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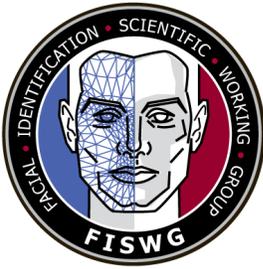
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## Face Recognition Systems Operation Assurance: Image Quality Assessment

### 26 **Purpose**

27 This document provides guidelines and techniques to help administrators of automated  
28 face recognition systems (FRS) produce advanced and accurate recognition statistics on  
29 the face recognition systems.

30 The intended audience of this document is system owners, system users, and system  
31 administrators of existing automated face recognition systems. Outside the scope of this  
32 document are include, but not necessarily limited to system setup, system tuning,  
33 workflow management and improvement, and proof of concept pilots.

34 This document is a follow on from the FISWG documents:

- 35 • “Understanding and Testing for Face Recognition Systems Operation Assurance”
- 36 • “Face Recognition Systems Operation Assurance: Identity Ground Truth”

37 The issues presented in this document form a base for other considerations and  
38 advanced topics when testing (e.g., system setup and tuning) which will be covered in  
39 future FISWG documents.

### 40 **1. Scope**

41 1.1 The scope of this document is to provide a detailed process and examples of

42 how testing for variations in facial image quality can be used in adjusting operational  
43 workflows. Testing for and verifying scoring variations when image quality varies is key  
44 so that facial search system workflows can be properly adjusted.

## 45 2. Referenced Documents

46 2.1 ANSI/NIST-ITL Standard Homepage:

47 [http://www.nist.gov/itl/iad/ig/ansi\\_standard.cfm](http://www.nist.gov/itl/iad/ig/ansi_standard.cfm)

48 2.2 P. Grother, M. Ngan, K. Hanaoka “NISTIR 8271 DRAFT SUPPLEMENT Face  
49 Recognition Vendor Test (FRVT) Part 2: Identification”

50 [https://pages.nist.gov/frvt/reports/1N/frvt\\_1N\\_report.pdf](https://pages.nist.gov/frvt/reports/1N/frvt_1N_report.pdf)

## 51 3. Terminology

52 3.1 *Definitions:*

53 3.1.1 *doppelganger, n*—an apparition or double of a living person.

54 3.1.2 *false accept rate, n*—see definition false match rate definition.

55 3.1.3 *false match rate, n*—the proportion of the completed biometric non-mated  
56 comparison trials that result in a false match. This is also referred to as false acceptance  
57 rate and does not include errors from images which do not create valid templates.

58 3.1.4 *false non-match rate, n*—the proportion of the completed biometric mated  
59 comparison trials that result in a false non-match. This is also referred to as false reject  
60 rate and does not include errors from images which do not create valid templates.

- 61 3.2 *Acronyms:*
- 62 3.2.1 *CMC–cumulative match characteristic*
- 63 3.2.2 *DET–detection error tradeoff*
- 64 3.2.3 *FAR–false accept rate*
- 65 3.2.4 *FMR–false match rate*
- 66 3.2.5 *FNMR–false non-match rate*
- 67 3.2.6 *FR–facial recognition*
- 68 3.2.7 *FRR–false reject rate*
- 69 3.2.8 *FRS–facial recognition system*
- 70 3.2.9 *ROC–receiver operating characteristics*

## 71 **4. Background**

72 4.1 When doing *accuracy* profiling, a *critical* step is to understand how image quality  
73 affects the processing of the imagery to be enrolled or searched. Regardless of the  
74 specific facial biometric algorithm used, there will be some facial imagery which causes  
75 facial localization errors which then produce biometric templates of no operational value.

76 4.2 Most of the work in these *processes* is on creating the testing frameworks and  
77 understanding how to *repeatedly* run tests, make corrections, and do retesting with what  
78 has been learned. Once the frameworks and the processing are understood, then the  
79 agency can make diligent progress, but it takes time and focus. The outcomes are worth  
80 the time spent as you begin to understand how the data interacts with the algorithms  
81 which give the agency the ability to trust the solution and not just assume the data is  
82 invalid.

83 4.3 Setting up frameworks to do *enrollment* and searching recording results is fairly  
84 mechanical as you learn the facial algorithms and the data sets to develop proper  
85 profiling. Understanding the data and building frameworks to analytically qualify the  
86 results is not trivial but must be done so effective operational metrics can be derived and  
87 applied.

## 88 5. Data Set

89 5.1 Care should be taken in selecting data sets to profile. It is recommended to  
90 select data sets which:

91 5.1.1 Have operational relevancy;

92 5.1.2 Have consistent image quality aspects: type of capture, size of images, subject  
93 poses, etc.;

94 5.1.3 Have sufficient identities and images to test with. This decision will be agency  
95 specific; and

96 5.1.4 Include associated identity truth information.

97 5.2 The data set used for this document is the Labeled Faces in the Wild (LFW) data  
98 set available at: <http://vis-www.cs.umass.edu/lfw/> See section “LFW Data Set  
99 Information” at <http://vis-www.cs.umass.edu/lfw/#information> and the referenced  
100 paragraph 2.2 document (NISTIR 8271 DRAFT SUPPLEMENT Face Recognition Vendor  
101 Test (FRVT) Part 2: Identification) for more information. Conceptually any other facial  
102 data set with identity ground truth can be used.

103 5.3 LFW is a widely used open source data set which will work well for this specific  
104 document. Information on this data set includes:

105 5.3.1 *Has smaller but consistent image sizes and file formats;*

106 5.3.2 *Has over 5,700 identities and over 13,000 images;*

107 5.3.3 *Has a wide range of subjects: sex, pose, lighting, etc.; and*

108 5.3.4 *Has stated identity ground truth errors.*

## 109 **6. Image Quality Assessment Process**

110 6.1 **Step 1** - Ensure the data set to use has verified ground truth.

111 6.2 **Step 2** - Extract the facial image quality from all images to enroll and search.

112 The image quality metrics to use will be vendor dependant. Care should be taken to  
113 select the quality metric(s) which have the largest value in doing an image quality  
114 assessment. Vendor specific quality metrics will have different numeric ranges and will  
115 have vendor specific usability ranges. Consulting with the provider of the facial algorithm  
116 is recommended.

117 6.3 **Step 3** - Enroll the facial images into a facial gallery for searching.

118 6.4 **Step 4** - Search the facial images against the facial gallery. The number of  
119 candidates returned for this document was 50. This number may vary with agency  
120 specifics. Do not use any scoring thresholds.

121 6.5 **Step 5** - Segment facial search results based on the facial image quality of the  
122 probe.

123 6.6 **Step 6** - For each segmentation group of the search results plot the biometric  
124 performance results and compare them. This document uses these plots:

125 6.6.1 FAR: False accept scoring

126 6.6.2 FRR: False reject scoring

127 6.6.3 DET: Detection error tradeoff

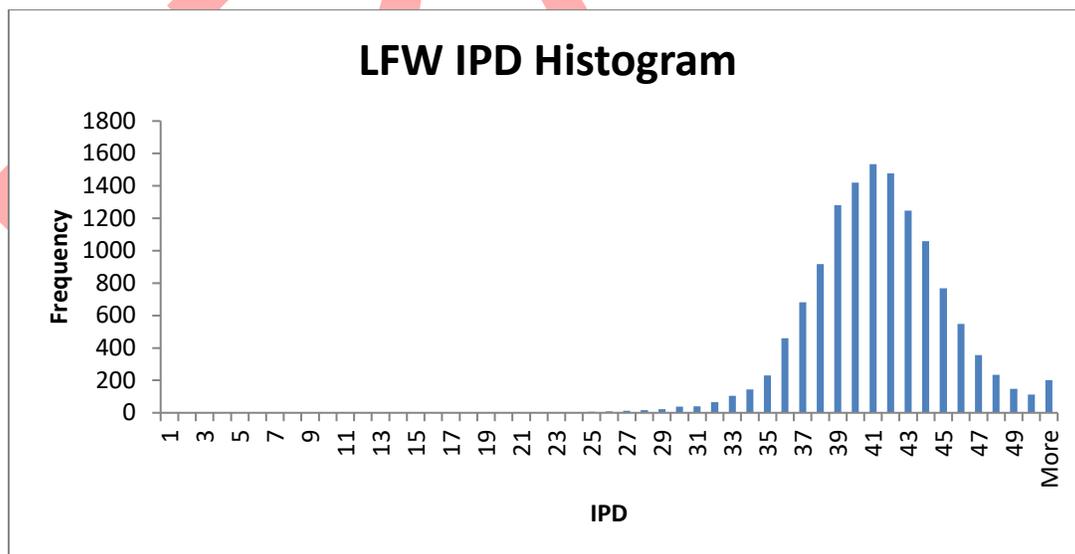
128 6.6.4 CMC: Cumulative Match Curve

129 6.7 **Step 7:** Determine if targeted image quality assessments can be done before  
 130 enrollment or searching to see if improper facial localization has occurred which caused  
 131 facial templates to be created which were of no operational value and should have either  
 132 been manually reviewed or simply not processed.

## 133 7. Image Quality Assessment Process Step Outputs

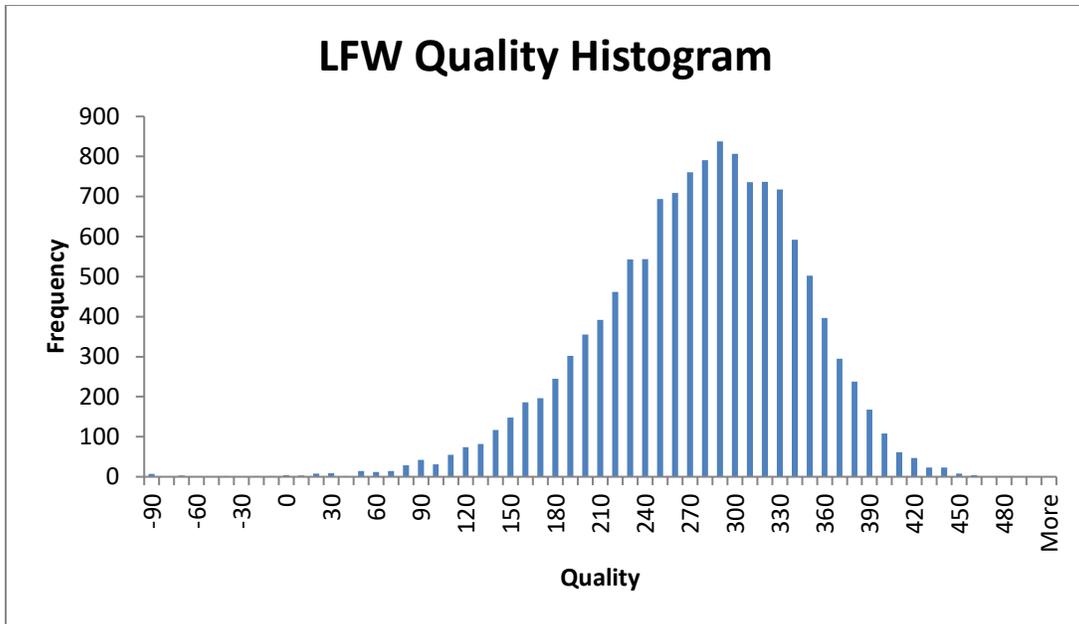
### 134 7.1 Step 2 Outputs:

135 7.1.1 When this was done with the LFW data set the following facial metrics were  
 136 recorded:



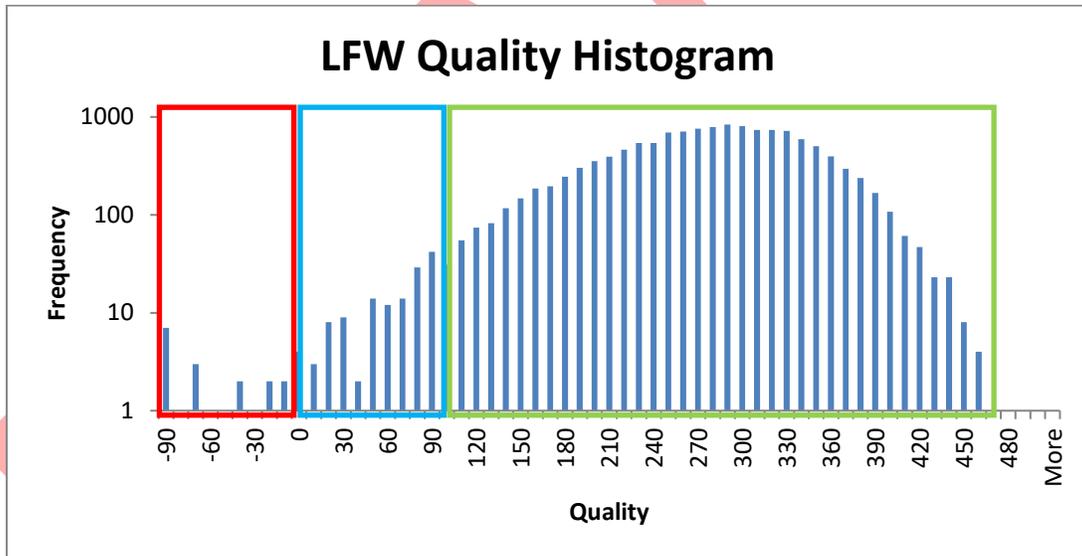
137  
138

Figure 1: LFW Interpupillary Distance (pixels)



139  
140  
141

Figure 2: LFW Quality (linear Y axis)



142  
143

Figure 3: LFW Quality (logarithmic Y axis)

144 7.1.2 The LFW images were then enrolled and searched. Based on the image quality  
 145 histograms three segments were selected:

146 7.1.2.1 All search results

147 7.1.2.2 All searches with a quality less than 0

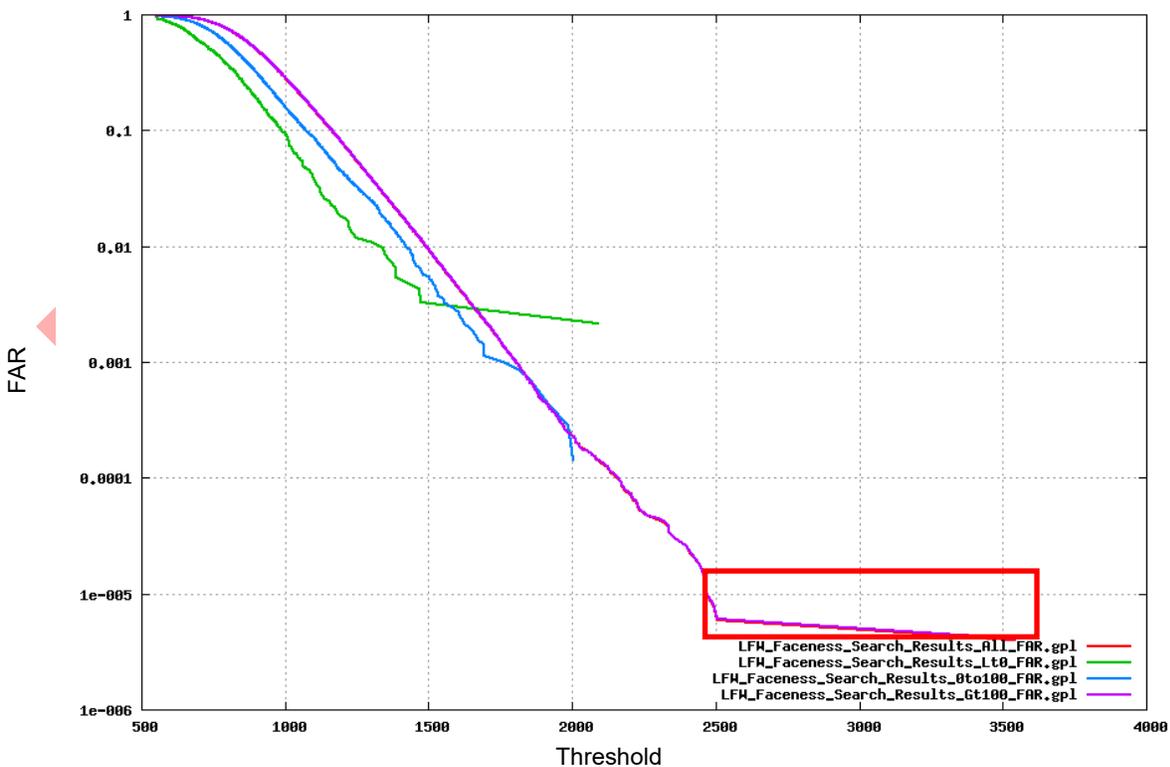
148 7.1.2.3 All searches with a quality between 0 and 100

149 7.1.2.4 All searches with a quality greater than 100

150 7.2 Step 6 Outputs:

151 7.2.1 Once the accuracy plots were created, they could be analyzed.

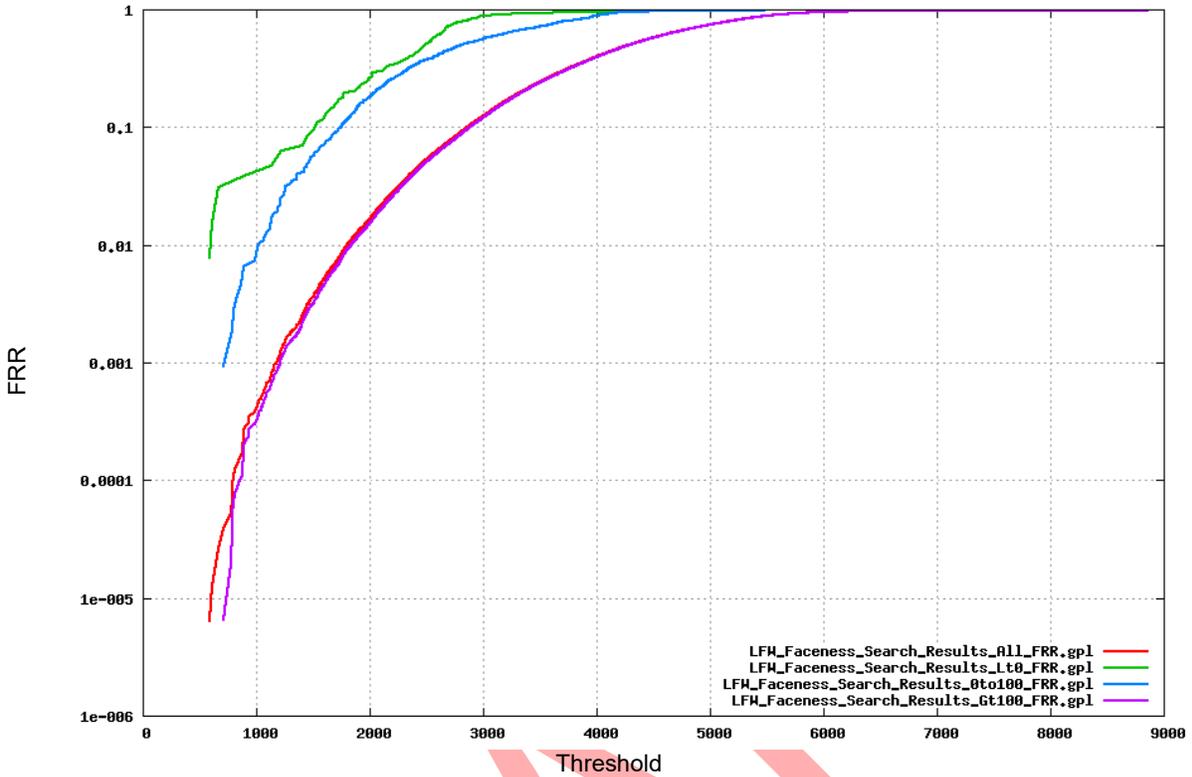
152 7.2.2 FAR and FRR were analyzed first.



153

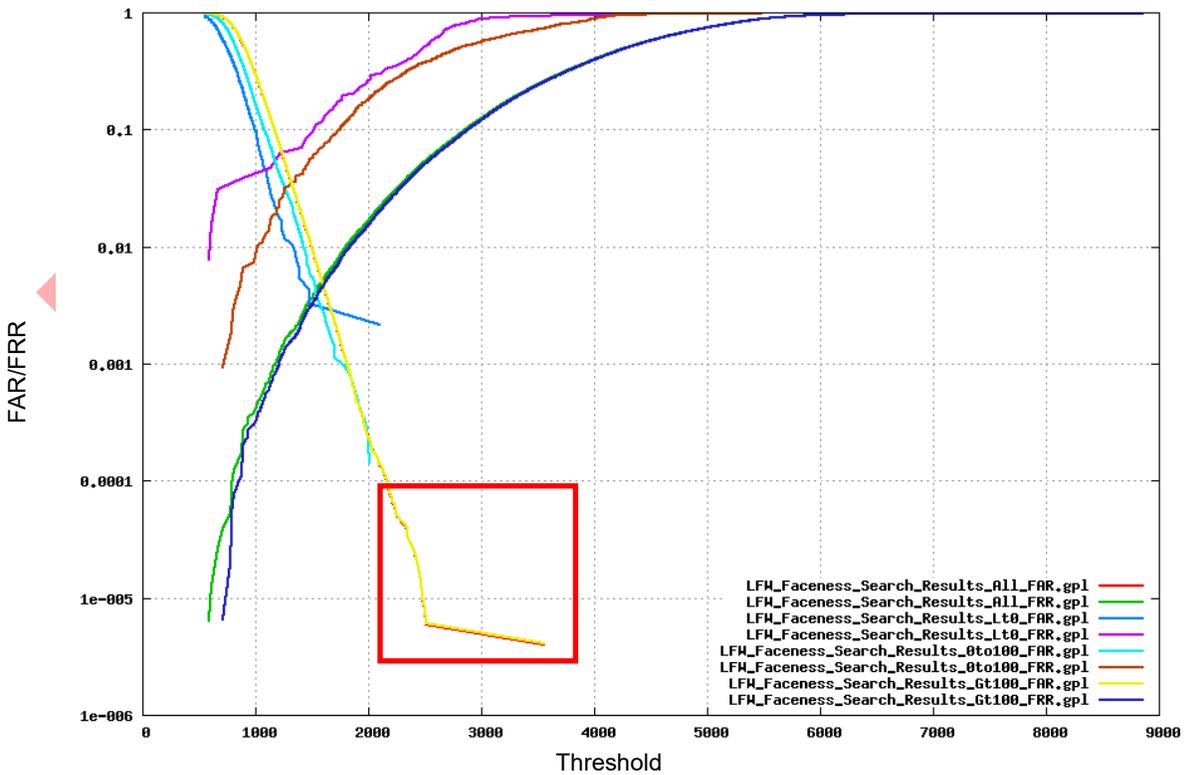
154

Figure 4: LFW FAR Imposter Scores (logarithmic Y axis)



155  
156

Figure 5: LFW FRR Mate Scores (logarithmic Y axis)



157  
158

Figure 6: LFW Mate and Imposter Scores (FAR and FRR)

159        **7.2.3 Notes on Step 6 FAR and FRR analysis:**

160        7.2.3.1 The overall imposter (FAR) scoring in Figures 4 and 6 was relatively  
161 consistent regardless of probe image quality with high values of ~2500.

162        7.2.3.2 The higher imposter (FAR) scoring in Figures 4 and 6 (red box) was due to  
163 the presence of known twins and several siblings and doppelgangers.

164        7.2.3.3 The highest imposter (FAR) score of ~3500 in Figures 4 and 6 (red box) was  
165 due to known twins in the LFW data set.

166        7.2.3.4 The overall mate scoring (FRR) in Figures 5 and 6 did vary with probe image  
167 quality with the lowest values of ~750.

**LFW Twin**



168

### LFW High Scoring Imposters

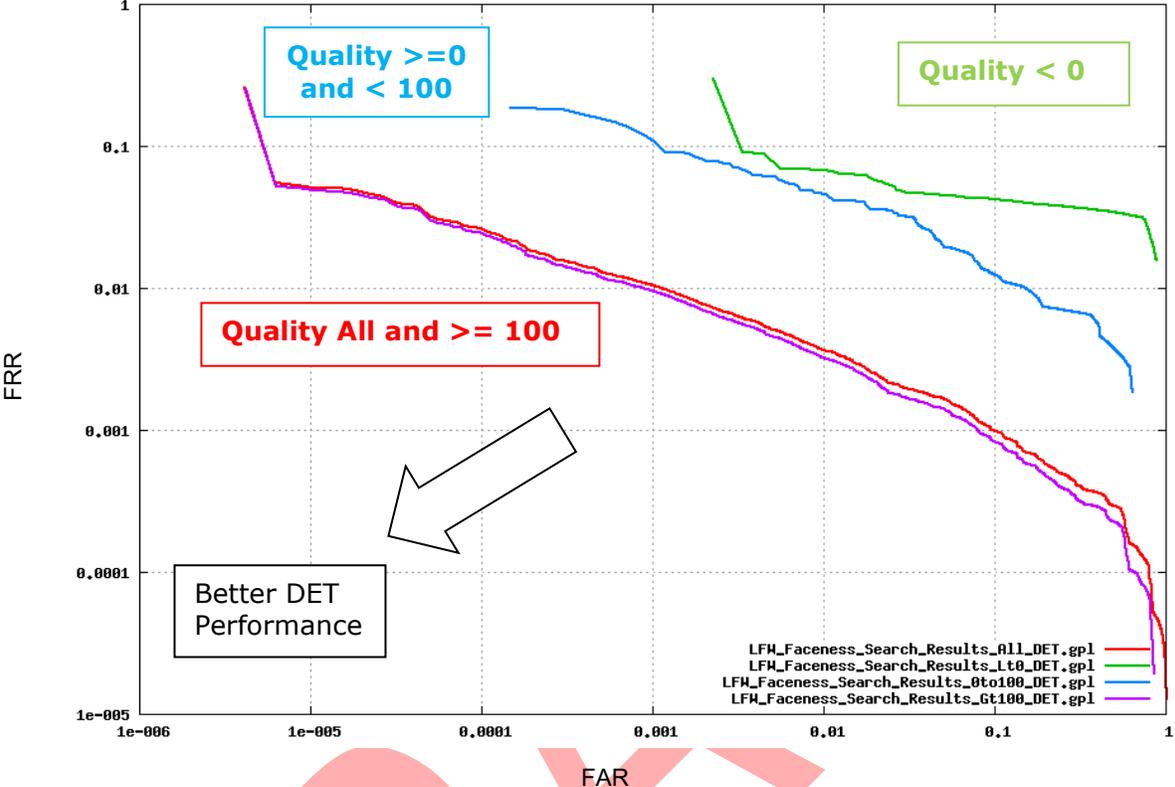


The images above are siblings



The images above are doppelgangers

170 7.2.4 DET was analyzed next.



171 FAR  
172 **Figure 7: LFW DET Curve**

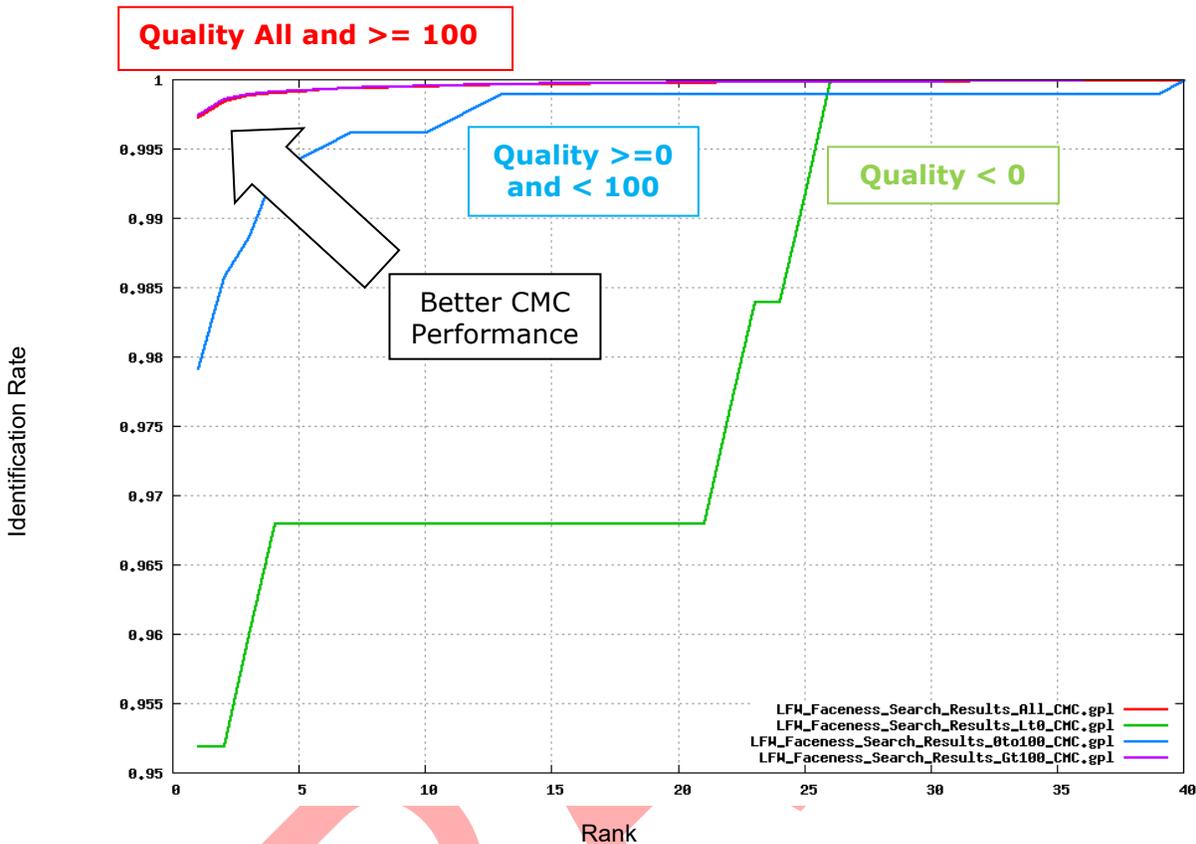
173 7.2.5 Notes on Step 6 DET analysis:

174 7.2.5.1 The DET curve performance in Figure 7 was better with higher probe image  
175 quality.

176

177 7.2.6 CMC was analyzed next.

178



179

180

Figure 8: LFW CMC Curve

181 7.2.7 Notes on Step 6 CMC analysis:

182 7.2.7.1 The CMC curve performance in Figure 8 was better with higher probe image  
183 quality.

184 7.2.8 Perhaps the largest outcome of this testing is presented in the CMC curves and  
185 can be described as follows:

186 7.2.8.1 The lowest image quality plots (< 0) have a 100% CMC point at a search  
187 results candidate of ~27.

188 7.2.8.2 The medium image quality plots (>= 0 and < 100) has a 100% CMC point at

189 a search results candidate of ~13.

190 7.2.8.3 The highest image quality plots ( $\geq 100$ ) has a 100% CMC point at a search  
191 results candidate of ~10.

192 7.2.8.4 So, in order to achieve the 100% CMC point for all image quality variations  
193 the number of candidates a human examiner must review is ~27.

194 7.2.8.5 If the overall image quality can be improved through proper image quality  
195 assessment during enroll and searching, then it would be possible to lower number of  
196 candidates a human examiner must review to achieve a desired 100% CMC point.

### 197 7.3 Step 7 Outputs:

198 7.3.1 The search probes were then analyzed to locate low quality imagery so the  
199 candidate lists returned could be reviewed to see if any mates were returned.

This probe had 51 mates in the gallery. Only one was returned at rank 3. The probe quality was -186.



This mate which was returned at a rank 3 with a score of 1213. Note the bad eye locations which were similar. The probe quality was -163.



This probe had 0 mates in the gallery. The probe quality was -164. This example shows how the incorrect face in the image was localized.



This probe had 0 mates in the gallery. The probe quality was -123. This example shows how improper facial localization occurred.



<p>This probe had 7 mates in the gallery. Only two were returned. The probe quality was -84.</p>	
<p>This mate which was returned at a rank 26 with a score of 575.</p>	
<p>This probe had 530 mates in the gallery. No mates were returned. The probe quality was -19.</p>	

This probe had 8 mates in the gallery.  
No mates were returned. The probe  
quality was 22.



## 200 8. Image Quality Assessment Outcomes

201 8.1 Based on this data set and the testing process documented here:

202 8.1.1 Facial search performance is affected by facial image quality which will be  
203 vendor dependent. Various image quality metrics needed to be analyzed to profile the  
204 facial image quality into specific ranges of interest.

205 8.1.2 The ability to segment search results based on facial image quality is shown.

206 8.1.3 Mates and imposter scoring profiles can change with mates being more  
207 susceptible to image quality issues.

208 8.1.4 FAR, FRR, DET and CMC curves were utilized in these processes.

209 8.1.5 How the facial image quality scoring and resultant operational ramifications  
210 need to be profiled to properly locate low scoring mates in search results which have  
211 lower image quality.

212 8.1.6 Detecting and correcting low quality facial imagery which could be manually  
213 corrected could improve overall facial accuracy.

214 8.1.7 Enrolling and searching all facial imagery without attention to image quality will

215 negatively affect facial workflows.

216

FISWG documents can be found at: [www.FISWG.org](http://www.FISWG.org)

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