

Fingerprint Recognition II

COMP 388-002/488-002 Biometrics

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Fall 2025



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Today we will...

Get to know

Fingerprint acquisition and enhancement.

Today's Attendance

Please fill out the form

forms.gle/XHPxsvDNKa7xGzih8

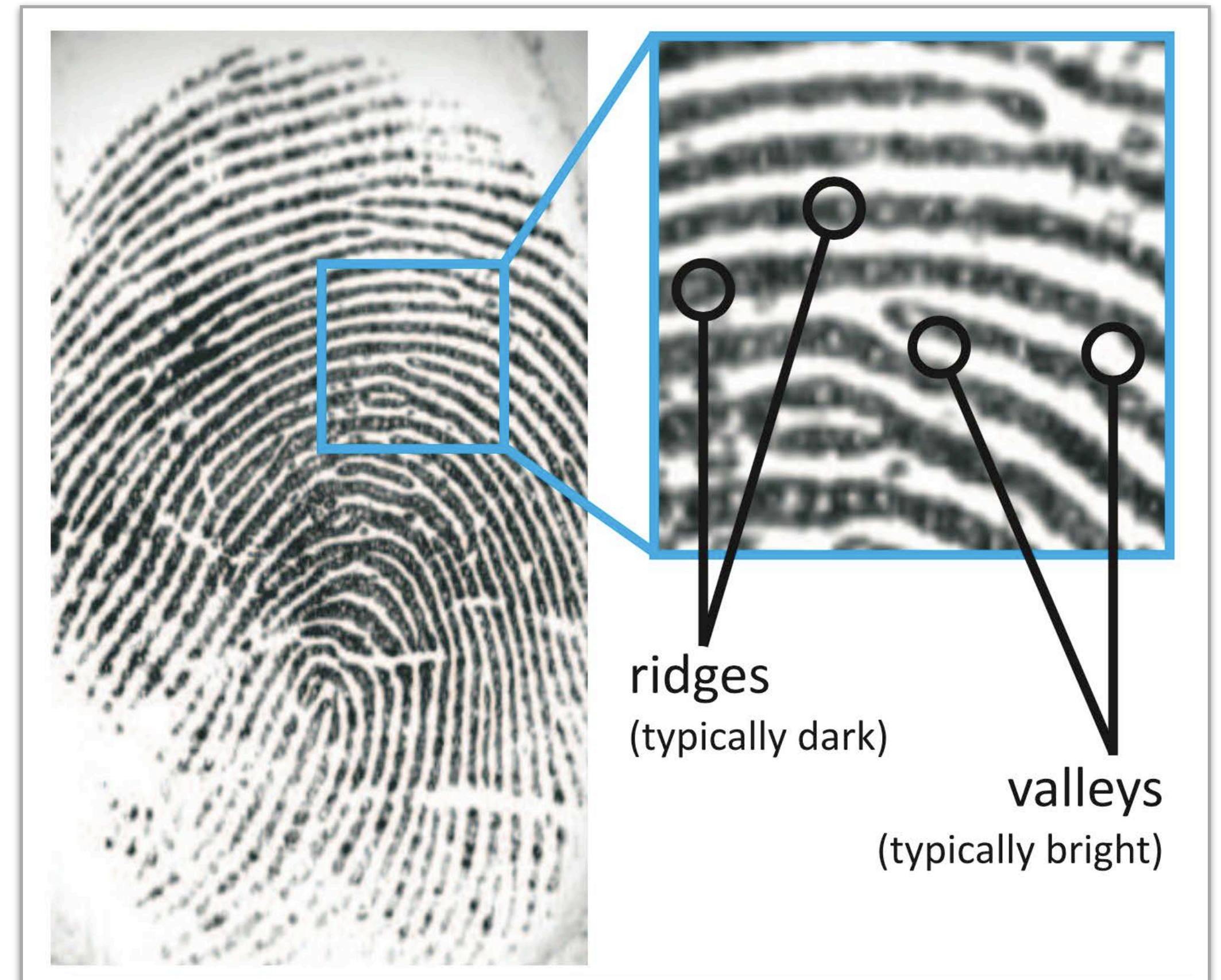


Features

What do we observe in fingerprints?

Ridges and Valleys

Embryology hypothesis:
Ridges appear as a result of the stresses in the womb during the growth of the fetus.



Source: Dr. Adam Czajka



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Features

**What do we observe
in fingerprints?**

Beyond Ridges and Valleys

Three types of features,
from coarse to fine levels:

- Level-1 Features
- Level-2 Features
- Level-3 Features



Features

What do we observe
in fingerprints?

Beyond Ridges and Valleys

Three types of features,
from coarse to fine levels:

- **Level-1 Features**
- Level-2 Features
- Level-3 Features



Features

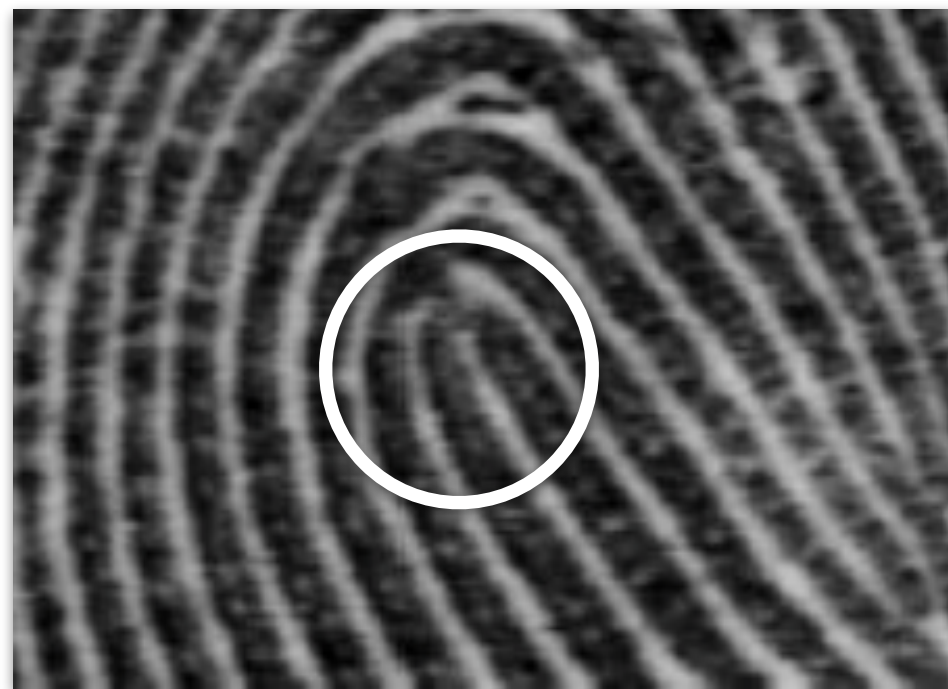
RECAP

Level-1 Features

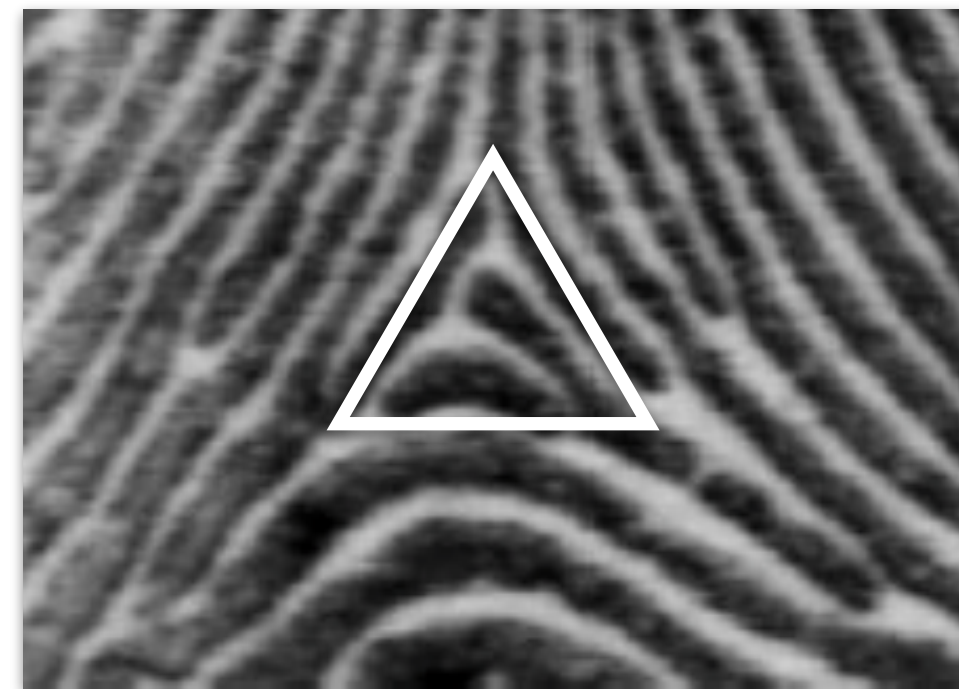
Observe singular points and core.

Useful capture resolution: 250 ppi (pixels per inch)

Singular Points



loop



delta

Jain, Ross, and Nadakumar
Introduction to Biometrics
Springer Books, 2011

Core

Up-most singular point

or (in case of no singular point)

Point of maximum ridge curvature.



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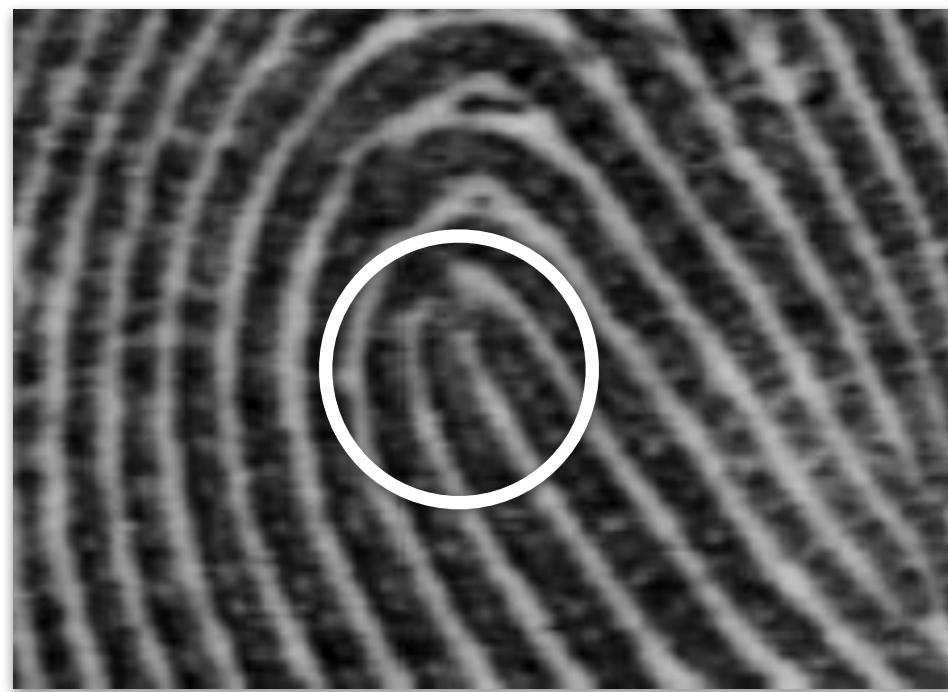
Features

RECAP

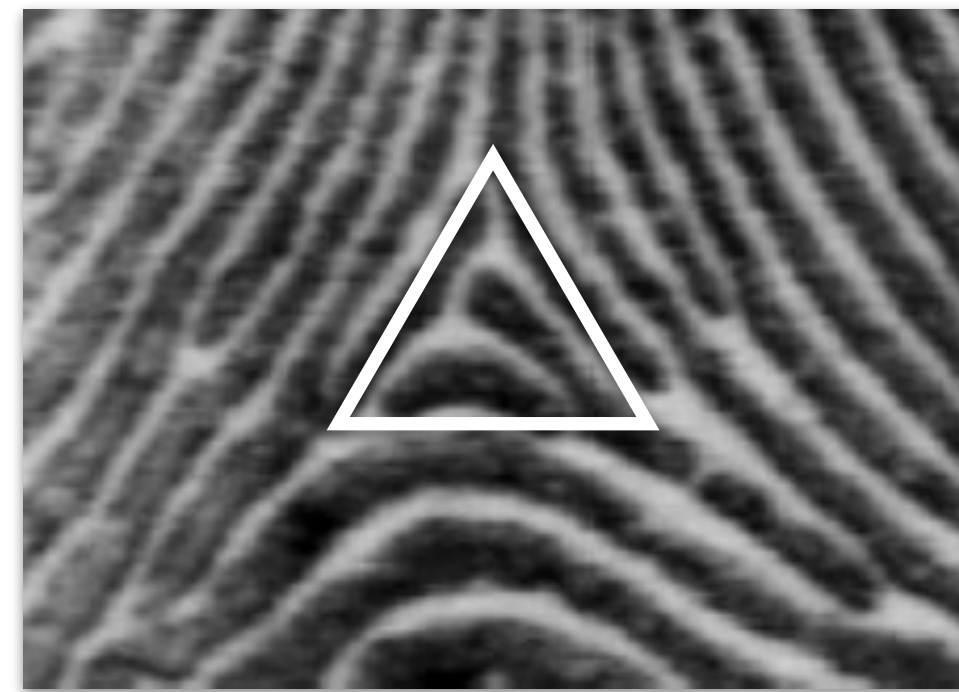
Level-1 Features

Observe singular points and core.

Usage of Singular Points and Core



loop



delta

Jain, Ross, and Nadakumar
Introduction to Biometrics
Springer Books, 2011

Alignment of two samples.
Fingerprint classification.

Features

RECAP

Fingerprint Classification

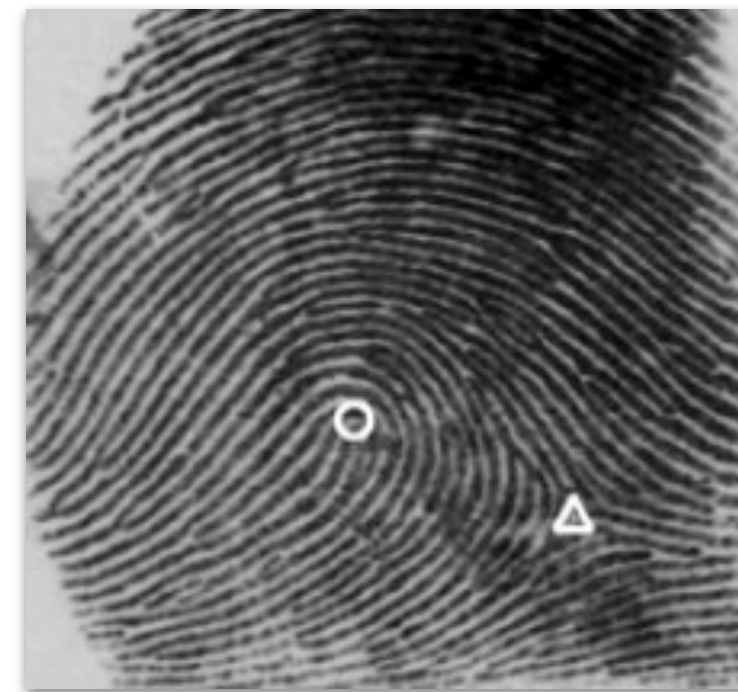
Jain, Ross, and Nadakumar
Introduction to Biometrics
Springer Books, 2011



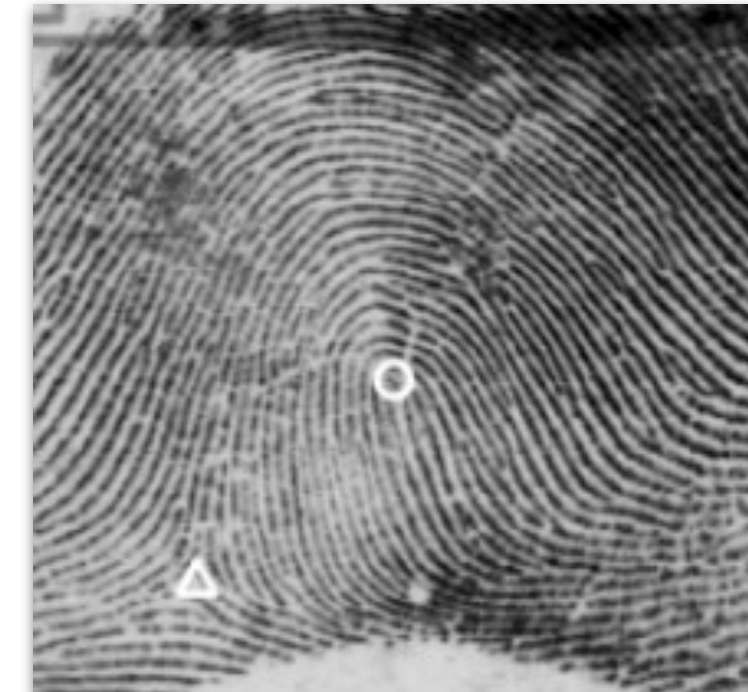
plain arch
4%



tented arch
3%

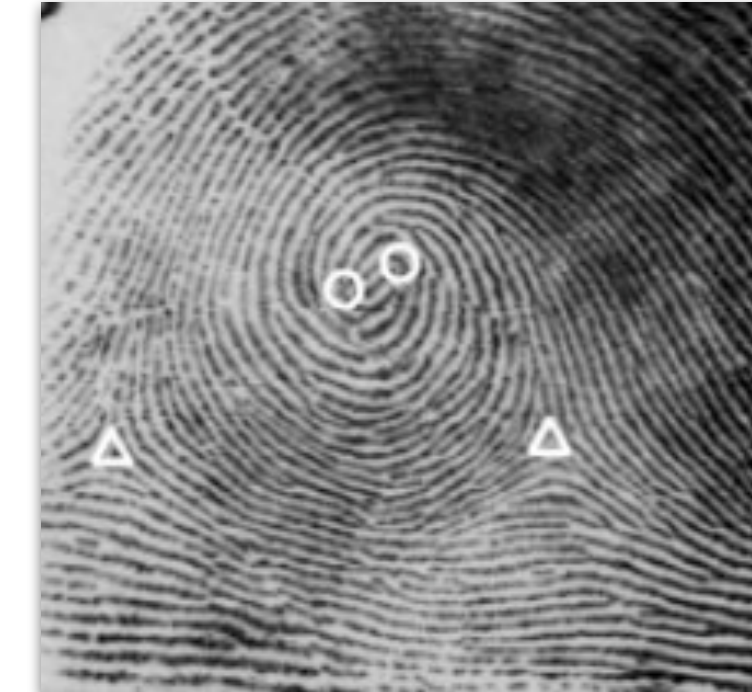


left loop

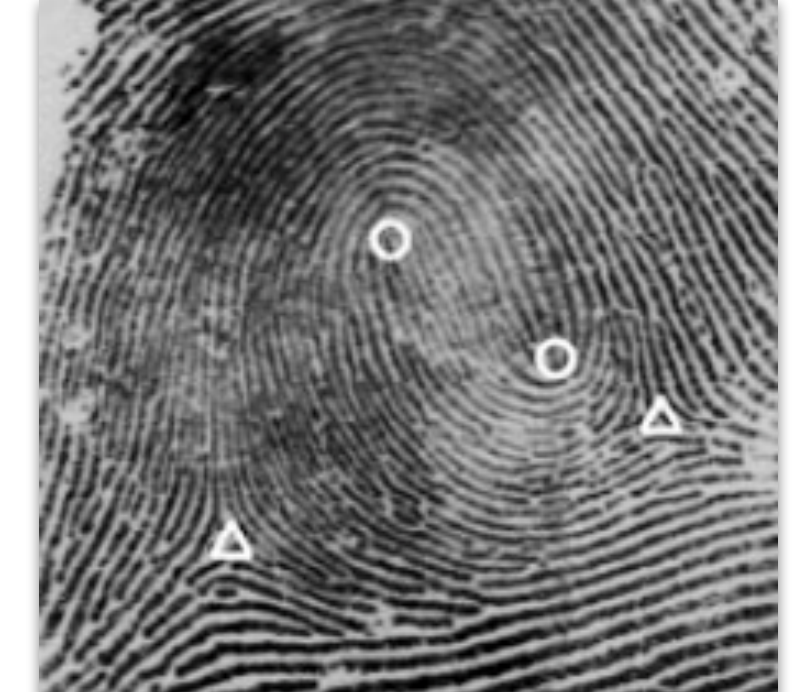


right loop

65%



whorl
24%



twin loop
4%

Percentages: frequencies of observation.

Features

RECAP

How useful are level-1 features?

FBI Automated Fingerprint Identification system (AFIS)

More than 200 million dactyloscopy cards.
Varied quality of samples.

Estimated: one untrained person
would spend **67 years** to search
1.7 million cards.

P.D. 154, 10-24

NAME *Hickhouse, William (Scas Puce)* F. P. FORMULA *10:5* MALE

F. P. No. *341* BER. No. *1000000* REF. *1000000*

PRISON *Filly* No. *1000000* *Alvin W. Brown*

1. RIGHT THUMB 2. R. FORE FINGER 3. R. MIDDLE FINGER 4. R. RING FINGER 5. R. LITTLE FINGER

6. LEFT THUMB 7. L. FORE FINGER 8. L. MIDDLE FINGER 9. L. RING FINGER 10. L. LITTLE FINGER

LEFT HAND PLAIN IMPRESSIONS OF THE FOUR FINGERS TAKEN SIMULTANEOUSLY

RIGHT HAND PLAIN IMPRESSIONS OF THE FOUR FINGERS TAKEN SIMULTANEOUSLY

IDENTIFICATION SECTION
Department of Police
CHICAGO

Impressions taken by *1*

Classified at *1*

Tested at *1*

Date *1 JAN 1974*

Date *19*

Date *19*

Date *19*

Date *19*

Features

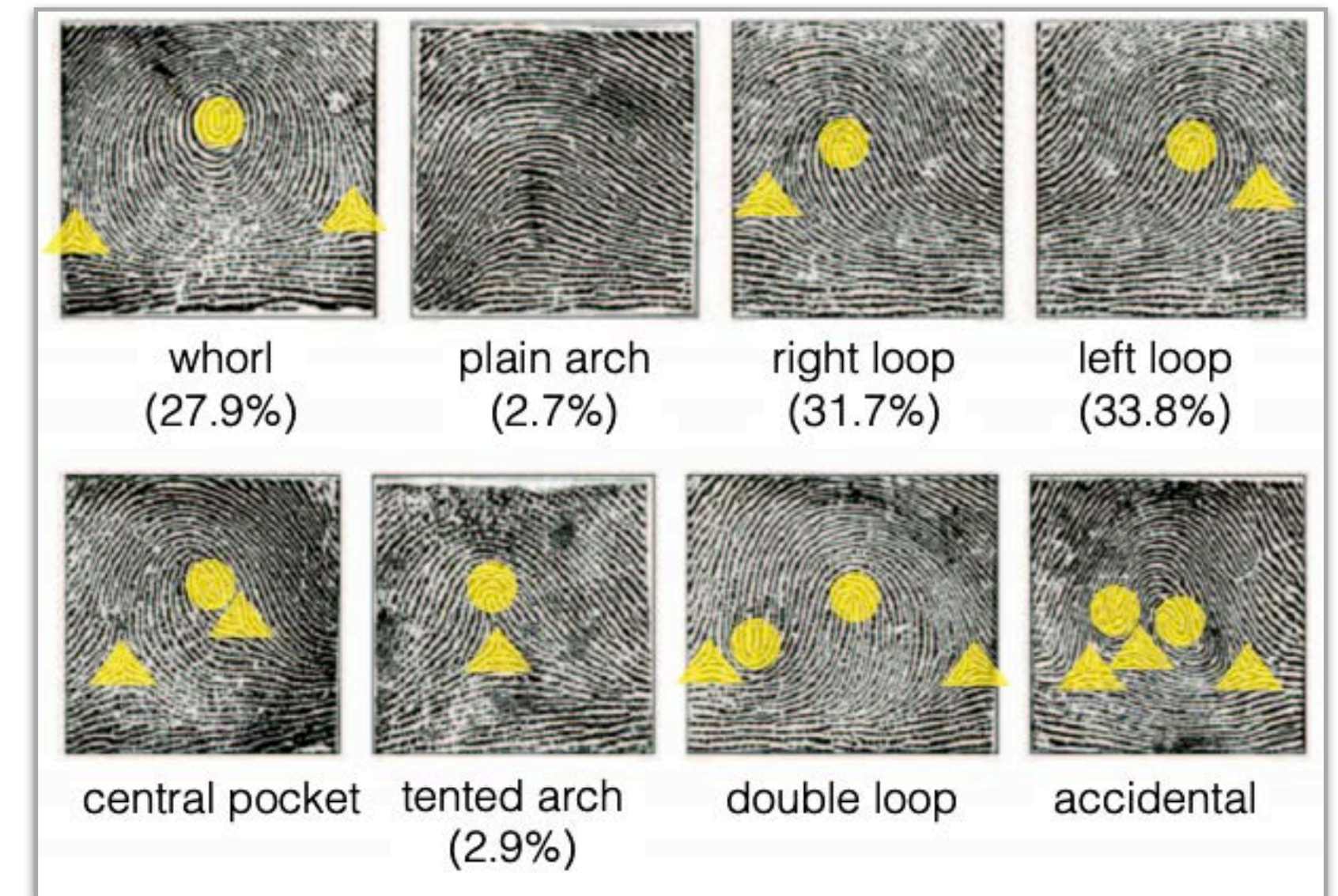
RECAP

How useful are level-1 features?

FBI Automated Fingerprint Identification system (AFIS)

More than 200 million dactyloscopy cards.
Varied quality of samples.

Thanks to fingerprint classification through level-1 features, this time is reduced to **20 min.**



Henry's features, an alternative classification of level-1 features with 8 classes.

Features

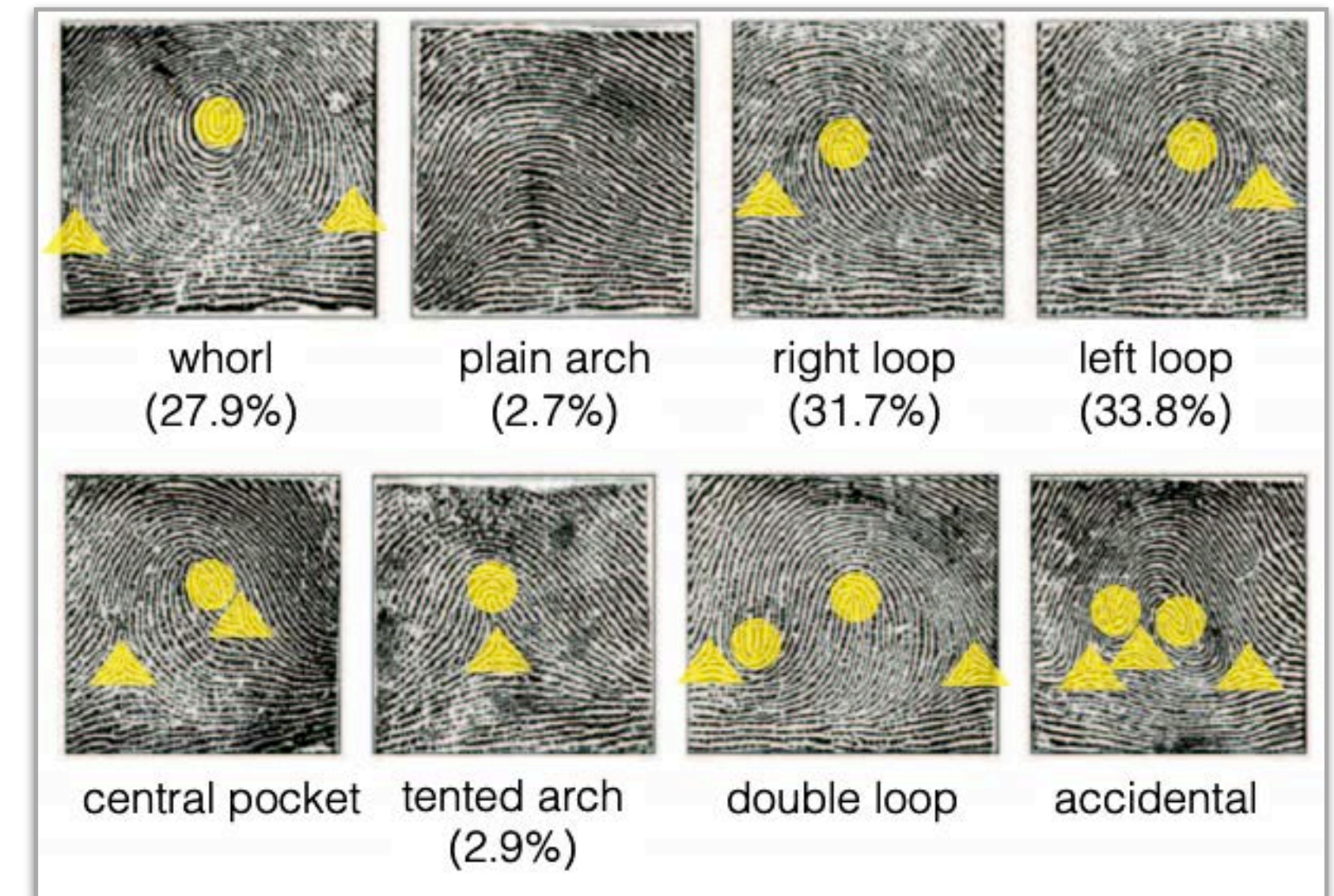
RECAP

How useful are level-1 features?

FBI Automated Fingerprint Identification system (AFIS)

More than 200 million dactyloscopy cards.
Varied quality of samples.

And a computer-based solution
can do it in seconds, benefitting from
the same features.



Henry's features, an alternative
classification of level-1 features
with 8 classes.

Features

What do we observe
in fingerprints?

Beyond Ridges and Valleys

Three types of features,
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- Level-1 Features
- **Level-2 Features**
- Level-3 Features



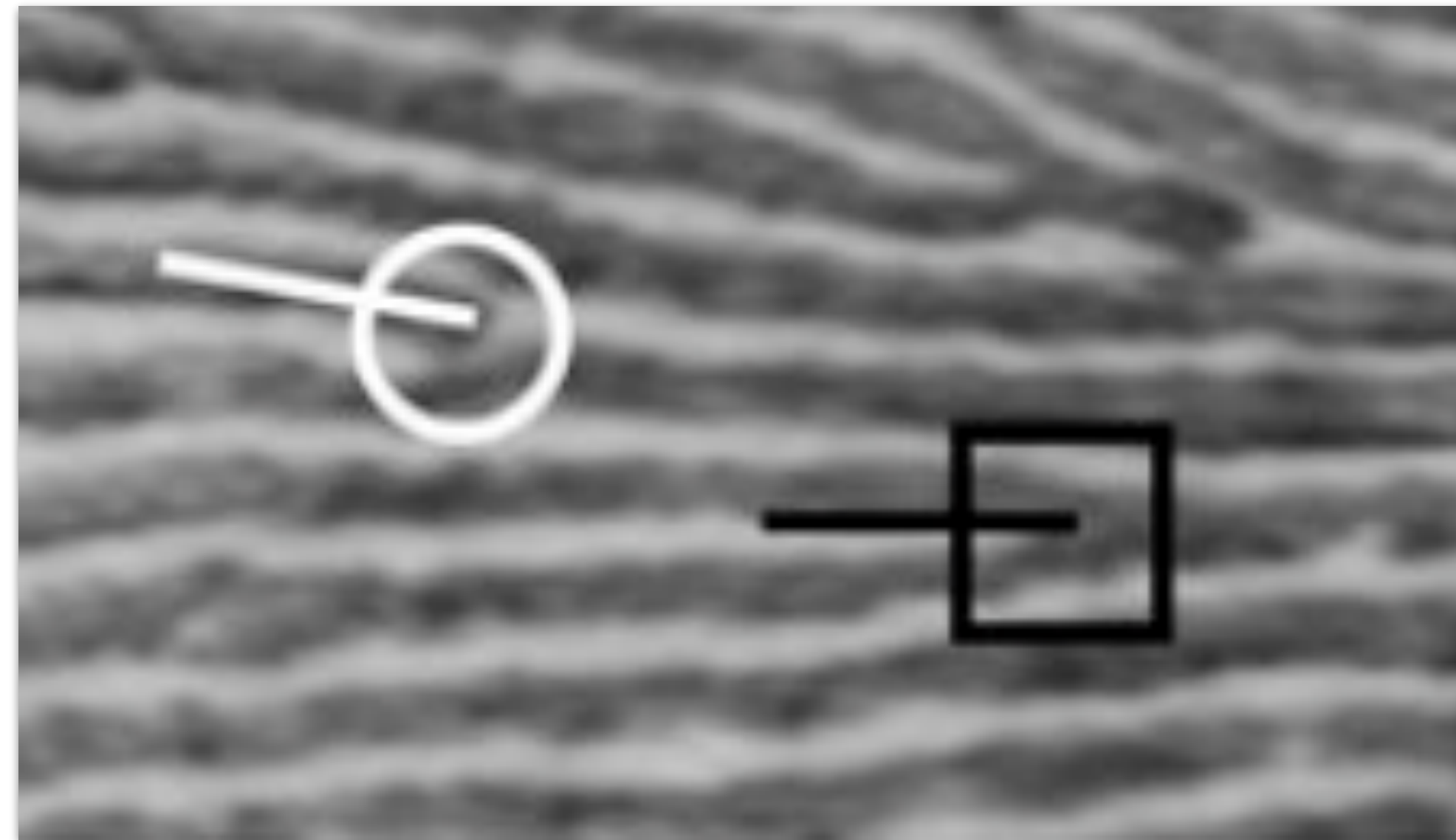
Features

Level-2 Features

Observe minutiae (Galton's details).

Useful capture resolution: 500 ppi

Ridge Ending



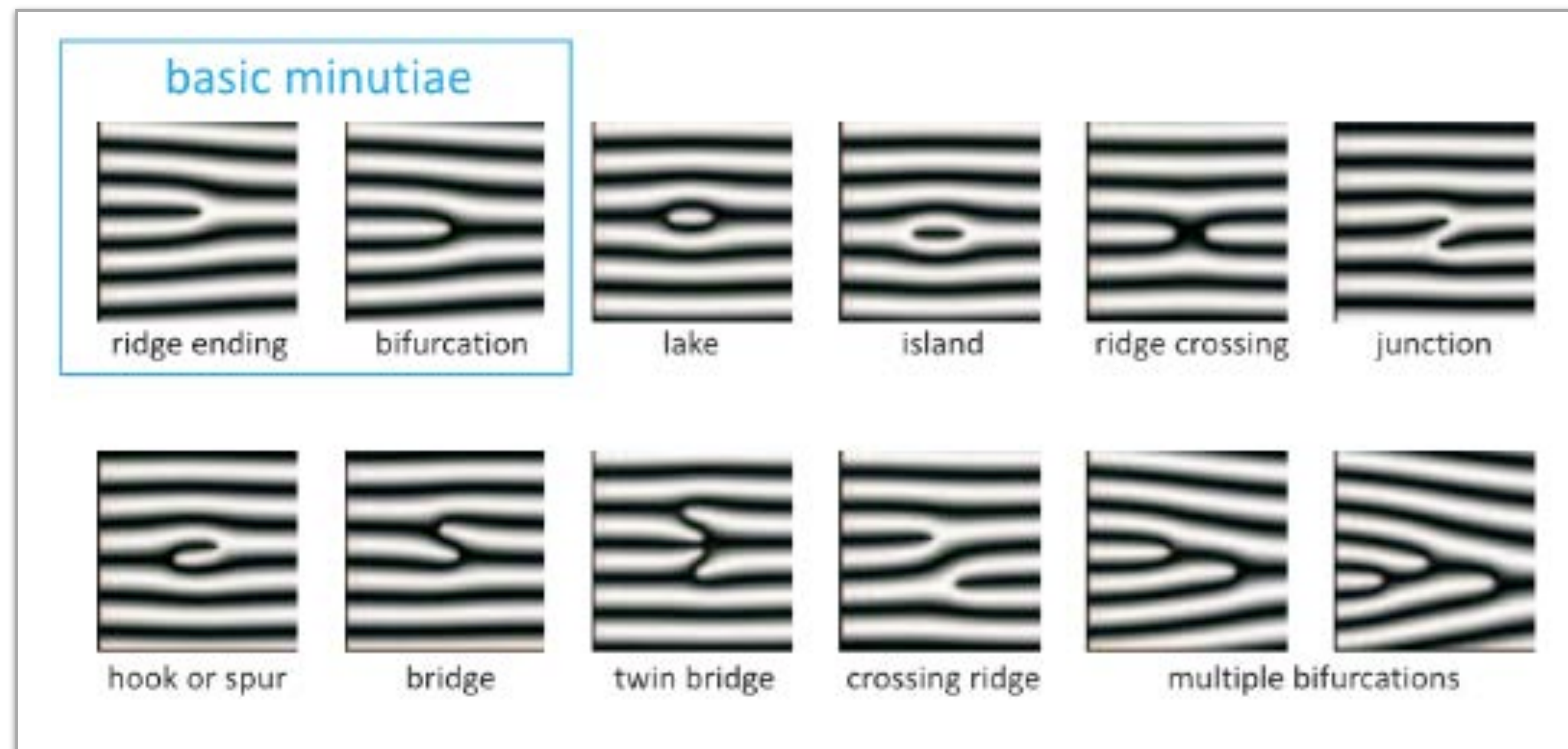
Ridge Bifurcation

Jain, Ross, and Nadakumar
Introduction to Biometrics
Springer Books, 2011

Features

Level-2 Features

Alternative minutiae.



Source:
www.optel.com.pl

Features

Level-2 Features

Usage of minutiae
Fingerprint matching.

More details on **how** to do it
in the upcoming classes.

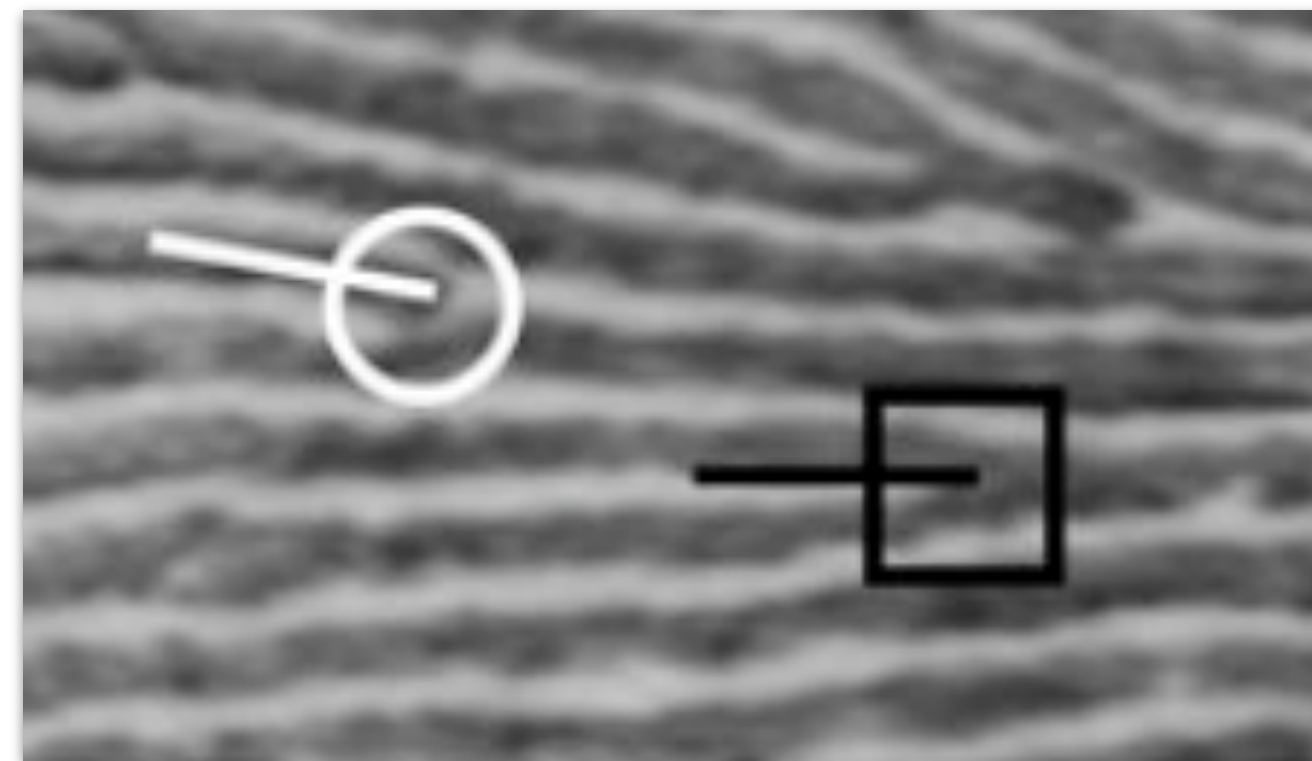


Features

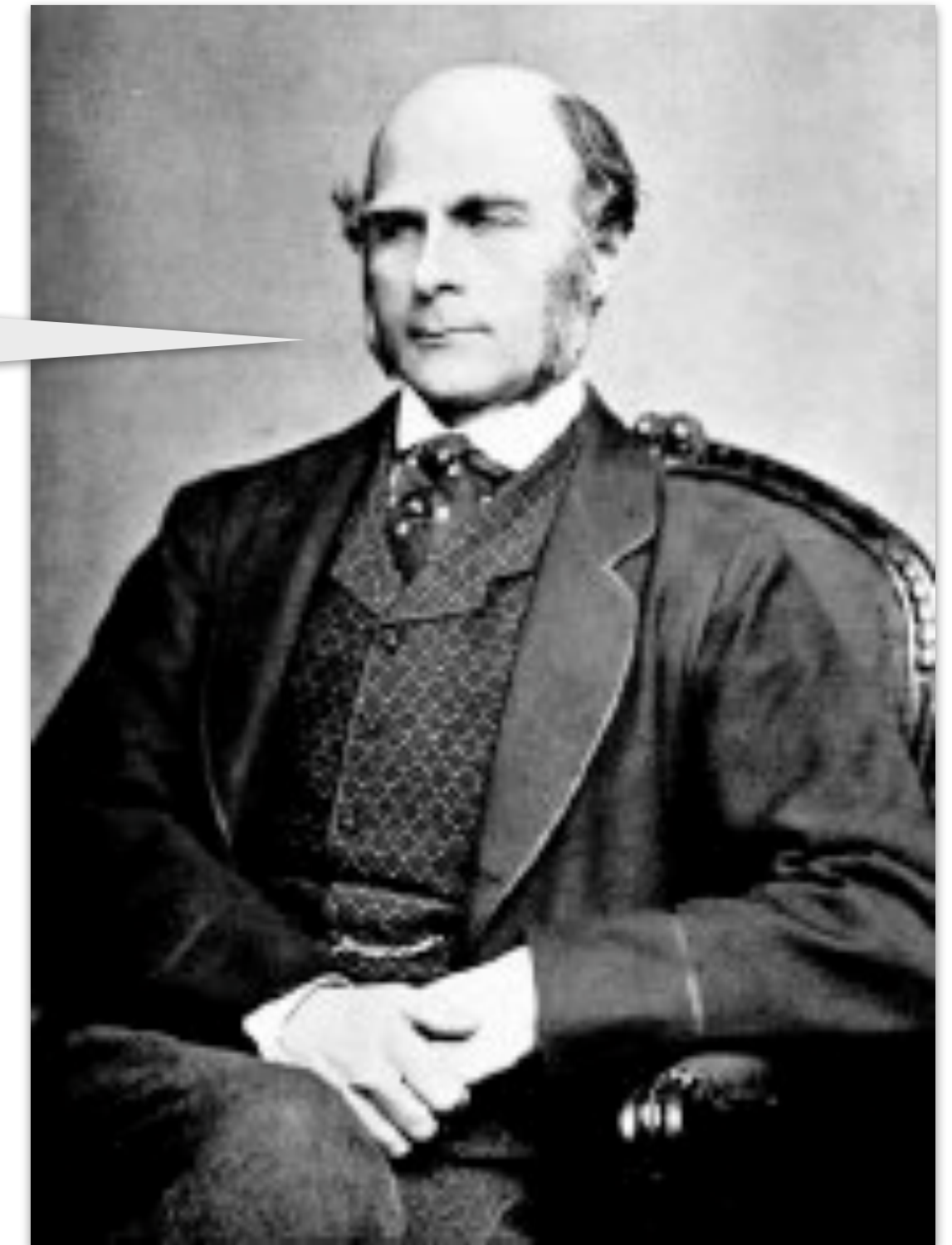
Level-2 Features

Galton's Estimate

Given 2 similar fingerprints,
what is the chance they come
from different people?
I'll tell you: 1 in 64 billion.



Jain, Ross, and Nadakumar
Introduction to Biometrics
Springer Books, 2011



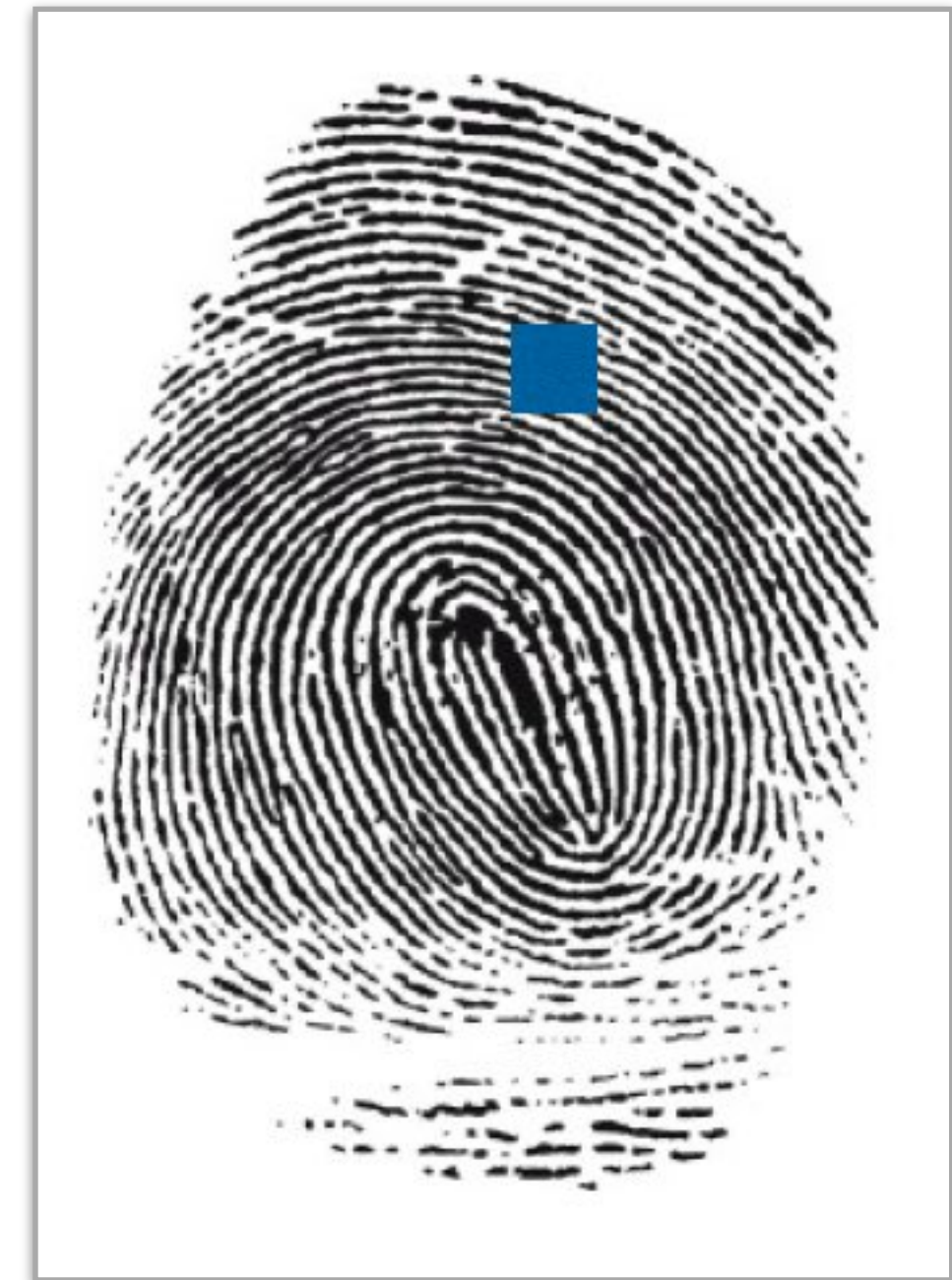
Features

Level-2 Features

Galton's Estimate

Rationale

What would be the smallest portion of a fingerprint leading to a 1/2 chance of being correctly guessed as belonging to a particular individual?



Source:
Dr. Walter Scheirer

Features

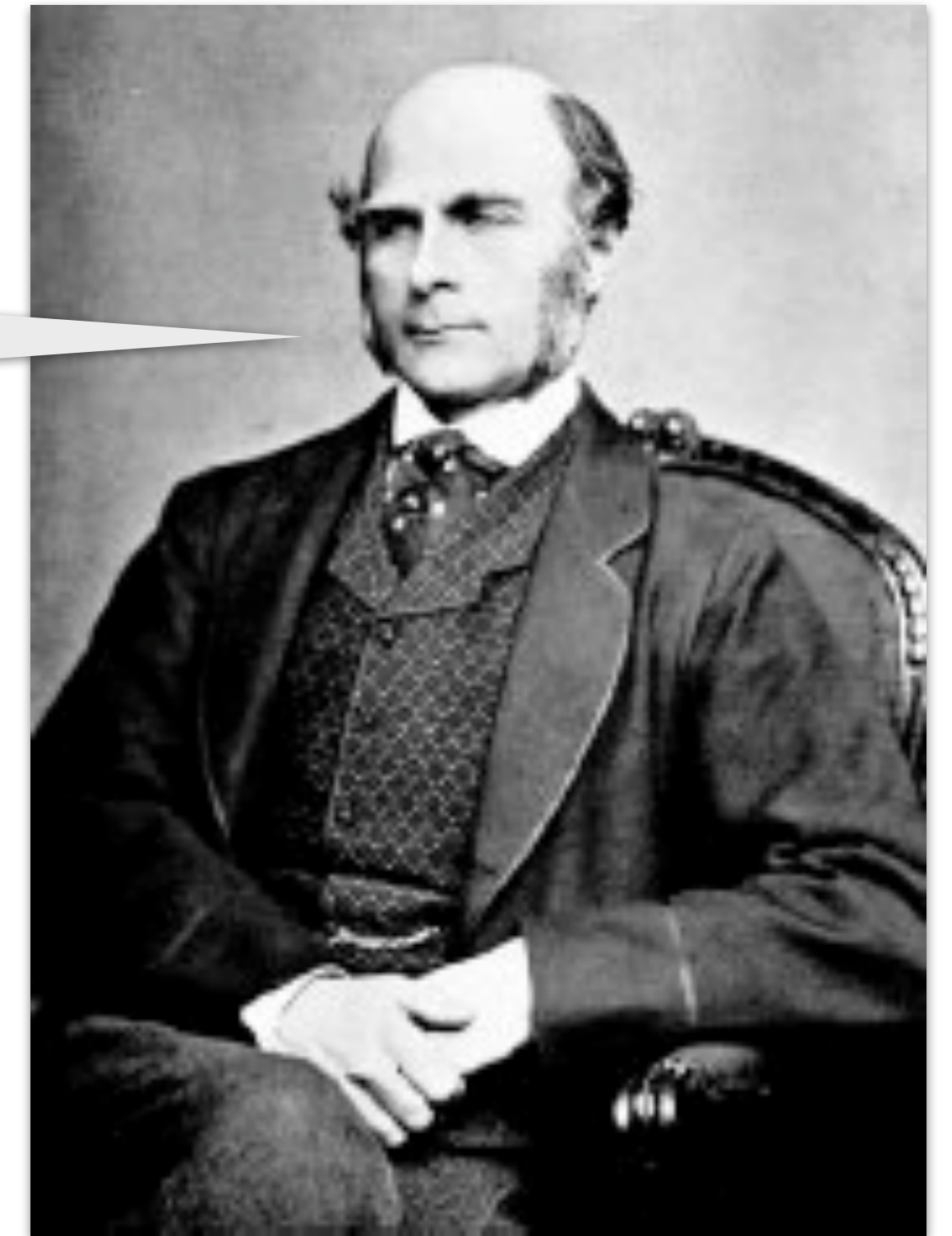
Level-2 Features

Galton's Estimate

After a few trials, let me say:
A square containing 5-6 ridges.

Rationale

What would be the smallest portion of a fingerprint leading to a 1/2 chance of being correctly guessed as belonging to a particular individual?



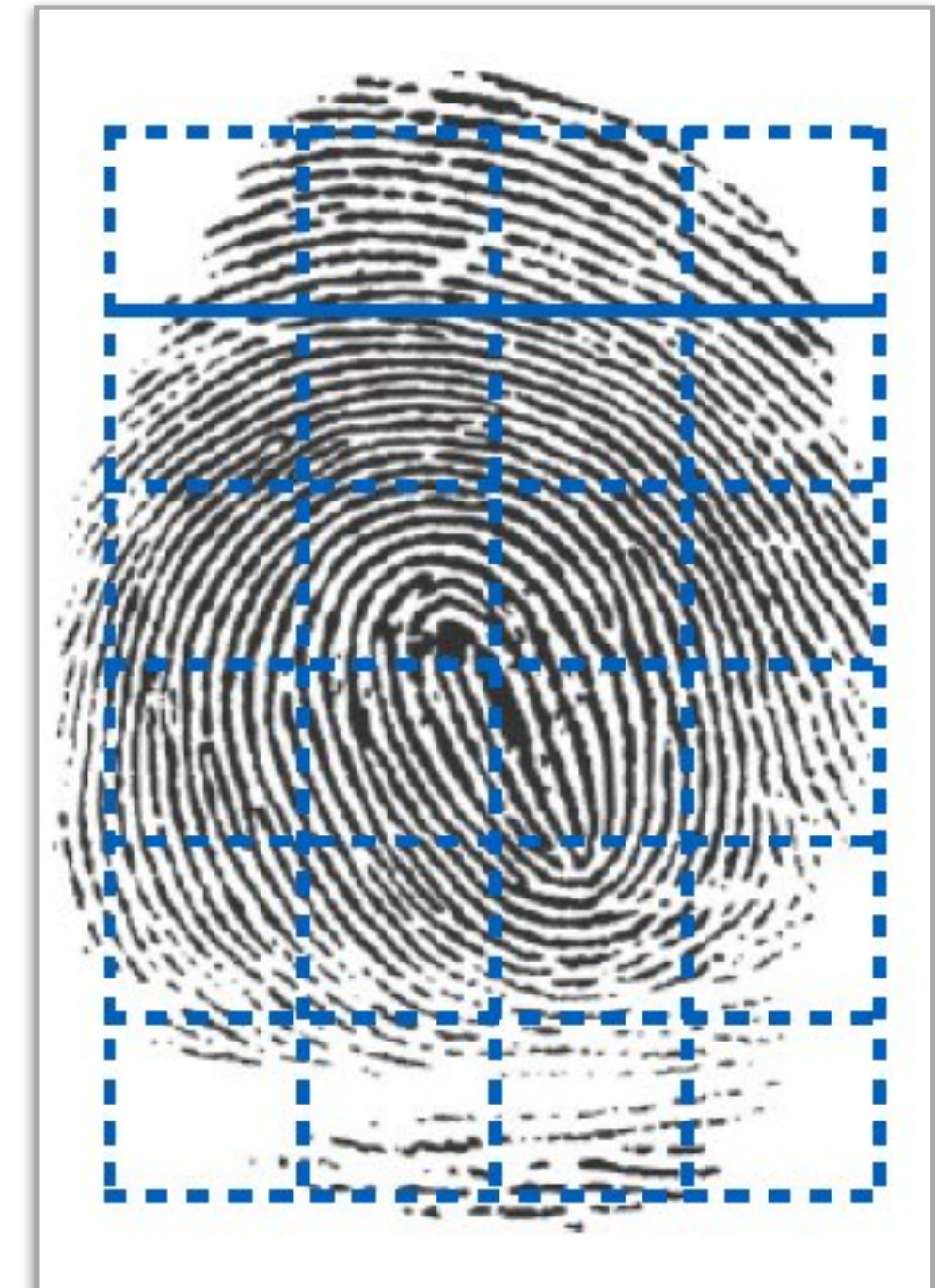
Features

Level-2 Features

Galton's Estimate

A typical fingerprint consists of 24 six-ridge squares.

Hence, the chance of correct full fingerprint guess: $1/2^{24}$



Source:
Dr. Walter Scheirer

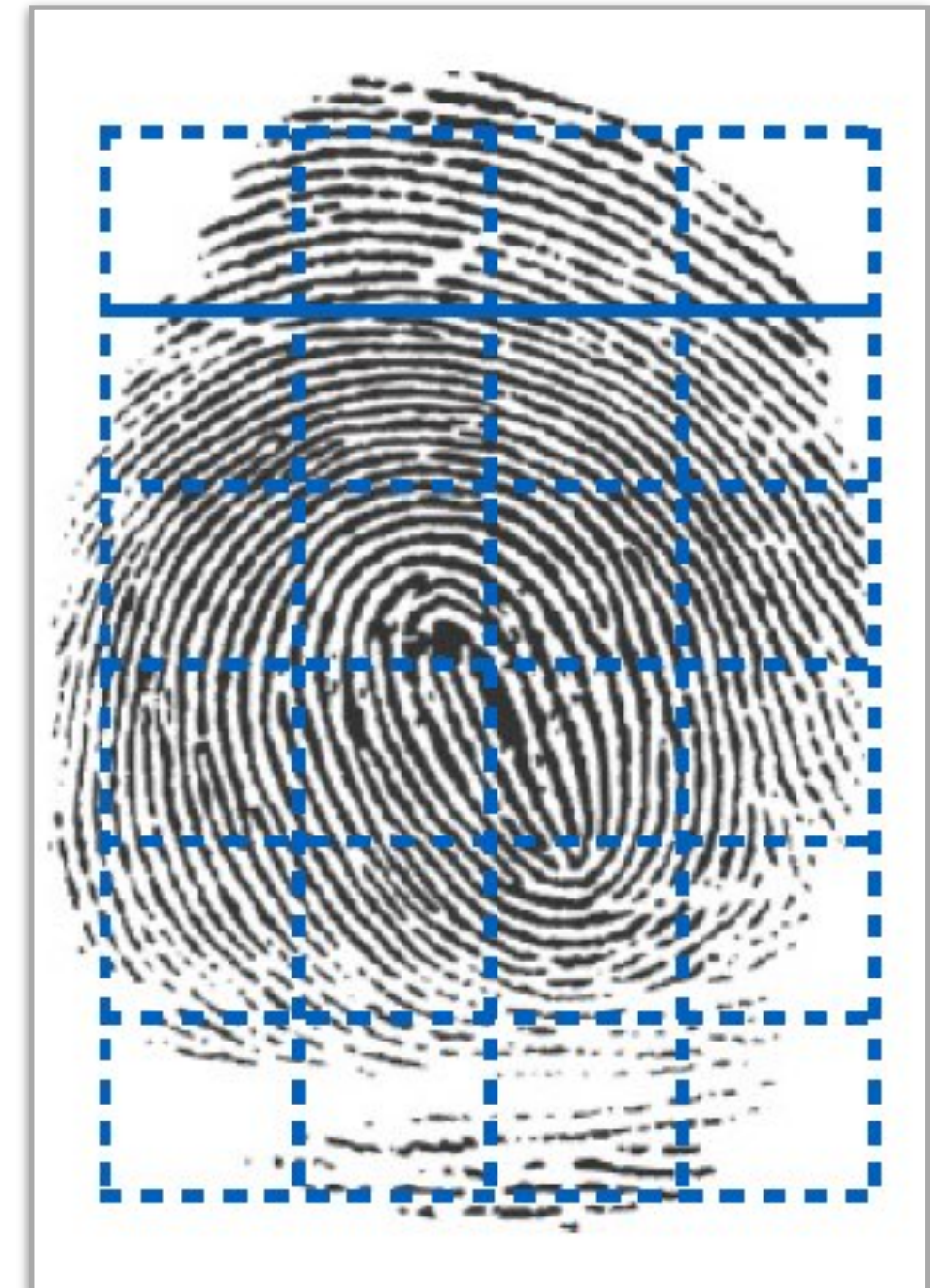
Features

Level-2 Features

Galton's Estimate

Chance of correct guess of
squares' disposition: $1/2^{12}$

considering the spatial restrictions



Source:
Dr. Walter Scheirer



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Features

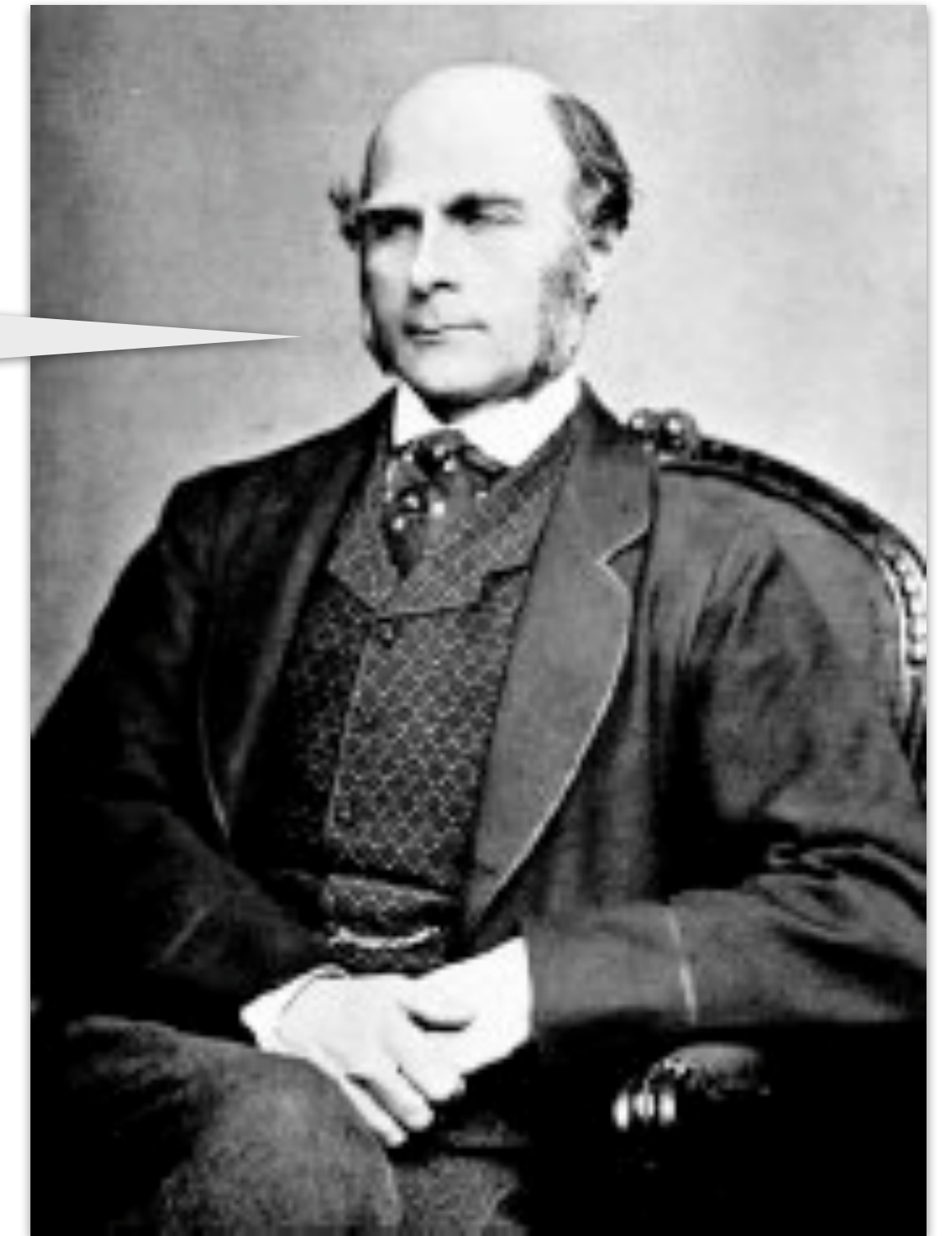
Level-2 Features

Galton's Estimate

1 in 64 billion

Total chance of a random fingerprint match a particular one:

$$1/2^{24} \times 1/2^{12} = 1/2^{36}$$



Features

Level-2 Features

Galton's Estimate

Total chance of a random fingerprint match a particular one:

$$1/2^{24} \times 1/2^{12} = 1/2^{36}$$

How many humans have ever lived?



107 billion

<https://www.bbc.com/news/magazine-16870579>

Features

What do we observe
in fingerprints?

Beyond Ridges and Valleys

Three types of features,
from coarse to fine levels:

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- **Level-3 Features**

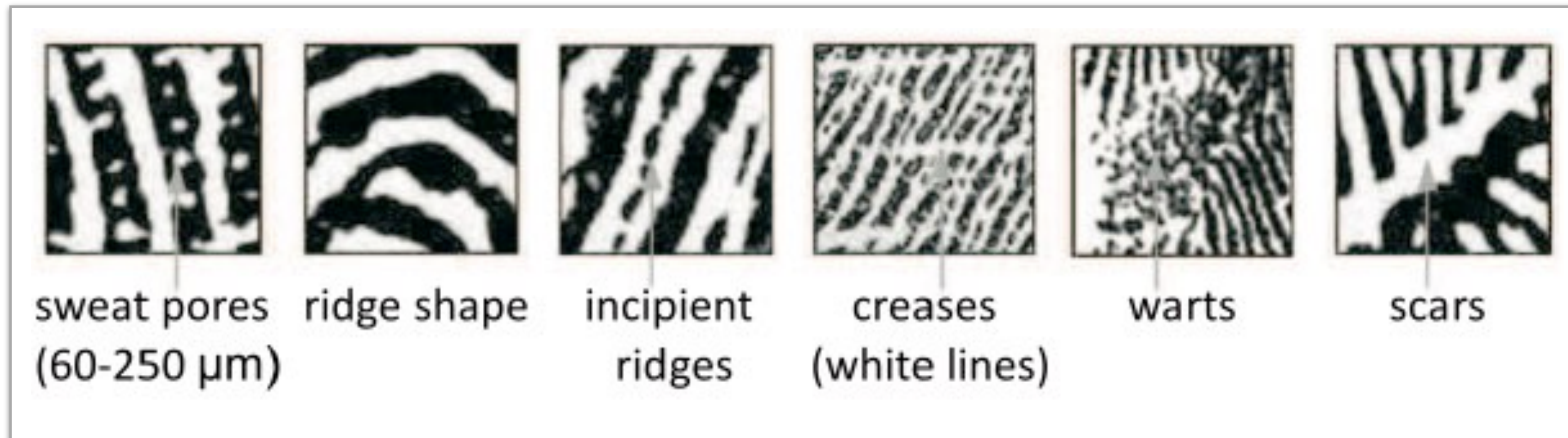


Features

Level-3 Features

Observe sweat pores, ridge shape, and lifetime acquired marks.

Useful capture resolution: 1000 ppi



Jain, Chen, and Demirkus
Pores and Ridges: High-Resolution Fingerprint Matching Using Level 3 Features
IEEE T-PAMI, 2007

Features

Level-3 Features

Observe sweat pores, ridge shape, and lifetime acquired marks.

Usage of Level-3 Features

Fingerprint liveness detection.

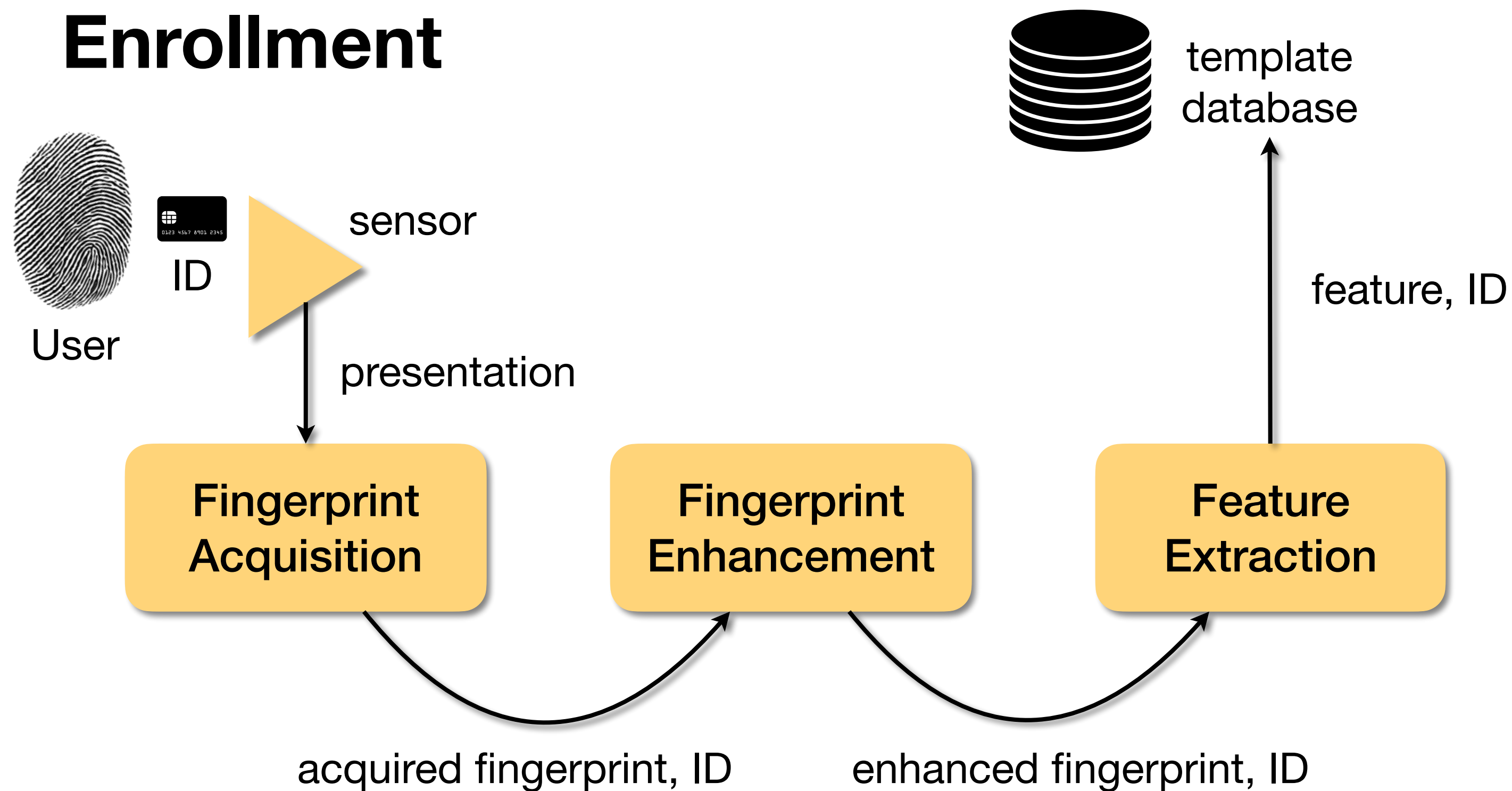
Rule-out questioned fingerprint matches.



<https://www.bbc.com/news/world-latin-america-21756709>

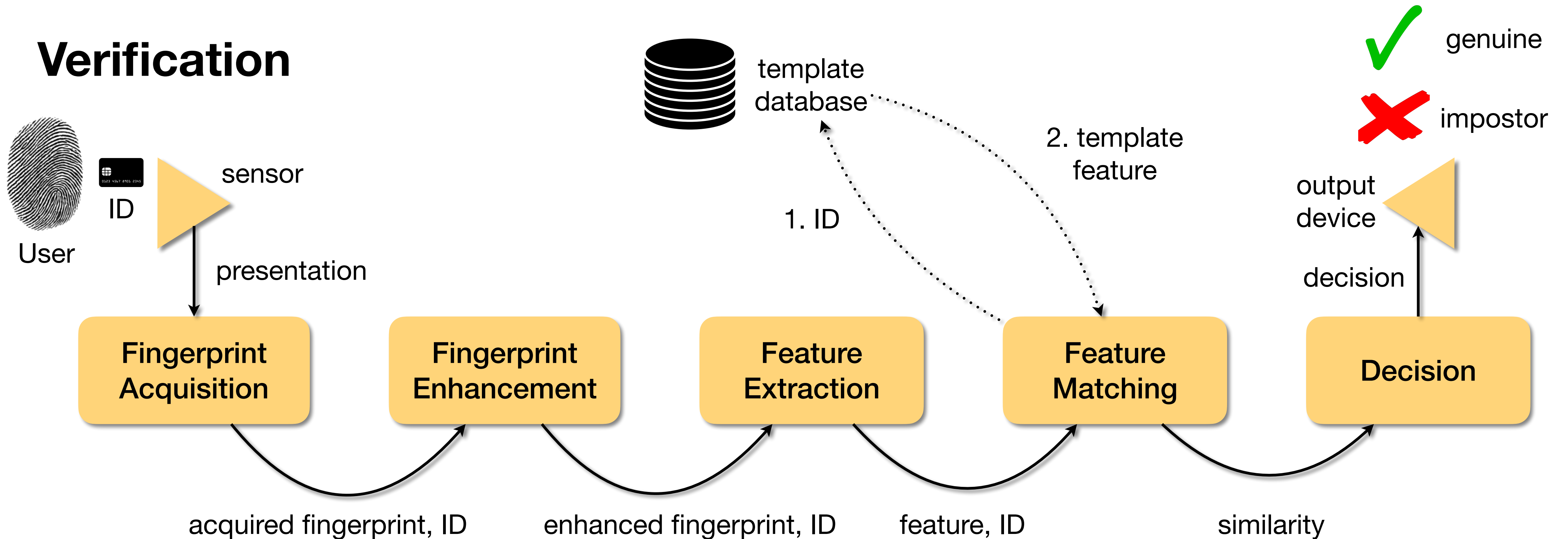
Fingerprint Recognition

Enrollment



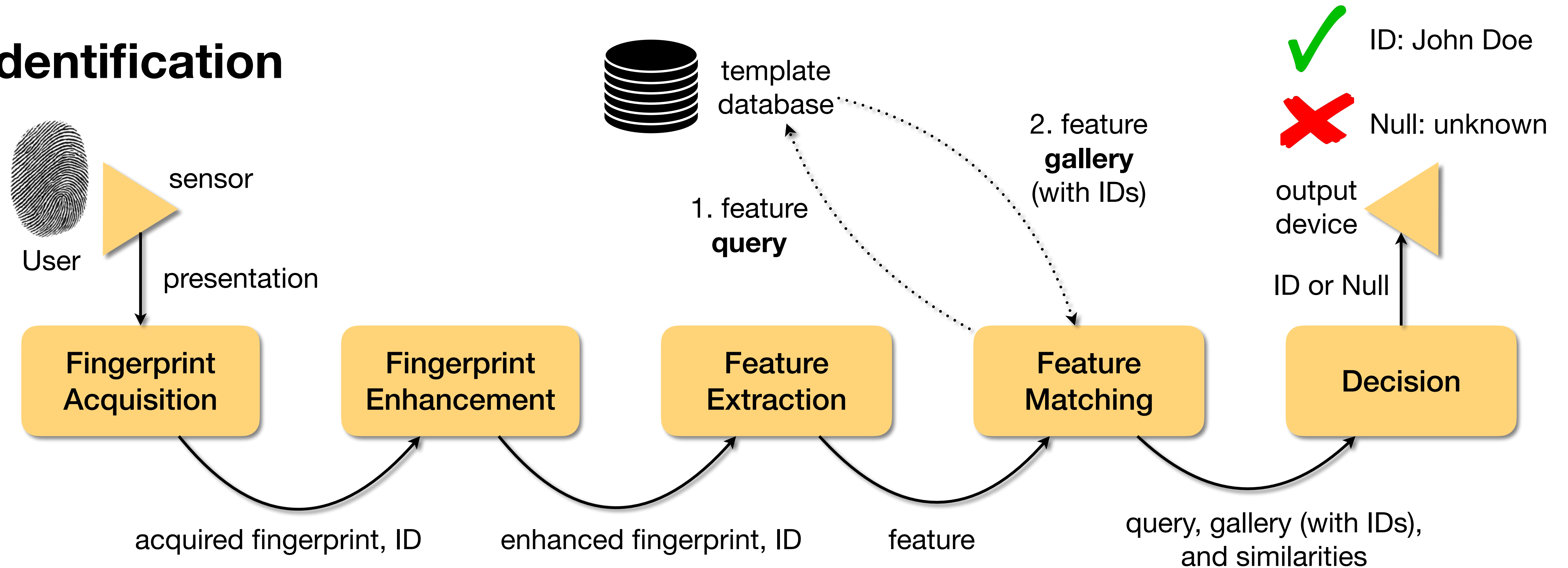
Fingerprint Recognition

Verification

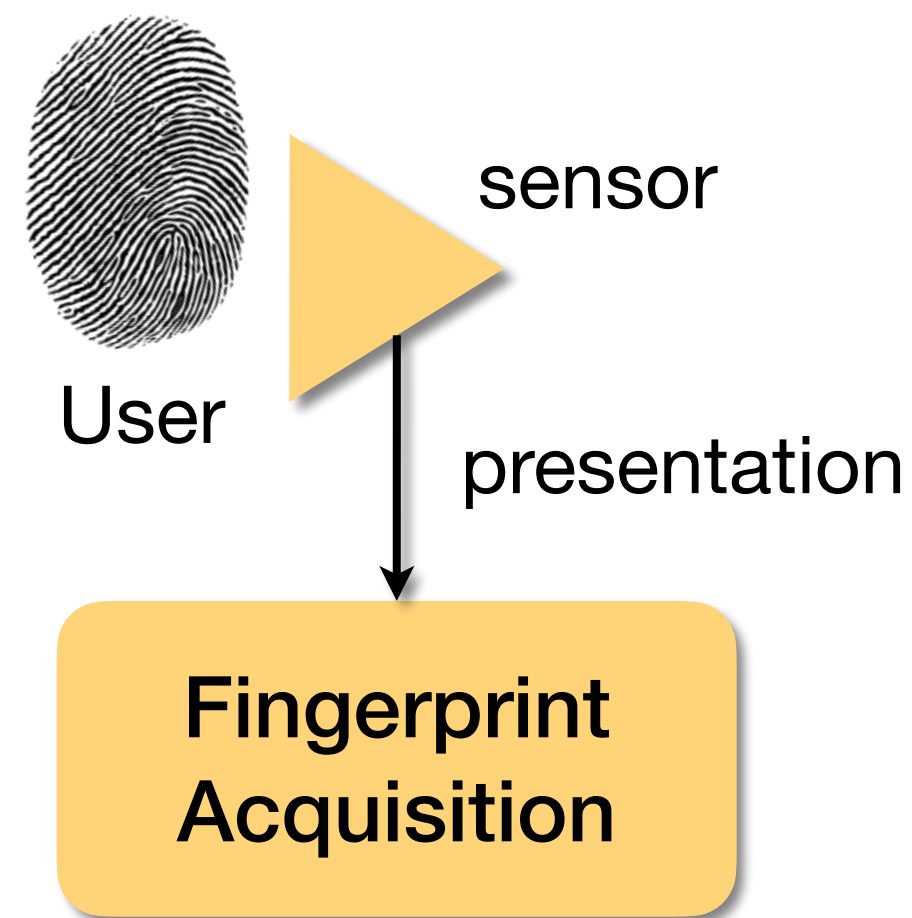


Fingerprint Recognition

Identification



Fingerprint Recognition



Acquisition

Off-line versus On-line



Acquisition

Off-line Acquisition Same fingerprint.

Jain, Ross, and Nadakumar
Introduction to Biometrics
Springer Books, 2011



rolled inked fingerprint



slap inked fingerprint

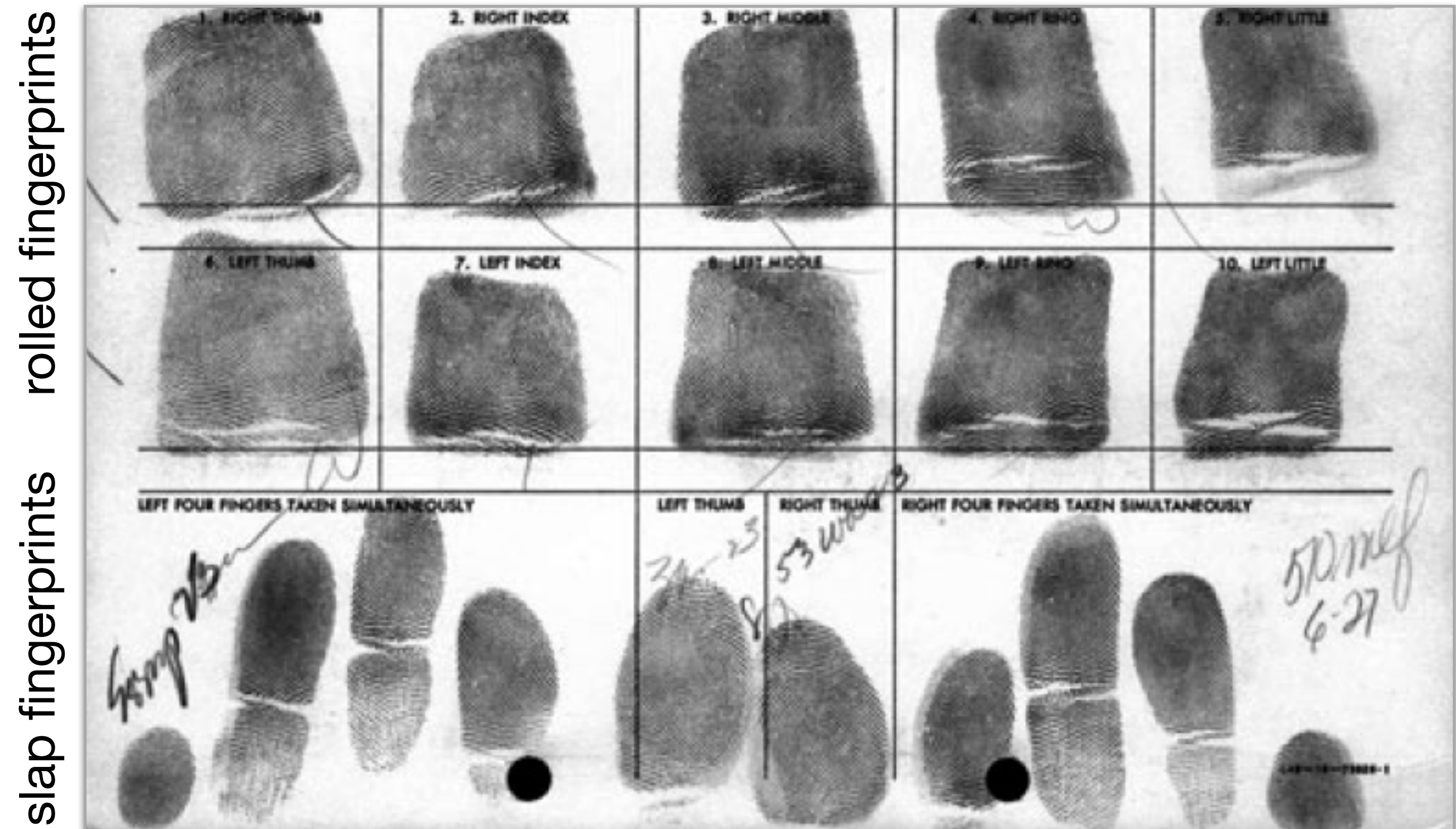


latent fingerprint

Acquisition

Off-line Acquisition

Scanning of
dactyloscopy cards.



Jain, Ross, and Nadakumar
Introduction to Biometrics
Springer Books, 2011

Acquisition

Off-line Acquisition

Photographing of
latent fingerprints.



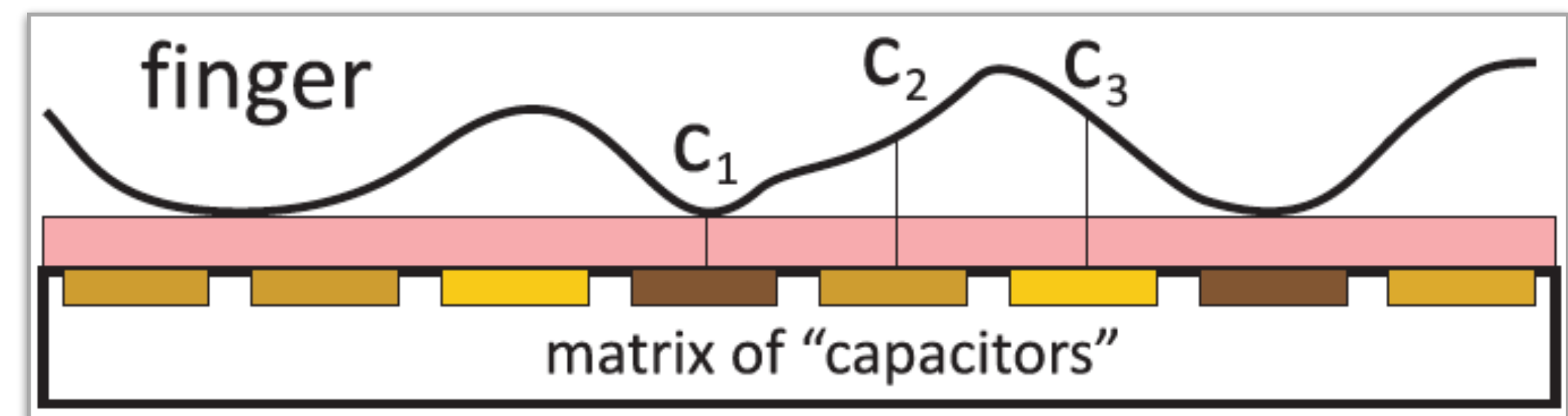
Source: Dr. Adam Czajka

Acquisition

On-line Acquisition

Capacitive sensors (1/6)

Ridges and valleys will generate different charges C_n , which will form different image segments.



Source: Dr. Adam Czajka

Low cost, but sensitive to dirt and moistness.

Typical resolution: 300 dpi (dots per inch).

Acquisition

On-line Acquisition

Capacitive sensors (1/6) Device and sample.



Precise Biometrics
Source: Dr. Adam Czajka



Source: <http://bias.csr.unibo.it/fvc2002/>

Acquisition

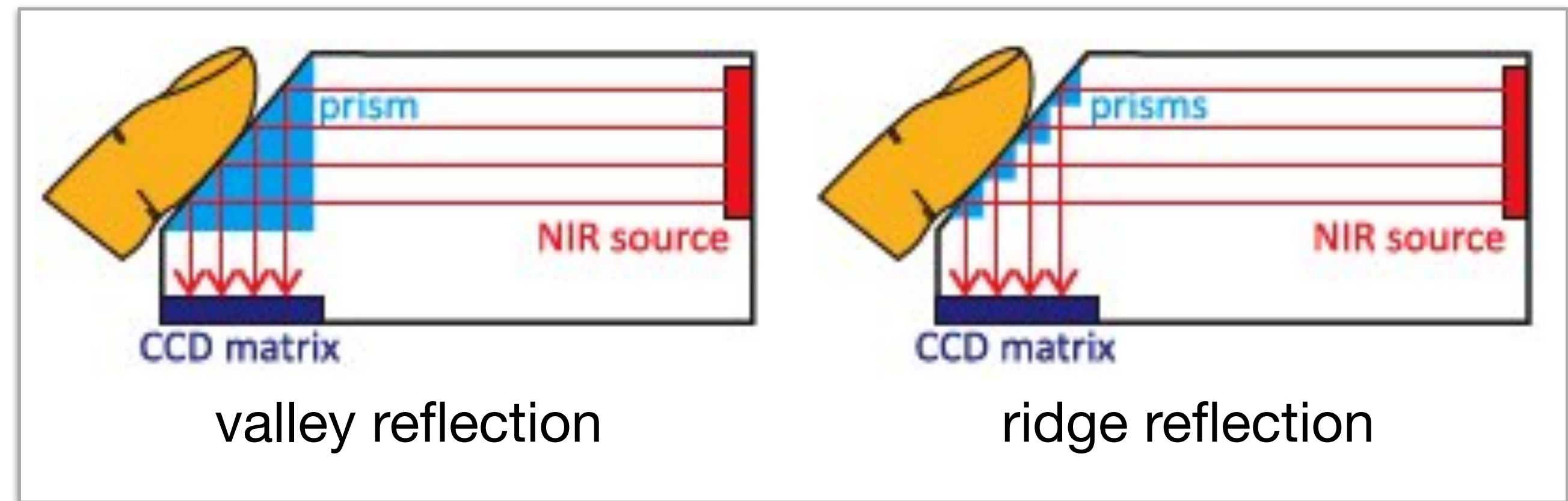
On-line Acquisition

Optical sensors (2/6)

Ridges won't be reflected on charge-coupled device (CCD) matrix, contrary to valleys, leading to darker image segments.

Typical resolution: 400-1000 dpi.

Source: Dr. Adam Czajka



Acquisition

On-line Acquisition

Optical sensors (2/6) Devices.



Identix

Source: Dr. Adam Czajka



Guardian

Acquisition

On-line Acquisition

Optical sensors (2/6) - Samples.

Source: Dr. Adam Czajka



slap

Biometrika FX2000



rolled

CrossMatch LS320



thumbs

L1 TP4100



little, ring, middle, and index

L1 TP4100

Acquisition

On-line Acquisition

Pressure sensors (3/6)

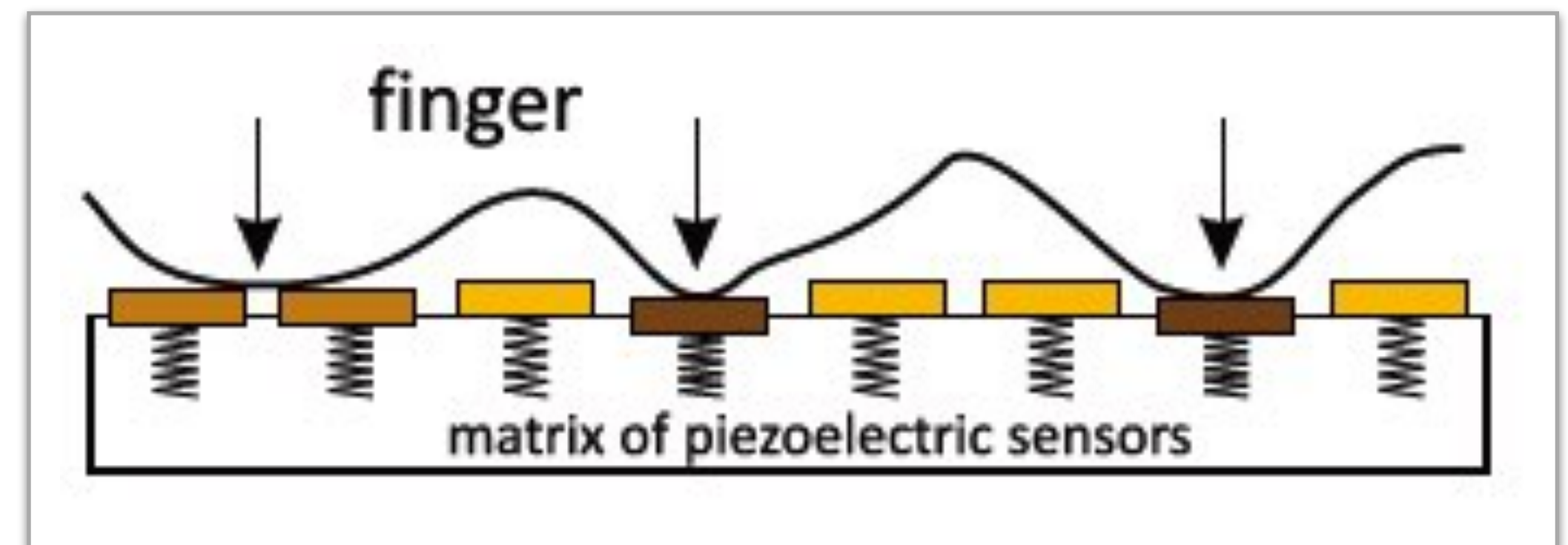
Also known as piezoelectric.

Ridges will cause stronger pressure than valleys, forming different image segments.

Robust to moistness.

Typical resolution: 400 dpi.

Source: Dr. Adam Czajka



Acquisition

On-line Acquisition

Pressure sensors (3/6) Device and sample.

Source: Dr. Adam Czajka



BMF/Hitachi

Source: Dr. Adam Czajka



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Acquisition

On-line Acquisition

Thermal sensors (4/6)

Based on surface temperature.

Ridges will transfer a different amount of heat when compared to valleys, leading to different image segments.



Acquisition

On-line Acquisition

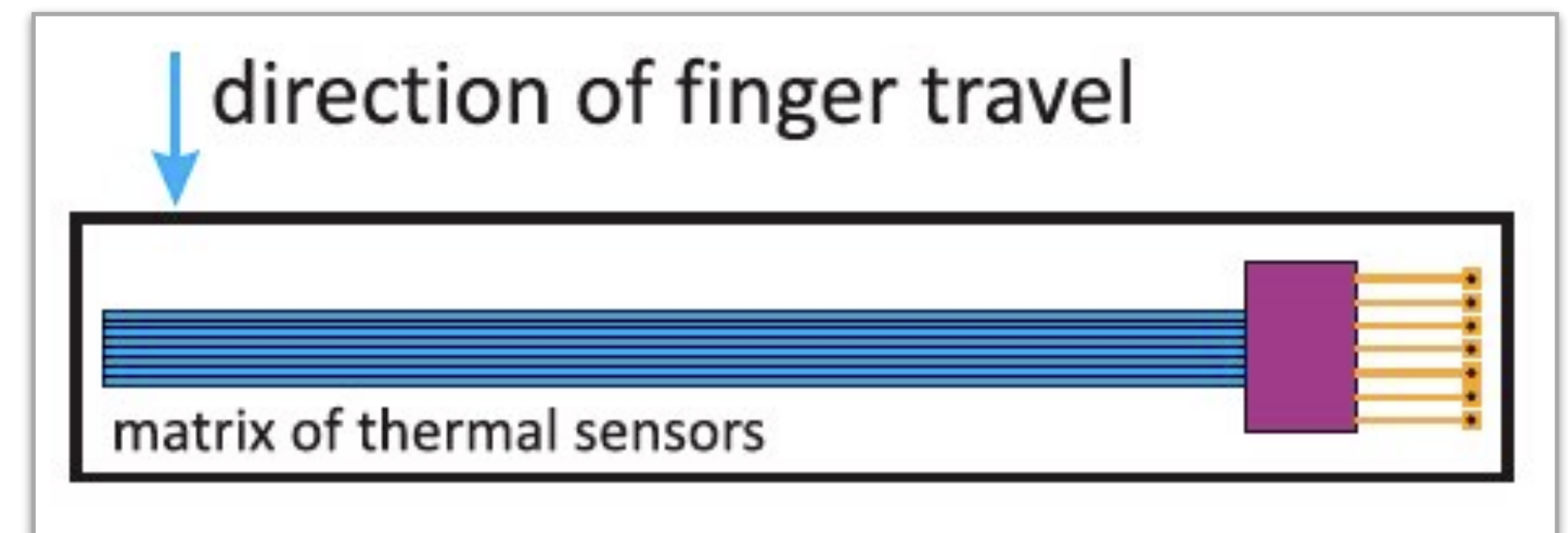
Thermal sensors (4/6)

Example: Atmel FingerChip

Finger is swept onto the sensor.

Thin sensor but high resolution
(typically 500 dpi).

While finger is swept, temperature is collected
at discrete time intervals.



Source: Dr. Adam Czajka

Acquisition

On-line Acquisition

Thermal sensors (4/6)

Example: Atmel FingerChip
Sample generation.



Source: Atmel FingerChip

finger sweep

discrete collection

fingerprint reconstruction

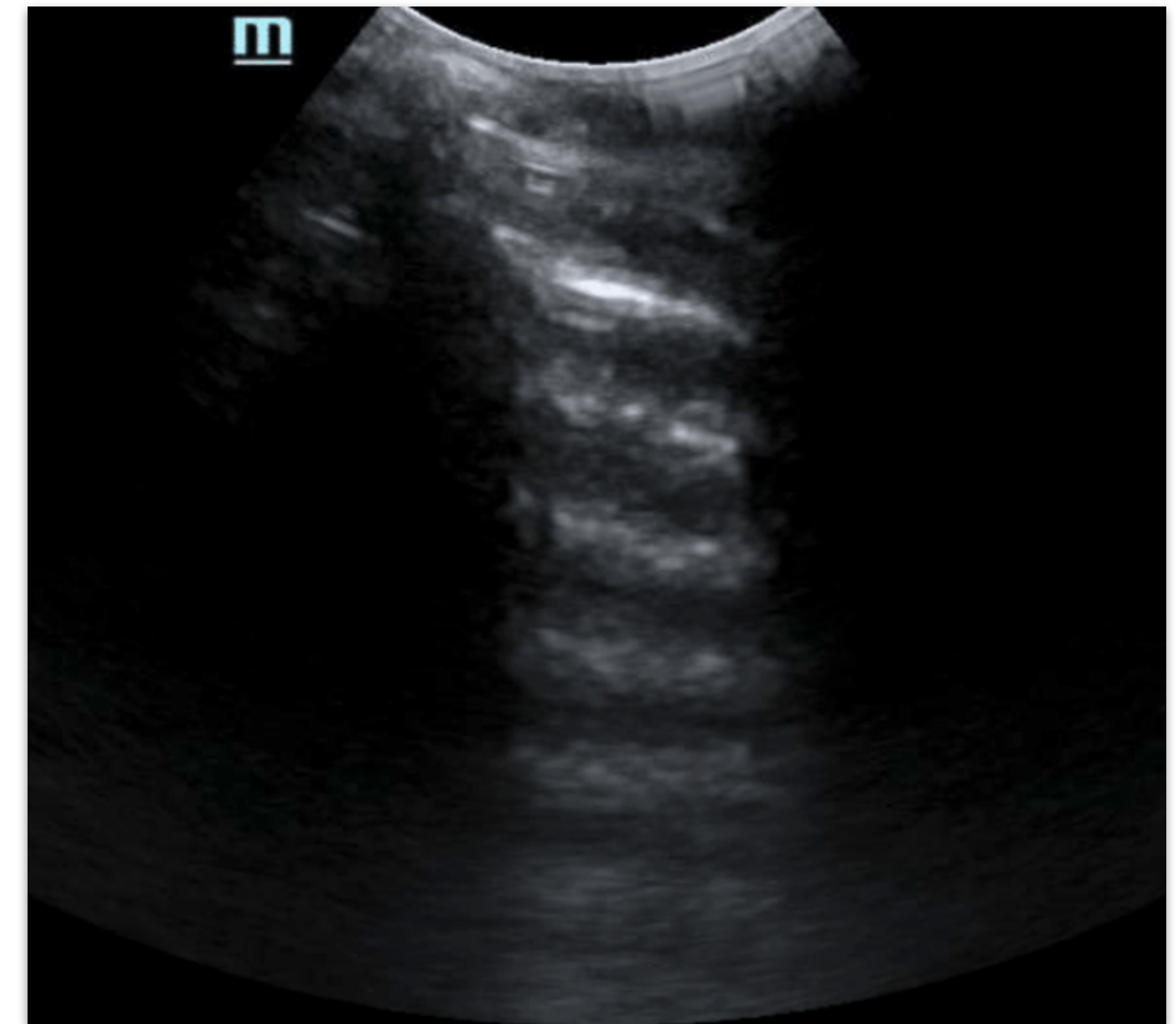
Acquisition

On-line Acquisition

Ultrasound sensors (5/6)

Measures the scattering of sound waves over the finger surface.

Ridges and valleys will produce different scattering, leading to different image segments.



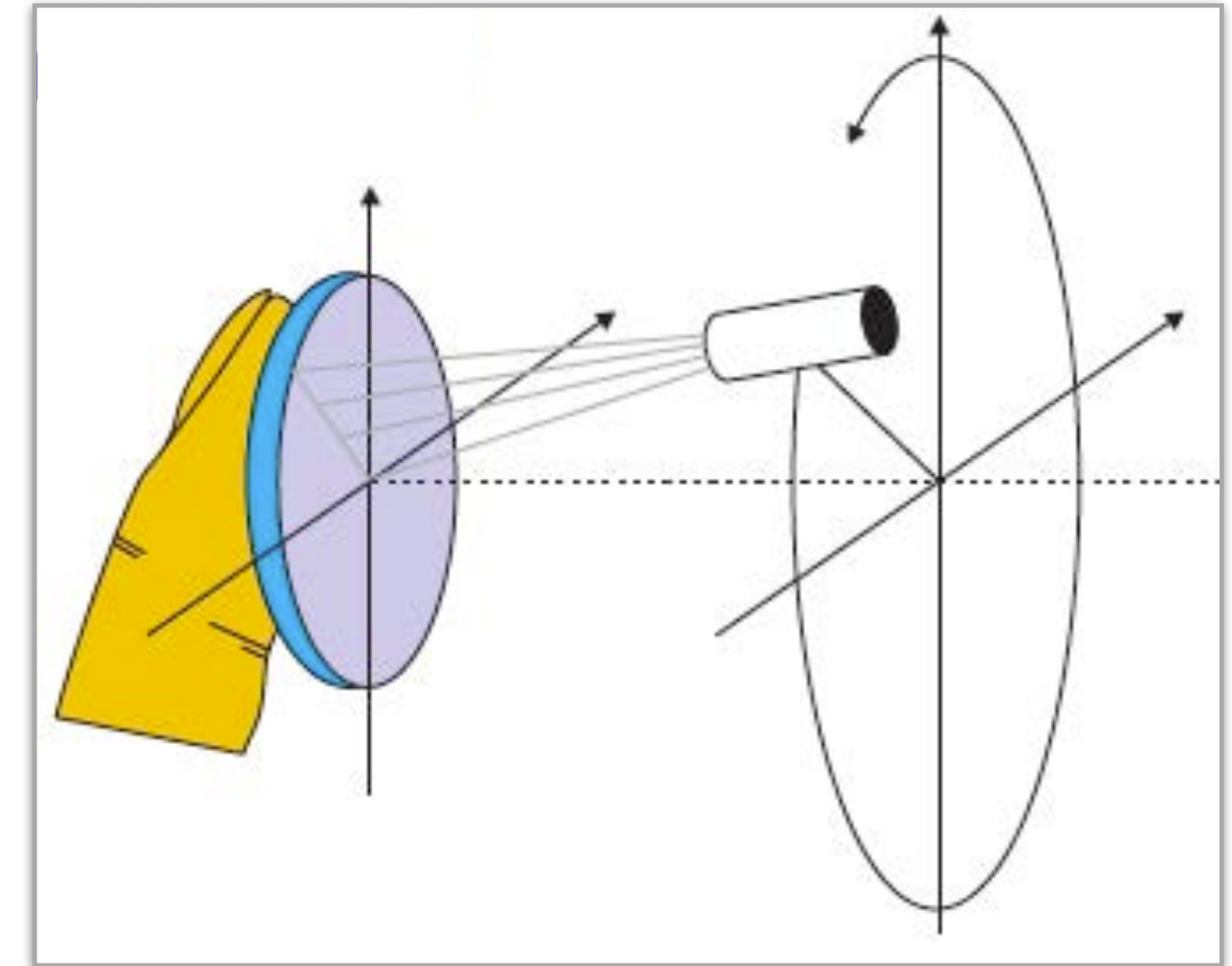
Acquisition

On-line Acquisition

Ultrasound sensors (5/6)

Example: Optel

Transducer moves along a circular trajectory whose central axis is perpendicular to the fingertip.



Source: Dr. Adam Czajka

More expensive. Typical resolution: 250 dpi.
Harder to be spoofed (due to ultrasounds penetration).

Acquisition

On-line Acquisition

Ultrasound sensors (5/6)

Example: Optel
Device and sample.



Source: www.optel.com.pl

Acquisition

On-line Acquisition

Ultrasound sensor (5/6)

Example: Qualcomm Fingerprint

Sensor embedded into the device display.



Source: mashable.com

Acquisition

On-line Acquisition

Touchless sensor (6/6)

3D imaging with CCD sensor.

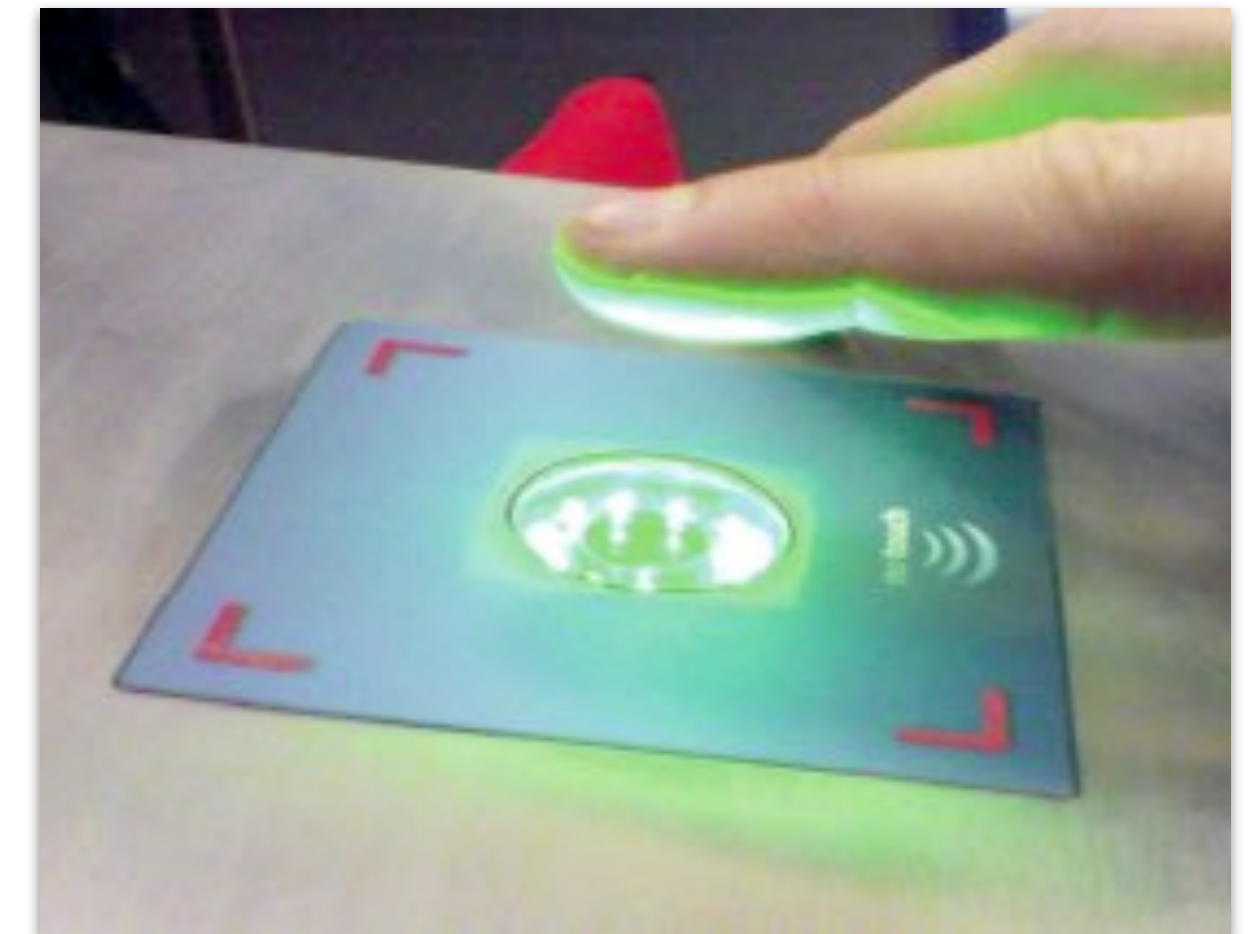
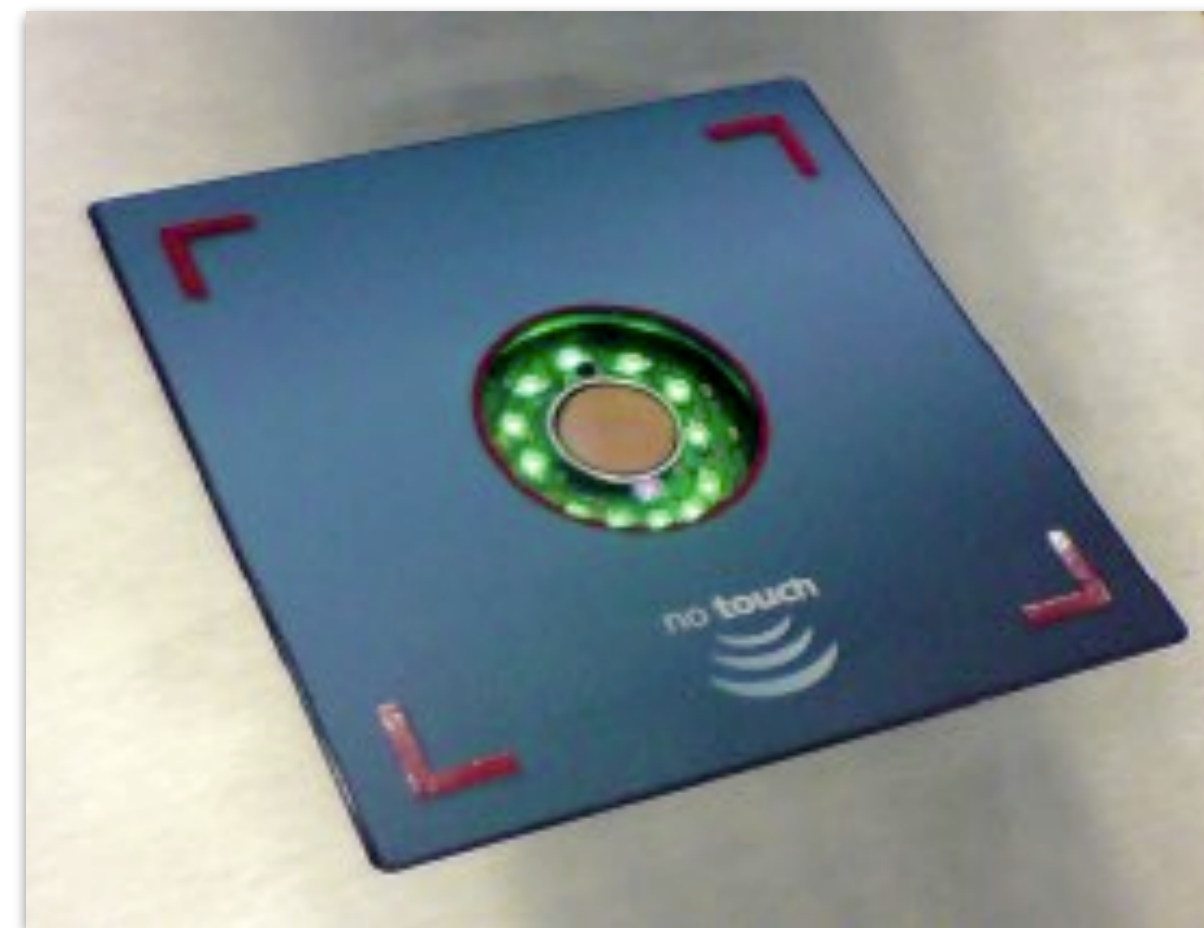


Acquisition

On-line Acquisition

Touchless sensor (6/6)

Example: TST Biometrics Device.

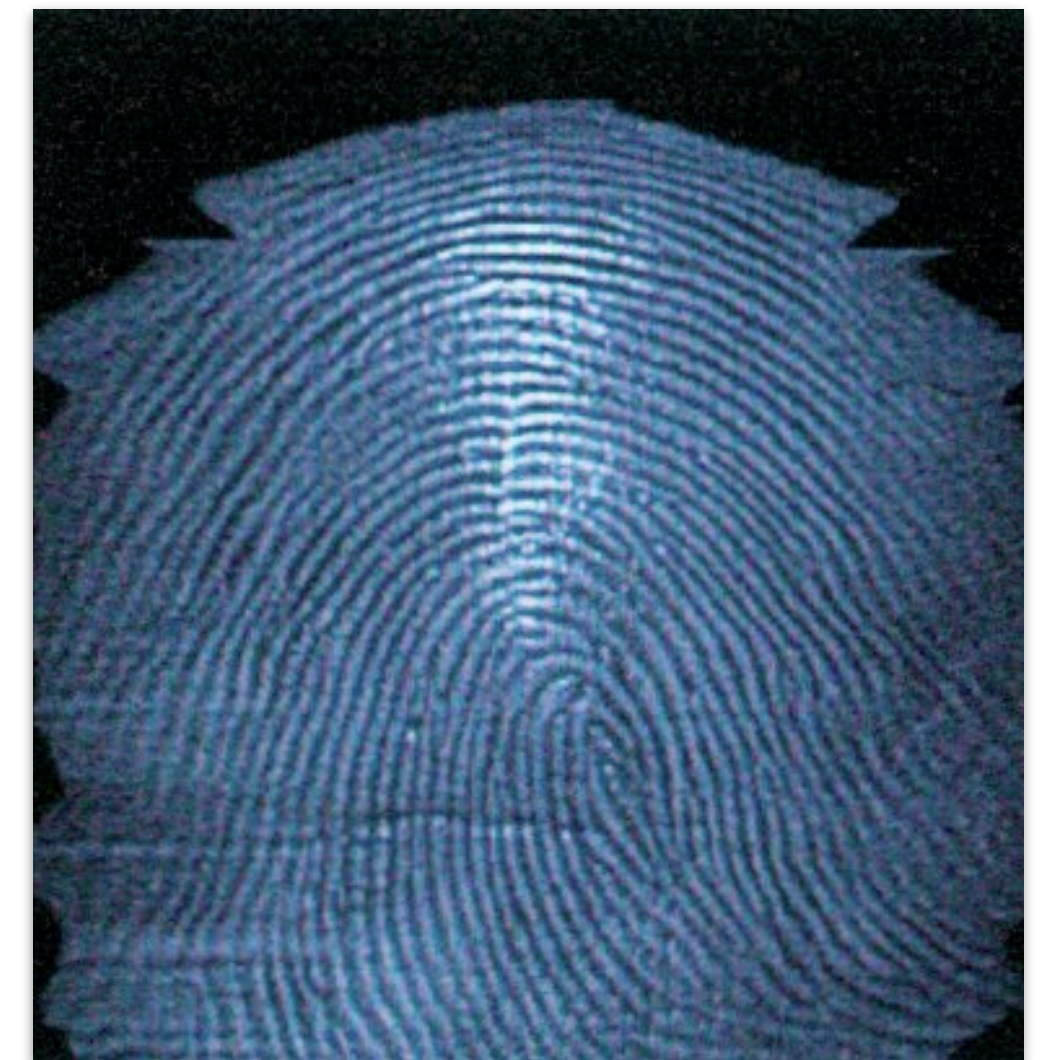


Source: Dr. Adam Czajka

Acquisition

On-line Acquisition

Touchless sensor (6/6)
Example: MorphoWave
Device and sample.



Source: Dr. Adam Czajka

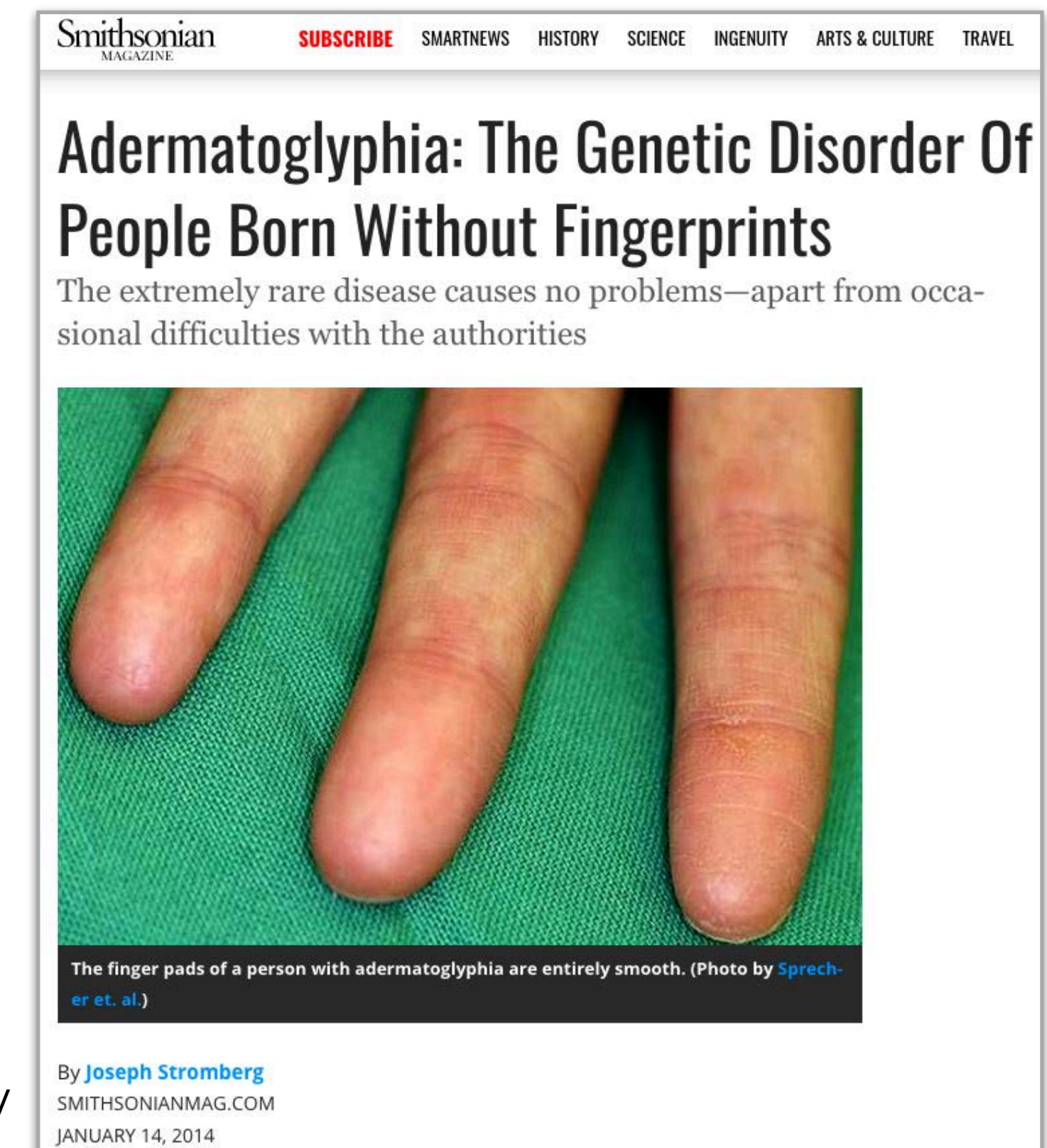
Acquisition

Problems

Adermatoglyphia

Leads to failure to acquire (FTA)
and failure to enroll (FTE).

<https://www.smithsonianmag.com/science-nature/adermatoglyphia-genetic-disorder-people-born-without-fingerprints-180949338/>



Acquisition

Problems

Presentation Attack

Techniques to generate fake fingerprints:

- Paper printouts.

- Clay or latex molds, plus wood-glue, gelatin, or silicone mold filling.



Source: Dr. Adam Czajka

Objectives: spoofing and obfuscation.

Faking Fingerprints



Available at: <https://www.youtube.com/watch?v=KdycMYILTr0>

Acquisition

Problems

Presentation Attack

How robust might be the different sensors?

Capacitive, Pressure, and Thermal

May be fooled, if synthetic material presents similar skin properties.
Not enough resolution for level-3 features.

Optical

May be fooled, including paper printout.
Larger resolution will allow the use of level-3 features.

Acquisition

Problems

Presentation Attack

How robust might be the different sensors?

Ultrasound

May be robust if ultrasound penetration is used.

Touchless

Flat fake samples may not work due to 3D detection.

Acquisition

Problems

Presentation Attack
How about humans?



Fake or authentic?

From capacitive sensor



Fake or authentic?

From capacitive sensor

Matsumoto, T.
Importance of Open Discussion on Adversarial Analyses for Mobile Security Technologies---A Case Study for User Identification---
ITU-T Workshop on Security, Seoul, 2002



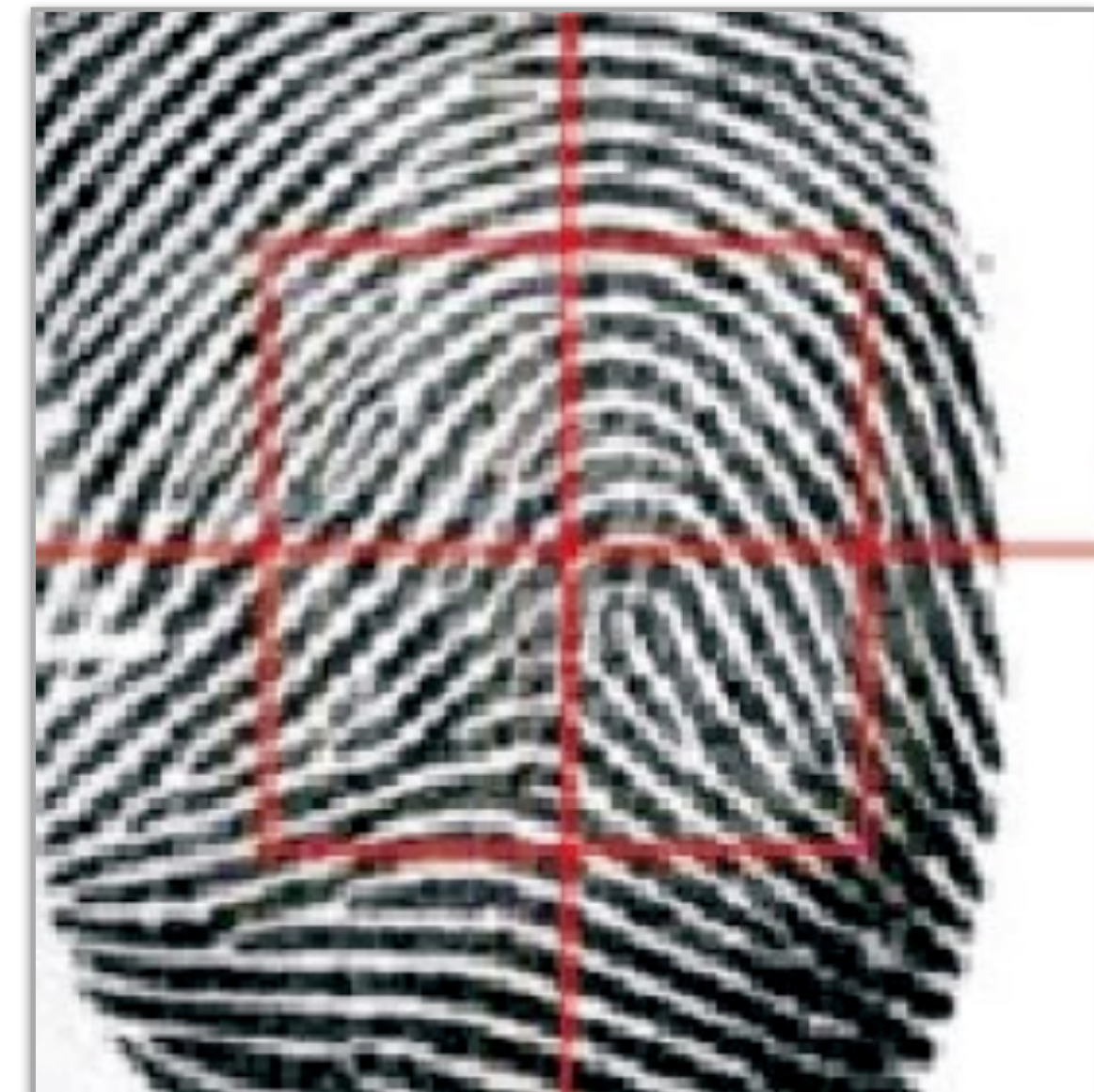
authentic



gelatin

Fake or authentic?

From optical sensor



Fake or authentic?

From optical sensor



authentic



silicone



gelatin

Matsumoto, T.
*Importance of Open Discussion on Adversarial Analyses for Mobile
Security Technologies---A Case Study for User Identification---*
ITU-T Workshop on Security, Seoul, 2002

Fake or authentic?

From optical sensor



Fake or authentic?

From optical sensor

Source: Dr. Adam Czajka

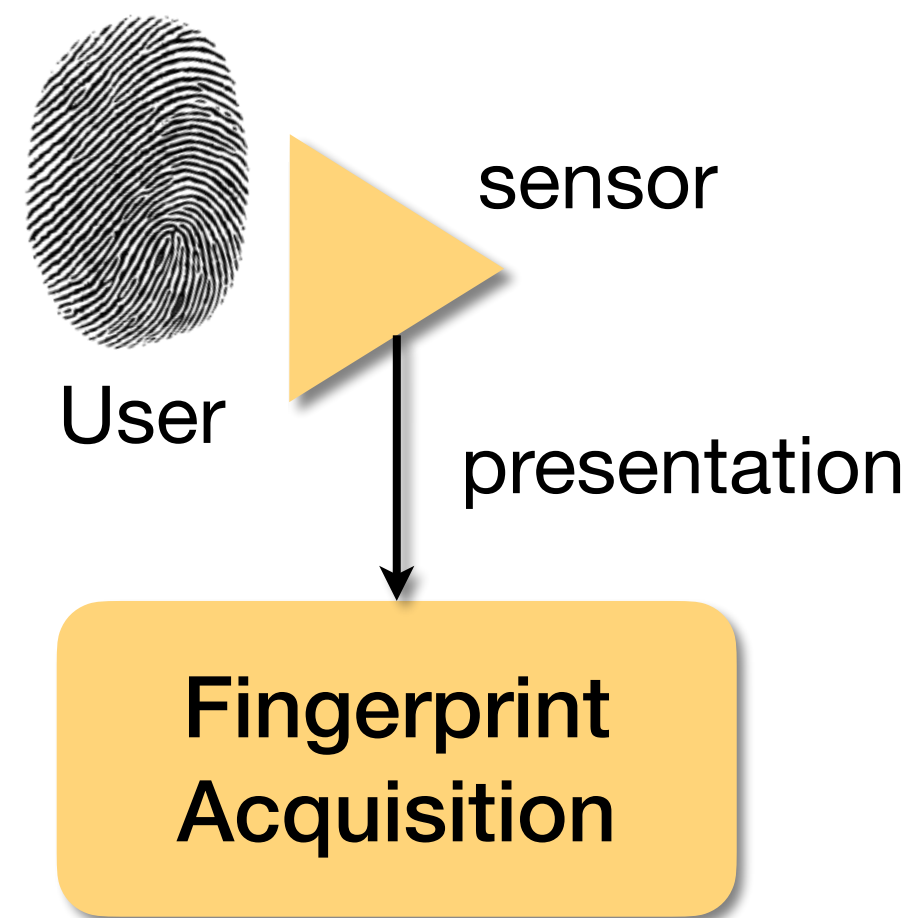
wood glue



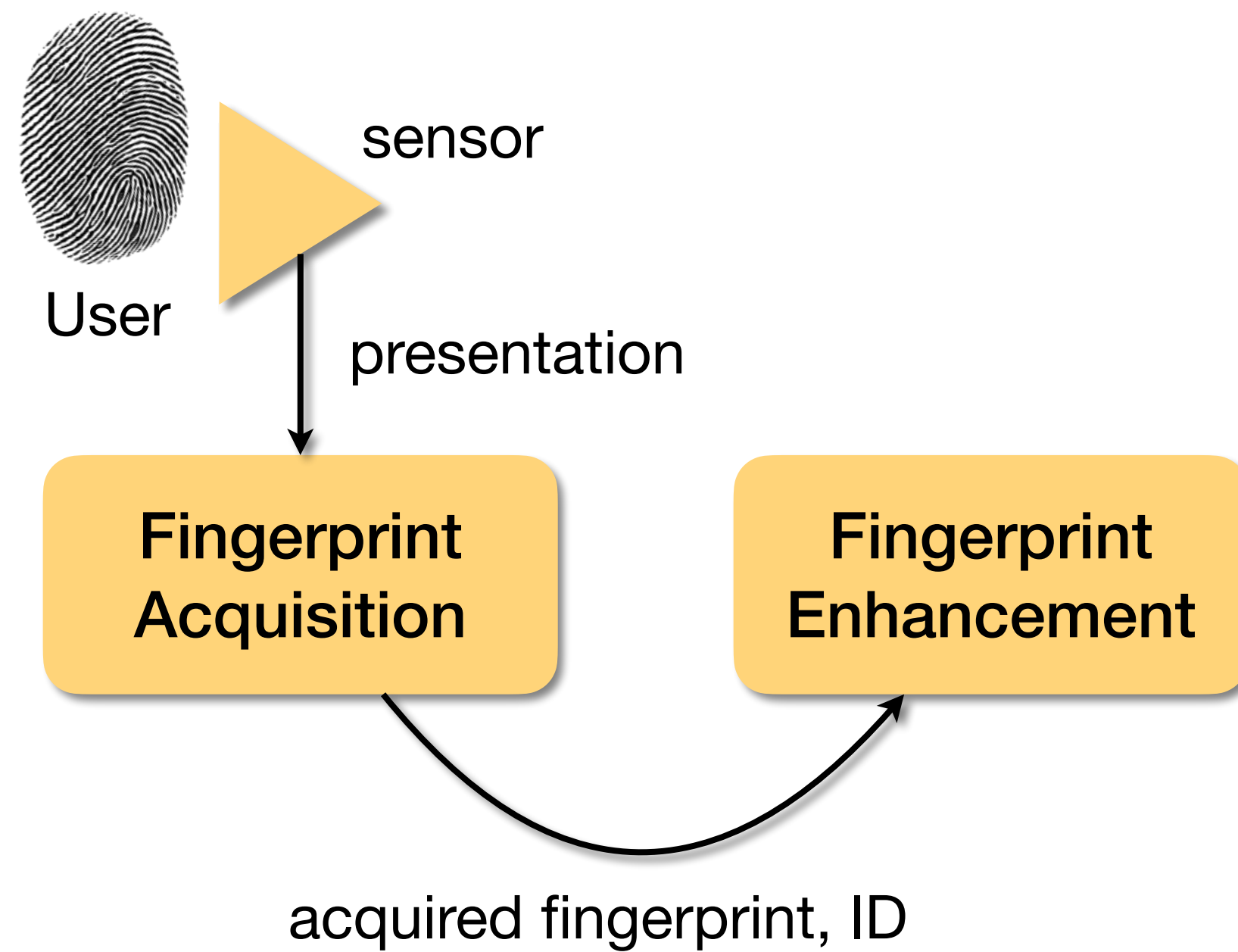
authentic



Fingerprint Recognition



Fingerprint Recognition



Enhancement

Objectives

Noise removal.

Keep only essential information.

Reduce intra-class variation.

Why do we need to enhance?

Poor illumination conditions.

Careless fingerprint presentation.

Limited sensor accuracy.

Sensor dirtiness.

Skin condition.



Enhancement

Capture Condition



too bright



too dark

Enhancement

Skin Condition

Maltoni et al.
Handbook of Fingerprint Recognition
Springer Books, 2009



normal



dry



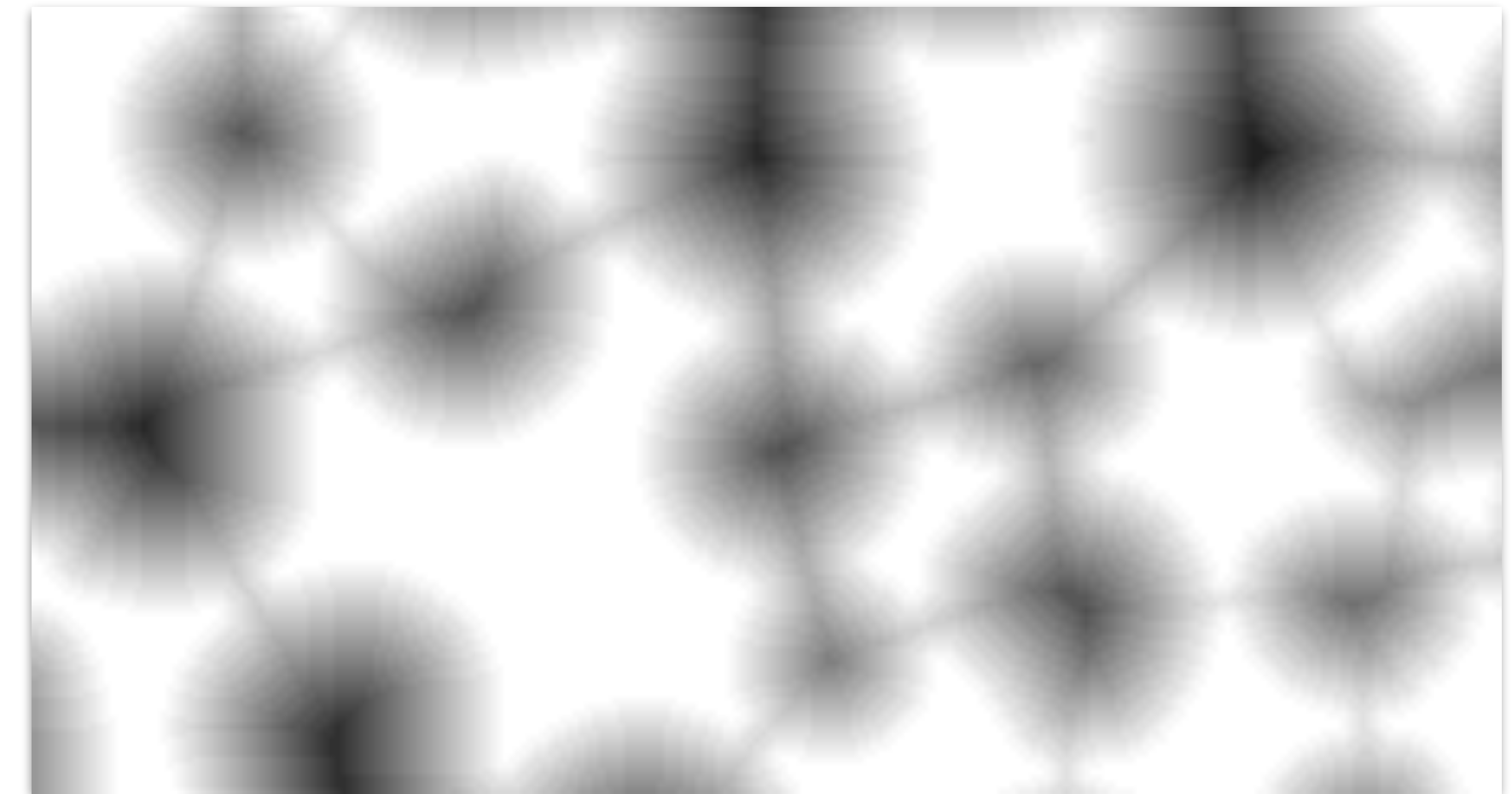
wet

Enhancement

Image Processing Solutions

Tasks

Enhancement of image contrast.
Enhancement of ridges and valleys.
Content segmentation.
Others.



Enhancement

Image Processing Solutions

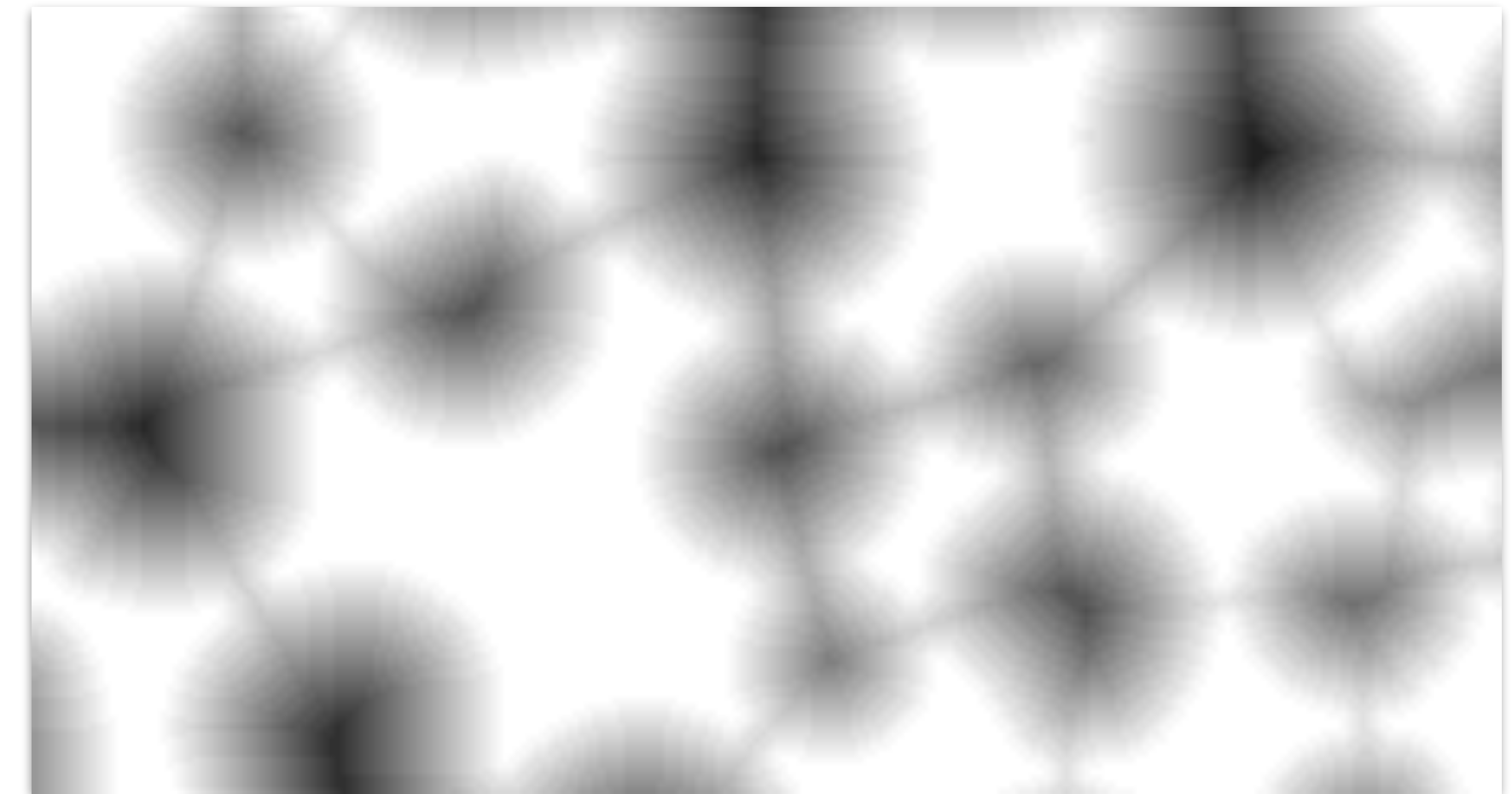
Tasks

Enhancement of image contrast.

Enhancement of ridges and valleys.

Content segmentation.

Others.



Enhancement

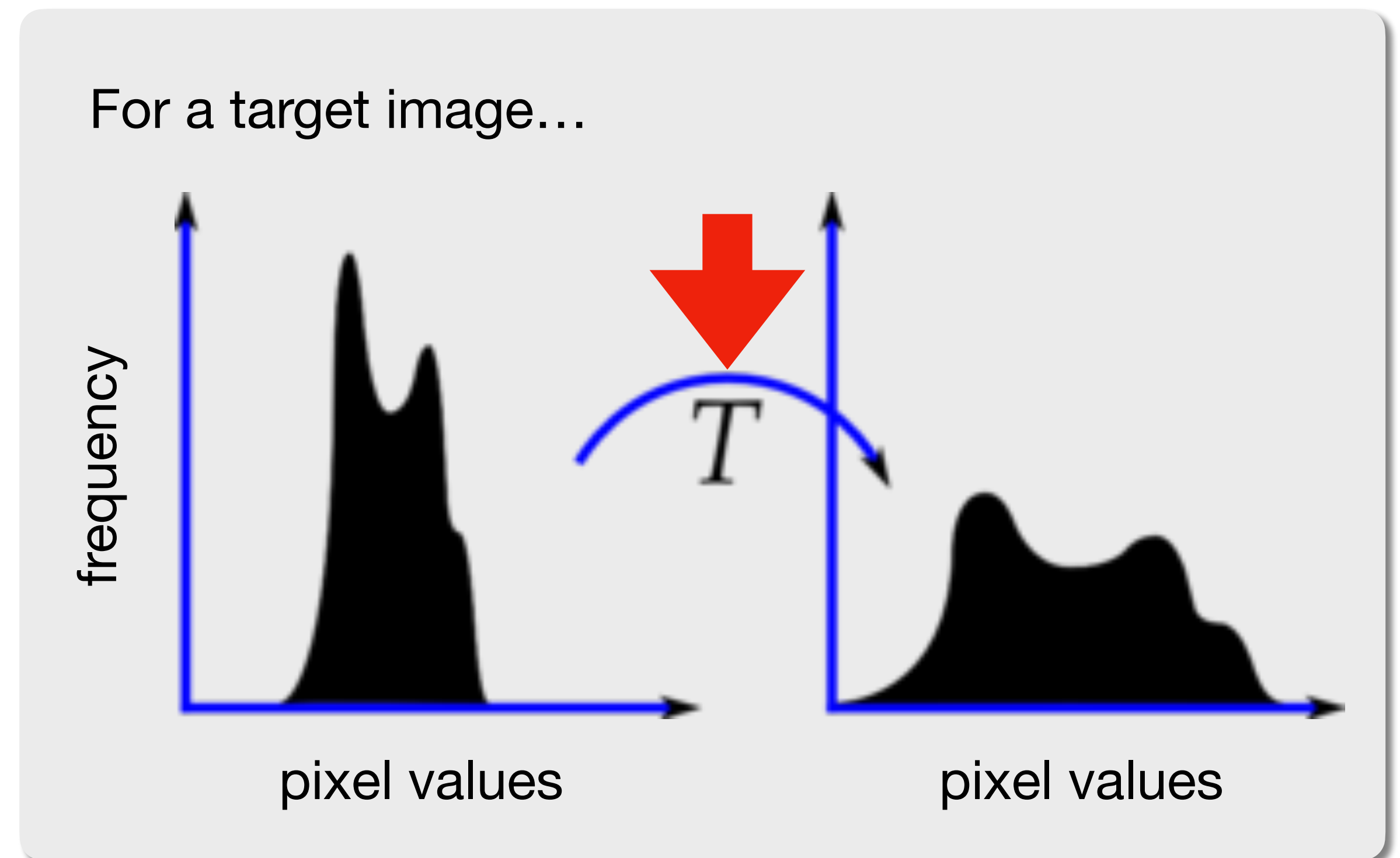
Image Contrast

Example:

Color histogram equalization.

Useful when pixel values are confined to a specific range (too bright or too dark images).

Stretching the color histogram will improve the contrast.



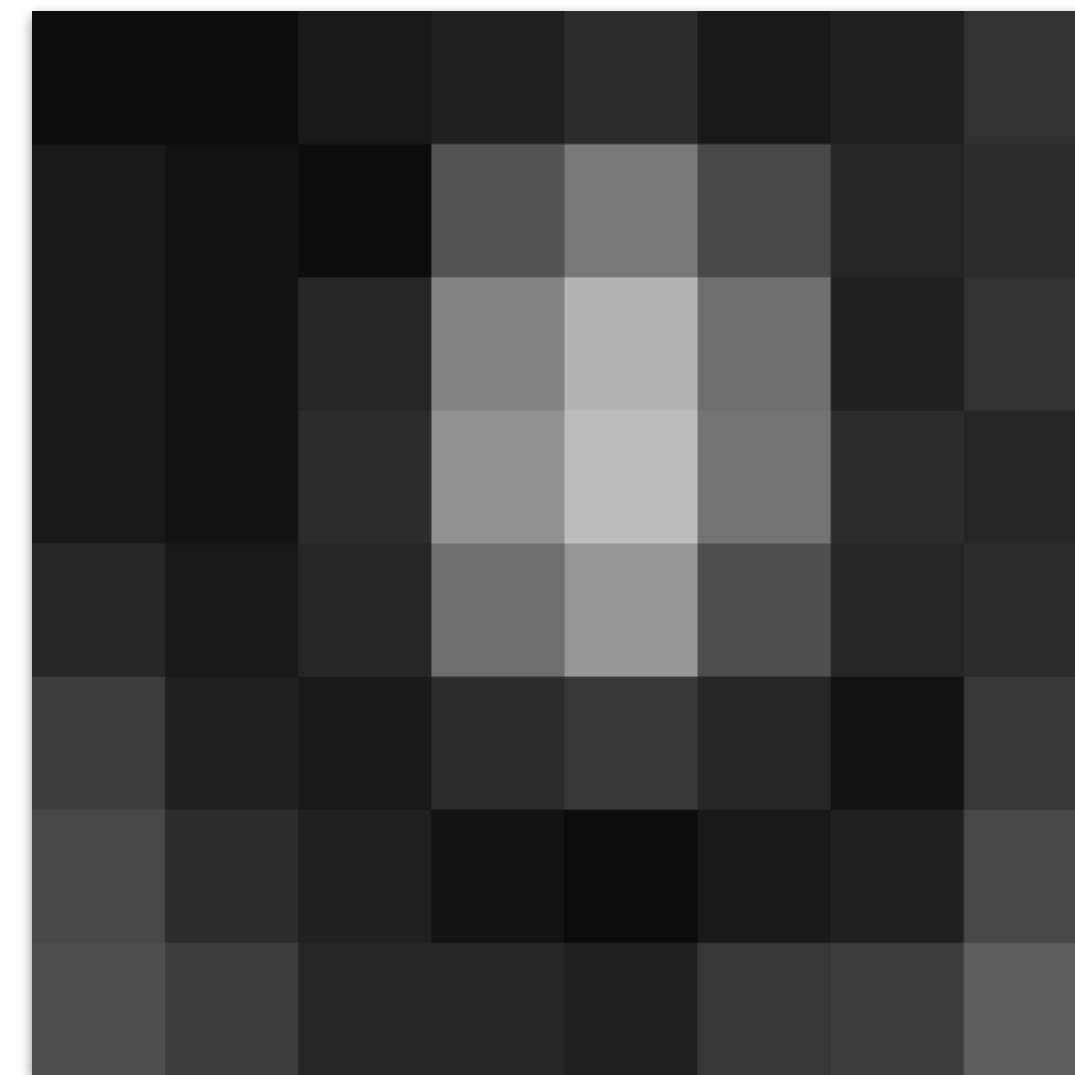
Enhancement

Color Histogram Equalization

Simple implementation

Toy Case

| | | | | | | | |
|----|----|----|-----|-----|-----|----|----|
| 52 | 55 | 61 | 59 | 79 | 61 | 76 | 61 |
| 62 | 59 | 55 | 104 | 94 | 85 | 59 | 71 |
| 63 | 65 | 66 | 113 | 144 | 104 | 63 | 72 |
| 64 | 70 | 70 | 126 | 154 | 109 | 71 | 69 |
| 67 | 73 | 68 | 106 | 122 | 88 | 68 | 68 |
| 68 | 79 | 60 | 70 | 77 | 66 | 58 | 75 |
| 69 | 85 | 64 | 58 | 55 | 61 | 65 | 83 |
| 70 | 87 | 69 | 68 | 65 | 73 | 78 | 90 |



Source: https://en.wikipedia.org/wiki/Histogram_equalization

Enhancement

Color Histogram Equalization

Simple implementation

Toy Case

1. Compute cumulative distribution function (CDF)

| | | | | | | | |
|----|----|----|-----|-----|-----|----|----|
| 52 | 55 | 61 | 59 | 79 | 61 | 76 | 61 |
| 62 | 59 | 55 | 104 | 94 | 85 | 59 | 71 |
| 63 | 65 | 66 | 113 | 144 | 104 | 63 | 72 |
| 64 | 70 | 70 | 126 | 154 | 109 | 71 | 69 |
| 67 | 73 | 68 | 106 | 122 | 88 | 68 | 68 |
| 68 | 79 | 60 | 70 | 77 | 66 | 58 | 75 |
| 69 | 85 | 64 | 58 | 55 | 61 | 65 | 83 |
| 70 | 87 | 69 | 68 | 65 | 73 | 78 | 90 |

color histogram

| Value | Count | Value | Count | Value | Count | Value | Count | Value | Count |
|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|
| 52 | 1 | 64 | 2 | 72 | 1 | 85 | 2 | 113 | 1 |
| 55 | 3 | 65 | 3 | 73 | 2 | 87 | 1 | 122 | 1 |
| 58 | 2 | 66 | 2 | 75 | 1 | 88 | 1 | 126 | 1 |
| 59 | 3 | 67 | 1 | 76 | 1 | 90 | 1 | 144 | 1 |
| 60 | 1 | 68 | 5 | 77 | 1 | 94 | 1 | 154 | 1 |
| 61 | 4 | 69 | 3 | 78 | 1 | 104 | 2 | | |
| 62 | 1 | 70 | 4 | 79 | 2 | 106 | 1 | | |
| 63 | 2 | 71 | 2 | 83 | 1 | 109 | 1 | | |

Source: https://en.wikipedia.org/wiki/Histogram_equalization

Enhancement

Color Histogram Equalization

Simple implementation

Toy Case

1. Compute cumulative distribution function (CDF)

| | | | | | | | |
|----|----|----|-----|-----|-----|----|----|
| 52 | 55 | 61 | 59 | 79 | 61 | 76 | 61 |
| 62 | 59 | 55 | 104 | 94 | 85 | 59 | 71 |
| 63 | 65 | 66 | 113 | 144 | 104 | 63 | 72 |
| 64 | 70 | 70 | 126 | 154 | 109 | 71 | 69 |
| 67 | 73 | 68 | 106 | 122 | 88 | 68 | 68 |
| 68 | 79 | 60 | 70 | 77 | 66 | 58 | 75 |
| 69 | 85 | 64 | 58 | 55 | 61 | 65 | 83 |
| 70 | 87 | 69 | 68 | 65 | 73 | 78 | 90 |

| v, Pixel Intensity | cdf(v) |
|--------------------|--------|
| 52 | 1 |
| 55 | 4 |
| 58 | 6 |
| 59 | 9 |
| 60 | 10 |
| 61 | 14 |
| 62 | 15 |
| 63 | 17 |
| 64 | 19 |

Source: https://en.wikipedia.org/wiki/Histogram_equalization

Enhancement

Color Histogram Equalization

Simple implementation

Toy Case

1. Compute cumulative distribution function (CDF)

2. Perform min-max normalization
[0, 255] interval

| | | | | | | | |
|----|----|----|-----|-----|-----|----|----|
| 52 | 55 | 61 | 59 | 79 | 61 | 76 | 61 |
| 62 | 59 | 55 | 104 | 94 | 85 | 59 | 71 |
| 63 | 65 | 66 | 113 | 144 | 104 | 63 | 72 |
| 64 | 70 | 70 | 126 | 154 | 109 | 71 | 69 |
| 67 | 73 | 68 | 106 | 122 | 88 | 68 | 68 |
| 68 | 79 | 60 | 70 | 77 | 66 | 58 | 75 |
| 69 | 85 | 64 | 58 | 55 | 61 | 65 | 83 |
| 70 | 87 | 69 | 68 | 65 | 73 | 78 | 90 |

| v, Pixel Intensity | cdf(v) | h(v), Equalized v |
|--------------------|--------|-------------------|
| 52 | 1 | 0 |
| 55 | 4 | 12 |
| 58 | 6 | 20 |
| 59 | 9 | 32 |
| 60 | 10 | 36 |
| ... | | |
| 120 | 52 | 247 |
| 144 | 63 | 251 |
| 154 | 64 | 255 |

Source: https://en.wikipedia.org/wiki/Histogram_equalization

Enhancement

Color Histogram Equalization

Simple implementation

Toy Case

1. Compute cumulative distribution function (CDF)

2. Perform min-max normalization
[0, 255] interval

| | | | | | | | |
|----|----|----|-----|-----|-----|----|----|
| 52 | 55 | 61 | 59 | 79 | 61 | 76 | 61 |
| 62 | 59 | 55 | 104 | 94 | 85 | 59 | 71 |
| 63 | 65 | 66 | 113 | 144 | 104 | 63 | 72 |
| 64 | 70 | 70 | 126 | 154 | 109 | 71 | 69 |
| 67 | 73 | 68 | 106 | 122 | 88 | 68 | 68 |
| 68 | 79 | 60 | 70 | 77 | 66 | 58 | 75 |
| 69 | 85 | 64 | 58 | 55 | 61 | 65 | 83 |
| 70 | 87 | 69 | 68 | 65 | 73 | 78 | 90 |

| | | | | | | | |
|-----|-----|-----|-----|-----|-----|-----|-----|
| 0 | 12 | 53 | 32 | 190 | 53 | 174 | 53 |
| 57 | 32 | 12 | 227 | 219 | 202 | 32 | 154 |
| 65 | 85 | 93 | 239 | 251 | 227 | 65 | 158 |
| 73 | 146 | 146 | 247 | 255 | 235 | 154 | 130 |
| 97 | 166 | 117 | 231 | 243 | 210 | 117 | 117 |
| 117 | 190 | 36 | 146 | 178 | 93 | 20 | 170 |
| 130 | 202 | 73 | 20 | 12 | 53 | 85 | 194 |
| 146 | 206 | 130 | 117 | 85 | 166 | 182 | 215 |

Source: https://en.wikipedia.org/wiki/Histogram_equalization

Enhancement

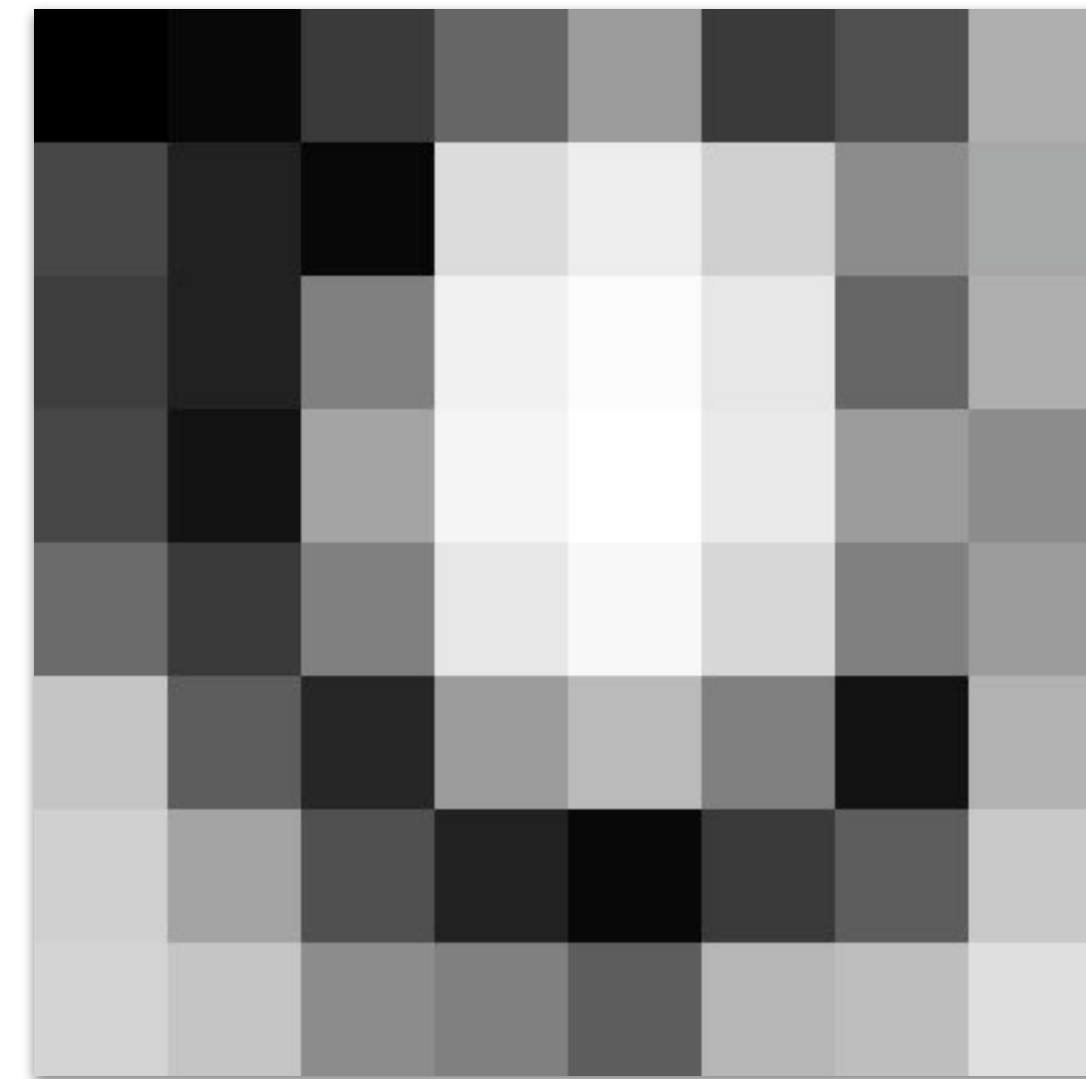
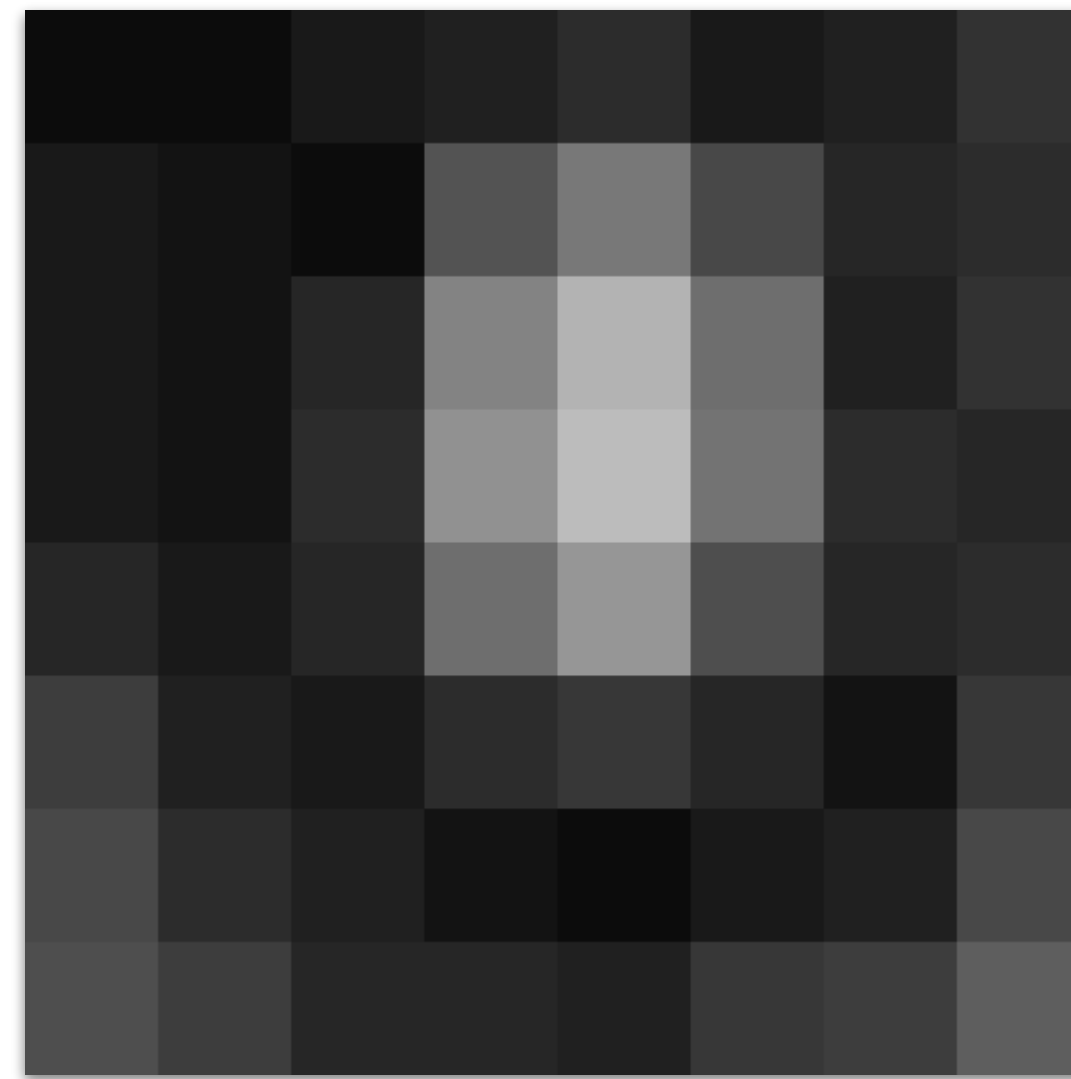
Color Histogram Equalization

Simple implementation

Toy Case

1. Compute cumulative distribution function (CDF)

2. Perform min-max normalization
[0, 255] interval



Source: https://en.wikipedia.org/wiki/Histogram_equalization

Enhancement

Image Contrast

Example:
Color histogram equalization.

Example: too bright capture.



before



after

Enhancement

Image Contrast

Example:
Color histogram equalization.

Example: too dark capture.



before



after

Enhancement

Image Processing Solutions

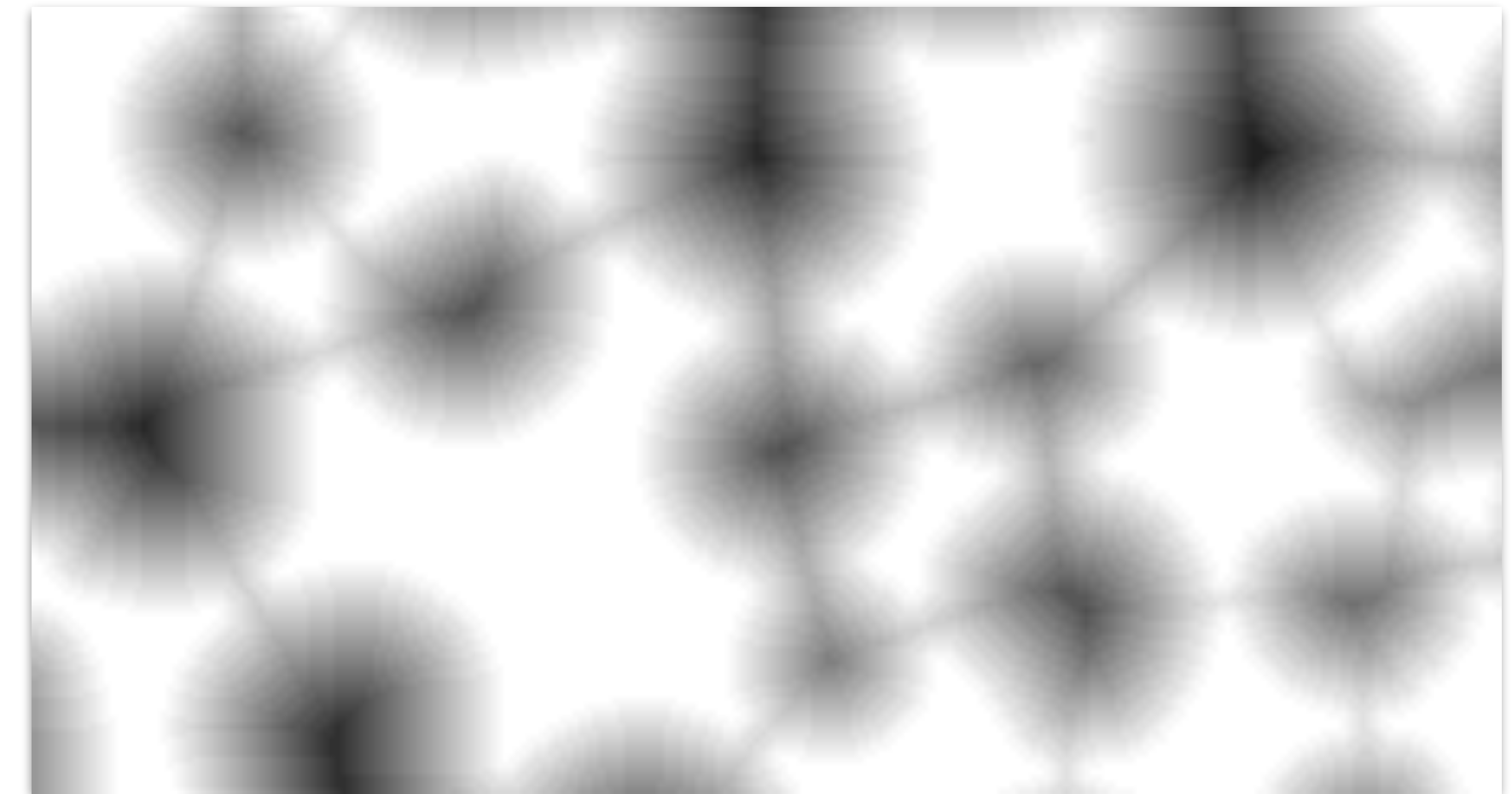
Tasks

Enhancement of image contrast.

Enhancement of ridges and valleys.

Content segmentation.

Others.



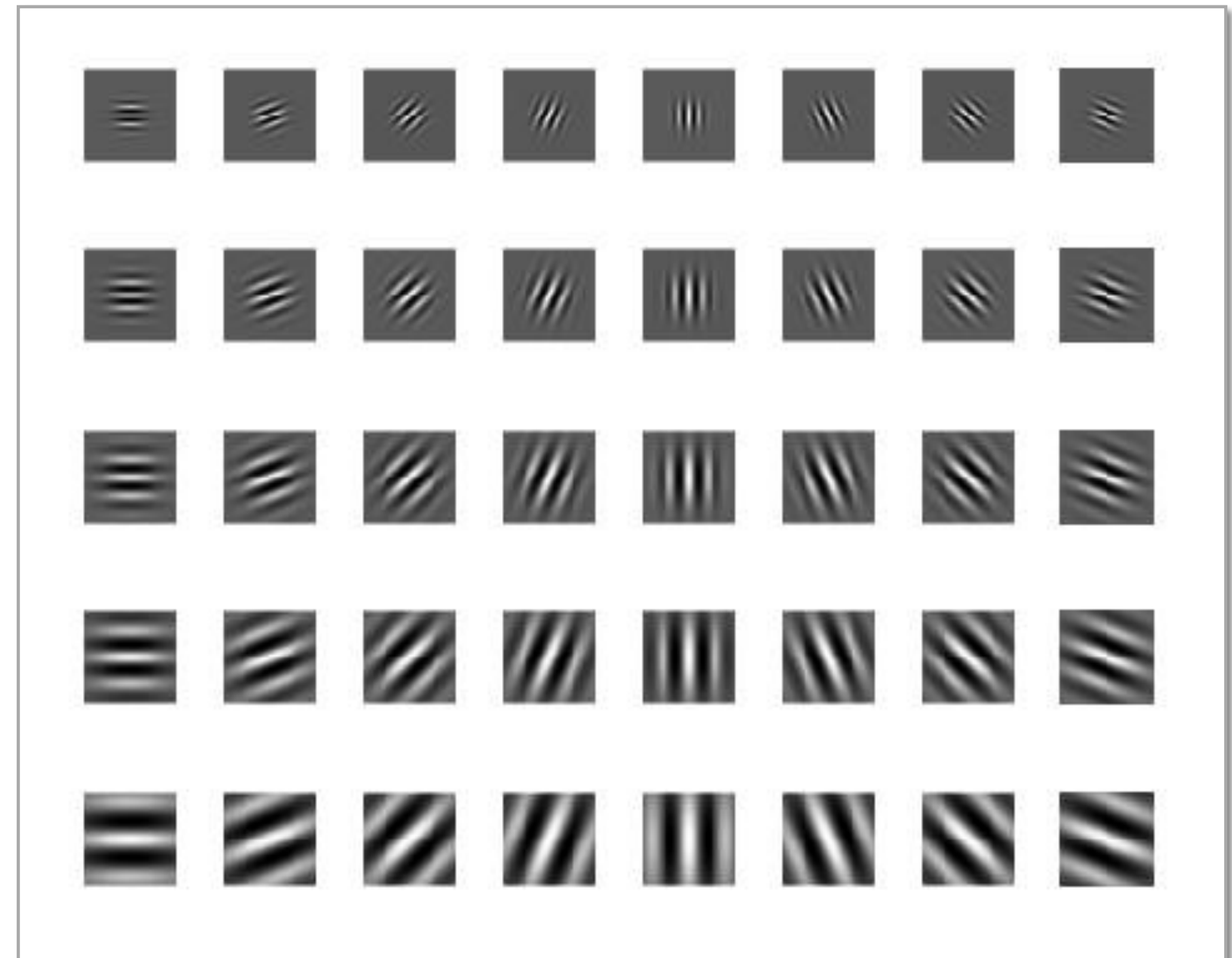
Enhancement

Ridges and Valleys

Example:

Image filtering with
Gabor filters.

Ridges and valleys may become
more prominent when a fingerprint
image is filtered by Gabor filters.



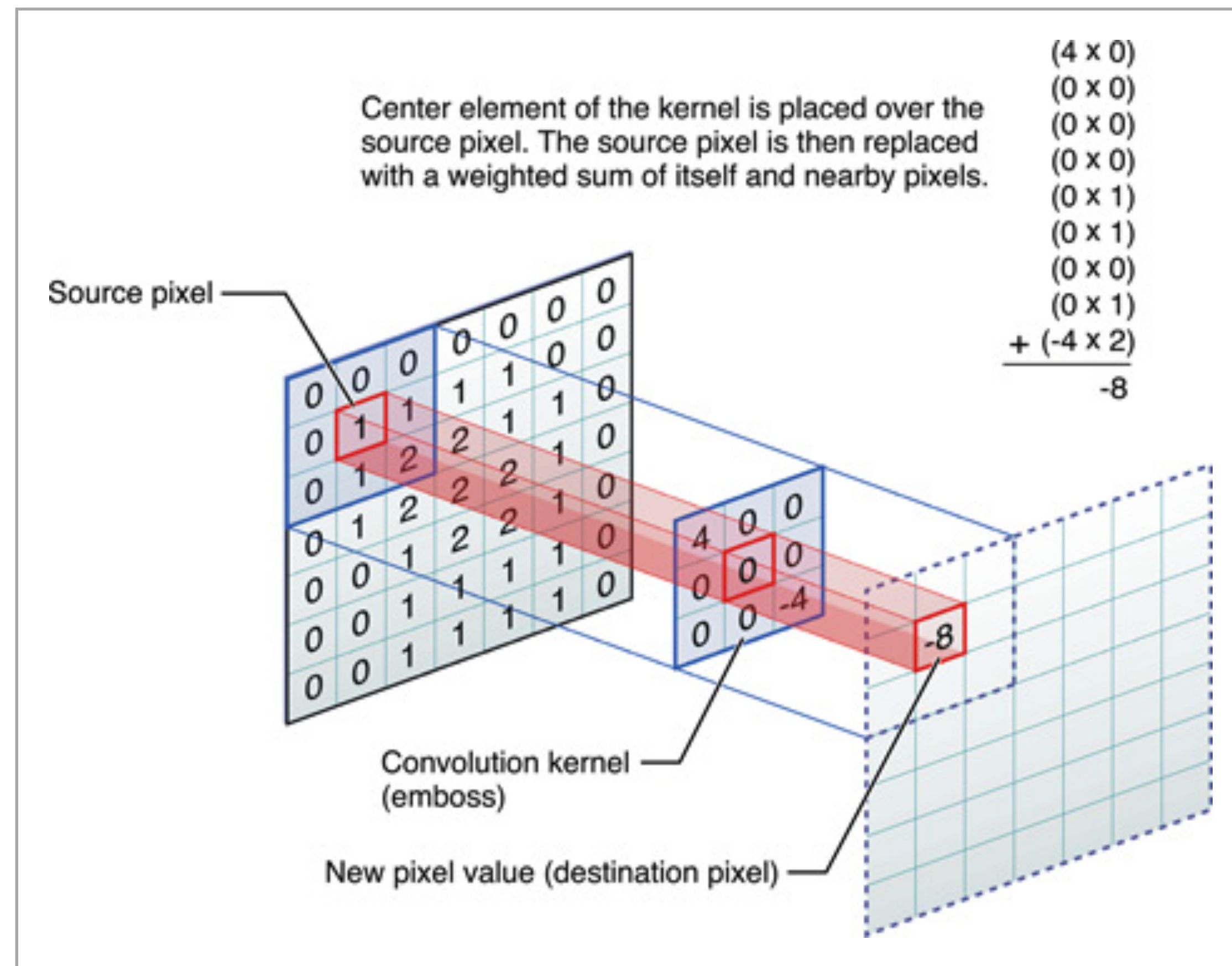
Enhancement

Ridges and Valleys

Example:

Image filtering with
Gabor filters.

Gabor filters may be
applied to an image
through convolutions.



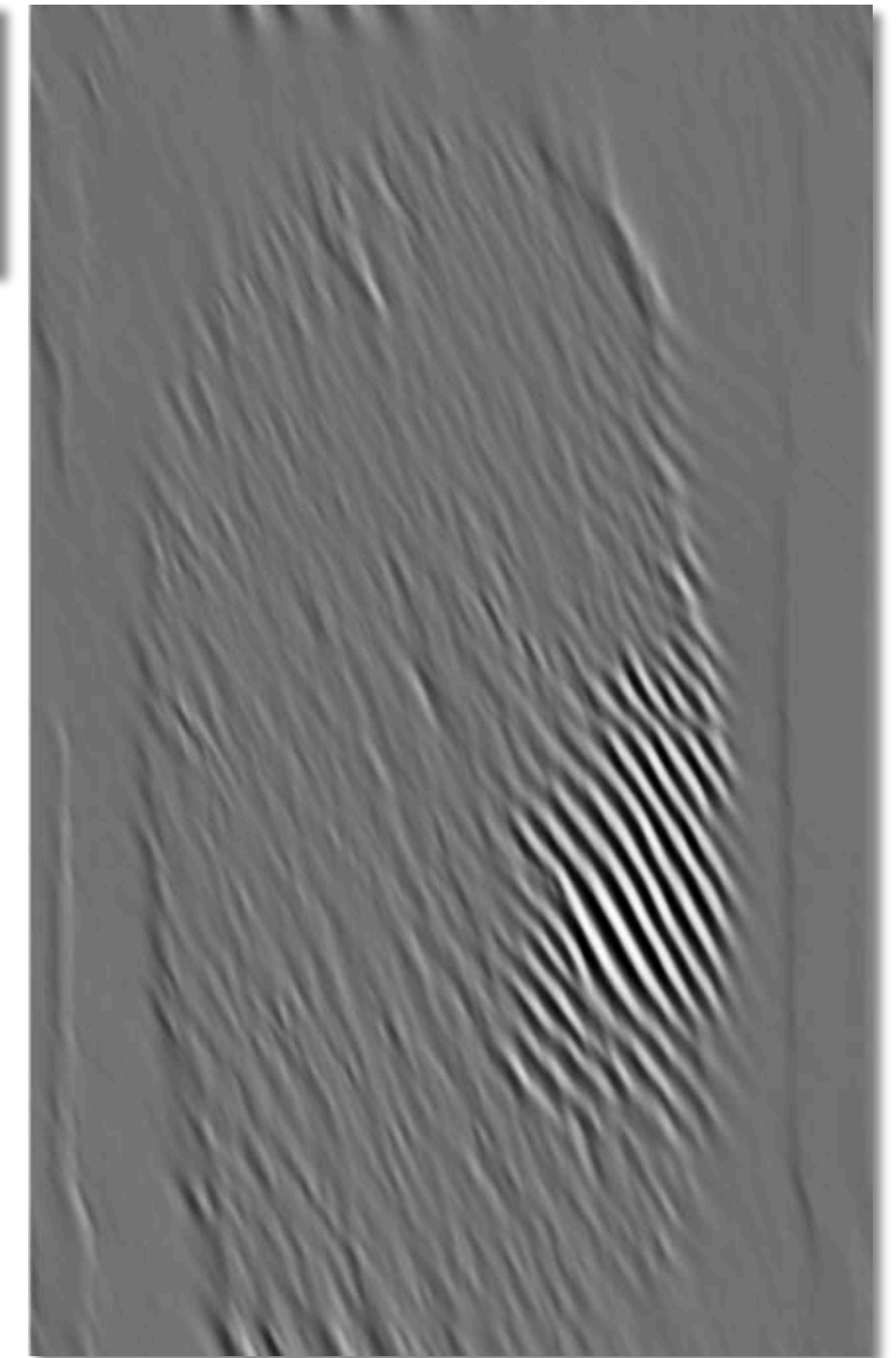
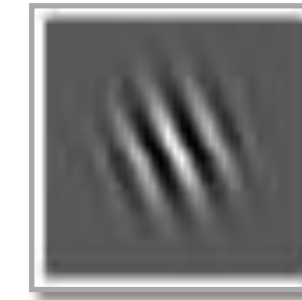
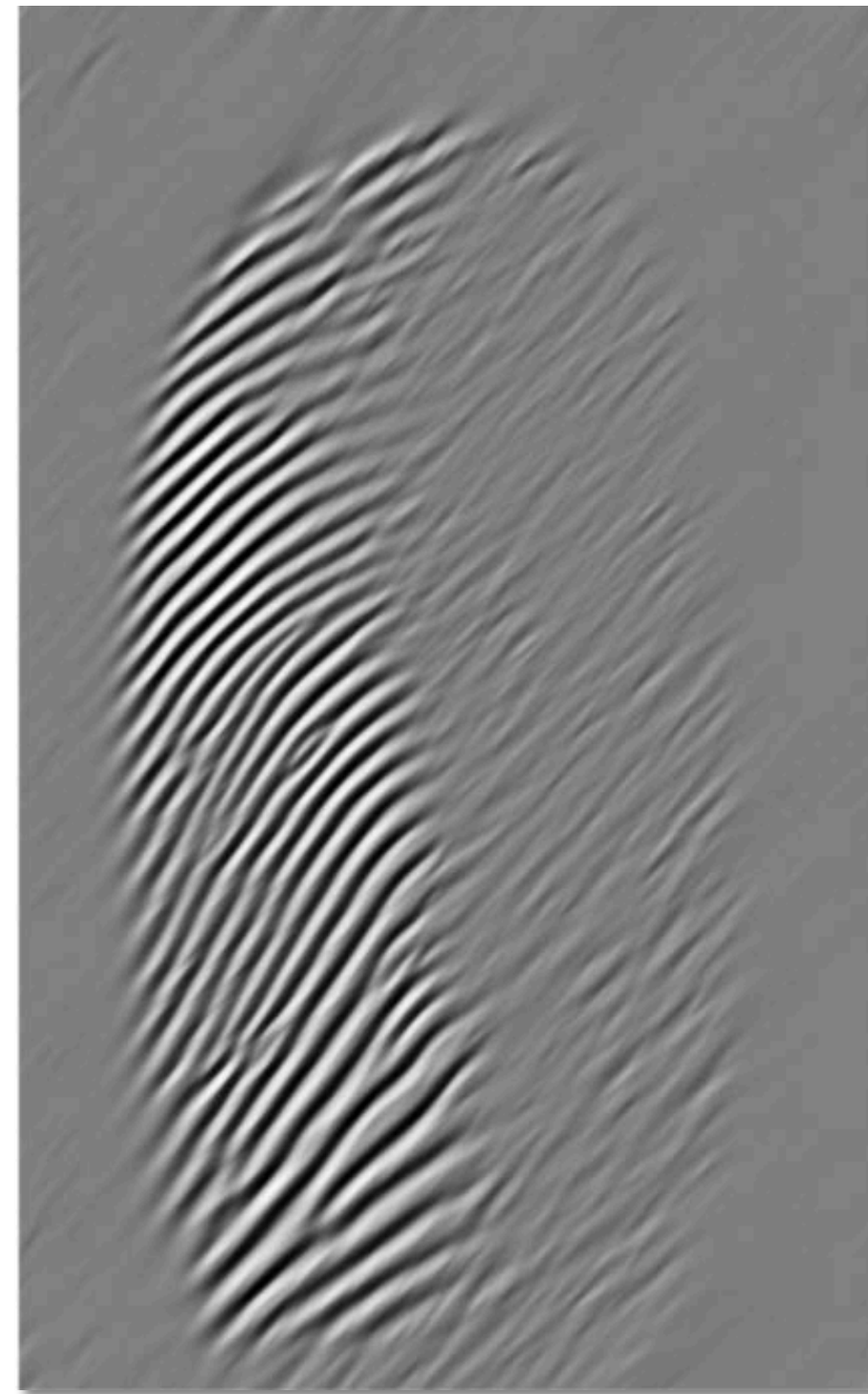
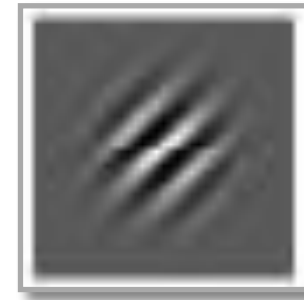
Source: <https://developer.apple.com/library/archive/documentation/Performance/Conceptual/vimage/ConvolutionOperations/ConvolutionOperations.html>

Enhancement

Ridges and Valleys

Example:

Image filtering with
Gabor filters.



Enhancement

Ridges and Valleys

Example:
Image filtering with
Gabor filters.

Maltoni et al.
Handbook of Fingerprint Recognition
Springer Books, 2009



before



after

Enhancement

Image Processing Solutions

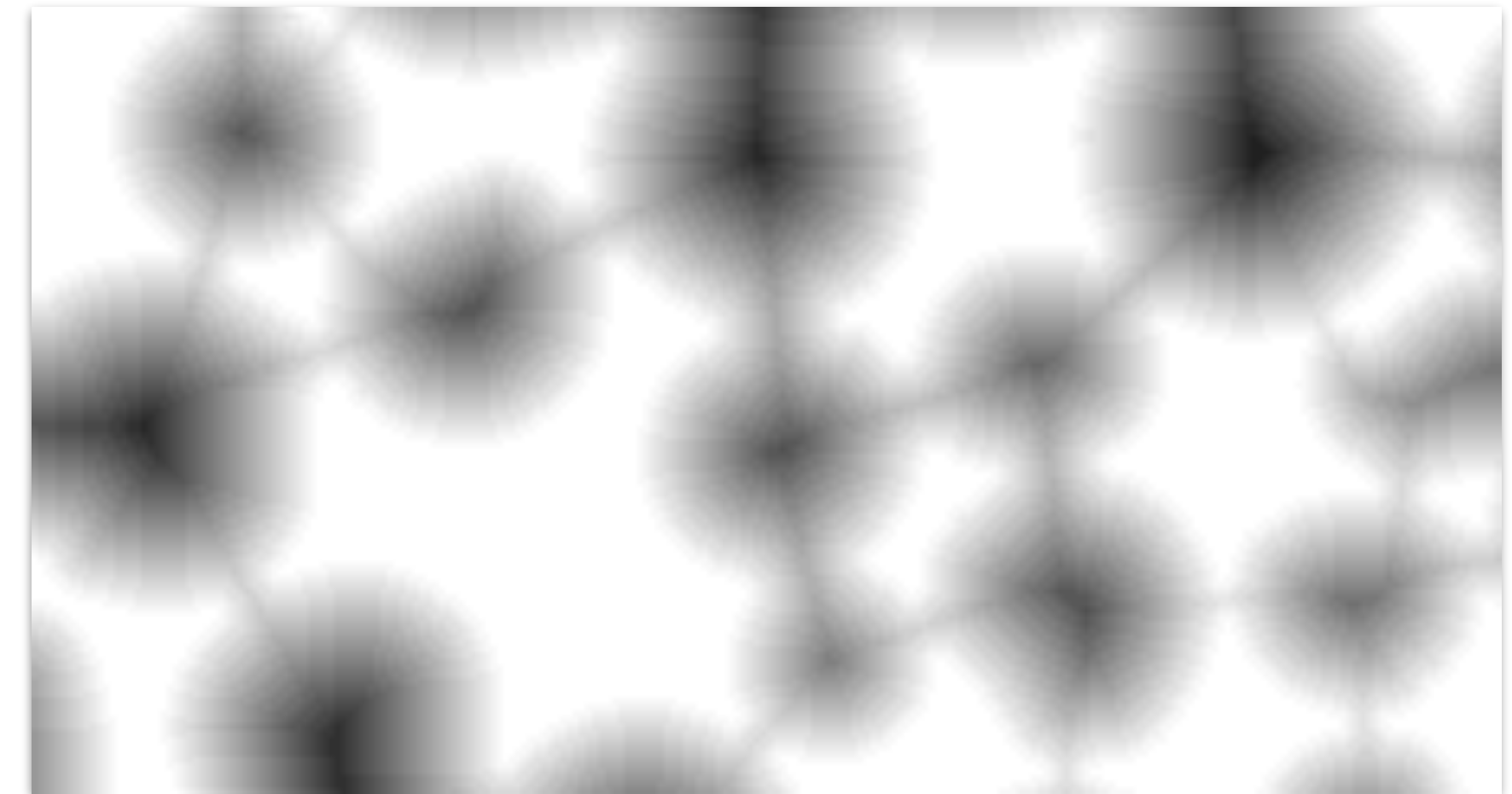
Tasks

Enhancement of image contrast.

Enhancement of ridges and valleys.

Content segmentation.

Others.



Enhancement

Segmentation

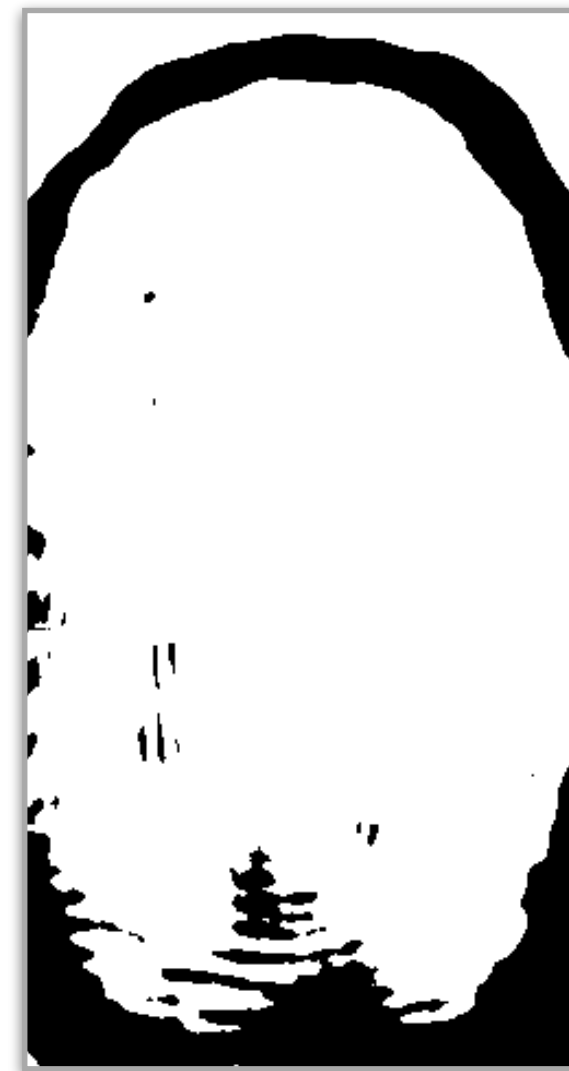
Example: blurring, thresholding, and morphological operations.



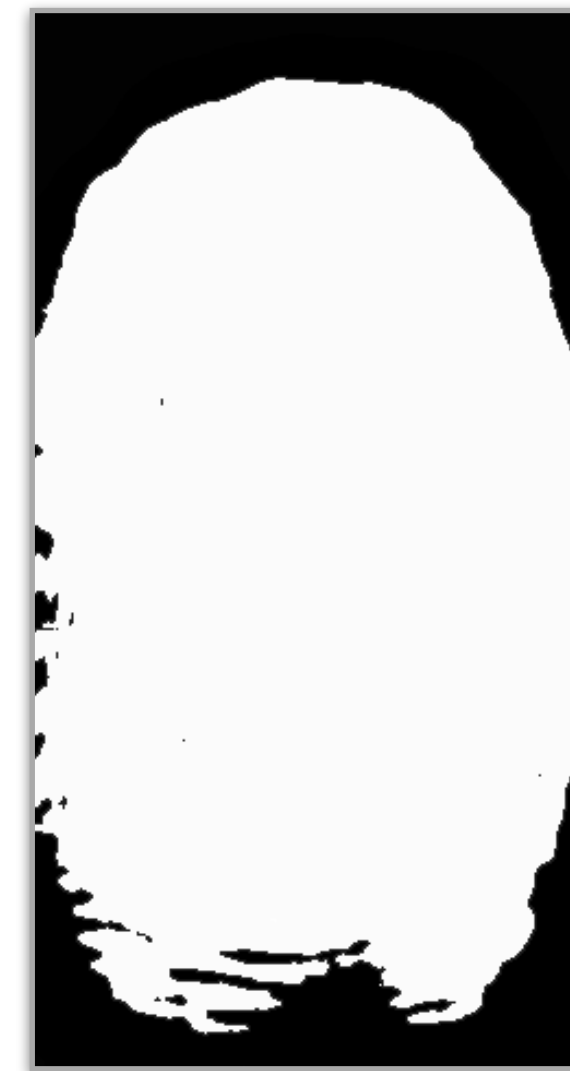
before



blur



threshold



open



after

Enhancement

Image Processing Solutions

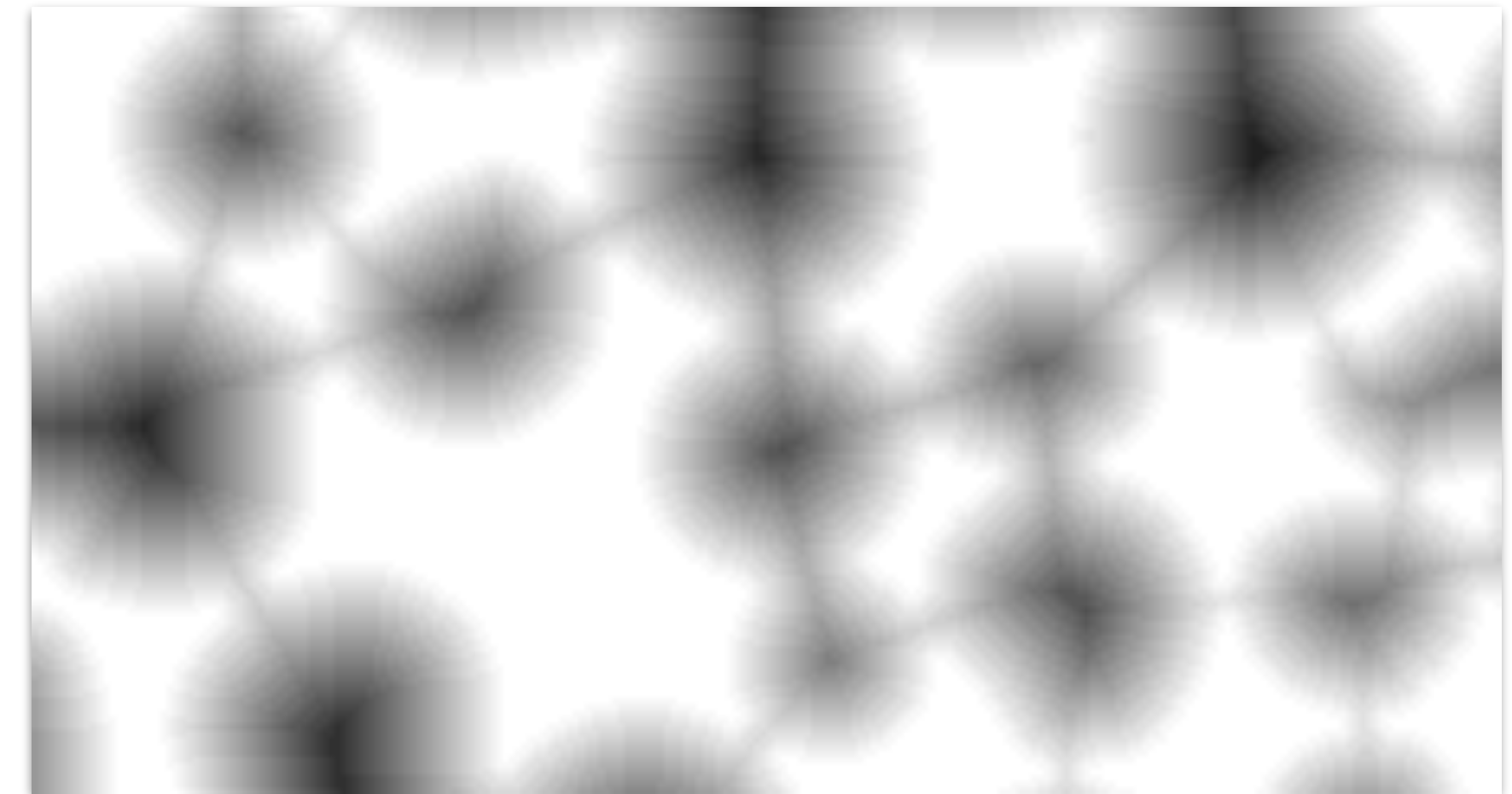
Tasks

Enhancement of image contrast.

Enhancement of ridges and valleys.

Content segmentation.

Others.



Enhancement

Image Processing Solutions

Be Aware

Besides the aforementioned techniques, there are much more sophisticated and effective ones.

We'll see some of them in practice and with more details during our next coding class.



Enhancement

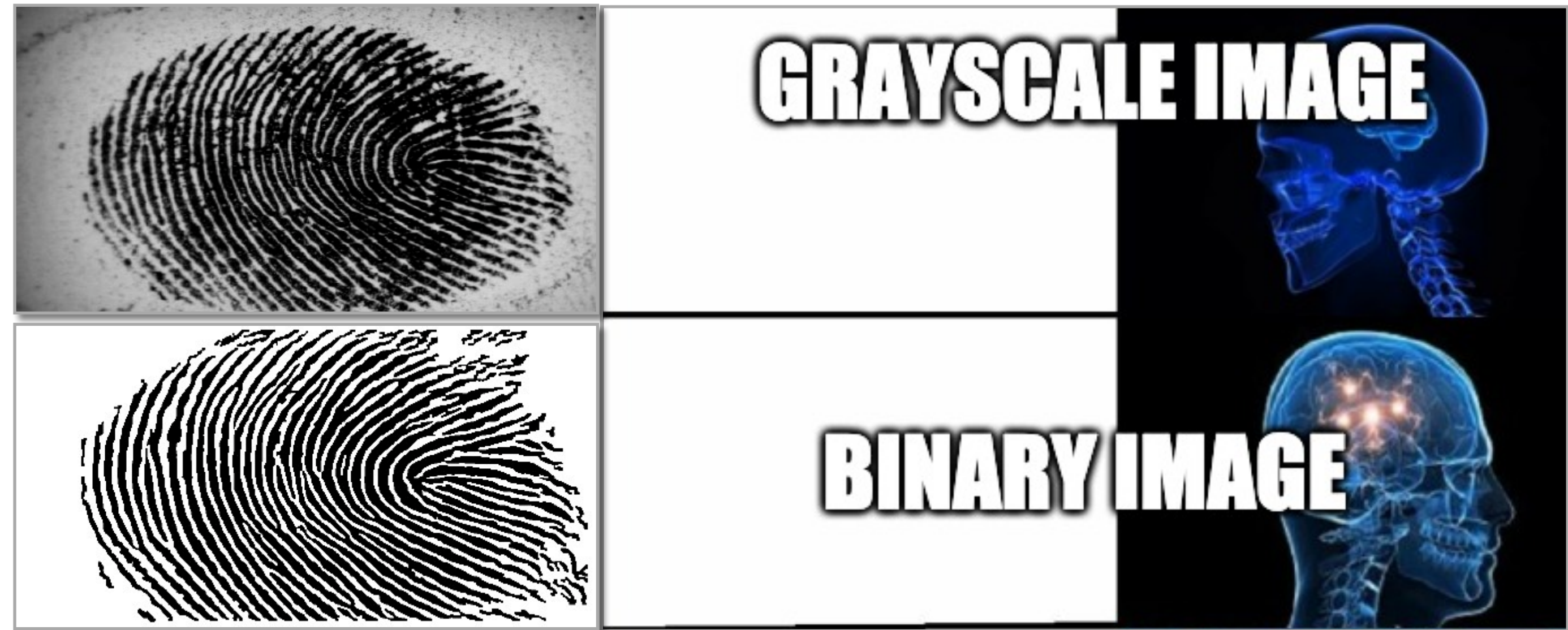
Other Strategies
Start from...



Enhancement

Other Strategies

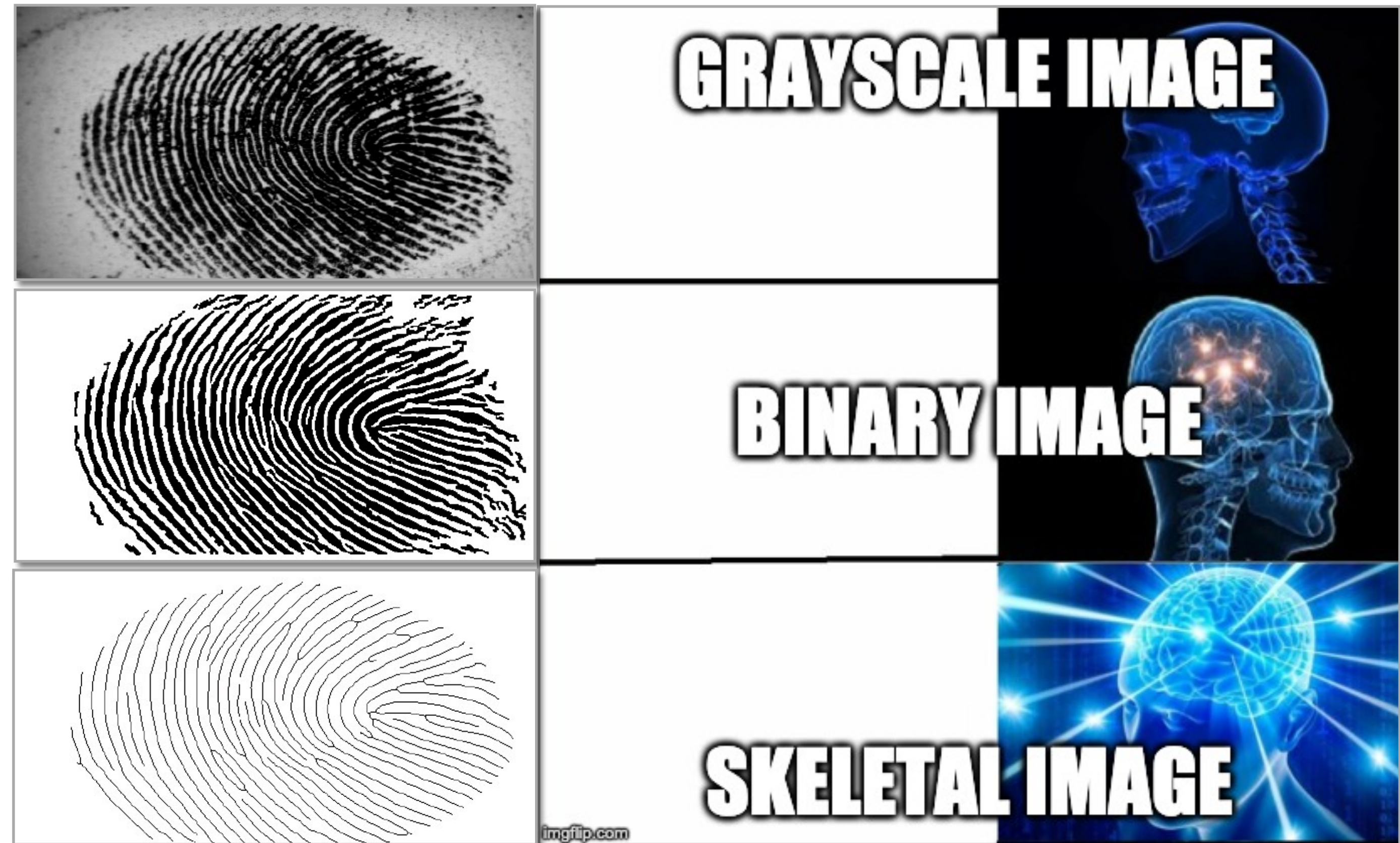
Start from...



Enhancement

Other Strategies

Start from...



Source: Dr. Adam Czajka

Enhancement

Other Strategies

Start from...

Each strategy has its own set of pros and cons, and will lead to different performance.



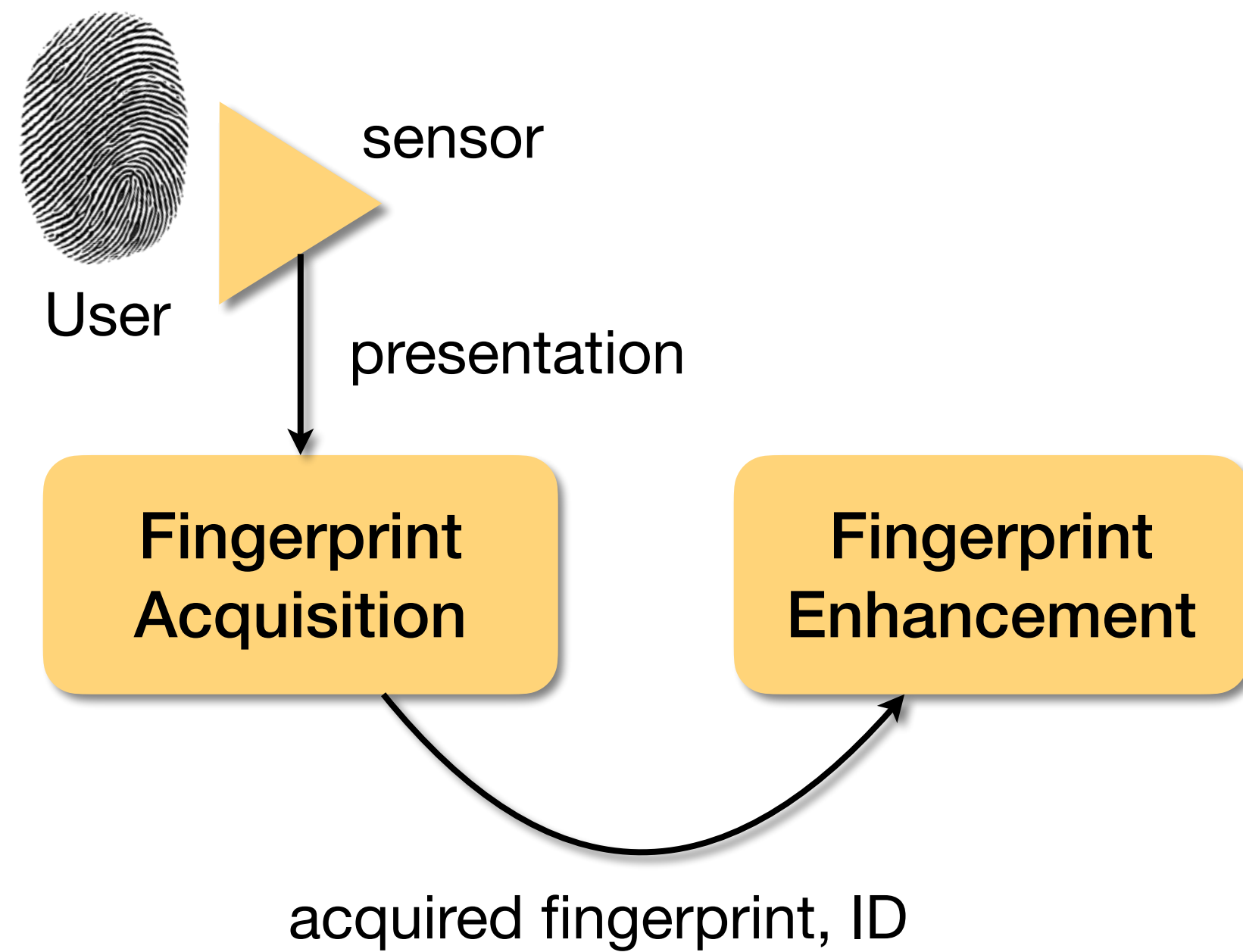
Source: Dr. Adam Czajka

Enhancement

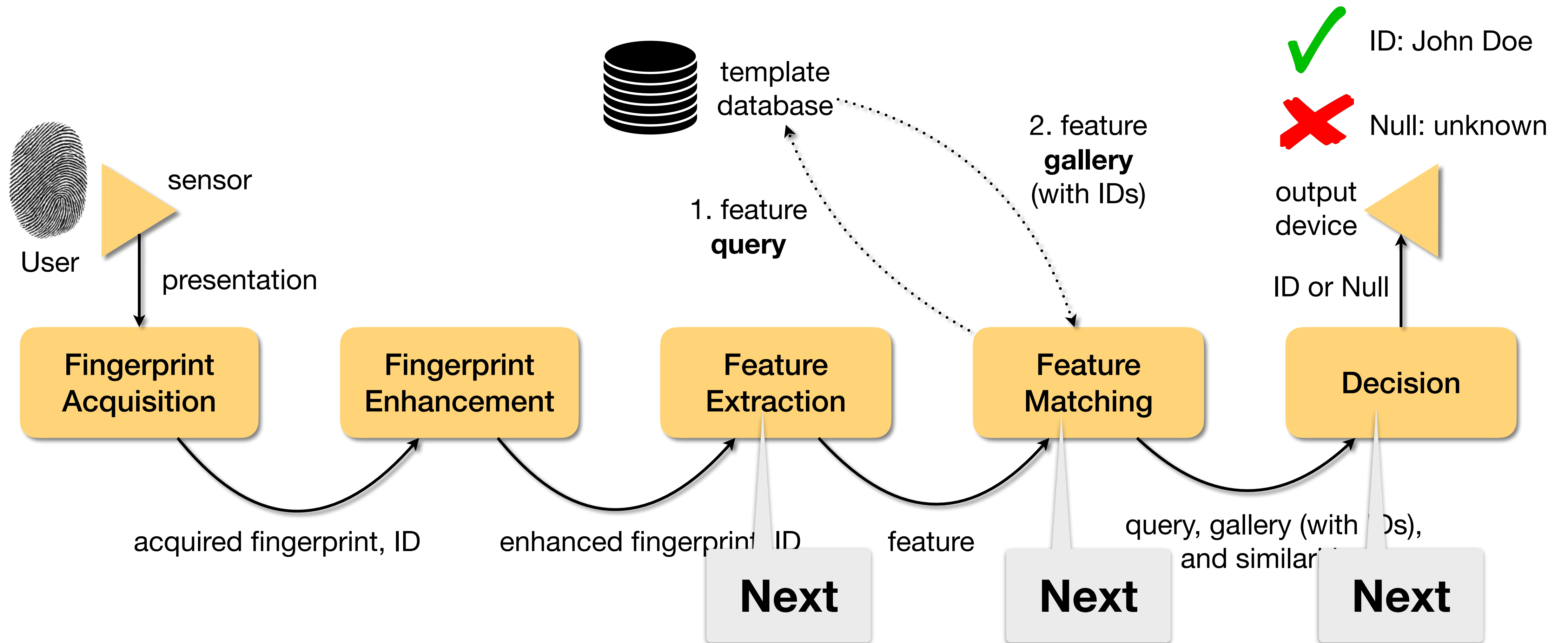


Source: Dr. Adam Czajka

Fingerprint Recognition



Fingerprint Recognition



What's Next?

Even more about fingerprints

Fingerprint feature extraction methods.
Fingerprint matching methods.

Fill out your *Today-I-missed* Statement

Please visit sakai.luc.edu/x/BCJs8K.

