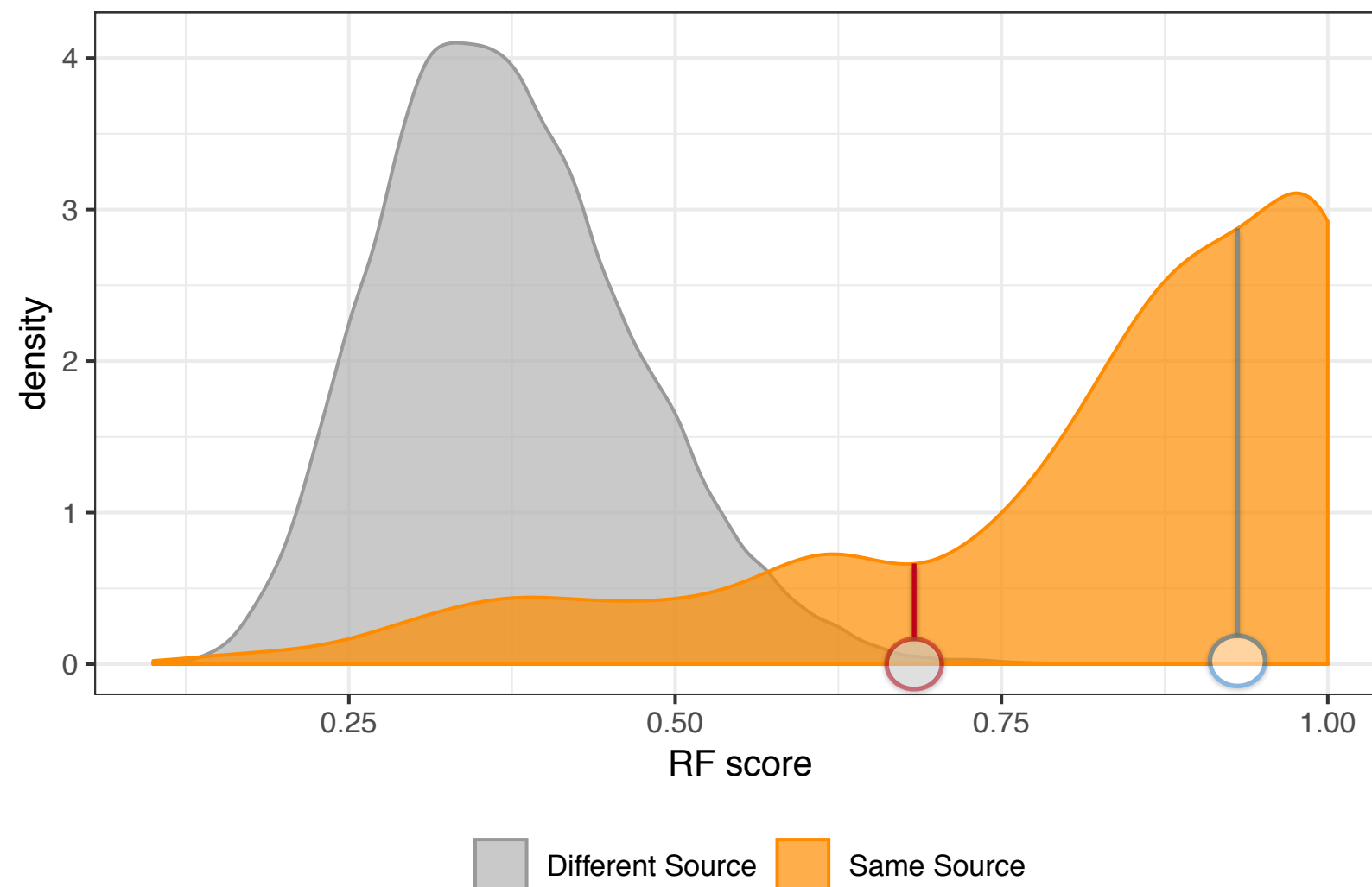
$2.77 / 1.6 \times 10^{-16} = 1.7 \times 10^{16}$
supports identification

○ 0.69

$0.67 / 0.05 = 13.4$

What does it mean?

★ let's assume that two bullet (land)s are compared, the model spits out a number ...



○ 0.91

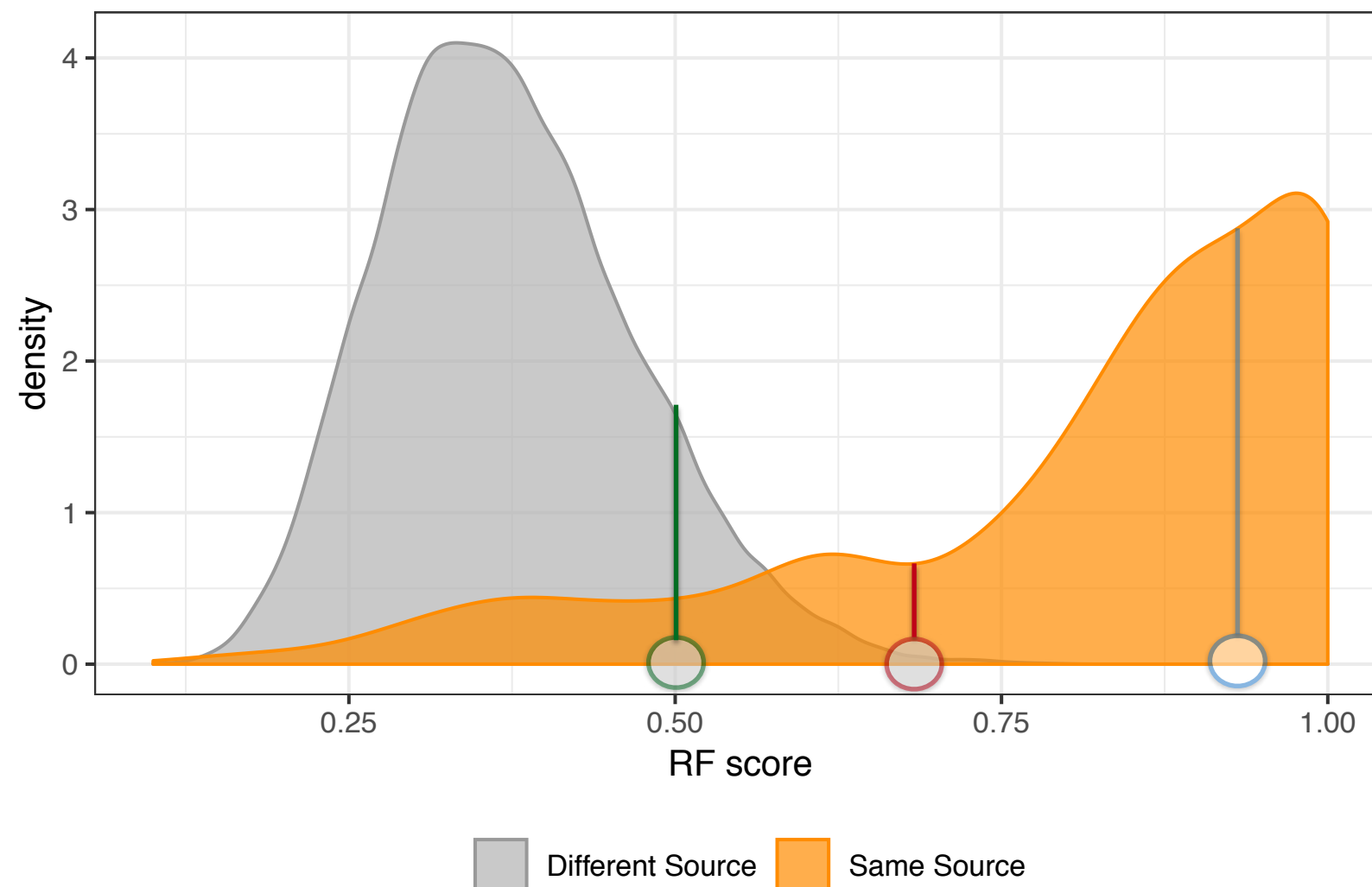
$2.77 / 1.6 \times 10^{-16} = 1.7 \times 10^{16}$
supports identification

○ 0.69

$0.67 / 0.05 = 13.4$
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What does it mean?

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○ 0.91

$2.77 / 1.6 \times 10^{-16} = 1.7 \times 10^{16}$
supports identification

○ 0.69

$0.67 / 0.05 = 13.4$
supports identification

○ 0.50

$0.43 / 1.65 = 0.26$
inconclusive

Shades of RF Scores

HS 224 to Clone HS 224

Random Forest Scores

- ★ How well do LEAs match between a bullet and its clone?

HS 224 Br 1 - Bullet 1 - Land 1



HS 224 Clone Br 1 - Bullet 1 - Land 1



HS 224 to Clone HS 224

Random Forest Scores

★ How well do LEAs match between a bullet and its clone?

HS 224 Br 1 - Bullet 1 - Land 1



HS 224 Barrel 1 - Bullet 1 - Land 2



HS 224 Clone Br 1 - Bullet 1 - Land 1



HS 224 Clone Br 1 - Bullet 1 - Land 2



HS 224 to Clone HS 224

Random Forest Scores

- ★ How well do LEAs match between a bullet and its clone?

Set	Clone		Original	
	Bullet	Bullet 1	Bullet 1	Bullet 2
Clone	Bullet 1	1.00	0.80	
	Bullet 2	0.80	0.83	
Original	Bullet 1		1.00	0.98
	Bullet 2		0.98	1.00

- ★ Original bullets match with higher RF score than clones

HS 224 to Clone HS 224

Random Forest Scores

- ★ How well do LEAs match between a bullet and its clone?

Set	Clone		Original	
	Bullet	Bullet 1	Bullet 1	Bullet 2
Clone	Bullet 1	1.00	0.80	0.97
	Bullet 2	0.80	0.83	0.78
Original	Bullet 1		1.00	0.98
	Bullet 2		0.98	1.00

- ★ Original bullets match with higher RF score than clones
- ★ Original bullets match well to clones, almost as well as to each other

HS 224 to Clone HS 224

Random Forest Scores

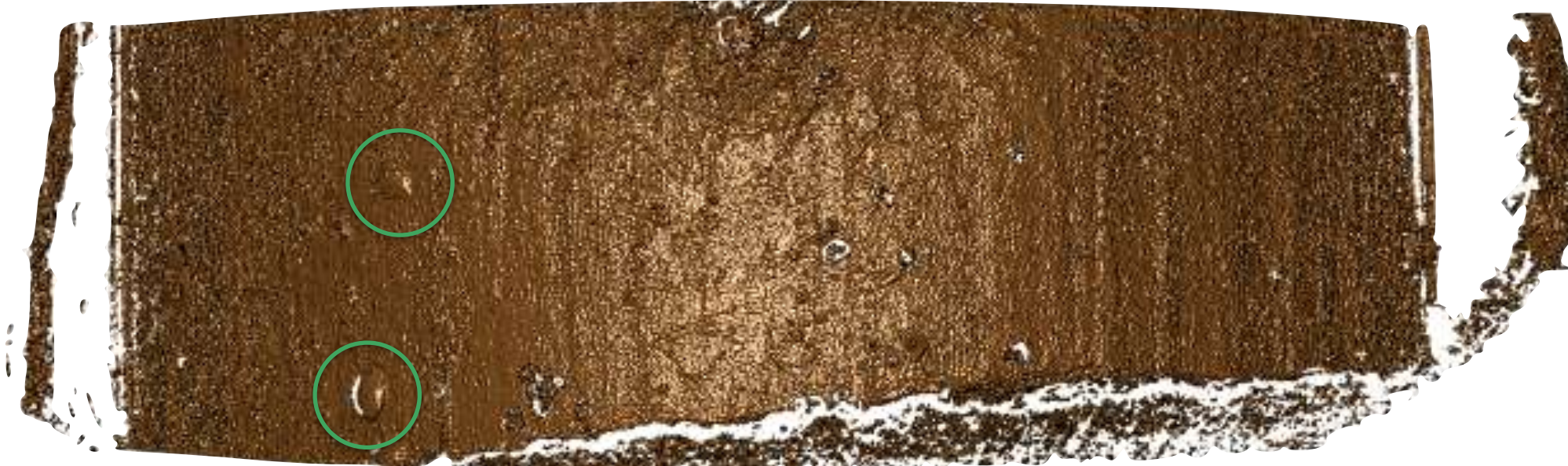
★ Some LEA scans of clones are showing artifacts:



HS 224 to Clone HS 224

Random Forest Scores

★ Some LEA scans of clones are showing artifacts:



HS 224 to Clone HS 224

Random Forest Scores

★ Some LEA scans of clones are showing artifacts:

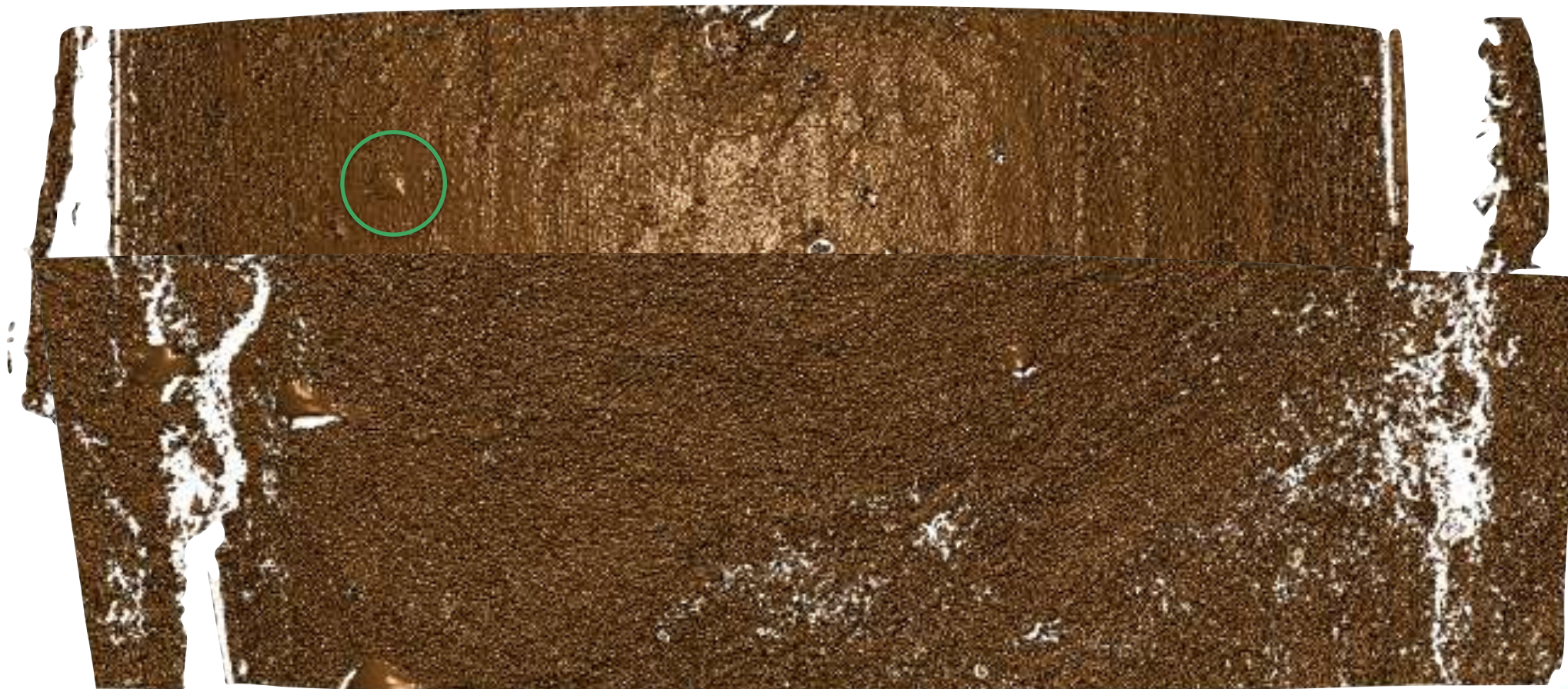


droplets

HS 224 to Clone HS 224

Random Forest Scores

★ Some LEA scans of clones are showing artifacts:

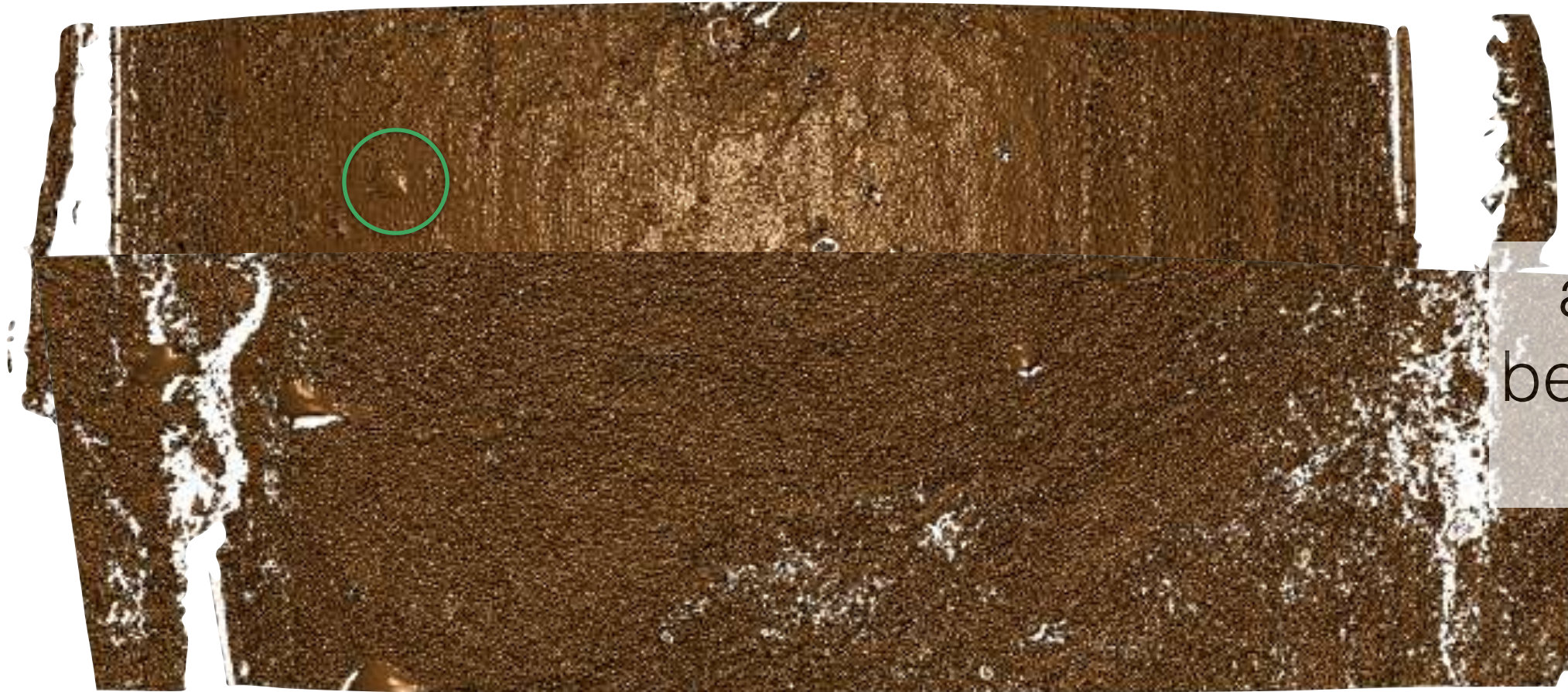


droplets

HS 224 to Clone HS 224

Random Forest Scores

★ Some LEA scans of clones are showing artifacts:



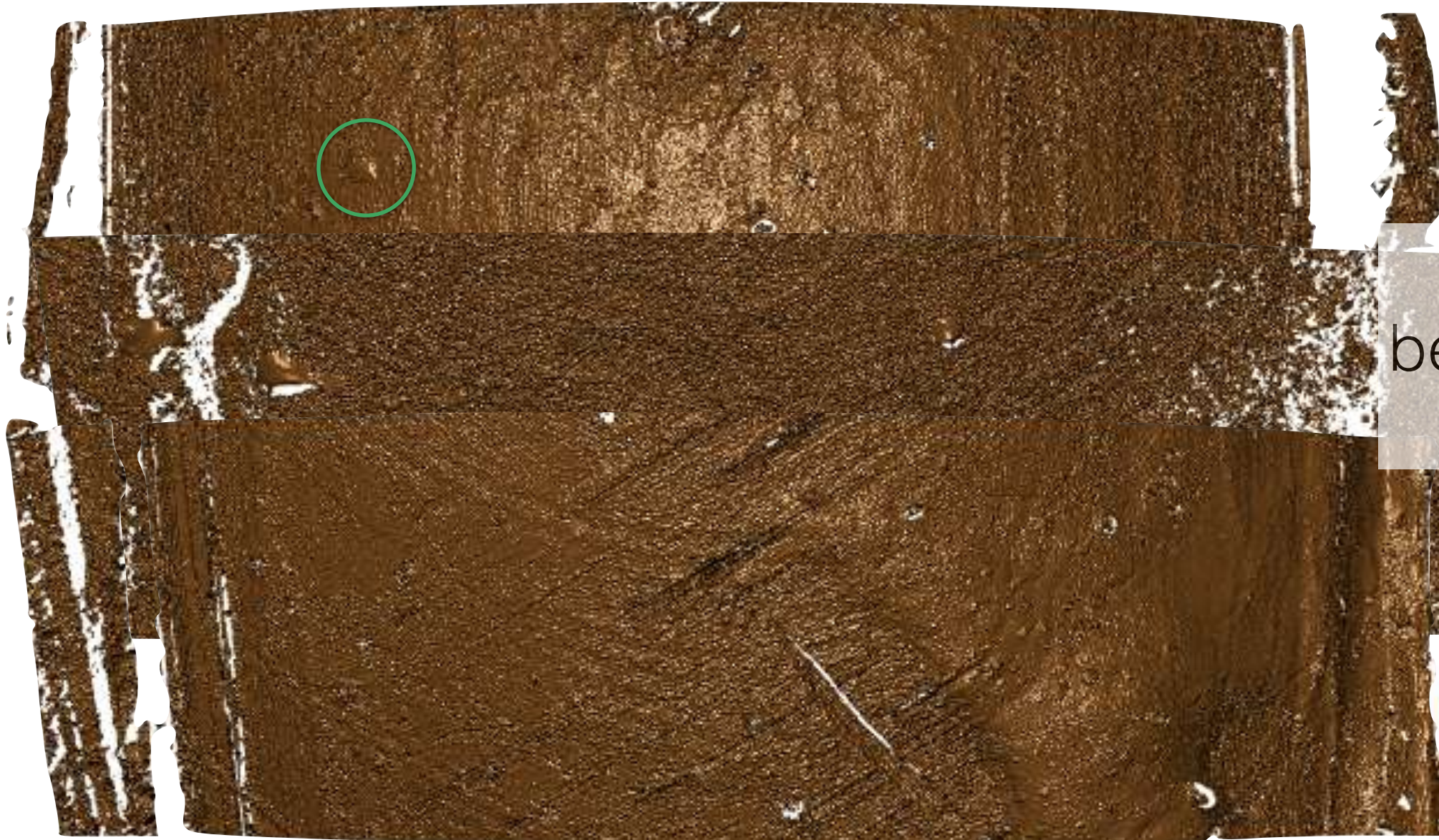
droplets

air pockets
between clone
and cast

HS 224 to Clone HS 224

Random Forest Scores

★ Some LEA scans of clones are showing artifacts:



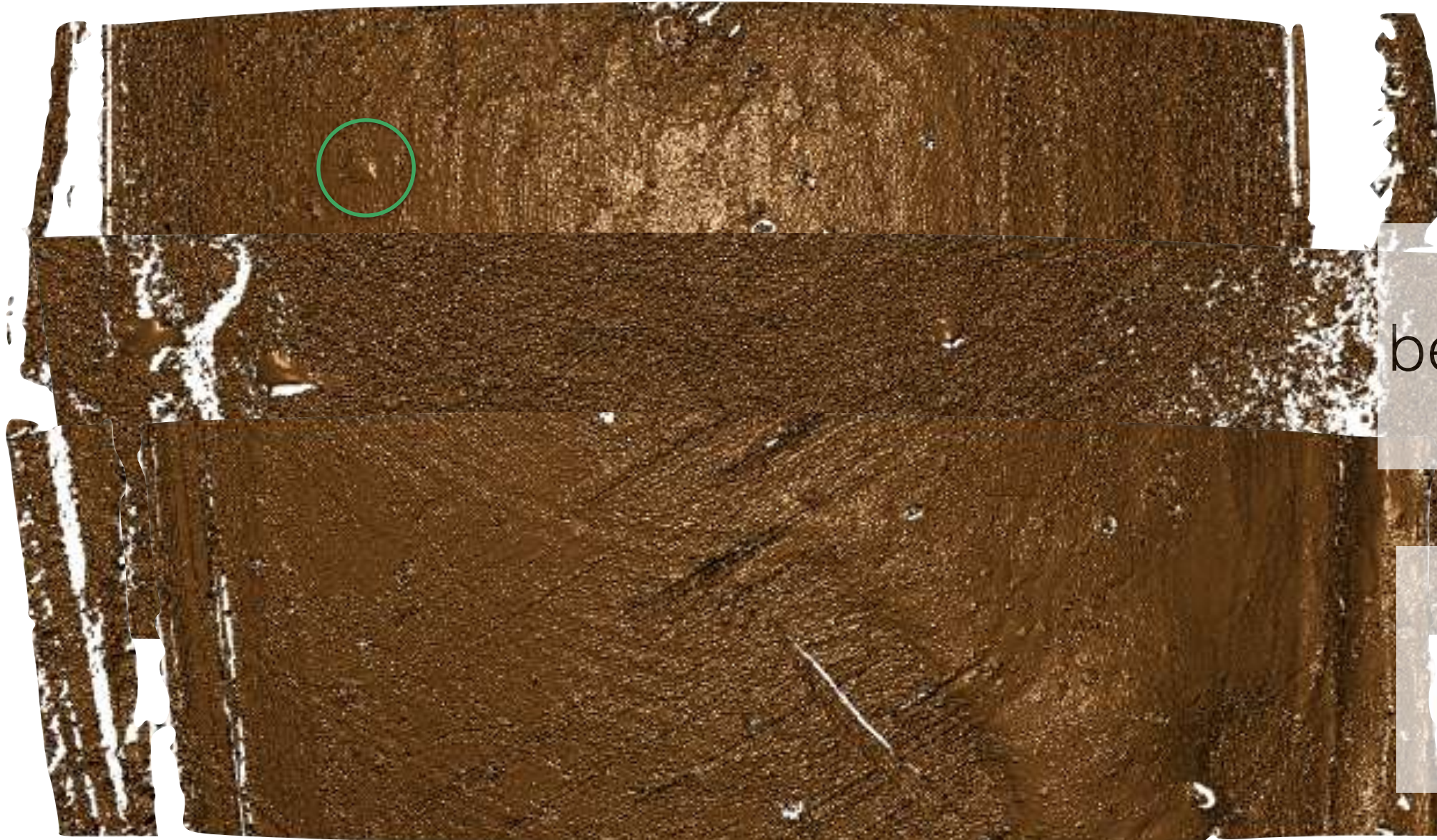
droplets

air pockets
between clone
and cast

HS 224 to Clone HS 224

Random Forest Scores

★ Some LEA scans of clones are showing artifacts:



droplets

air pockets
between clone
and cast

some other
structure but
no striae

HS 224 to Clone HS 224

Random Forest Scores

★ Some LEA scans of clones are showing artifacts:



droplets

air pockets
between clone
and cast

Automatic RF scores allow
us to assess clone quality
quantitatively

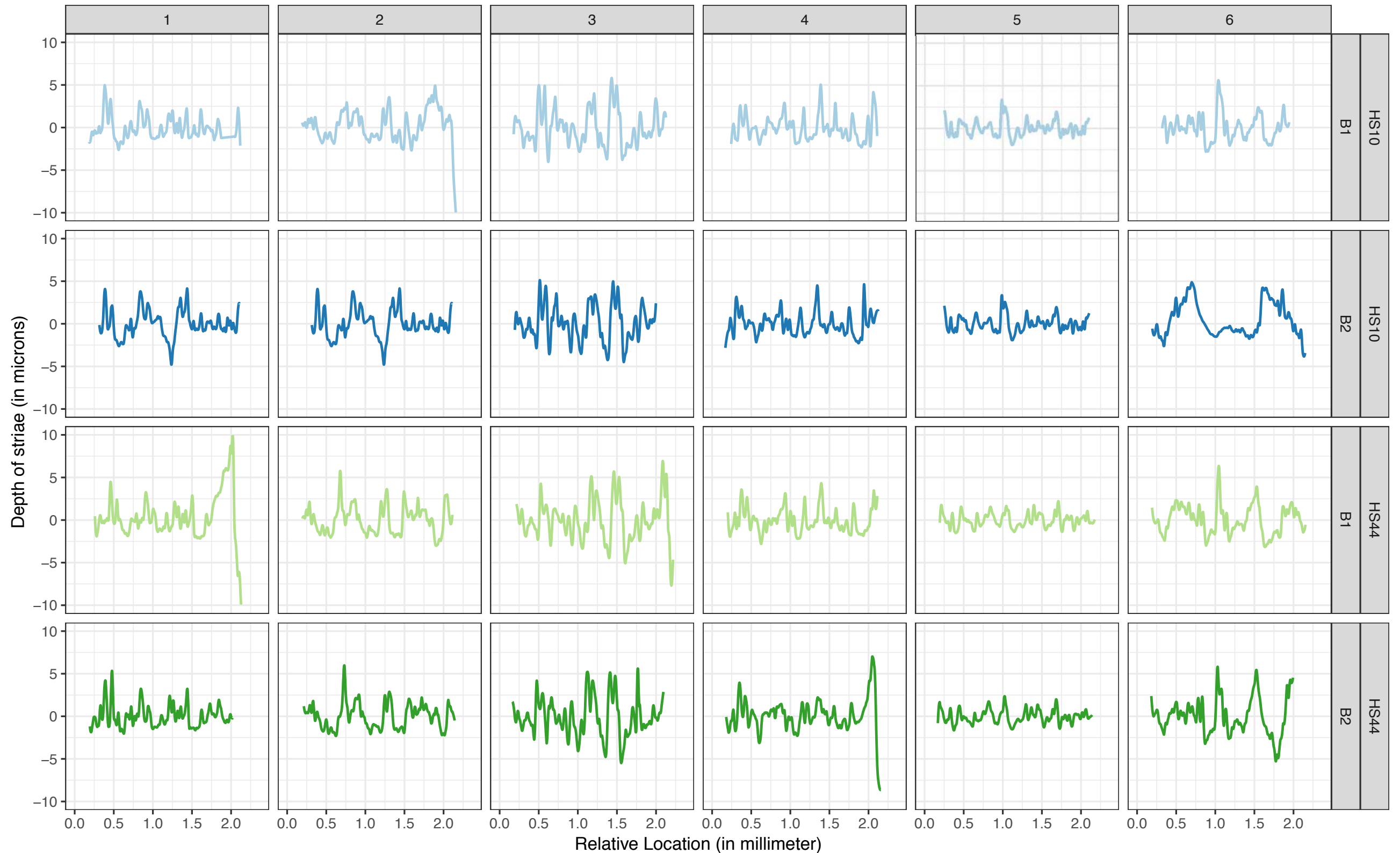
some other
structure but
no striae

Hamby Set 10 and 44

Random Forest Scores

- ★ Hamby Set 10 and Set 44 are from the same 10 barrels of Ruger P-85s
- ★ Bullets from Hamby Set 44 are fired some time later (~ 240 shots in each barrel between the sets)
- ★ Two Goals:
 - ★ **check** identifications between the sets
 - ★ **quantify** identifications

Hamby Set 10 and 44 Barrel 1



Set-Barrel-Bullet — Br1.B1.HS10 — Br1.B2.HS10 — Br1.B1.HS44 — Br1.B2.HS44

Hamby Set 10 and 44

Random Forest Scores

★ Identification between Sets 10 and 44 are possible

	HS10-B1	HS10-B2	HS44-B1	HS44-B2
HS10-B1	1.00	0.65		
HS10-B2	0.65	1.00		
HS44-B1			1.00	0.79
HS44-B2			0.79	1.00

Hamby Set 10 and 44

Random Forest Scores

★ Identification between Sets 10 and 44 are possible

	HS10-B1	HS10-B2	HS44-B1	HS44-B2
HS10-B1	1.00	0.65	0.56	0.59
HS10-B2	0.65	1.00	0.57	0.57
HS44-B1			1.00	0.79
HS44-B2			0.79	1.00

Hamby Set 10 and 44

Random Forest Scores

★ Identification between Sets 10 and 44 are possible

	HS10-B1	HS10-B2	HS44-B1	HS44-B2
HS10-B1	1.00	0.65	0.56	0.59
HS10-B2	0.65	1.00	0.57	0.57
HS44-B1			1.00	0.79
HS44-B2			0.79	1.00

★ RF scores between sets are not as high as within sets

Hamby Set 10 and 44

Random Forest Scores

★ Identification between Sets 10 and 44 are possible

	HS10-B1	HS10-B2	HS44-B1	HS44-B2
HS10-B1	1.00	0.65	0.56	0.59
HS10-B2	0.65	1.00	0.57	0.57
HS44-B1			1.00	0.79
HS44-B2			0.79	1.00

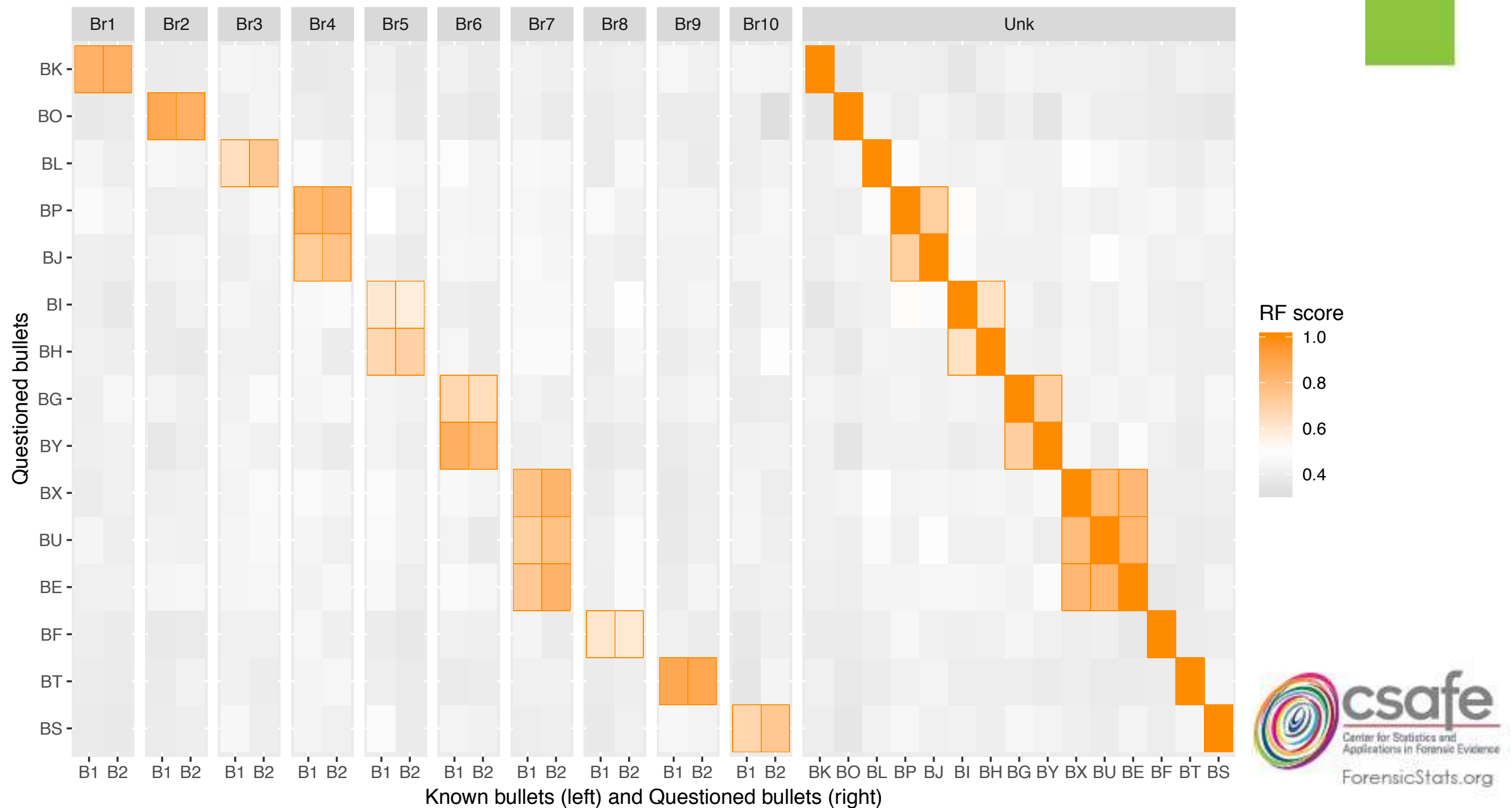
★ RF scores between sets are not as high as within sets

★ RF scores can be used to quantify the strength of an identification

RF Scores in context

Hamby Set 44

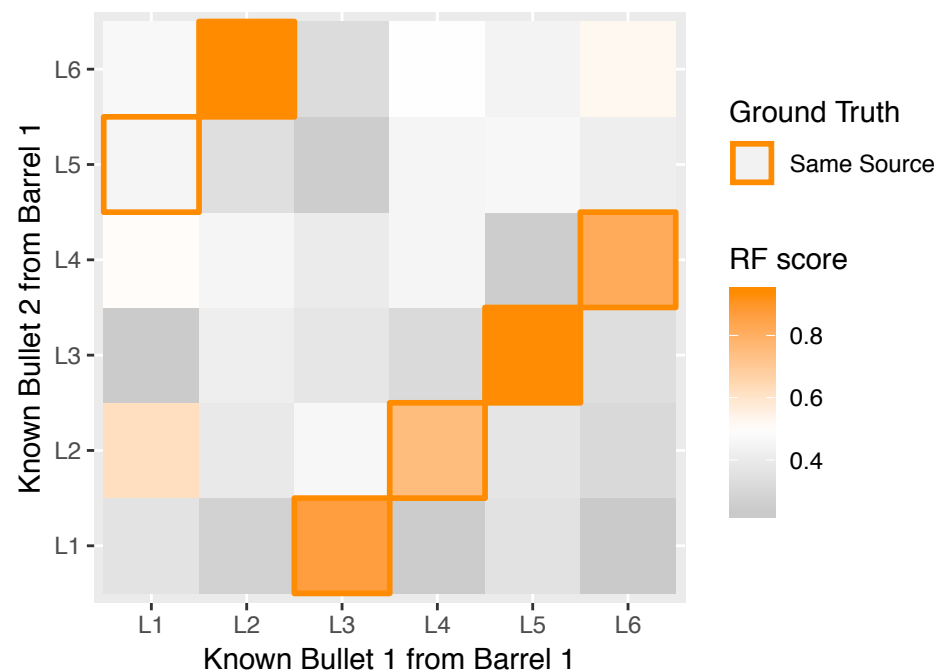
RF Scores in context of a study



Hamby Set 44

Random Forest Scores

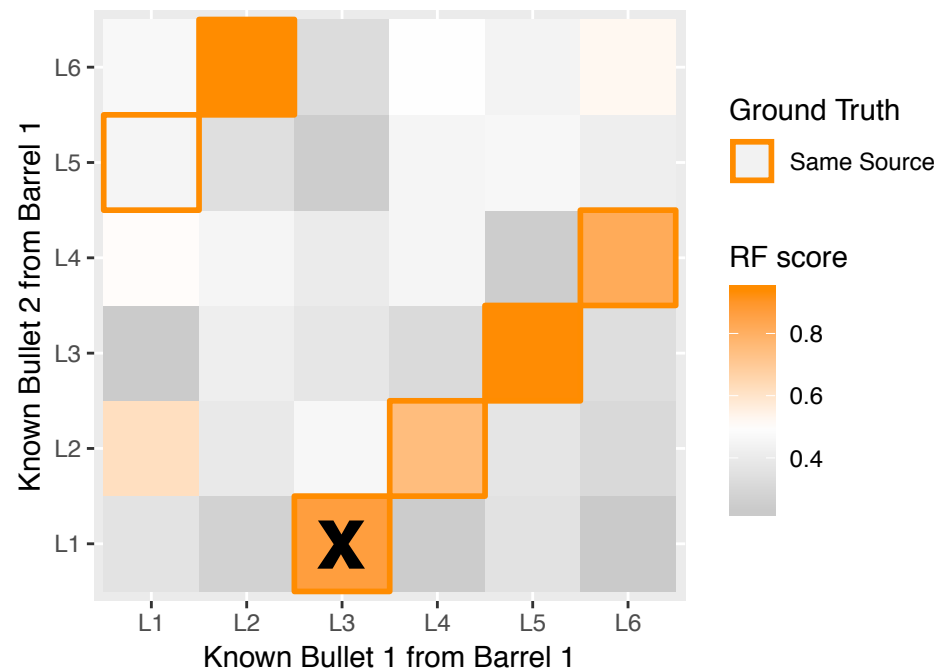
★ Land-to-land matches for bullets 1 and 2 from barrel 1



Hamby Set 44

Random Forest Scores

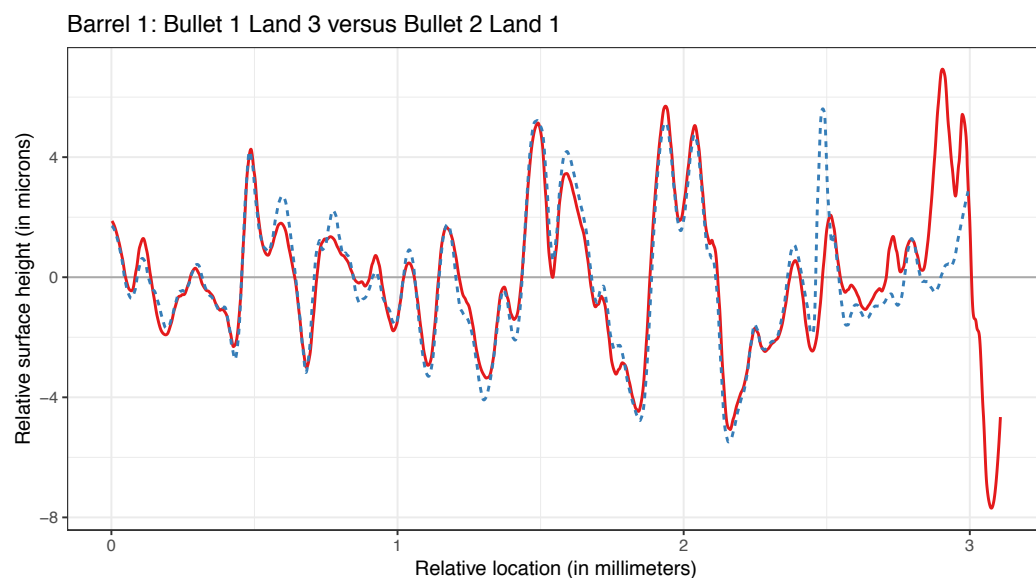
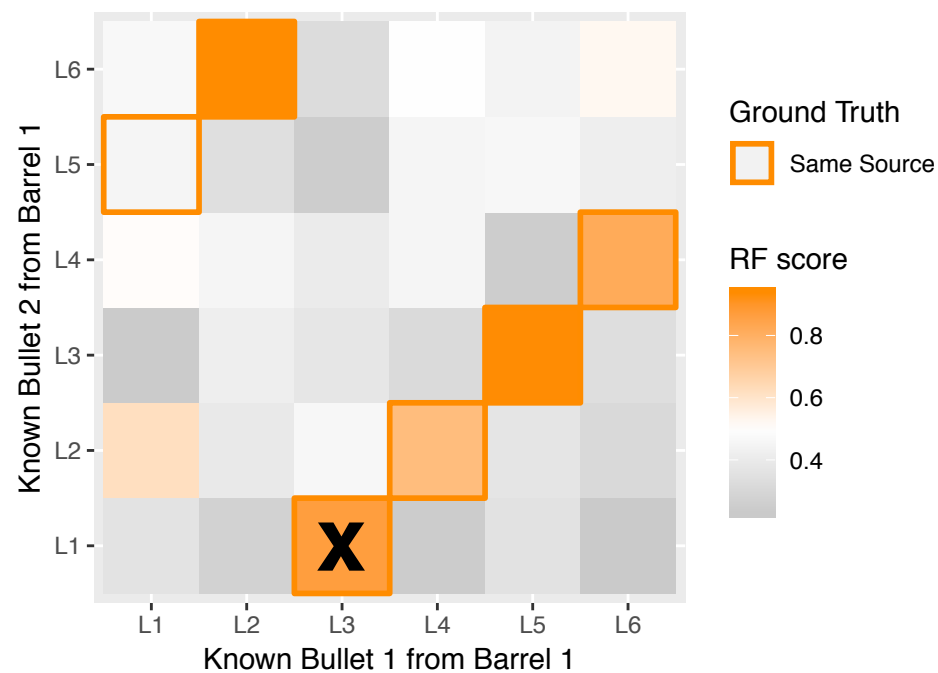
★ Land-to-land matches for bullets 1 and 2 from barrel 1



Hamby Set 44

Random Forest Scores

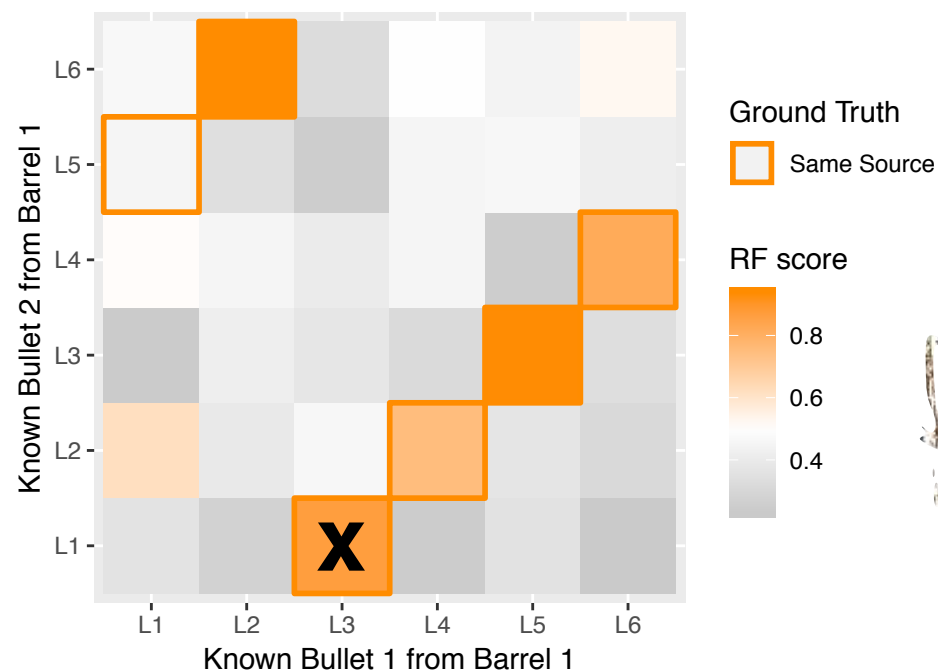
★ Land-to-land matches for bullets 1 and 2 from barrel 1



Hamby Set 44

Random Forest Scores

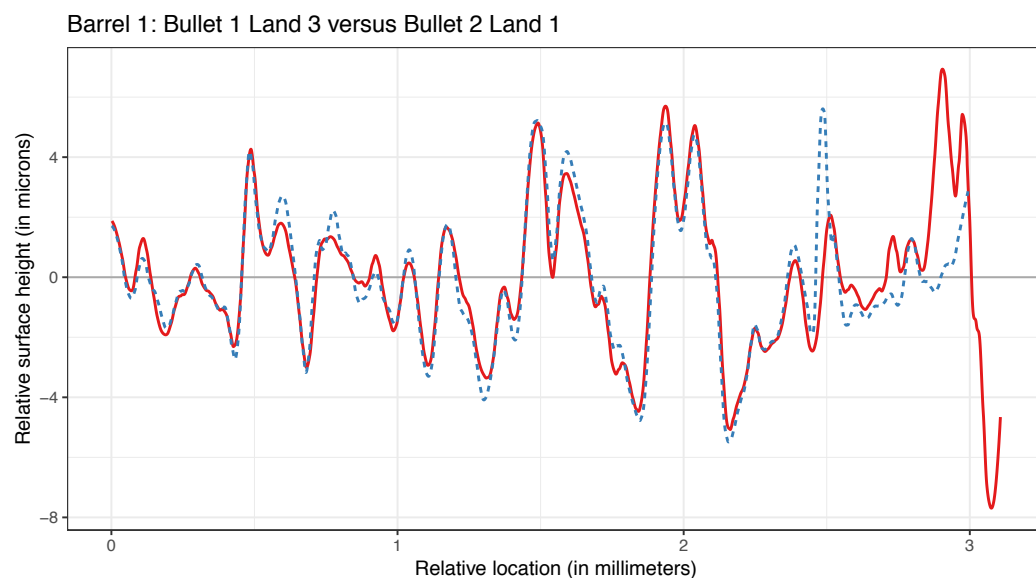
★ Land-to-land matches for bullets 1 and 2 from barrel 1



Barrel 1 - Bullet 1 - Land 3



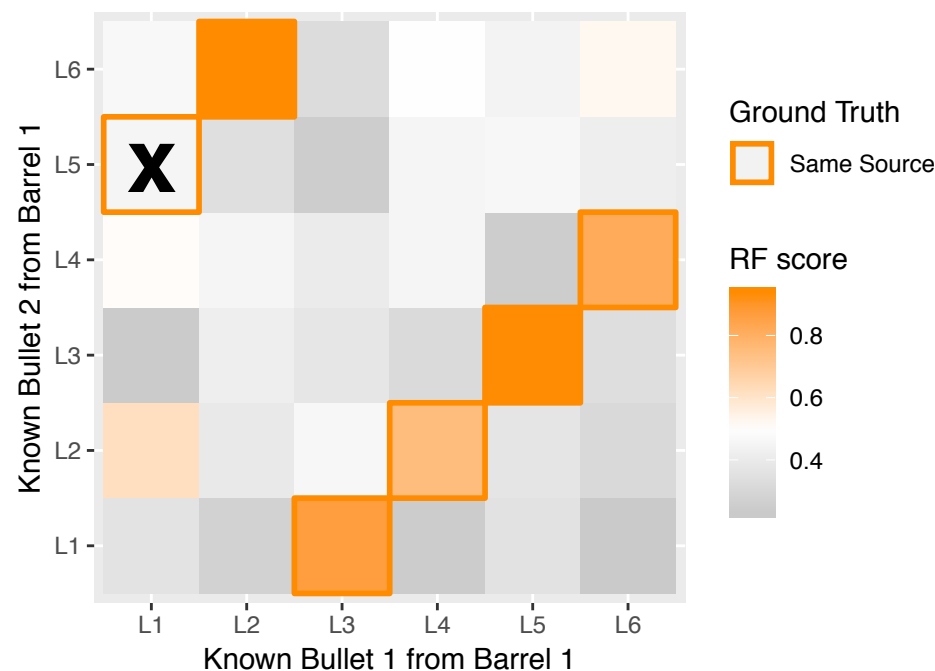
Barrel 1 - Bullet 2 - Land 1



Hamby Set 44

Random Forest Scores

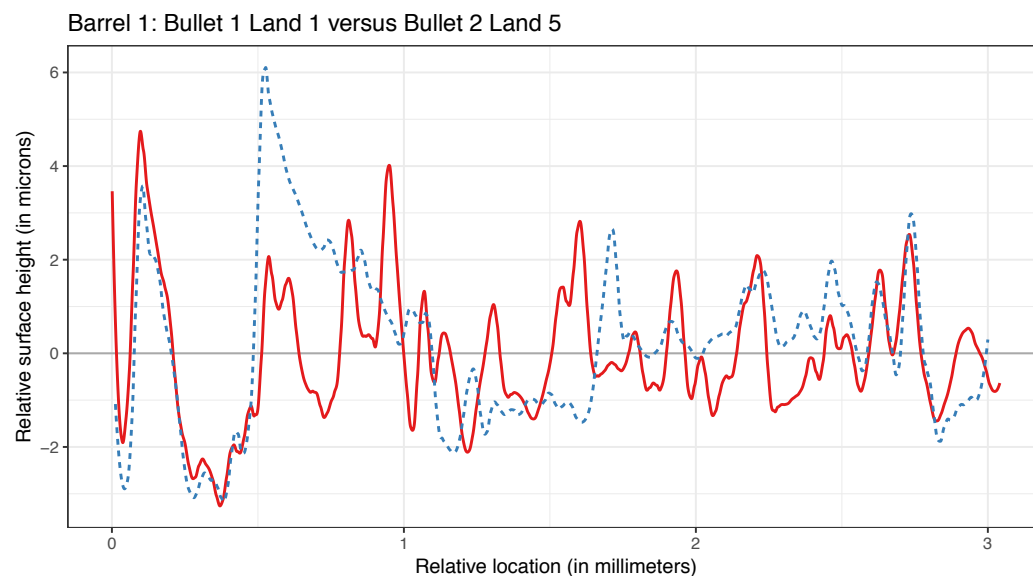
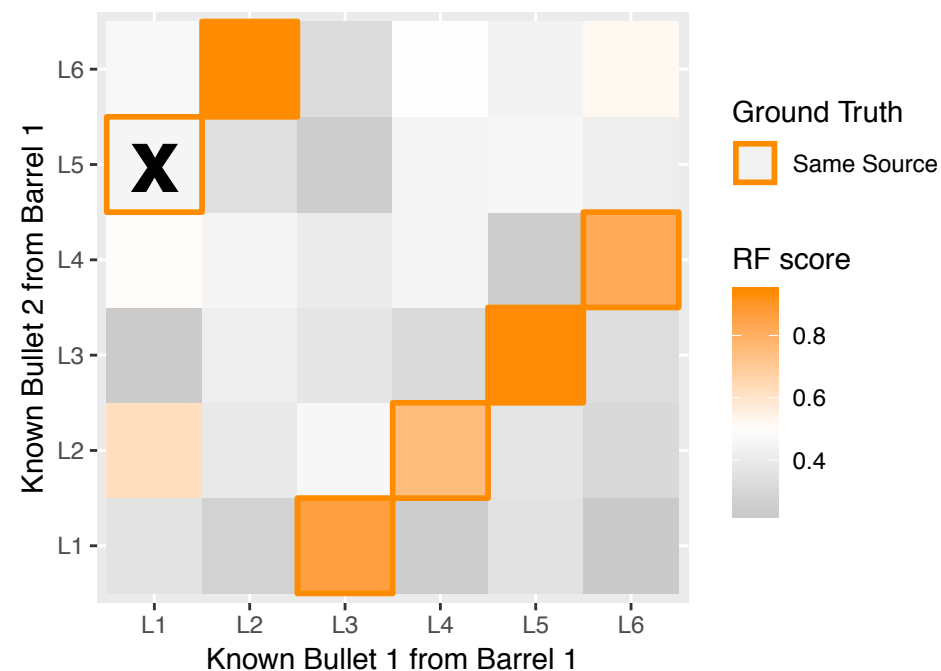
★ Land-to-land matches for bullets 1 and 2 from barrel 1



Hamby Set 44

Random Forest Scores

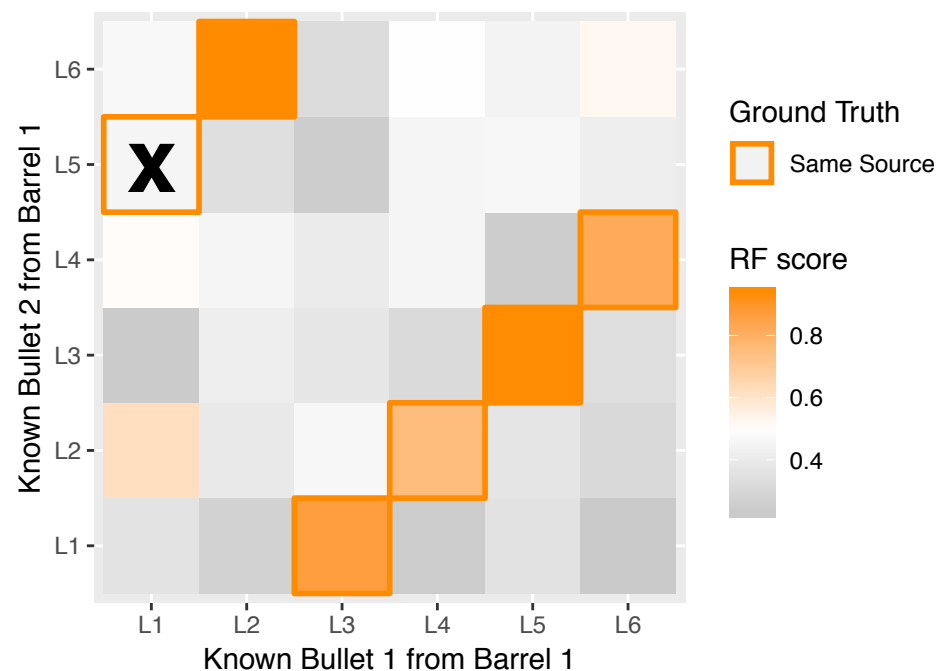
★ Land-to-land matches for bullets 1 and 2 from barrel 1



Hamby Set 44

Random Forest Scores

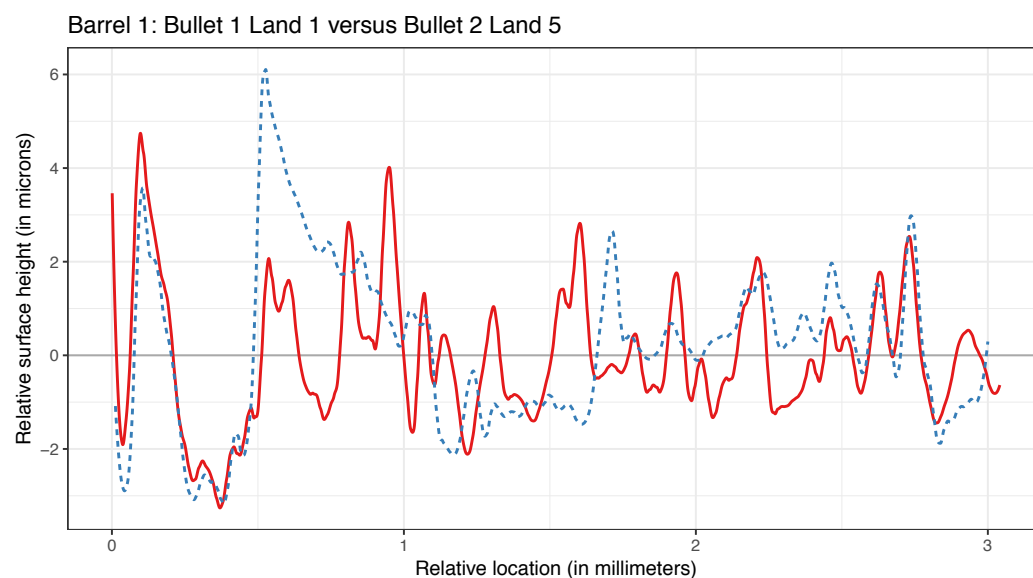
★ Land-to-land matches for bullets 1 and 2 from barrel 1



Barrel 1 - Bullet 1 - Land 1

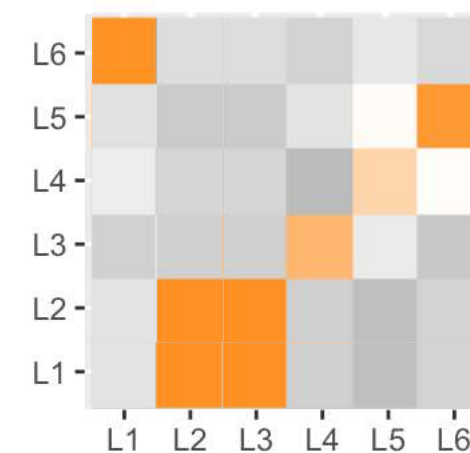
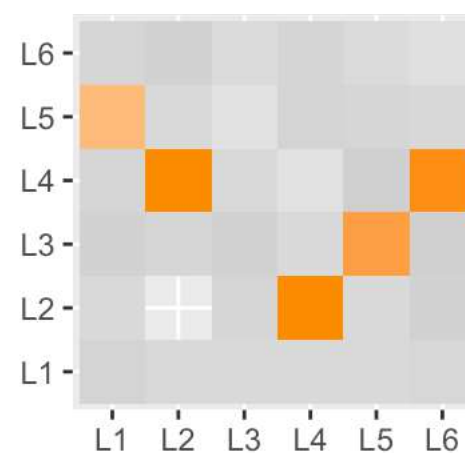
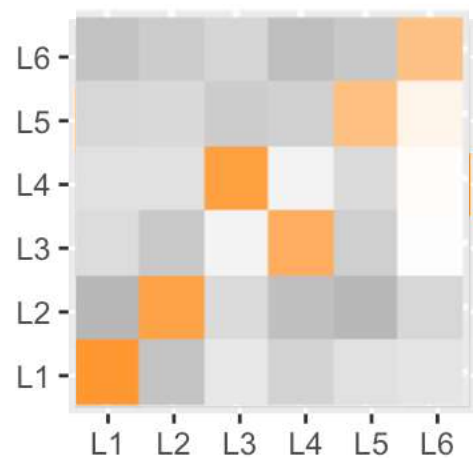


Barrel 1 - Bullet 2 - Land 5



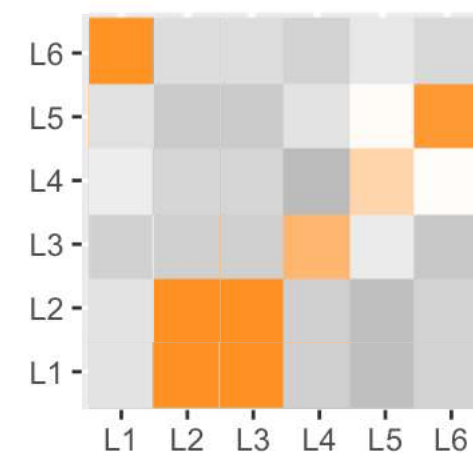
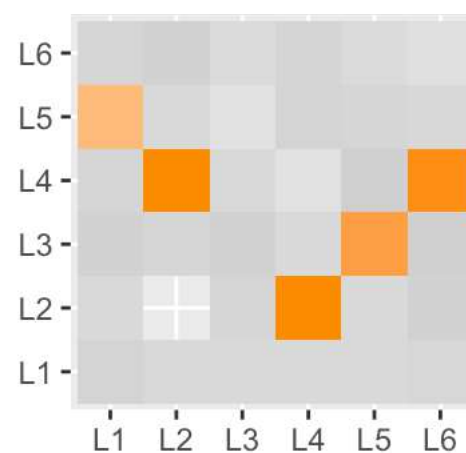
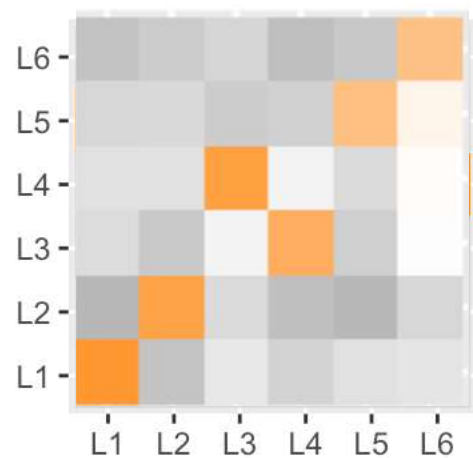
Using the RF Score for diagnostics

★ Land-to-land matches for two bullets show oddities



Using the RF Score for diagnostics

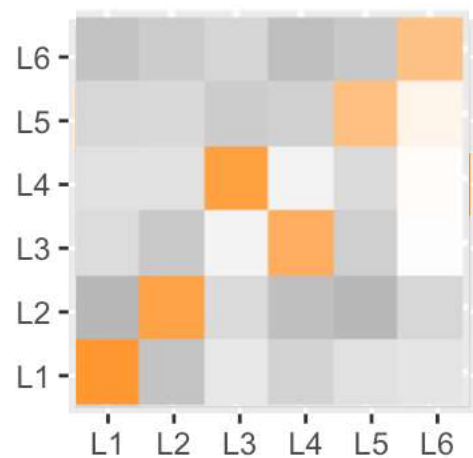
★ Land-to-land matches for two bullets show oddities



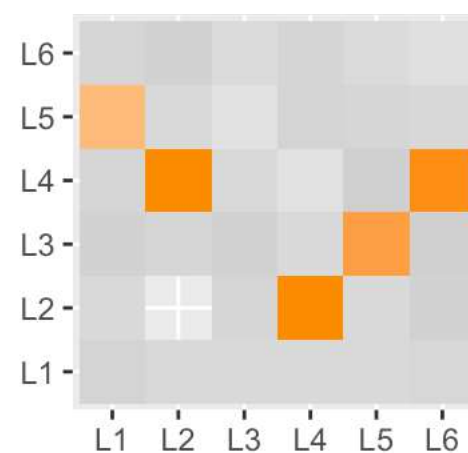
*scans for lands 3
and 4 were mis-
labelled for one
bullet*

Using the RF Score for diagnostics

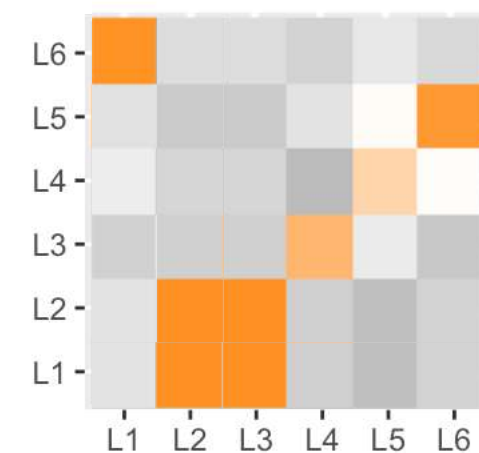
★ Land-to-land matches for two bullets show oddities



scans for lands 3 and 4 were mislabelled for one bullet

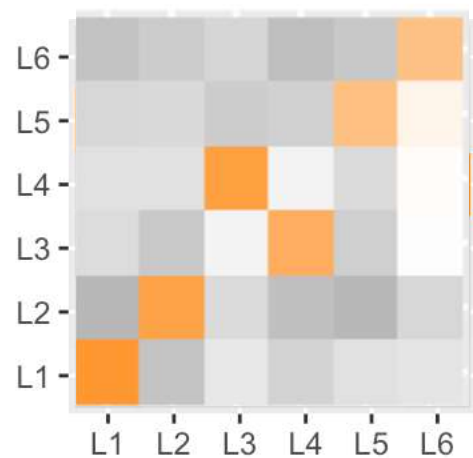


instead of turning right to scan land 6, scanner turned left and scanned land 4 again

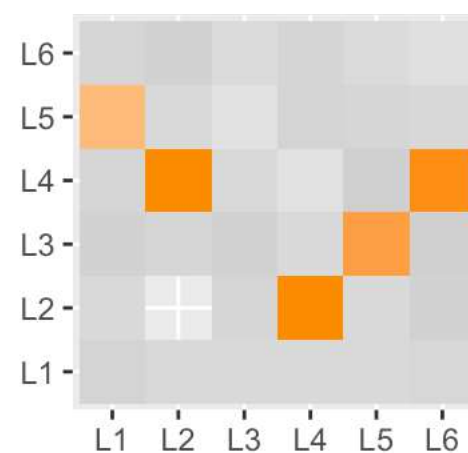


Using the RF Score for diagnostics

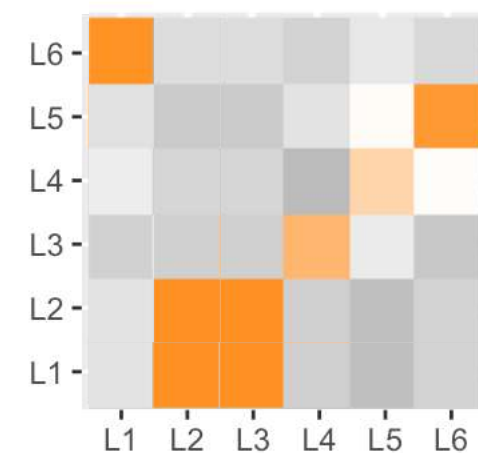
★ Land-to-land matches for two bullets show oddities



scans for lands 3 and 4 were mislabelled for one bullet



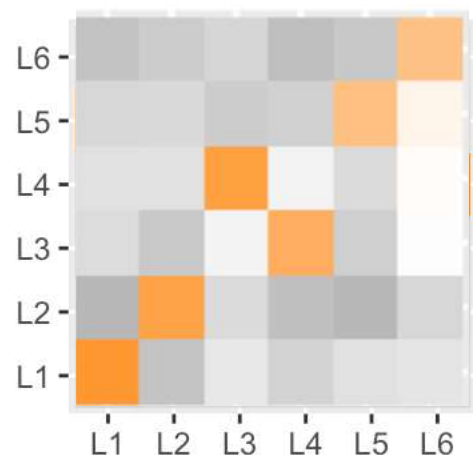
instead of turning right to scan land 6, scanner turned left and scanned land 4 again



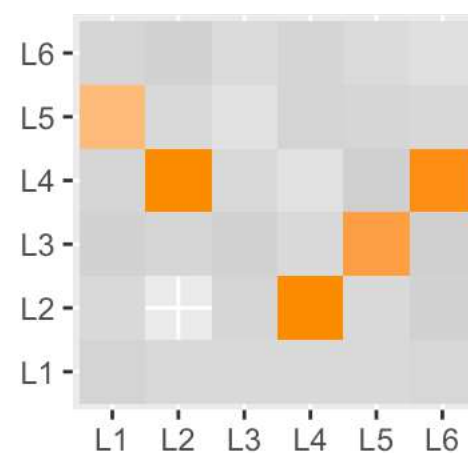
scans for land 2 and land 3 are identical

Using the RF Score for diagnostics

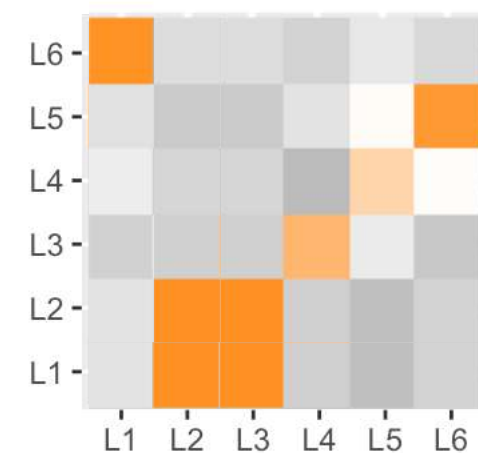
★ Land-to-land matches for two bullets show oddities



scans for lands 3 and 4 were mislabelled for one bullet



instead of turning right to scan land 6, scanner turned left and scanned land 4 again



scans for land 2 and land 3 are identical

These things went wrong in our scanning lab

Summary & further work

- ★ Random Forest Model provides interpretable and consistent scores for quantifying identifications
 - ★ matching clones to original bullets - assess clone quality
 - ★ matching bullets between Hamby sets
 - ★ quantification of bullet and scan deficiencies
- ★ Visualizations with increased context used for diagnostics
- ★ Limitations: traditional rifling only (LEAs are matched)

Thank You!

Questions?

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ISU CSAFE bullet team



<https://forensicstats.org/>

@csafe_coe 