

# *Fingerprint Quality Analysis: a PC-aided approach*

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A. Mattei, Ph.D,\*  
F. Cervelli, Ph.D,\*  
F. Zampa, M.Sc,\*  
F. Dardi, Ph.D<sup>§</sup>

\* *RaCIS, Italy*

§ *Innovation factory*

# Summary



- Motivation
- Forensic quality of images
- Generic purpose contrast index
- Human visual system contrast index
- Results
- Conclusions
- Future works



To evaluate different enhancement techniques:

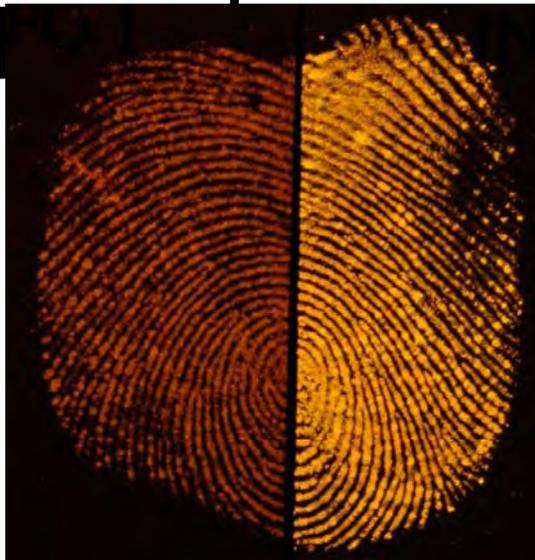
- Can we suggest an **objective way to compare** the results?
- Can we find an objective way to rank the **effectiveness of different development techniques** from the point of view of the **forensic expert**?

# Experimental Setup



- Fingerprints left on paper
- Paper cut in two, developed with different reagents and then

com



# Purpose of Comparison



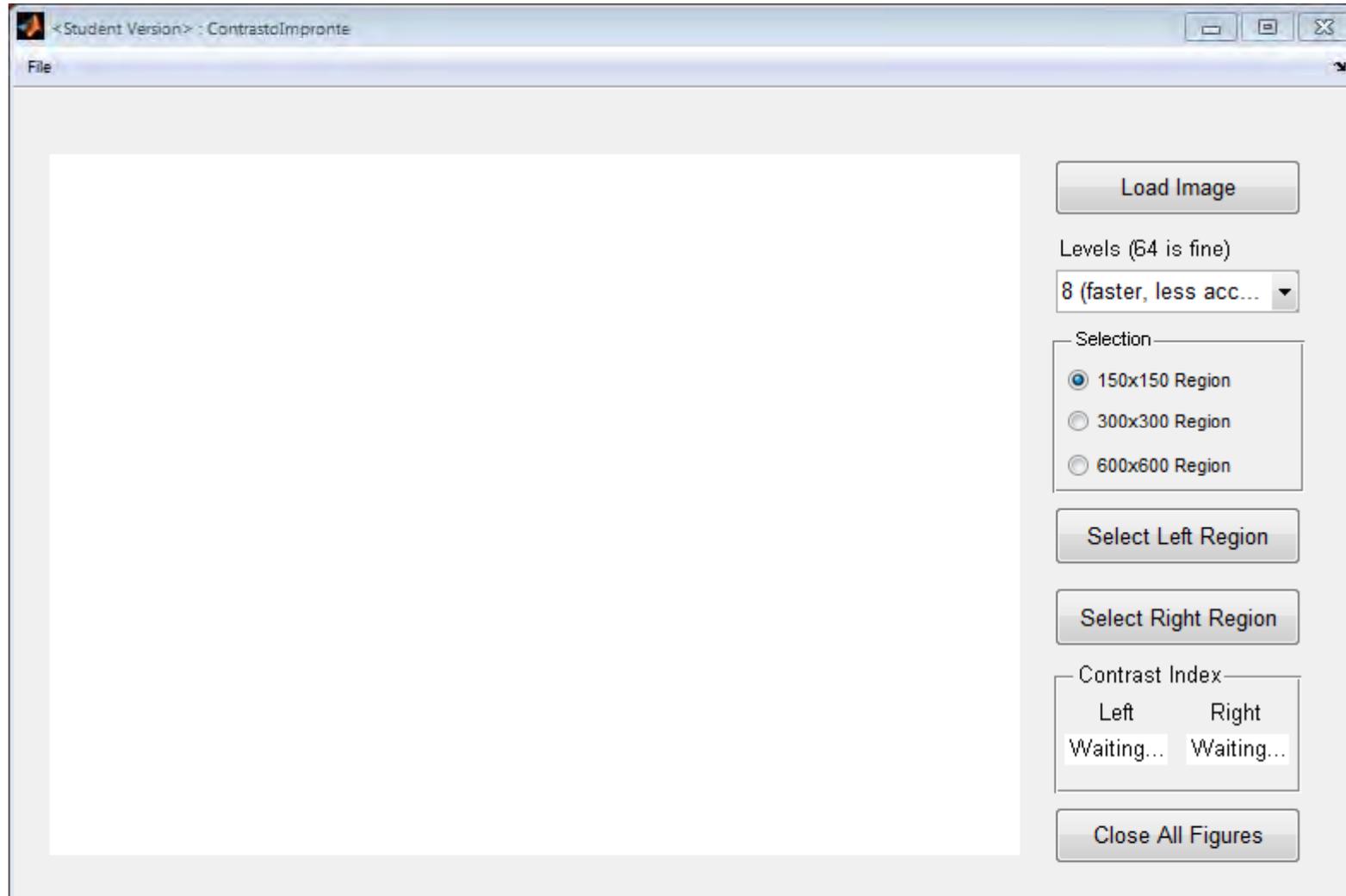
One to one comparison to see which half of the **same fingerprint** was developed "better"

# Test Set-up

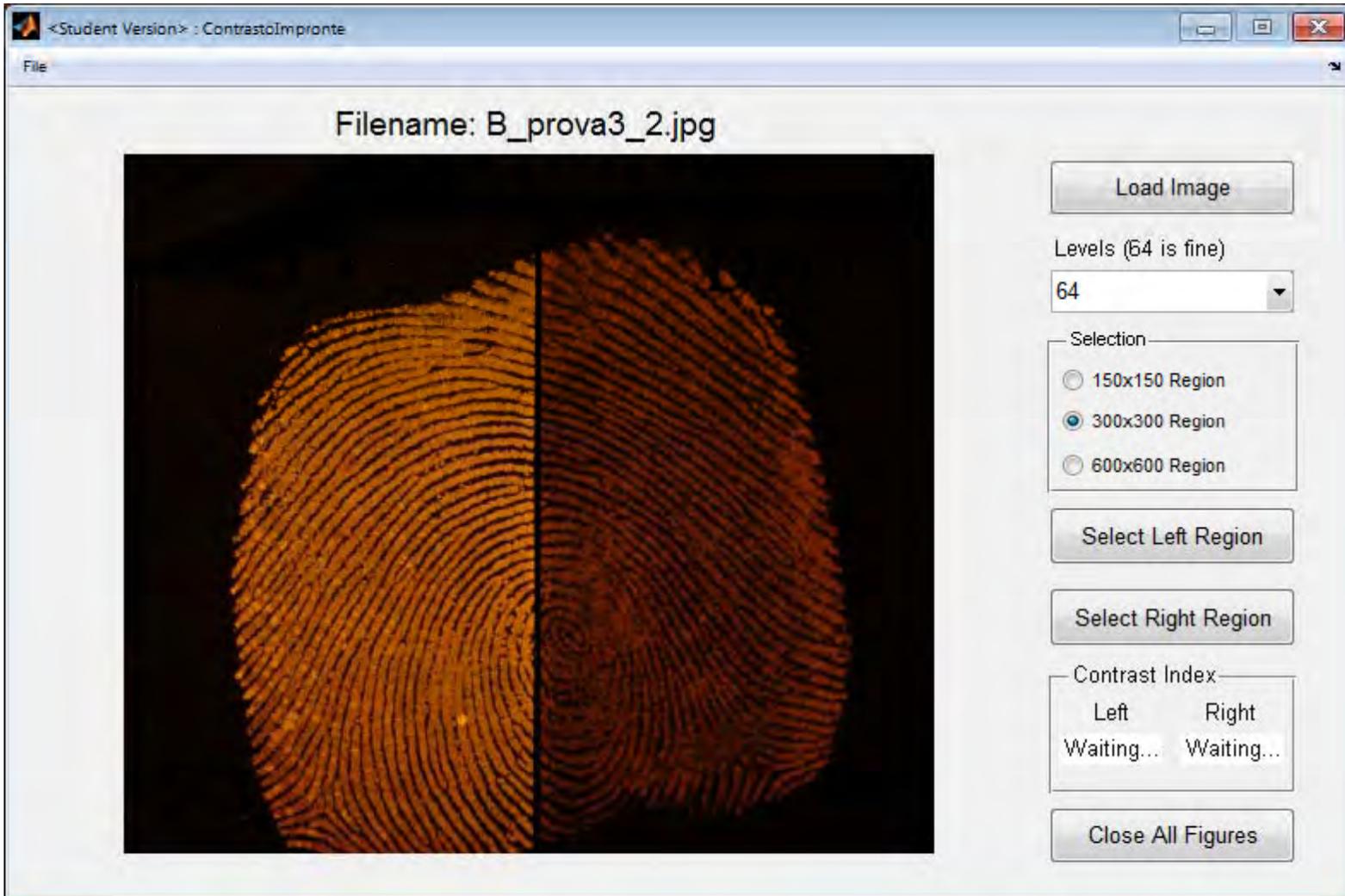


- All fingerprints acquired at a constant distance from the camera
- Camera settings and light for fluorescence are changed to the expert's opinion
- Each fingerprint halves are acquired together

# Fingerprints: How Does It Work?

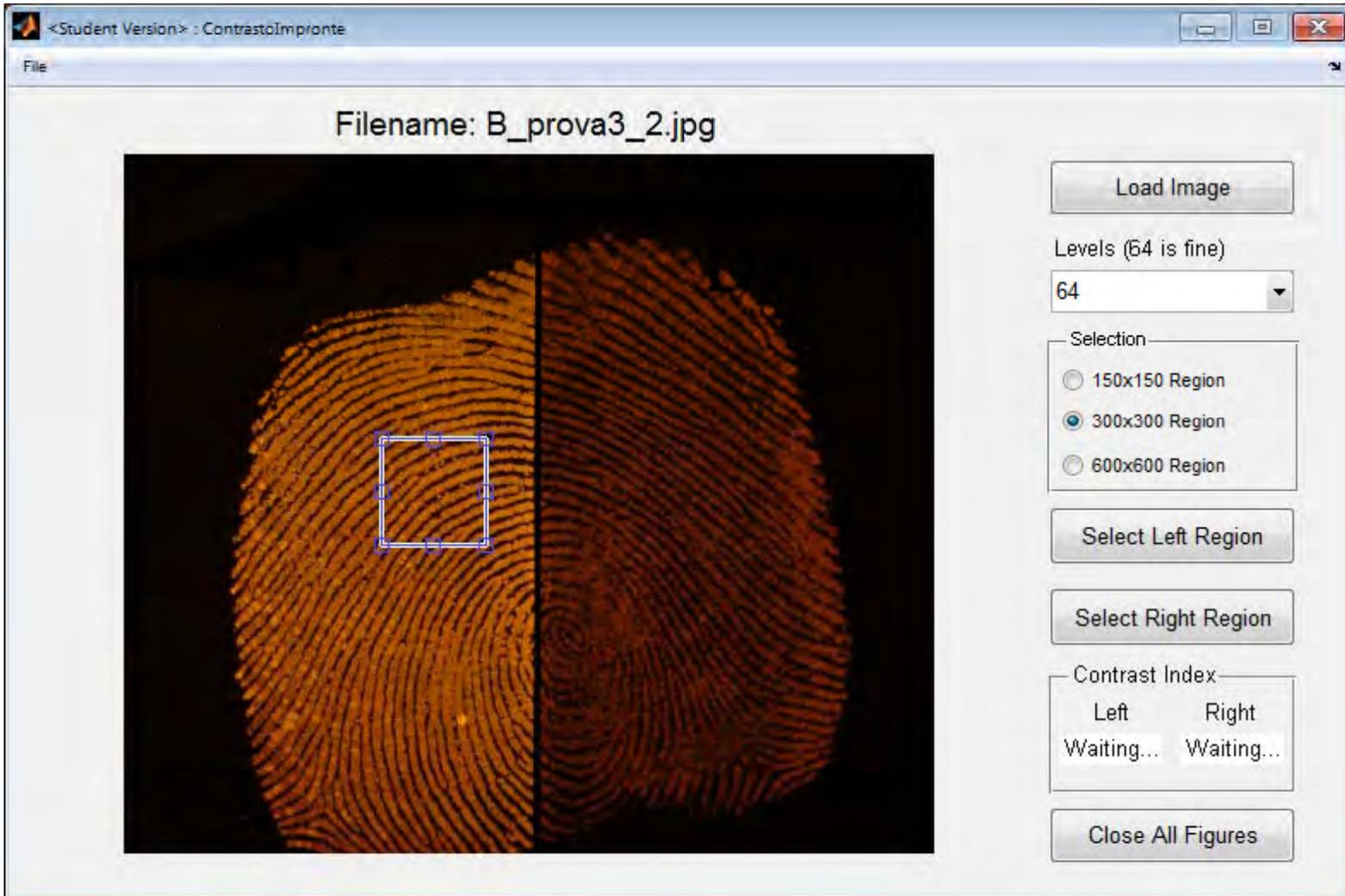


# Fingerprints: How Does It Work?

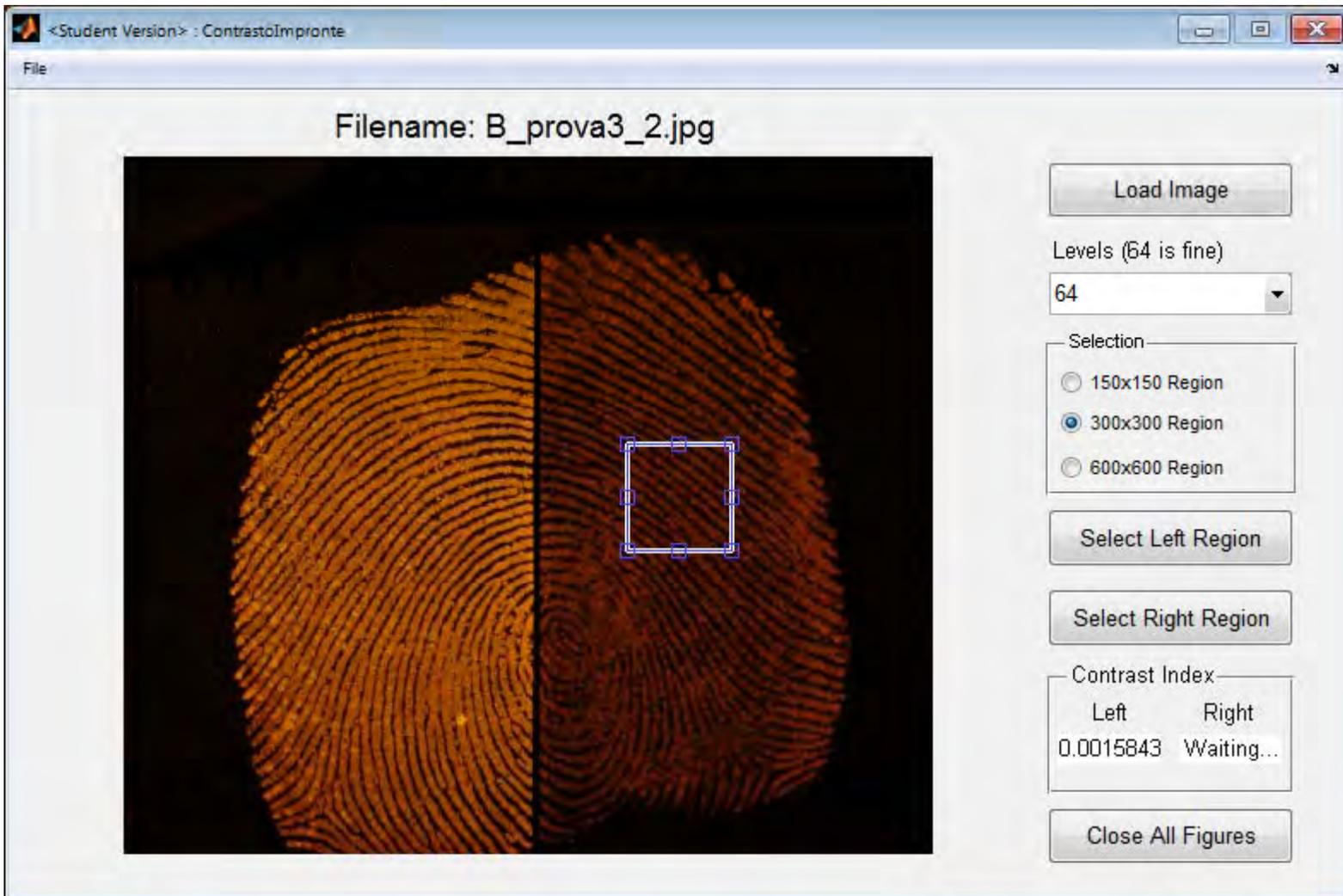




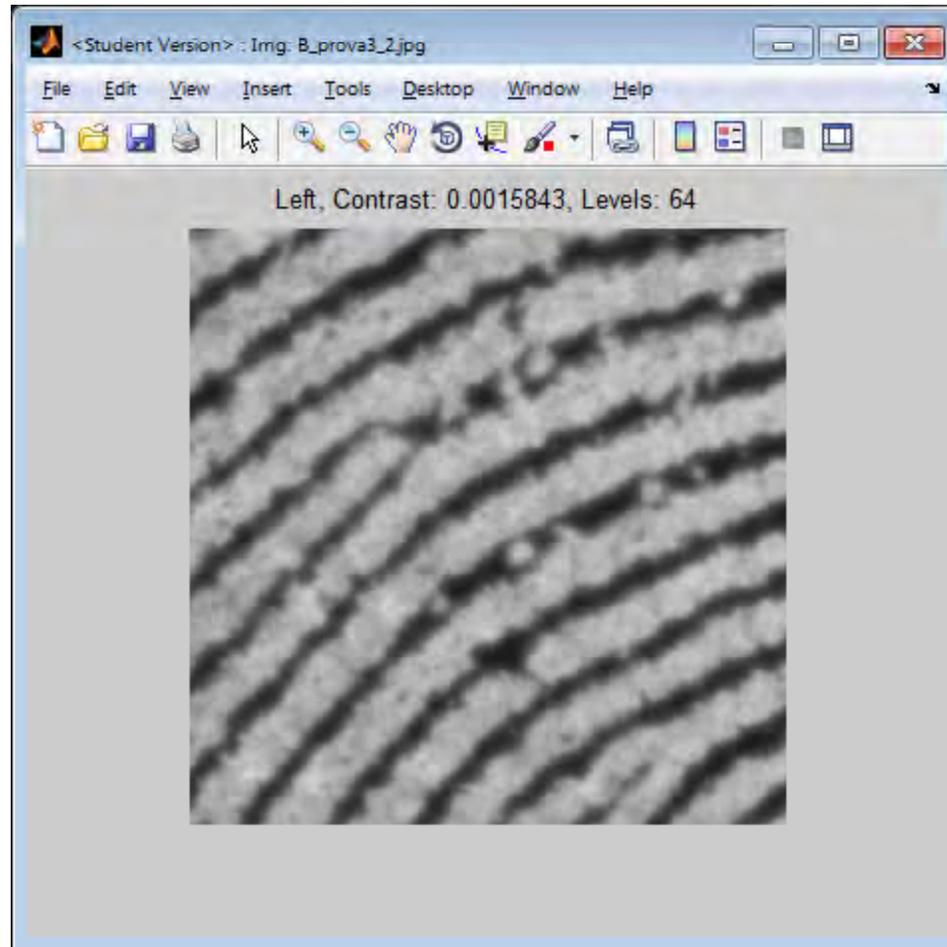
# Fingerprints: How Does It Work?



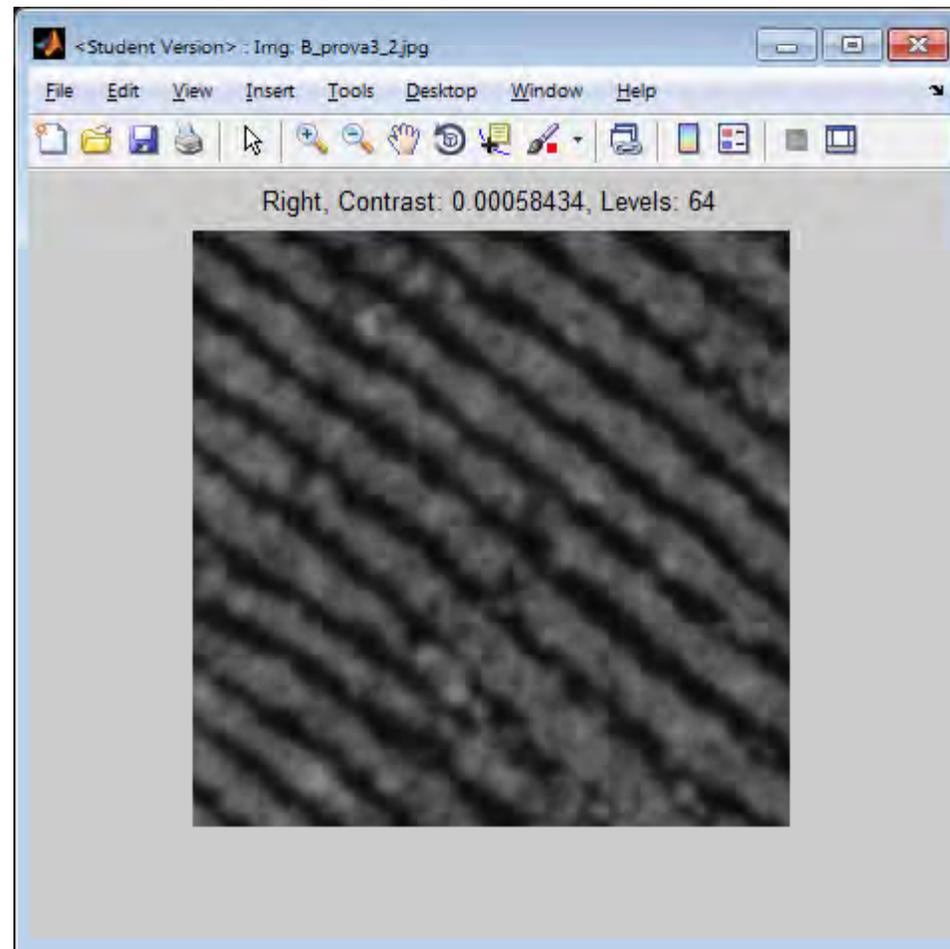
# Fingerprints: How Does It Work?



# Fingerprints: How Does It Work?



# Fingerprints: How Does It Work?



# Consequence



- We can compare fingerprints deposited at different times
- Then, we can compare the expert's opinion to the software outcome and see how they compare and **teach** the software how to **rank fingerprint quality**

# Consequence



- If done properly, this will be useful to assess the forensic quality of fingerprint well before they are even shown to the expert

# Extend the Concept



- Change the word "fingerprint" with the **forensic image of your choice**



# Easy? Maybe not ....



- Need to translate the concept of forensic quality in a PC computable quantity
- **Forensic quality**: usefulness for forensic analysis
- We chose to use **contrast in order to capture forensic quality**





- We have to choose a contrast computation method to evaluate the forensic quality of an image
- Methods fall in three main categories:
  - **general** purpose
  - image **specific** (knows the kind of image it is looking at)
  - human visual system (**HVS**) aware

# Forensic Quality: State of Art (Partial)



- Chen *et al.* “Fingerprint Quality Indices for Predicting Authentication Performance”, *Springer LNCS-3546*, p. 160 (2005).
- Tabassi *et al.* “A Novel Approach to Fingerprint Image Quality”, Proc. of ICIP 2005, p. 37 (2995).
- Fronthaler *et al.* “Automatic Image Quality Assessment with Application in Biometrics”, Proc. of *IEEE WB 2006*, p. 30 (2006).
- Vanderwee *et al.* “The Investigation of a Relative Contrast Index Model for Fingerprint Quantification” *FSI 204*, 74 (2011).

# Forensic Quality: State of Art Evaluation



- Mainly devoted to **fingerprint**, with no real mention to other forensic relevant imagery (faces, tool marks, shoe marks, tire marks)
- Interest in image quality effects on **AFIS performance**
- Interest in fingerprint quality after being acquired by dedicated, proper devices
- Few works care about the **expert's opinion**

# Used Methods



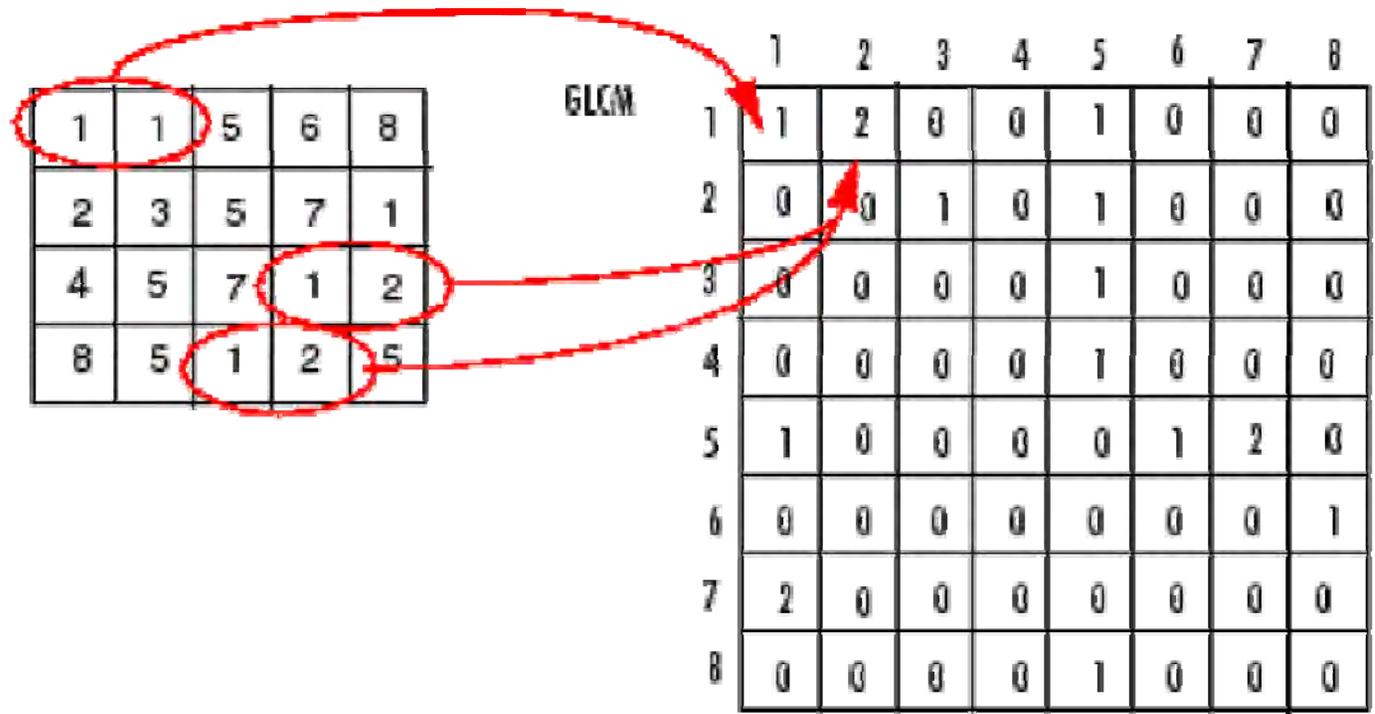
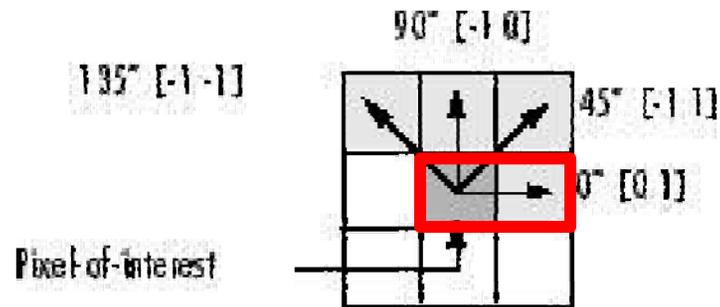
- We have used the following **two** methods:
  - **gray level co-occurrence matrix** (general purpose method)
  - number of **just noticeable difference** levels (HVS method)



## ■ Gray level co-occurrence matrix (GLCM):

- is a matrix created by calculating **how often** a pixel with grayscale intensity value  $i$  occurs horizontally (or vertically or diagonally) adjacent to a pixel with grayscale intensity value  $j$
- thus element  $(i,j)$  of GLCM specifies the **number of times** that the pixel with value  $i$  occurred horizontally (or vertically or diagonally) adjacent to a pixel with value  $j$

# GLC Matrix: Example



# GLCM: Contrast

- Many quantities can be defined by the GLCM:

- **contrast:**

$$\sum_{i,j} |i - j|^2 p(i, j)$$

assumes **0 value** for constant image  
need to be calculated **for every orientation**  
(horizontal, vertical and diagonal)

# GLCM: Properties



- Changes with **rotation**
- Changes with **scale**
- Doesn't know the **image structure**
  
- Need to:
  - renormalize images (so that they are the “same”)
  - be cautious in interpretation, as this is method is unaware of what a fingerprint is



# Number of Just Noticeable Different Levels



- The method quantifies the **perceptive contrast** experienced by the human eye
- Must be initialized with average physiological and viewing quantities:
  - screen size and resolution
  - distance of view
  - area of foveola (region of the retina where the focus of attention of the eye is situated)

# Number of Just Noticeable Different Levels



- Same luminance variation is **differently perceived** according to the average luminance
- For each value  $L$  of the luminance and its surrounding average  $S$  it is possible to calculate the luminance variation needed to **produce a perception of difference**
- This is called **just noticeable**



- In this work the **perceived contrast** between two luminance extremes  $L_{\min}$  and  $L_{\max}$  is assessed as the number of JNDs between them
- We look at the **JNDs distribution** to **try to deduce information** on the particular class of images that is analyzed

# JND: Properties

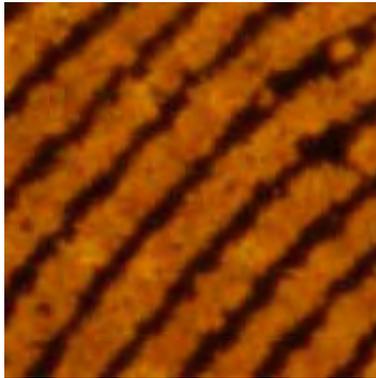


- ❑ Changes with viewing conditions
- ❑ Changes with processing
  
- ❑ Need to:
  - ❑ modify parameters to respect viewing conditions if comparison with others is needed

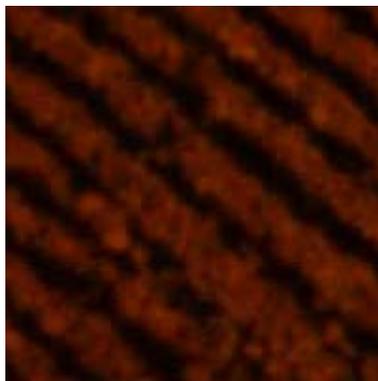
# JND: Examples



No processing



$N = 285$

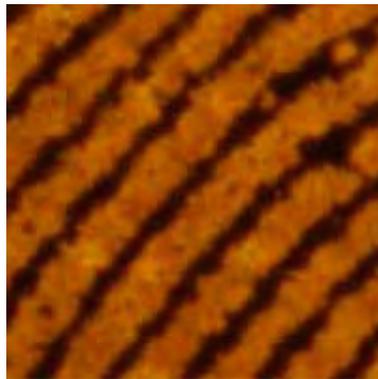


$N = 187$

# JND: Examples

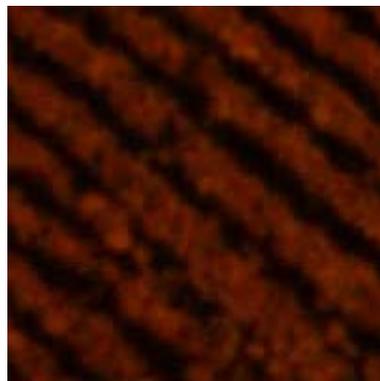


No processing



$N = 285$

Histogram equalization

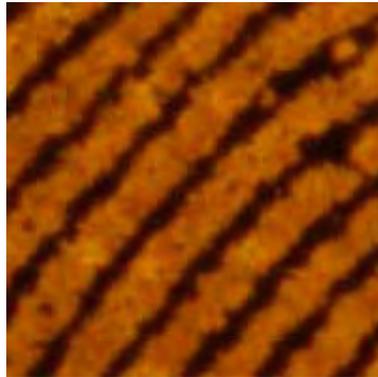


$N = 187$

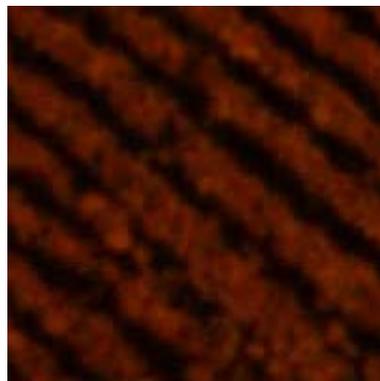
# JND: Examples



No processing

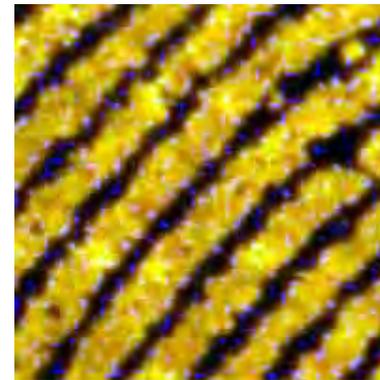


$N = 285$

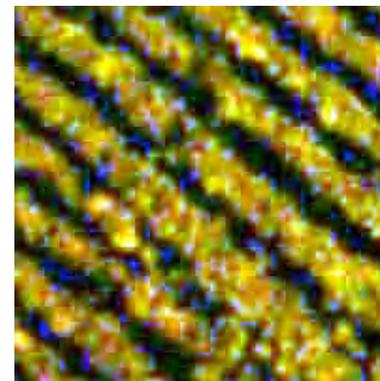


$N = 187$

Histogram equalization



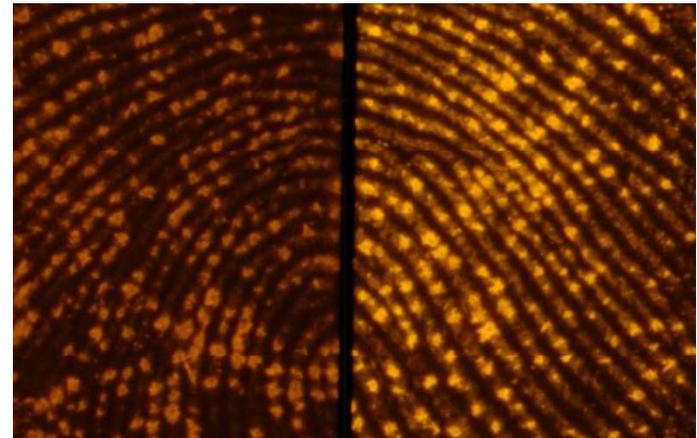
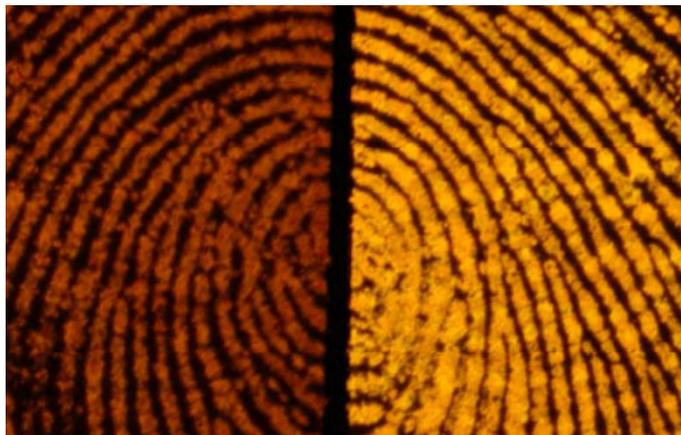
$N = 454$



$N = 444$

# Results

- GLCM method is able to rank only the quality of fingerprints with defined ridges (even if faint)



- HVS method is able to correctly rank all fingerprints and to detect automatically the dotted ones



# Fingerprint Quality: Comparison



- More than 400 fingerprints analyzed

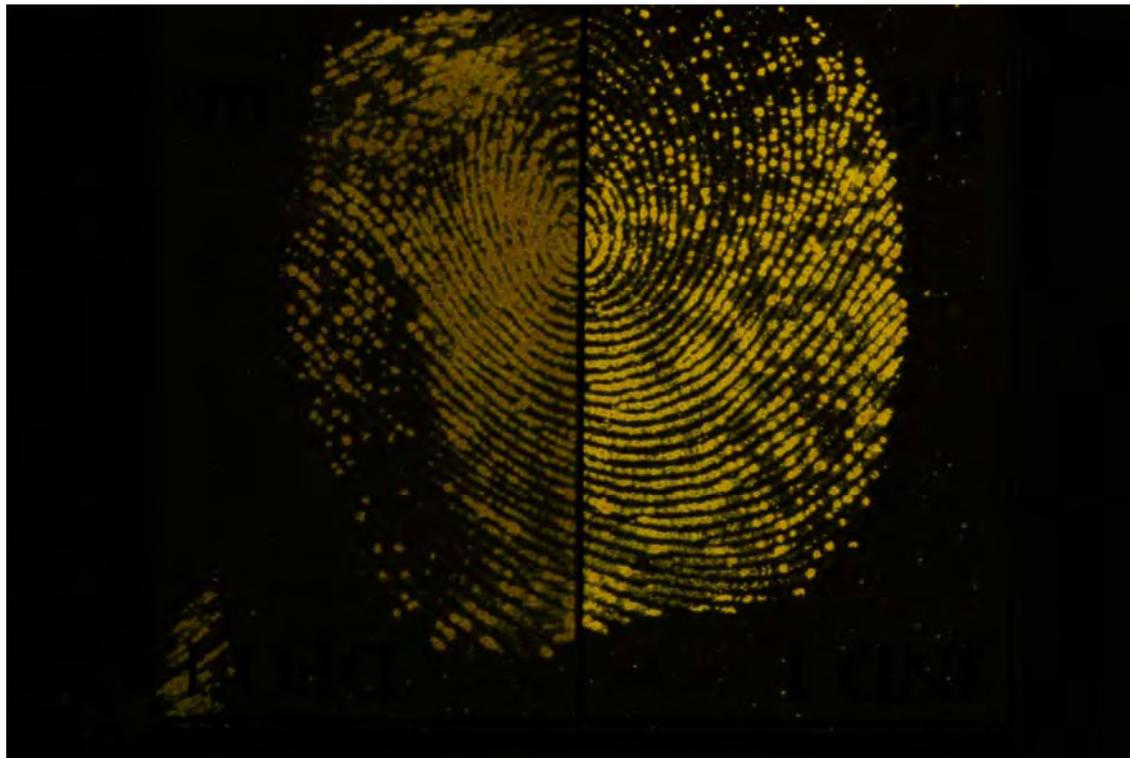
	A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R
1	impronta	SX MAX	SX fsm<10%	SX fsm 10-30	SX > 50%	DX max	DX fsm<10%	DX fsm 10-30	DX > 50%	Qualità SX	Metodo SX	Qualità DX	Metodo DX	Condizioni	x,y sinistra	x,y destra	Qualità Operatore	Note Operatore
120	B_prova4_5	205	0.03	0.48	0.13	205	0.01	0.4	0.09	0.0005065	IND	0.000581	DFO	150x150; 256	1358, 808	2133, 345	IND=DFO	nessuna
121	B_prova4_6	166	0.03	0.43	0.14	288	0.19	0.55	0.05	0.0003991	DFO	0.0001434	IND	150x150; 256	1103, 611	1792, 1027	DFO	nessuna
122	C_prova4_1	176	0.31	0.44	0.07	201	0.02	0.6	0.11	0.0002303	IND	0.0002997	DFO	300X300; 256	1204, 670	1622, 670	IND=DFO	DFO peggio contrastata ma estesa
123	C_prova4_2	210	0.04	0.62	0.11	251	0.32	0.34	0.15	0.0003377	DFO	0.0005184	IND	300X300; 256	1193, 670	1627, 622	IND=DFO	puntinate
124	C_prova4_3	122	0.14	0.71	0.01	180	0.01	0.8	0.02	0.0001481	IND	0.0001287	DFO	300X300; 256	1257, 665	1643, 638	IND=DFO	DFO peggio contrastata ma estesa
125	C_prova4_4	178	0.01	0.75	0.04	224	0.15	0.64	0.03	0.0001767	DFO	0.0002383	IND	300X300; 256	1225, 665	1643, 579	IND	nessuna
126	C_prova4_5	110	0.07	0.64	0.04	190	0.01	0.71	0.05	7.10E-05	IND	0.0002918	DFO	300X300; 256	1220, 446	1675, 409	DFO	IND si vede poco
127	C_prova4_6	187	0.02	0.76	0.04	163	0.37	0.5	0.01	0.0001806	DFO	0.0001141	IND	300X300; 256	1177, 500	1675, 563	IND	ai limiti utilità; DFO appi
128	D_prova4_1	208	0.3	0.39	0.1	219	0	0.4	0.27	0.0006196	IND	0.0005647	DFO	300X300; 256	1204, 191	1617, 830	DFO	puntinate
129	D_prova4_2	195	0.04	0.56	0.15	257	0.11	0.42	0.2	0.0004035	DFO	0.0006298	IND	300X300; 256	1257, 776	1654, 489	IND=DFO	puntinate
130	D_prova4_3	229	0.17	0.55	0.09	192	0.09	0.55	0.13	0.0001937	IND	0.0002079	DFO	300X300; 256	1188, 718	1585, 633	IND0,	nessuna
131	D_prova4_4	181	0.05	0.7	0.05	136	0.13	0.7	0.01	0.0001938	DFO	8.64E-05	IND	300X300; 256	1204, 1144	1574, 936	non utili	nessuna
132	D_prova4_5	187	0.15	0.35	0.24	222	0.02	0.41	0.0004465	IND	0.0007672	DFO	300X300; 256	1257, 547	1654, 766	DFO	puntinate	
133	D_prova4_6	183	0	0.56	0.08	176	0.21	0.59	0.03	0.0002658	DFO	0.0001696	IND	300X300; 256	1241, 1048	1611, 1000	non utili	nessuna
134	E_prova4_1	195	0.26	0.26	0.24	235	0.07	0.32	0.28	0.000635	IND	0.0008019	DFO	300X300; 256	1268, 505	1585, 414	IND=DFO	puntinate
135	E_prova4_2	235	0.13	0.47	0.11	207	0.16	0.47	0.13	0.0005055	DFO	0.0003919	IND	300X300; 256	1231, 473	1574, 398	DFO	puntinate
136	E_prova4_3	228	0.07	0.3	0.24	201	0	0.25	0.25	0.0007611	IND	0.0003321	DFO	300X300; 256	1247, 846	1585, 691	IND=DFO	nessuna
137	E_prova4_4	211	0	0.14	0.13	177	0.14	0.65	0.05	0.000348	DFO	0.0003109	IND	300X300; 256	1231, 537	1595, 516	DFO	leggera puntinatura
138	E_prova4_5	198	0.02	0.23	0.24	194	0	0.22	0.28	0.0005677	IND	0.000479	DFO	300X300; 256	1236, 723	1563, 723	IND=DFO	nessuna
139	E_prova4_6	236	0.03	0.35	0.12	242	0.14	0.58	0.02	0.0003798	DFO	0.0003672	IND	300X300; 256	1199, 633	1553, 649	IND=DFO	nessuna
140	F_prova4_1	215	0.37	0.25	0.16	257	0.11	0.45	0.2	0.0006102	IND	0.0006516	DFO	300X300; 256	1241, 798	1617, 723	DFO	puntinate
141	F_prova4_2	240	0.22	0.5	0.1	206	0.18	0.61	0.05	0.0004584	DFO	0.0002377	IND	300X300; 256	1220, 542	1595, 579	non utili	nessuna
142	F_prova4_3	205	0.31	0.48	0.03	176	0.02	0.66	0.07	0.0002636	IND	0.0002042	DFO	300X300; 256	1236, 595	1558, 590	IND	ai limiti utilità
143	F_prova4_4	228	0.04	0.55	0.16	220	0.3	0.39	0.1	0.0004004	DFO	0.0002848	IND	300X300; 256	1193, 686	1601, 617	DFO	ai limiti utilità
144	F_prova4_5	180	0.1	0.36	0.22	194	0.02	0.32	0.67	0.0005415	IND	0.0004872	DFO	300X300; 256	1241, 681	1590, 675	IND=DFO	leggera puntinatura
145	F_prova4_6	201	0.01	0.5	0.15	174	0.17	0.54	0.05	0.0002992	DFO	0.0002019	IND	300X300; 256	1172, 830	1585, 691	DFO	ai limiti utilità
146	B_prova5_1	242	0.01	0.14	0.57	316	0.02	0.21	0.44	0.0011548	IND	0.0012941	DFO	150x150; 256	1433, 180	2096, 303	DFO	nessuna
147	B_prova5_2	193	0.04	0.21	0.32	286	0.01	0.2	0.49	0.0004158	DFO	0.0014771	IND	150x150; 256	1114, 574	2133, 84	IND	nessuna
148	B_prova5_3	254	0	0.2	0.5	220	0	0.2	0.51	0.0010328	IND	0.0004554	DFO	150x150; 256	1390, 1394	1649, 1250	IND	nessuna
149	B_prova5_4	190	0	0.11	0.52	292	0	0.21	0.51	0.0003288	DFO	0.0009803	IND	150x150; 256	1369, 888	1728, 1197	IND	nessuna
150	B_prova5_5	314	0.01	0.2	0.5	215	0	0.15	0.47	0.0016865	IND	0.0005484	DFO	150x150; 256	1252, 936	1595, 1000	IND	zone di puntinatura
151	B_prova5_6	222	0.01	0.18	0.5	322	0.03	0.17	0.5	0.0005884	DFO	0.0019609	IND	150x150; 256	1343, 808	1617, 963	IND	zone di puntinatura

# Fingerprint Quality: Results



- Tested all fingerprints with two different quality assessment algorithms
- Comparison to fingerprint expert to see difference with algorithms and to tune them
- If done properly useful to assess forensic utility of fingerprint before showing them to the expert

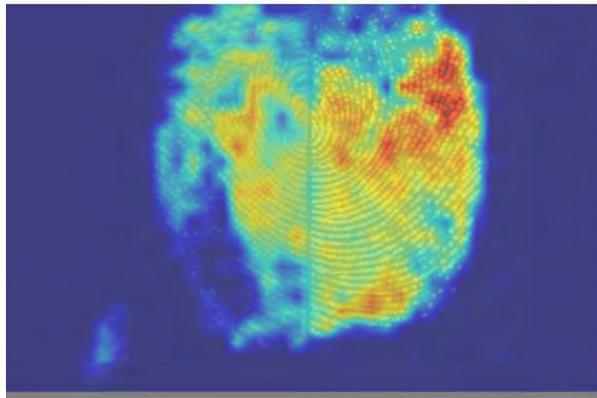
# Fingerprint Quality Maps



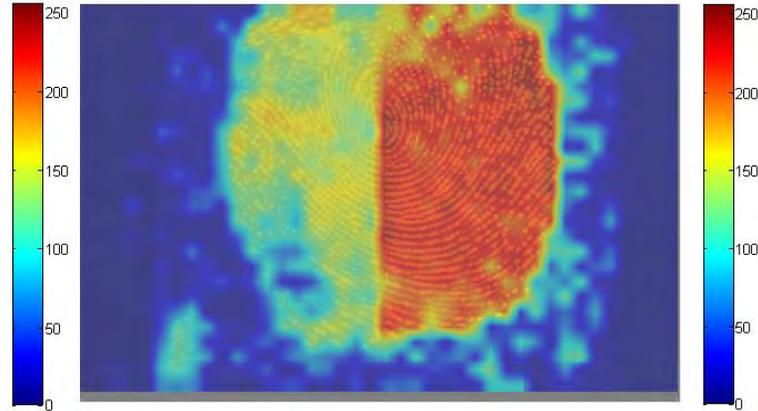
# Fingerprint Quality Maps



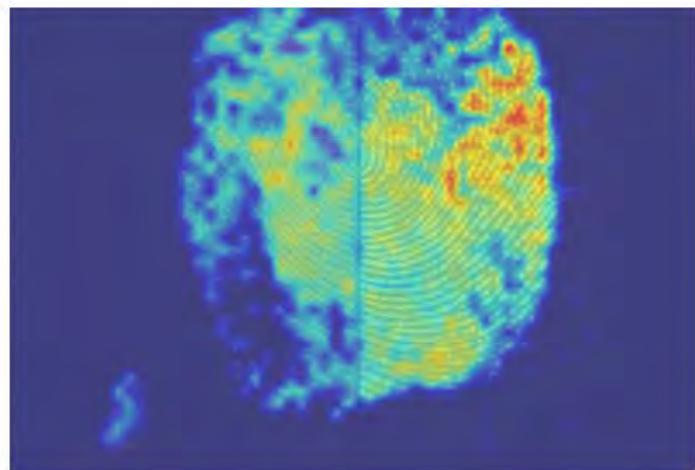
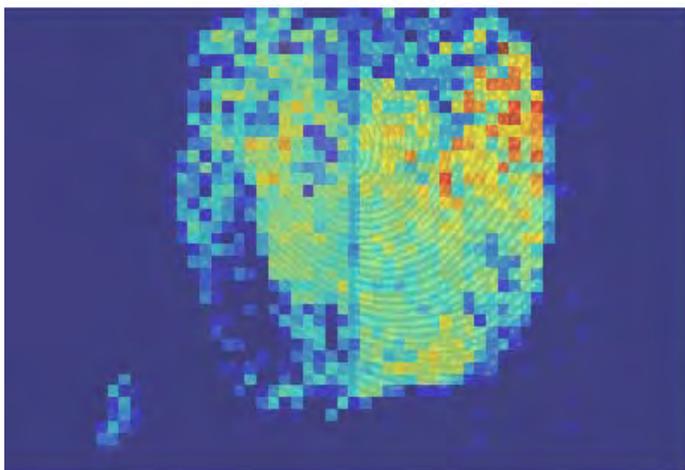
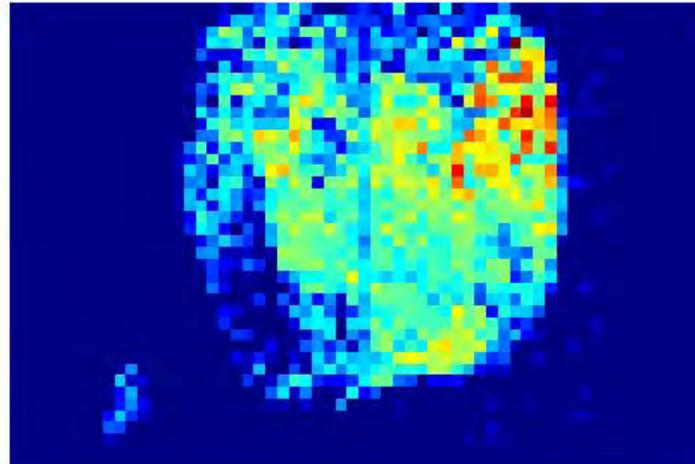
GLCM Map



HVS Map



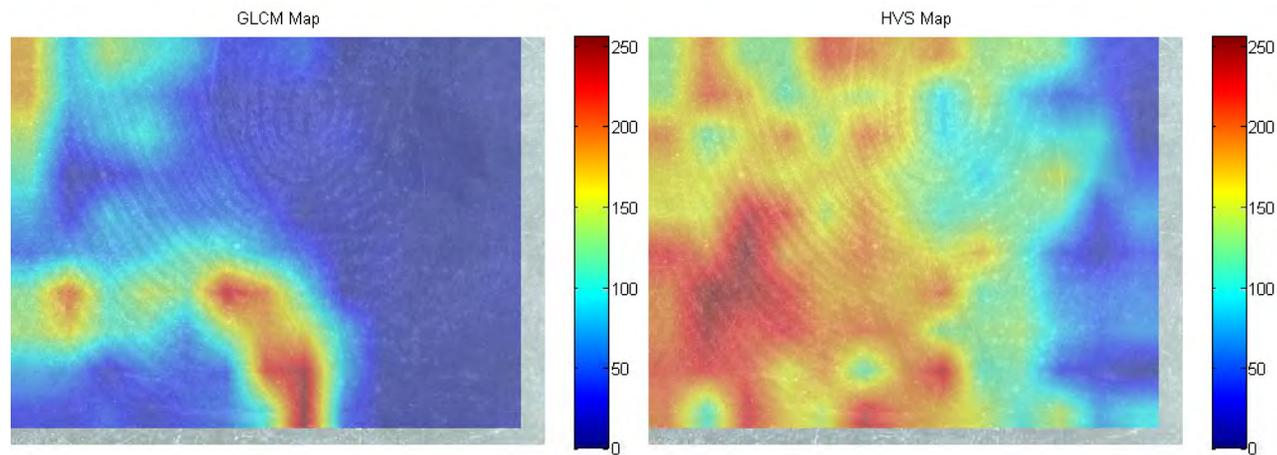
# Fingerprint Quality Maps



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# Fingerprint Quality Maps

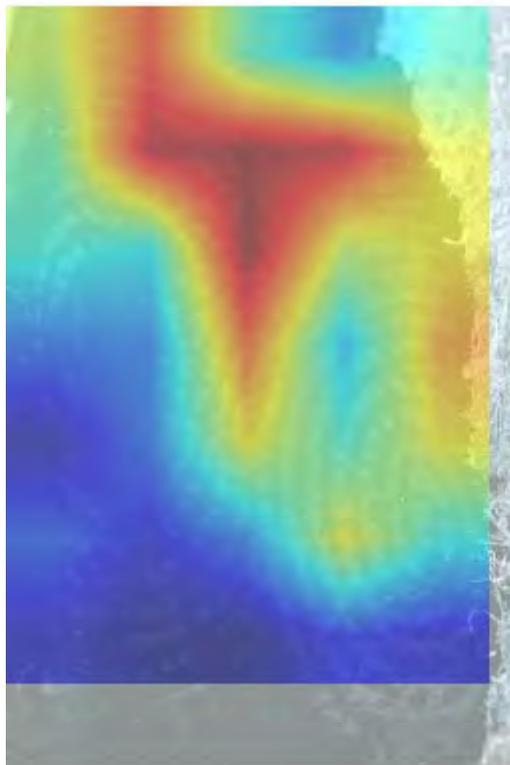




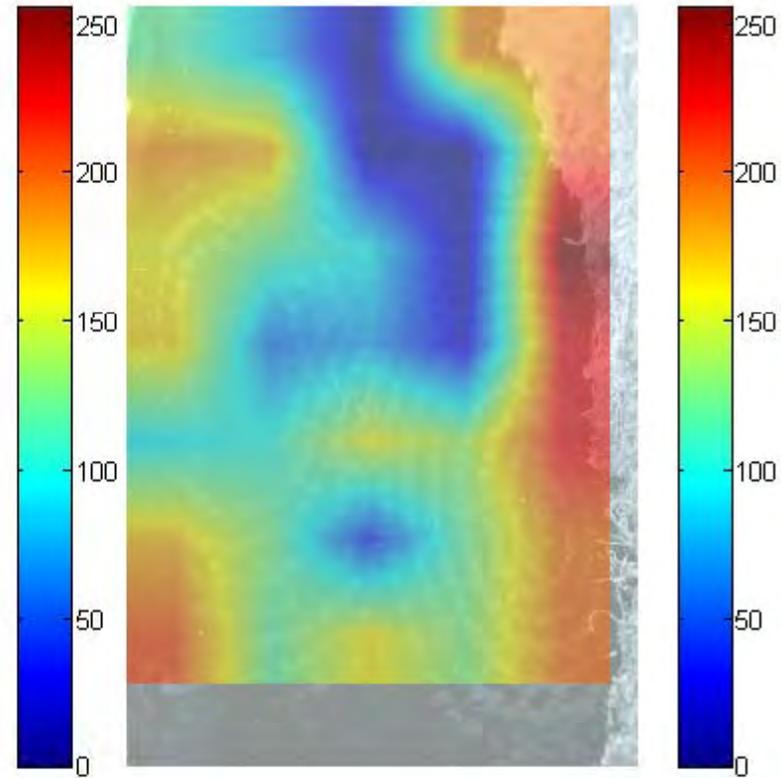
# Fingerprint Quality Maps



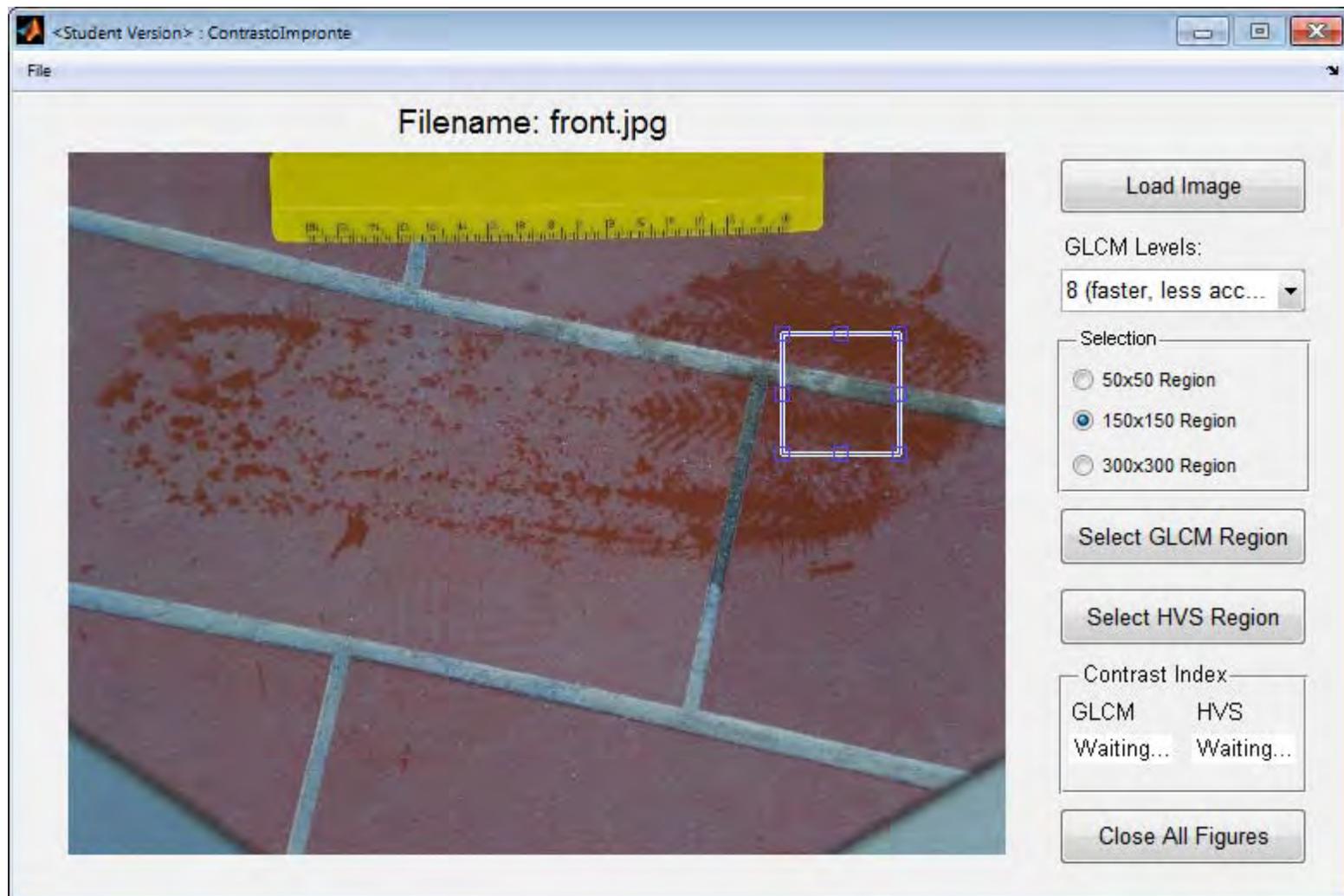
GLCM Map



HVS Map



# Other application: Shoemarks





- "No-reference measurement of perceptually significant blurriness in video frames", *Signal Image and Video Processing* **5**, 271-282 (2011)
- "A set of features for measuring blurriness in video frames", Melecon 2010, IEEE Mediterranean Electro-technical Conference, Valletta, Malta, 26-28 April 2010.
- "Blurriness estimation in video frames: a study on smooth objects and textures", in Proceeding of the SPIE Electronic Imaging Conference, San Jose (CA) USA, (2010).
- "Causes and visual experience of blurriness in video frames", submitted

# Conclusions



- The forensic quality (i.e. usefulness) of images **can be assessed by using some contrast definition** for images
- Generic purpose systems need to be **used with caution** if they do not allow teaching them the kind of object under analysis
- HVS systems can be used to **assess quality and degradation causes** of images
- This could **support** the expert's analysis

# Future Works



- Complete analysis of HVS distribution to **teach** the software extended features and what are the most common cause of quality degradation
- Try quality index tool to **other forensic fields** (shoes, faces, tool marks, tire marks, etc.)
- Notice that the system will be tuned using expert's opinions

# Future works: full system



# Contacts



[aldo.mattei@gmail.com](mailto:aldo.mattei@gmail.com)

[fcube@innovationfactory.it](mailto:fcube@innovationfactory.it)