

# Measuring Biometric Sample Quality By Biometric Information

Andy Adler<sup>1</sup>, Richard Youmaran<sup>2</sup>

<sup>1</sup>Carleton University, <sup>2</sup>University of Ottawa  
Ottawa, Canada

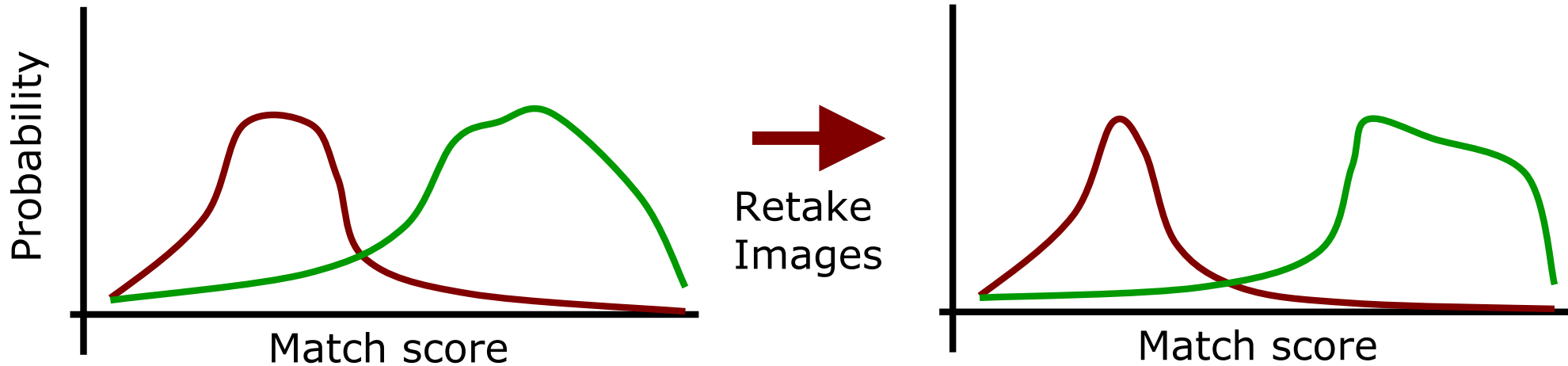
# Biometric Sample Quality

## Biometric Sample Quality views

- character
  - inherent features
- Fidelity
  - accuracy of features
- utility
  - predicted biometrics performance

INCITS, Biometric Sample Quality Standard Draft,  
M1/06-0003

# Utility



If images match better (ie. lower errors)  
then samples were better

# Utility

- ❑ Fairly simple conceptually
- ❑ Dependent on matching algorithm
- ❑ Doesn't allow quantification of "inherent" quality

# Character / Fidelity

Descriptions of “inherent” quality of a biometric sample

## □ Character

- Blur
- Shadows
- Poor lighting

## □ Fidelity

- A good image of the wrong part

# Example: *Character*

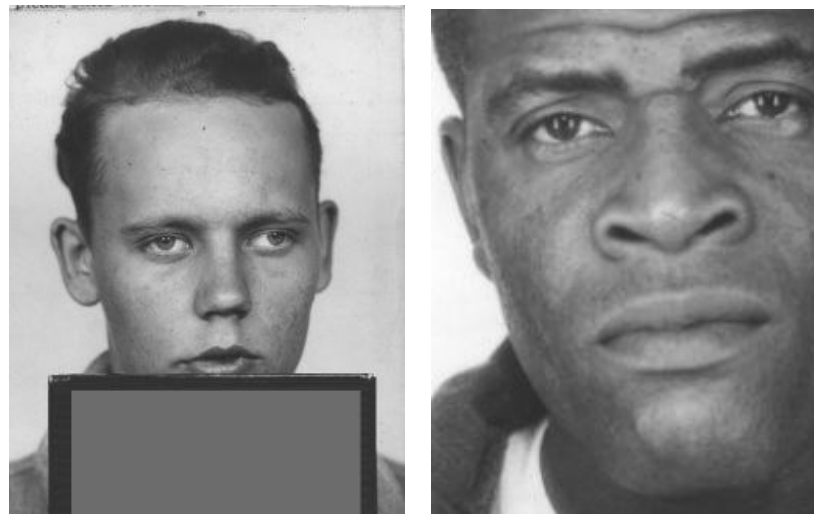
← Best Faces

**Human Selections**

Worst Faces →



# Example: *Fidelity*



# How can we measure quality

□ Probing question:

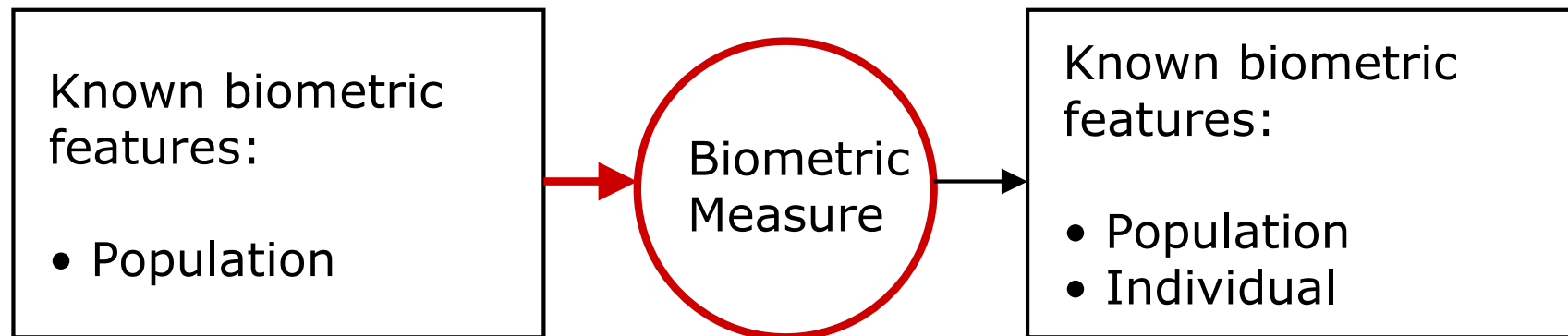
*Why do we worry about low quality images?*

□ Answer:

*They have less "biometric information"*

# Biometric Information

- We define “biometric information” as:  
*the decrease in uncertainty about the identity of a person due to a set of biometric measurements.*





# Biometric Information:

*relative entropy*  $D(p||q)$

$$BI = D(p || q) = \int p(\mathbf{x}) \log_2 \frac{p(\mathbf{x})}{q(\mathbf{x})} d\mathbf{x}$$

## □ Distributions

- Individual,  $p(\mathbf{x})$

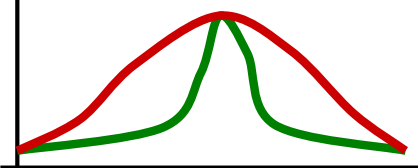
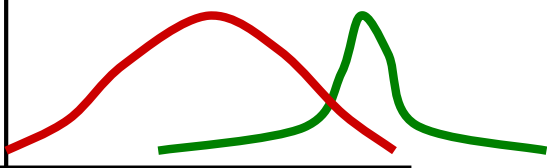
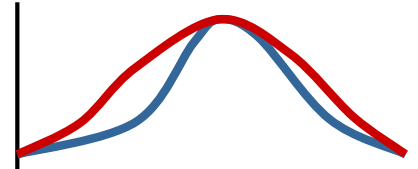
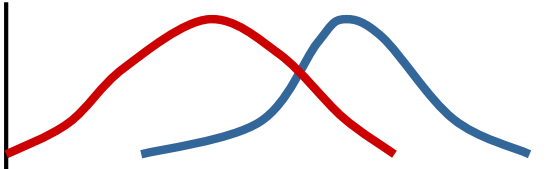
- Population,  $q(\mathbf{x})$

□  $D$  measures extra information in  $p$  than  $q$

## □ Distribution models

- Gaussian models, PCA features, regularization

# Example: Height

	Average (5½' tall)	Tall (6½' tall)
Measure #1 (doctor's office =accurate)	 $D = 0.23$ bits	 $D = 2.7$ bits
Measure #2 (via telescope =inaccurate)	 $D = 0.05$ bits	 $D = 1.1$ bits

What is the *quality difference* between measures?

# Quality of a biometric measure

- quality difference between “instruments”  $f$  and  $g$  is

$$\Delta BI = \frac{\frac{1}{N_f} \sum_{i=1}^{N_f} (D(p_{f_i} \| q_{f_i}) - D(p_{g_i} \| q_{g_i}))^2}{\frac{1}{N_f} \sum_{i=1}^{N_f} (D(p_{f_i} \| q_{f_i}))^2}$$

Number of features  $\rightarrow$

- Person#1:  $\Delta BI = (.23 - .05 / .23)^2 = .61$   
Person#2:  $\Delta BI = (2.7 - 1.1 / 2.7)^2 = .36$

# Application #2: Face Recognition

## Aberdeen Face database

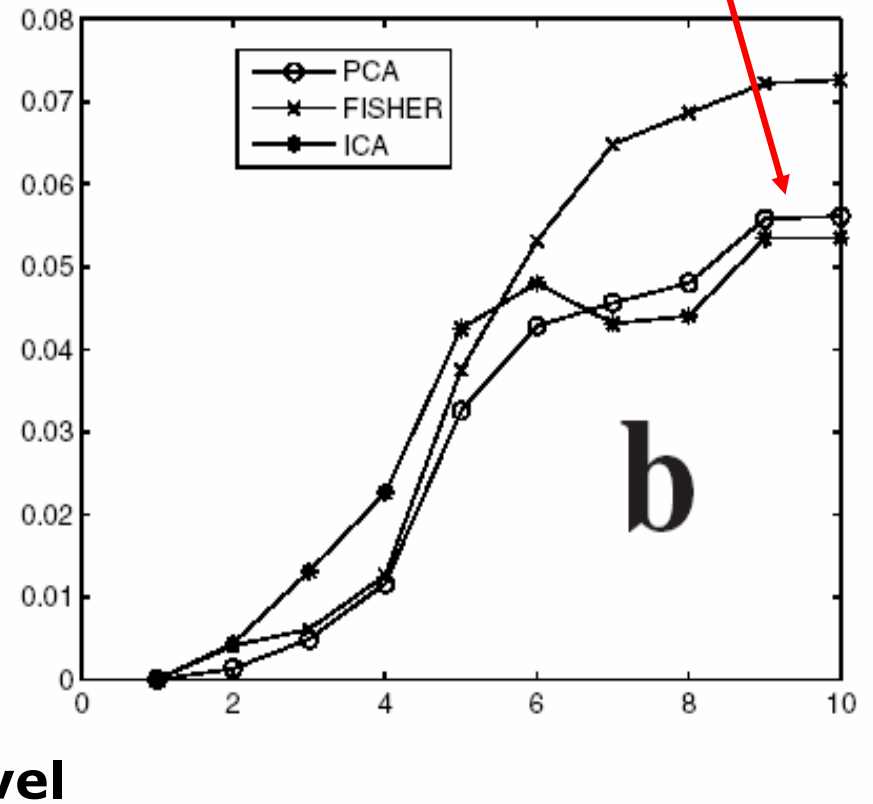
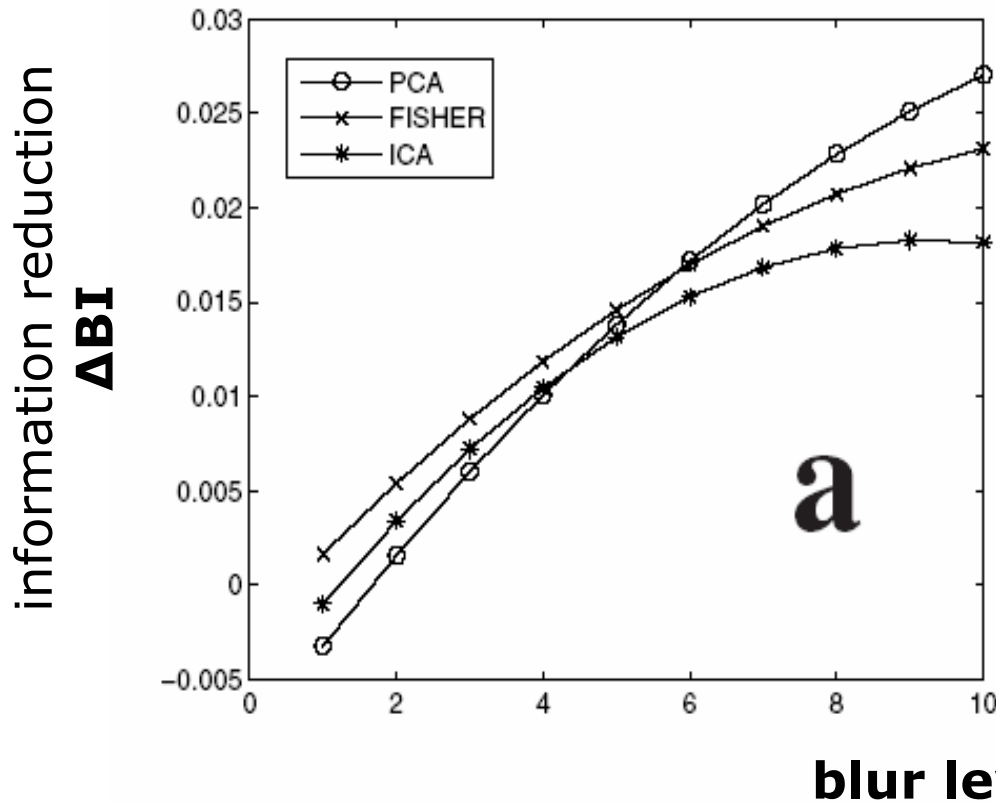
- 18 frontal images of 16 persons
- Variability in lighting and expression between images
- $D(p||q)$  computed for 100 features using
  - PCA (eigenface)
  - FLD (fisherface)

# PCA / FLD / PCA+FLD

<b>PCA</b>	<b>FLD</b>	<b>PCA+FLD</b>
<i>45.0 bits</i>	<i>37.0 bits</i>	<i>55.6 bits</i>

- Extra information from both PCA and FLD is small.
- Values seem reasonable.
  - Our extrapolations from FRVT2002 give 27.3 bits for lead algorithm

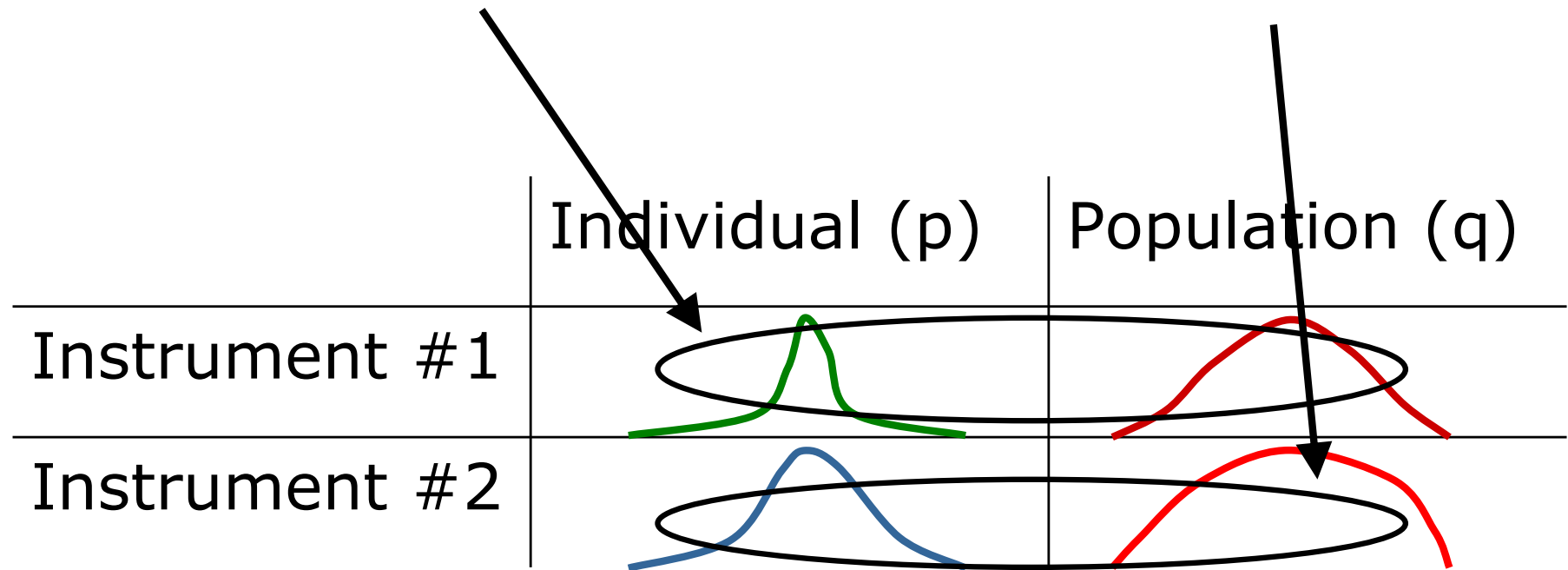
# Quality decrease with blur



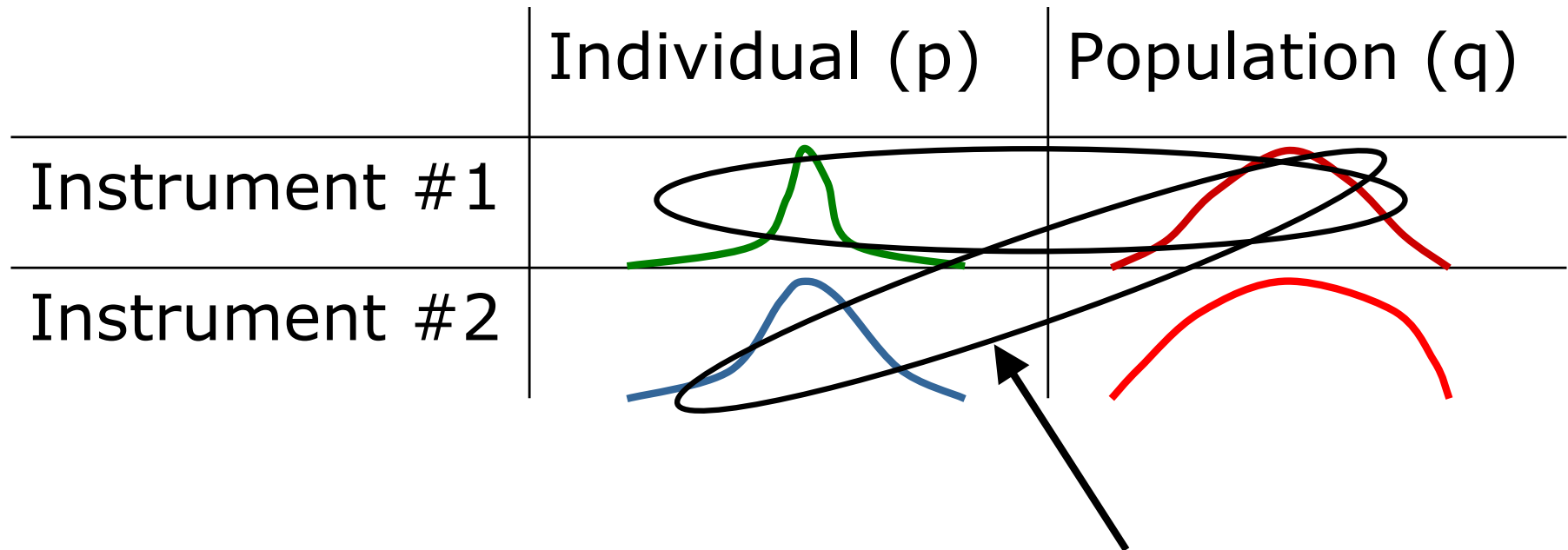
# Aside: our initial mistake

□ Must compare

(Instrument #1) – (Instrument #2)



# Our wrong calculation



- ❑ We compared blurry instrument to the clean population
- ❑ This increased biometric information
- ❑ The algorithm says: *I can recognize p. He always has a blurry face!*



# Summary

- relationship between *biometric quality* and *biometric information*
- A method to measure the quality change due to an image degradation
- *Limitation*: Can't measure quality of a single image

# Applications

- ❑ Clarify nature of biometric quality measures
- ❑ Help quantify limits of impact of quality on matcher performance
- ❑ Help quantify effects of biometric fusion with low quality data
- ❑ Privacy impact of approaches to de-identify face data

# Comment: *Quality*

- ❑ *Quality* is a value laden term
- ❑ Can we tell users this?



- ❑ Maybe we need another term:  
*Clarity?*