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Footwear Impression Research at NIST

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NIST

November 7, 2018

MOTIVATION

- 2009 NAS; 2016 PCAST:
 - Footwear identifications are largely subjective
 - Questions about reliability
 - Questions about scientific validity
 - Need for quantitative assessments of footwear evidence
 - Need for increased empirically-tested objectivity of footwear analysis
- Need to improve **quantitative analysis**
- Need for **algorithmic approaches for** the forensic footwear community



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GOALS

- Develop quantitative, objective methods for footwear impression comparisons
 - High degree of repeatability & reproducibility
 - Easier to measure accuracy with objective methods
- High performance – good discrimination power
- Provide prototype software tools to be evaluated for the following purposes
 - Use by practitioners in casework
 - Use by researchers to develop algorithms



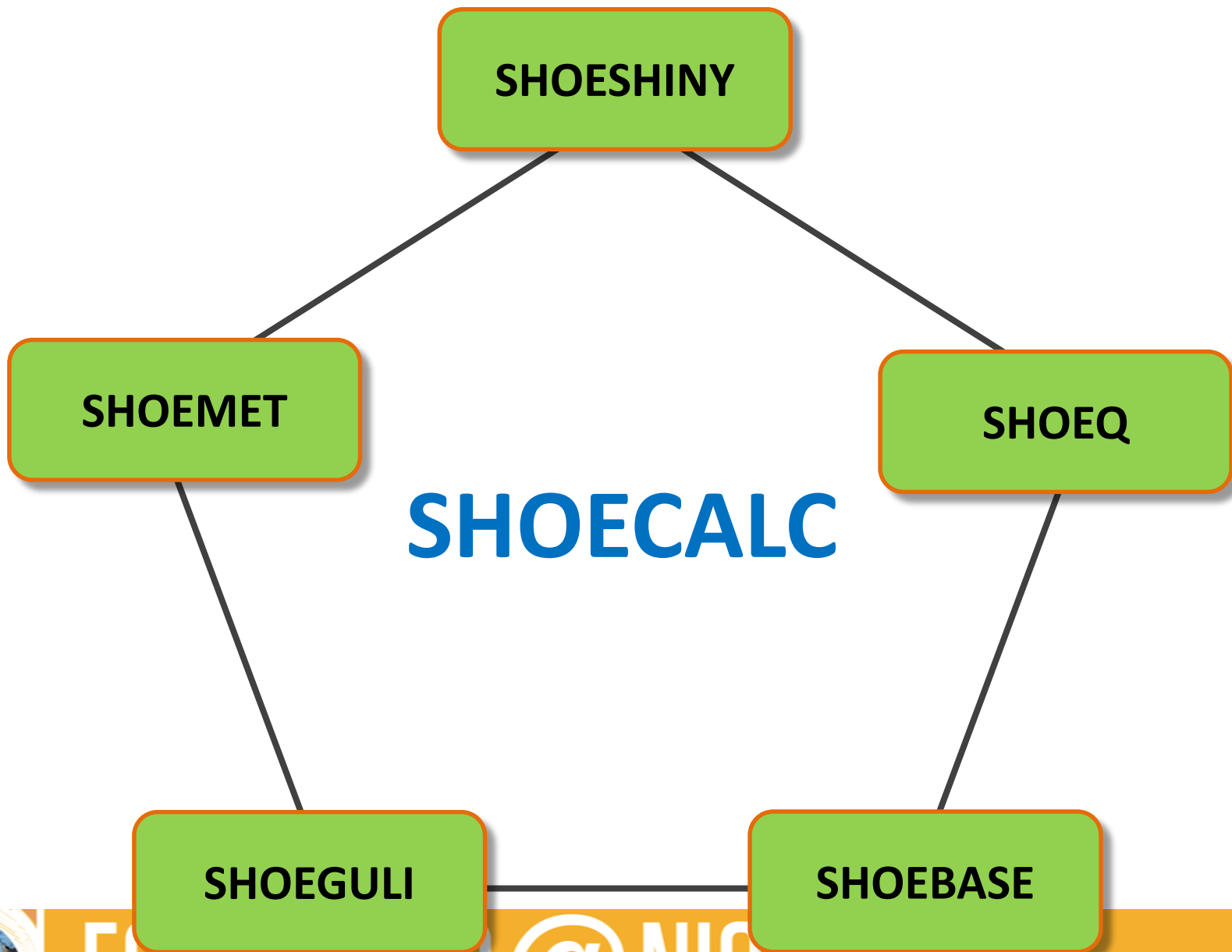
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SHOECALC

- A prototype system for footwear analysis that will allow
 - Researchers/developers to have a workbench for development of quantitative methods
 - Examiners to use these quantitative methods during casework
 - Development of this system is a long-term goal





SHO

Database consisting of

- Real & staged crime scene impressions and metadata
- Catalogue of outsole designs and metadata
- Test impressions from shoes of arrestees or research volunteers
- Catalogue of acquired characteristics (RACs) along with shape, size, location, brand, outsole design, etc.
- Interfaces and formats for submitting and maintaining footwear data

SHOEMET

SHO

SHOEGULI

SHOEBASE



Synthetic/augmented footwear impressions

- For research and testing, generates synthetic footwear impressions with user specified characteristics and with ground truth known
- Characteristics include outsole designs, wear amounts, sizes, and distributions of RACs; different matrix/substrate combinations
- Synthetic test & crime scene impressions
- Augmented data for research/tests

NY

ALC

SHOEQ

SHOEGULI

SHOEBASE



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SHOESHIN

Comparison Measures

- A workbench for experimentation with different comparison scores. Some scores lead to better discrimination between mated and non-mated pairs of images than others.
- User inputs a function for computing a comparison score and applies it to any given pair of images; numerical score is reported.
- Also uses SHOEGULI to conduct experiments and produce ROC charts for comparing with a catalog of known, high performance comparison scores.

SHOEMET

SHOEGULI

SHOEBASE



Quality Measures

- Measuring different characteristics that describe the degradation, distortion, completeness, number of features in the impression
- Input is any footwear image; output is a list of quality metrics
- May be used as a workbench for experimentation with different image quality metrics

SHOESHINY

SHOEQ

SHOEGULI

SHOEBASE



SHOESHINY

SHOEMET

SHOEQ

- GUI for user interaction with the other modules of SHOECALC
- Allows user to upload images for calculation of comparison and quality scores
- Examine various choices of comparison metrics, **scores** and their ROC charts, and select choices for reporting the information in the evidence
- Exploratory analysis of data, charts, etc.



Today's talks

1. *Towards an end-to-end system for quantitative footwear impression comparisons* – Martin Herman
 - End-to-end prototype system for use by examiners during casework
2. *Image Alignment and Feature Extraction for Shoeprint Matching* – Gautham Venkatasubramanian
 - As part of end-to-end system, alignment of questioned and known impressions, along with feature extraction to be used for image matching
3. *Deep Learning based Feature Extractors for Shoeprint Matching* – Sarala Padi
 - As part of end-to-end system, features learned in a DNN model are used for image matching
4. *Matching Randomly Acquired Characteristics (RACs) in Footwear Impressions* – Weiqing Chen
 - As part of the end-to-end system, RAC features are extracted and matched





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Towards an End-to-End System for Quantitative Footwear Impression Comparisons

Presented by:

Martin Herman

Other Core Team Members:

Hari Iyer, Steve Lund, Gunay Dogan, Yooyoung Lee

Information Technology Laboratory, NIST

November 7, 2018

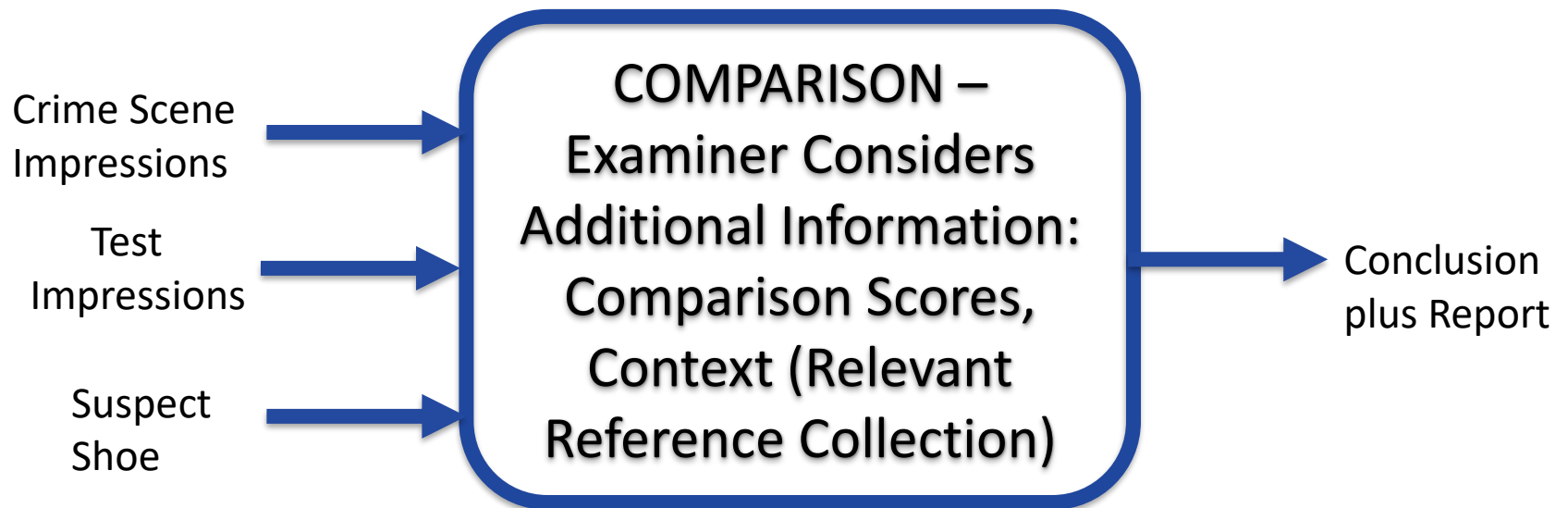
Use of SHOECALC: Quantitative Footwear Impression Comparisons

- For use by examiners in evidence evaluation
- FRStat for fingerprints (U.S. Defense Forensic Science Center) – currently in use

Current Examiner Comparison Process



Proposed Examiner Comparison Process

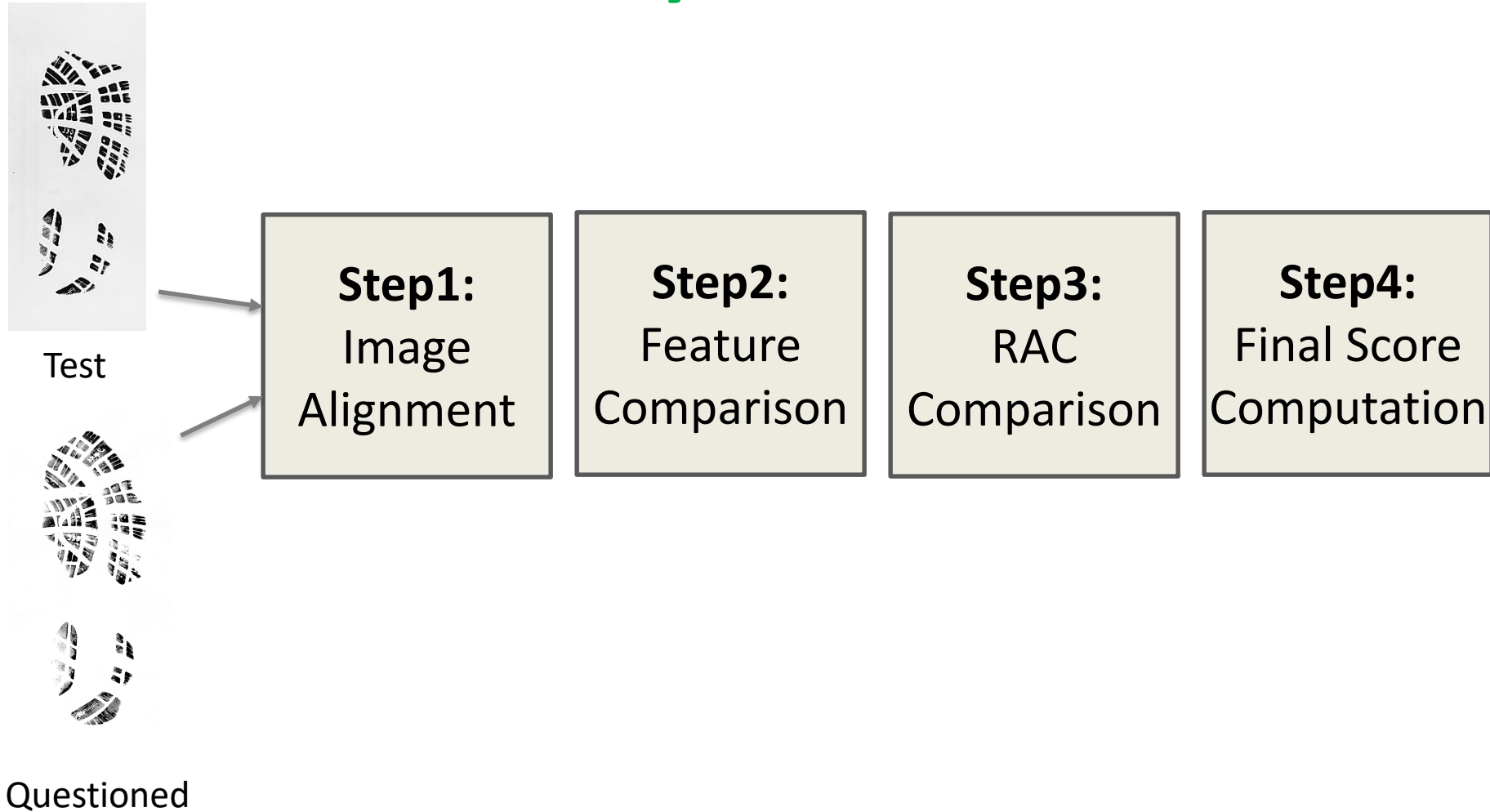


Elements of the Comparison Score

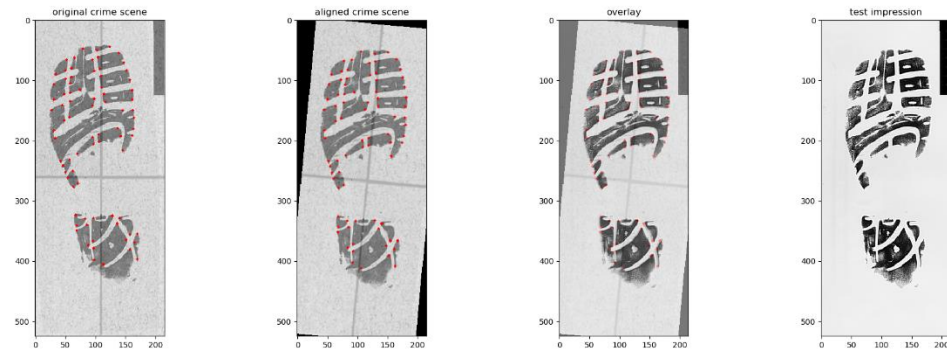
- Features considered in total score
 - Shoe size
 - Outsole design features
 - Wear features
 - RACs
- Transparency for examiner
 - Examiner should be able to understand how the score is related to features above
 - Our goal is for the examiner to be able to relate the score to SWGTREAD “Range of Conclusions Standard”



Workflow for End-to-End Scoring System



Step1: Image Alignment



- Determine best alignment of the two impression images (Questioned and Test)
- If images do not align well, then (optional) SCORE <- 0 & STOP

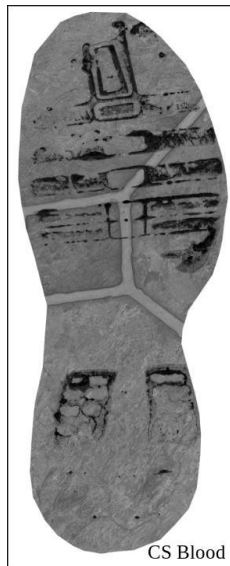
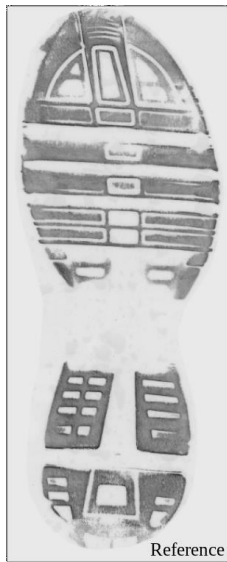
Described in talk later in session.



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Step 2: Feature Comparison



- Compares features based mainly on combination of design, wear and size. RAC features play only very small part.
- Score considers combined features inside a Region of Interest

Described in talk later in session.



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Step 3: RAC Comparison

Test



Questioned



- Compute score based on RACs
- RACs marked on test impression by examiner, then transformed to questioned impression after alignment.
- Then corresponding patches are compared.
- No marking of RACs in questioned impression.

[Described in talk later in session.](#)



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Step4: Computing Final Score

- Final score is combination of feature comparison & RAC comparison scores
 - Goal is to relate the individual feature and RAC scores, plus final score, to SWGTREAD conclusions scale
- The final score is computed using reference dataset of ground-truth-known mates and non-mates.
- Composite RAC score = combined Score-based Likelihood Ratio (SLR) of individual RAC SLRs
- Final score = SLR obtained from bivariate density of composite RAC score and feature score



End-to-End Score Computation: Examples



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Example 1

Close Non-Match (left shoe flipped)
Everspry EverOS Scanner

Questioned 1

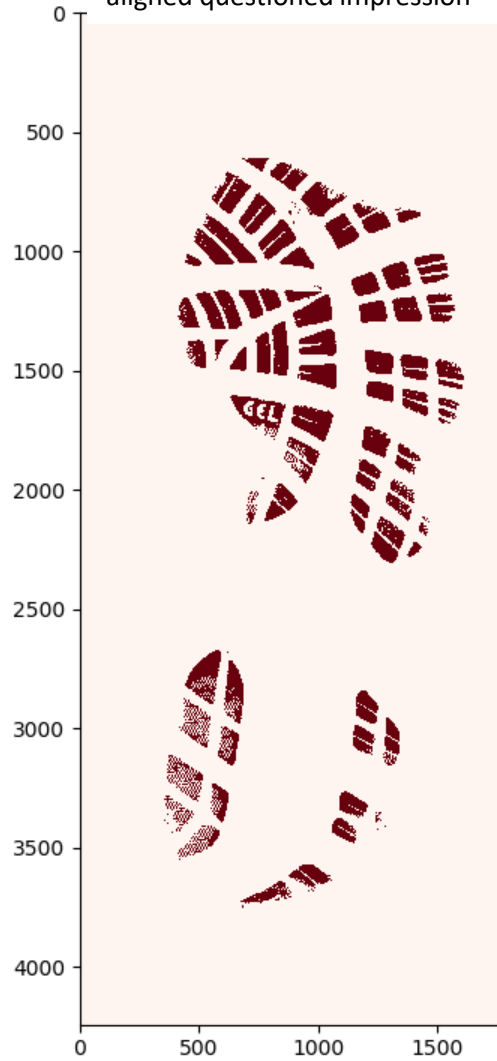


Test1

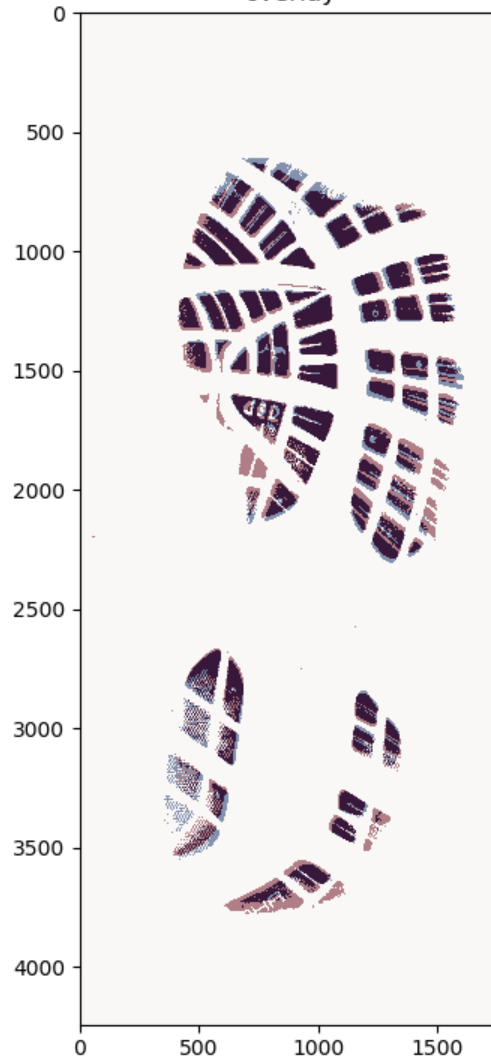


Alignment

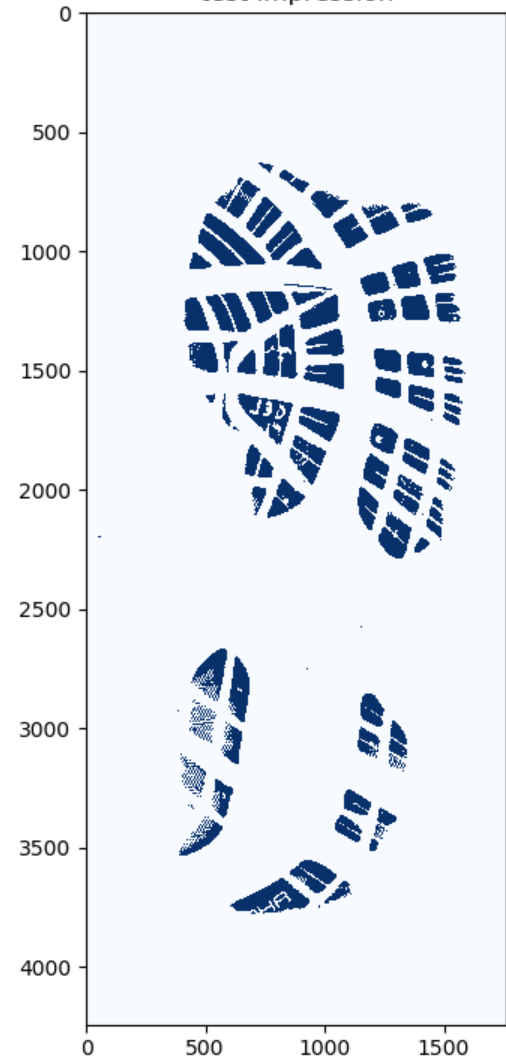
aligned questioned impression



overlay



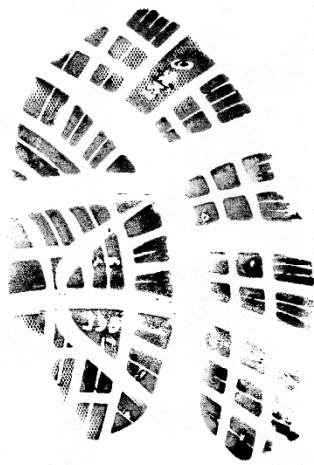
test impression



Example 2

Known Match

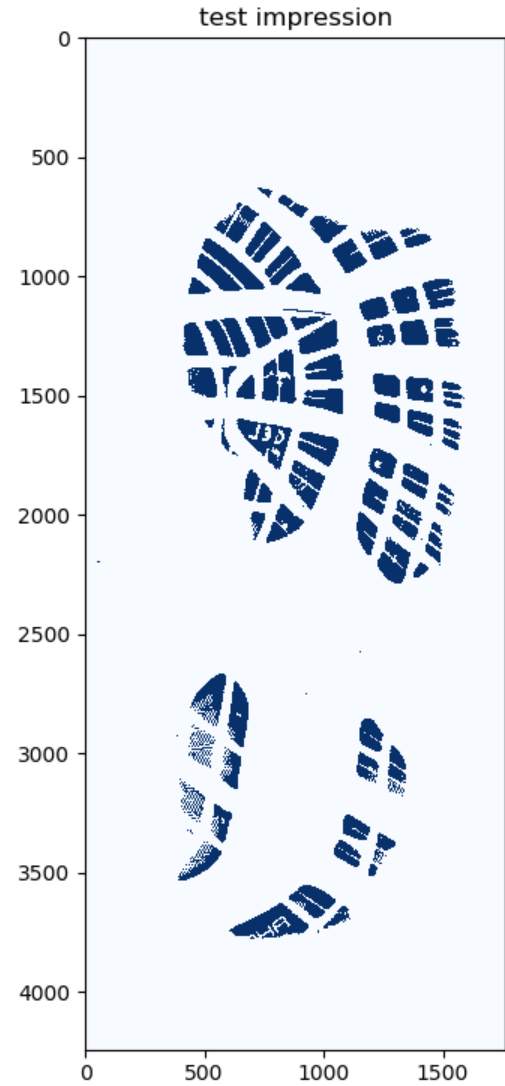
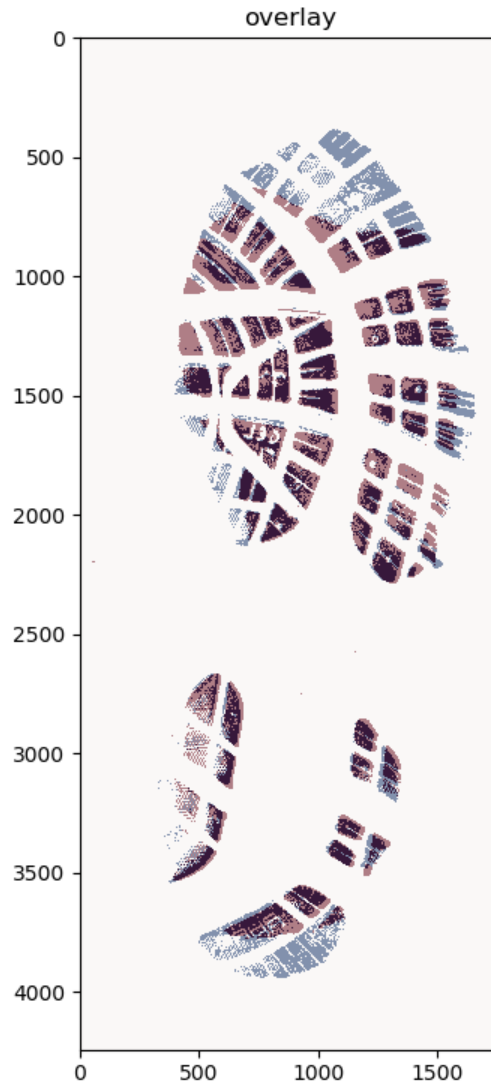
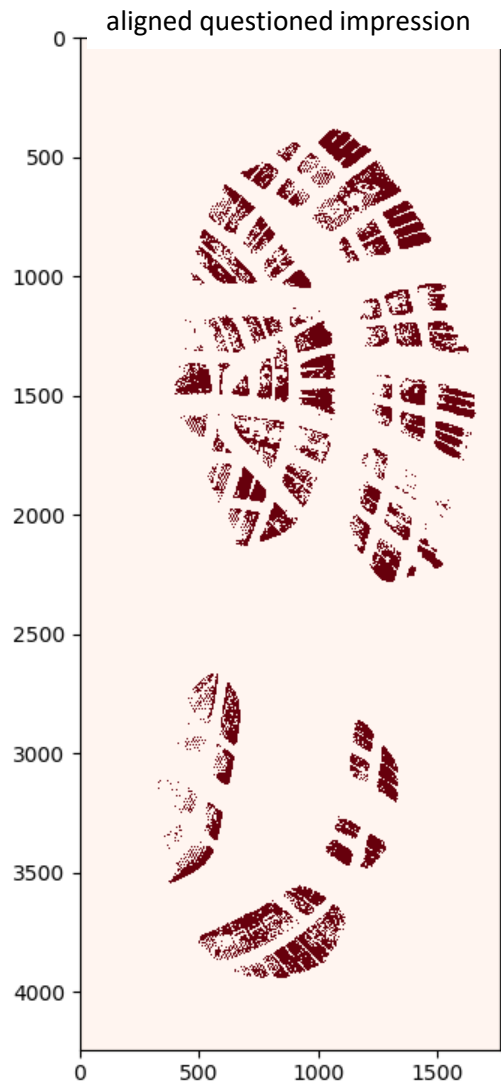
Questioned 2



Test1



Alignment



Example 3

Known Match

Dust Impression (Jacqueline Speir, WVU)

Questioned 3

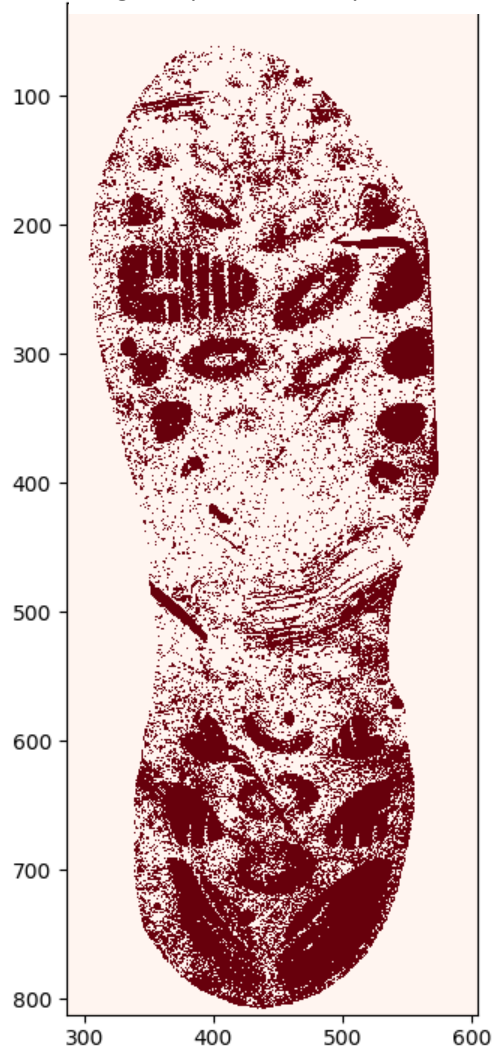


Test3

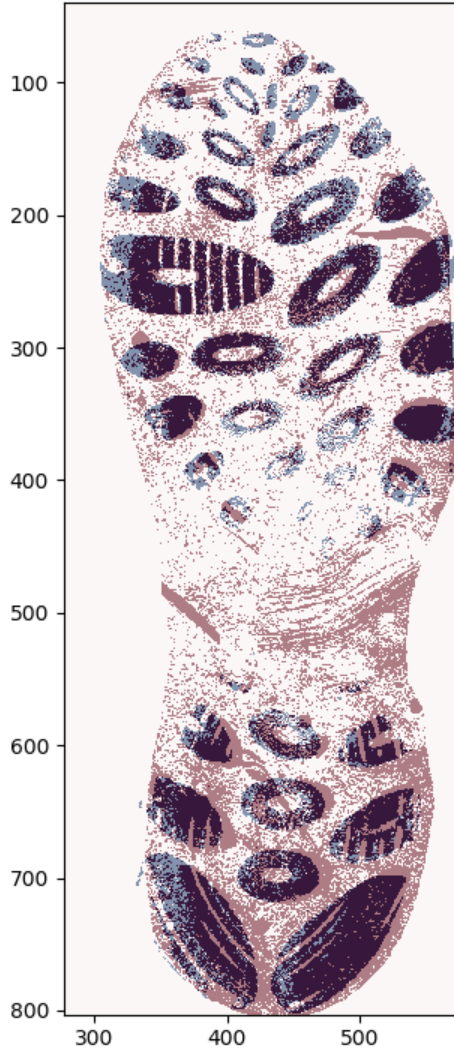


Alignment

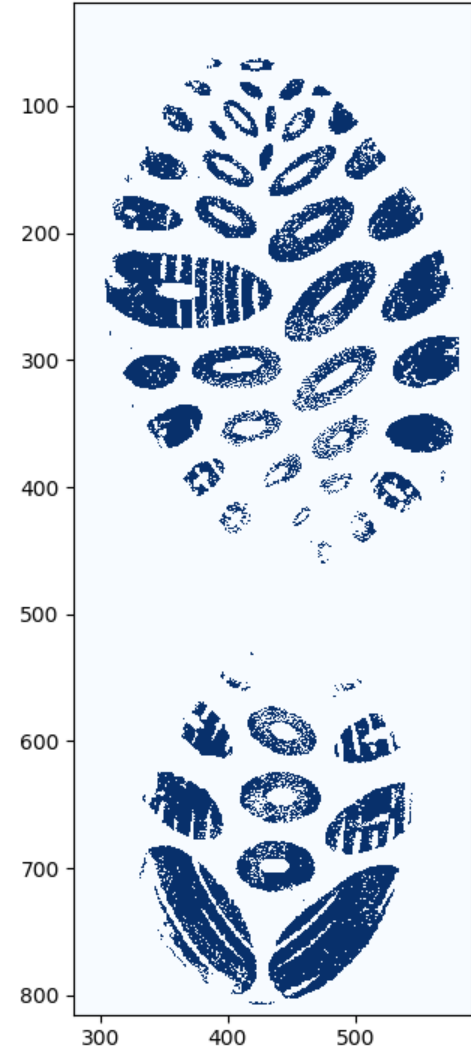
aligned questioned impression



overlay



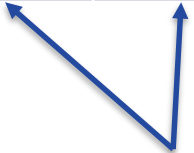
test impression



Comparison Scores

Feature Comparison Scores

Q1 vs Test1	Q2 vs Test1	Q3 vs Test3
0.8771	0.8009	0.8760



Close non-match score is greater than match score. But scores are very close.

RAC Comparison Scores

Rac No.	Q1 v Test1	Q2 v Test1	Q3 v Test3
1	0.0141	0.6002	0.2173
2	0.2042	0.5177	0.4651
3	0.0467	0.1392	0.3483
4	0.2992	0.5813	
5	0.5409	0.8777	
6	0.1849	0.2387	
7	0.5997	0.6835	
8	0.0938	0.7494	
9	0.0272	0.7950	
10	0.3495	0.7741	
11	0.3681	0.7558	
12	0.1567	0.6828	
13	0.4892	0.6302	

Final Comparison Scores

Q1 vs Test1	Q2 vs Test1	Q3 vs Test3
0.3831	0.6540	0.4501

Final scores for illustration only. Chosen without reference score distributions.

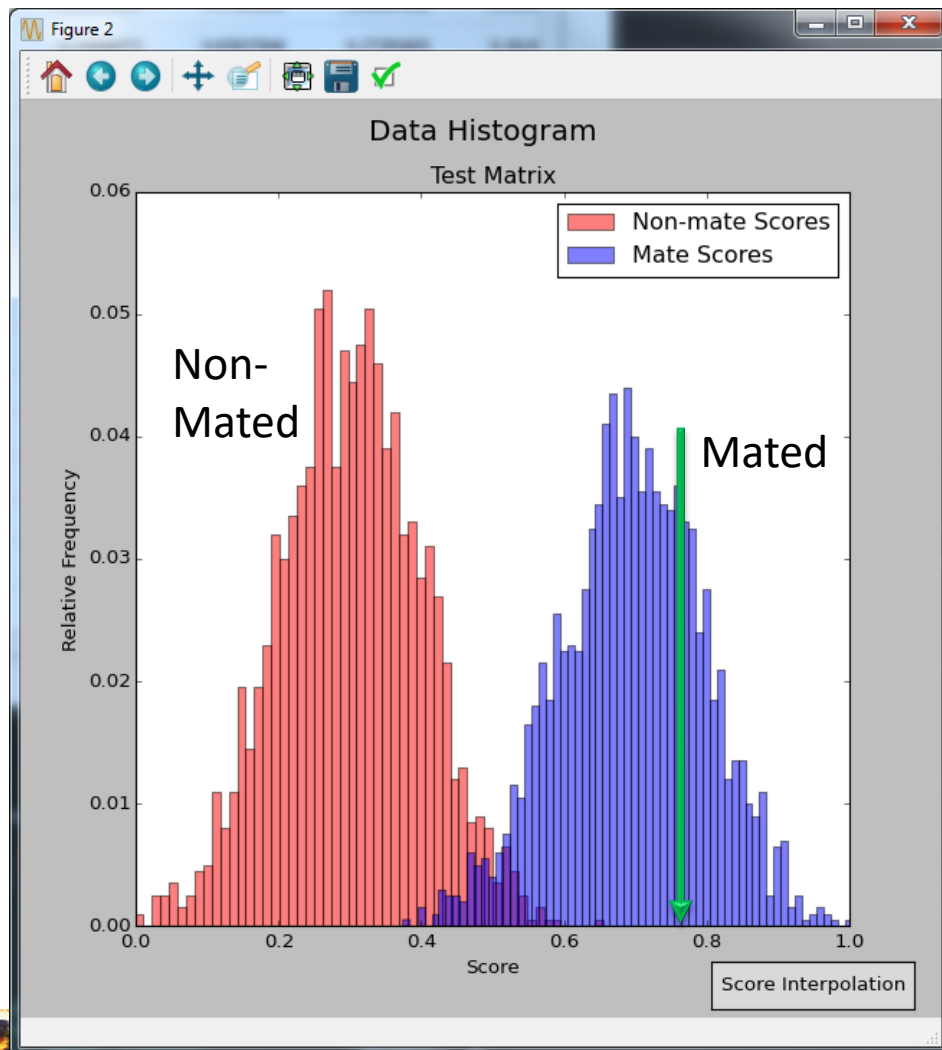
Q1 – close non-match; Q2 – known match; Q3 – known match

Context for Interpreting Comparison Scores - 1

- How do we determine what significance to give to any particular score?
- Answer: evaluate the score in the context of ground-truth-known mated and non-mated pairs that are representative of impressions obtained under conditions similar to the current crime scene.
 - E.g., same quality and quantity of information
- Provide context for
 1. Feature comparison score – size, design, wear
 2. RAC score
 3. Final score



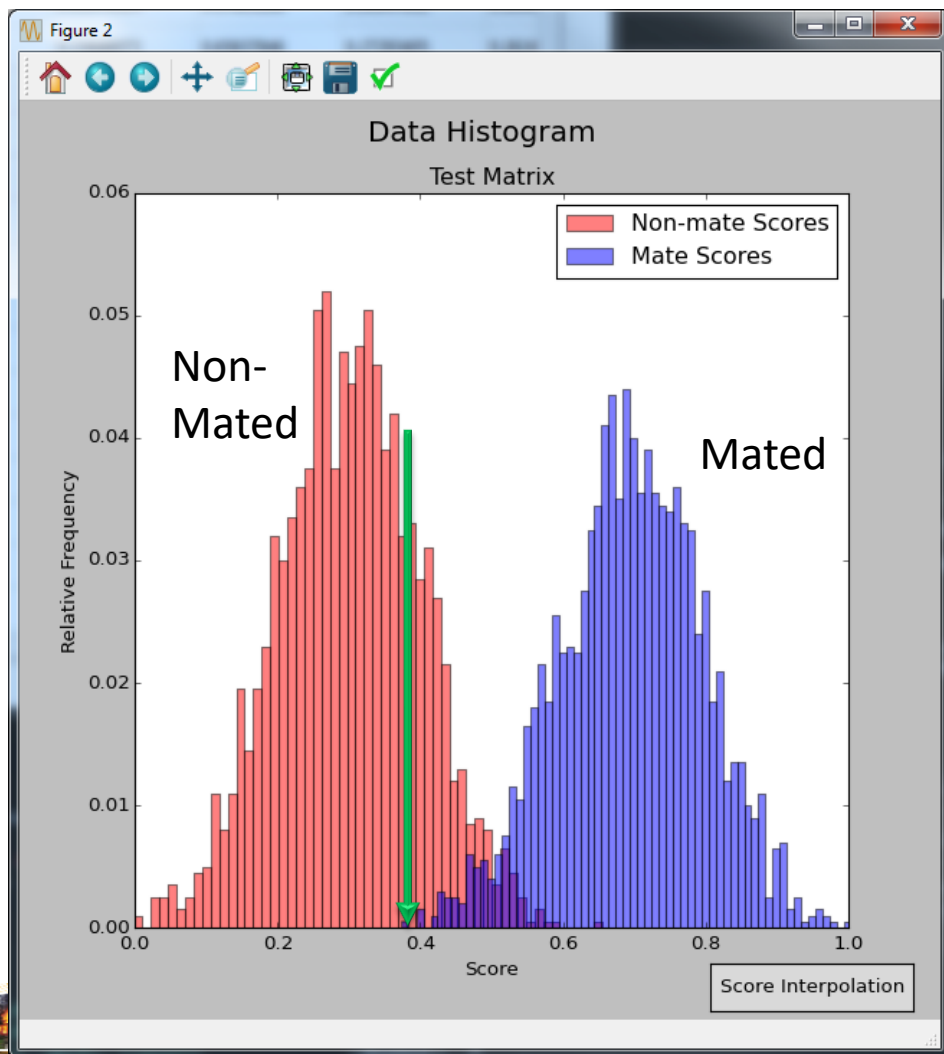
Context for Interpreting Comparison Scores - 2



- A score that lies mainly within mated pair scores indicates strong support for a match proposition.



Context for Interpreting Comparison Scores - 3

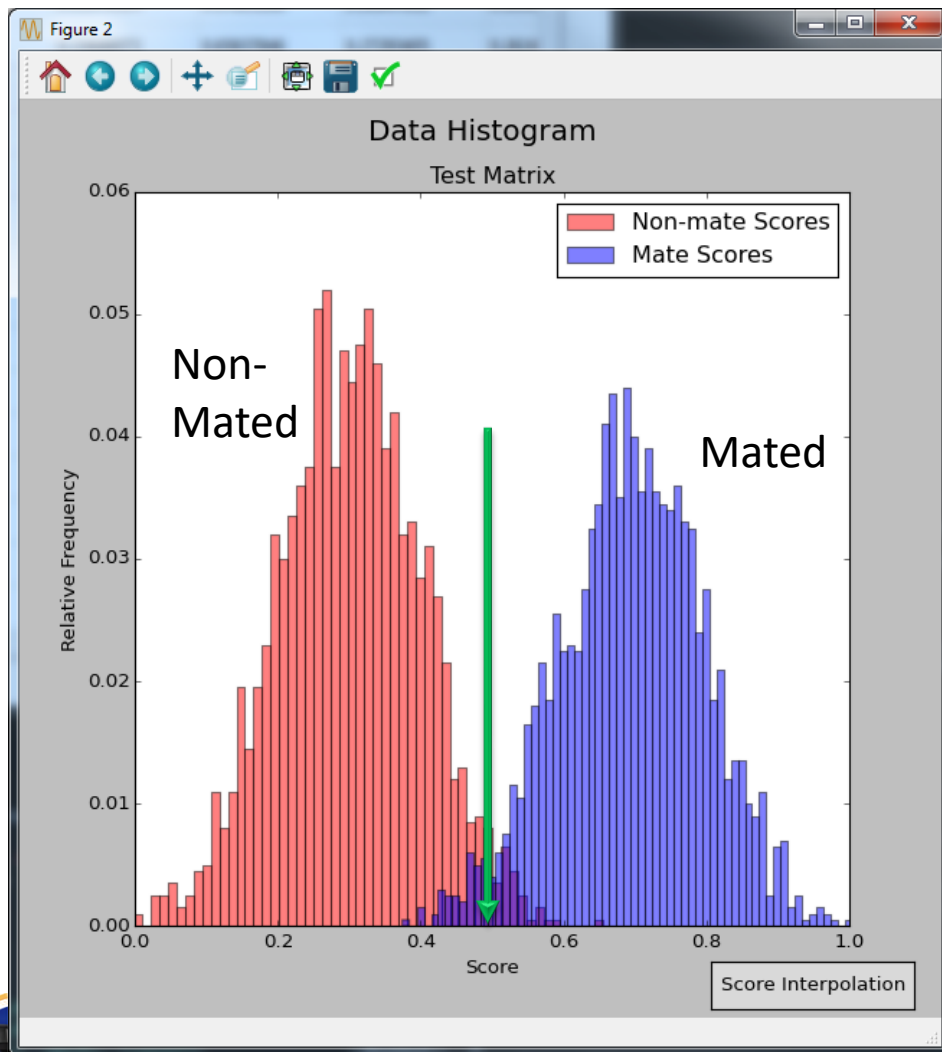


- A score that lies mainly within mated pair scores indicates strong support for a match proposition.
- A score that lies mainly within non-mated pair scores indicates strong support for a non-match proposition.

- Scores that represent significant pairs support neither proposition.



Context for Interpreting Comparison Scores - 4



- A score that lies mainly within mated pair scores indicates strong support for a match proposition.
- A score that lies mainly within non-mated pair scores indicates strong support for a non-match proposition.
- A score that occurs nearly equally often among mated and non-mated pairs does not provide support for either proposition.



Thank You

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