

# Fingerprint Recognition II

COMP 388-002/488-002 Biometrics

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Fall 2024



**LOYOLA**  
UNIVERSITY CHICAGO

# Today we will...

*Get to know*

Fingerprint acquisition and enhancement.

# Today's Attendance

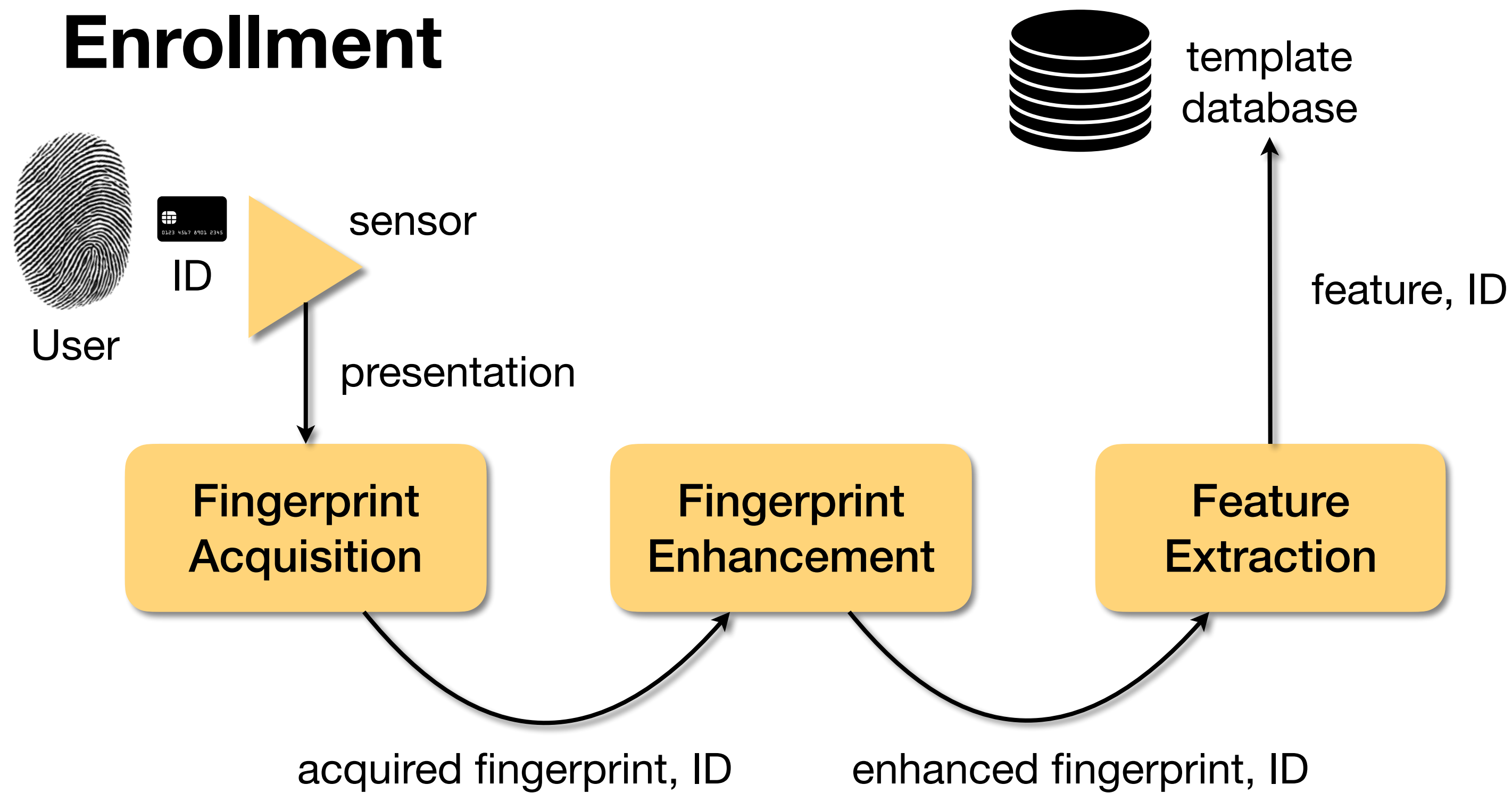
**Please fill out the form**

<https://forms.gle/JPq8kvJZACytZQjn8>



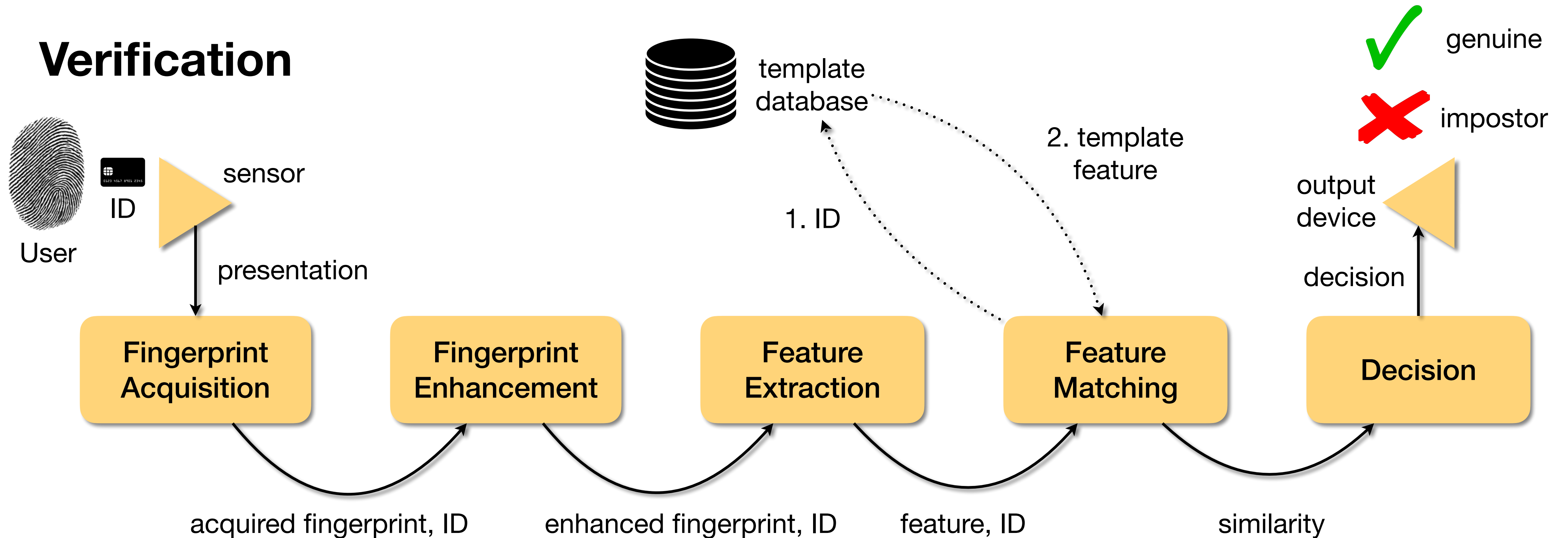
# Fingerprint Recognition

## Enrollment



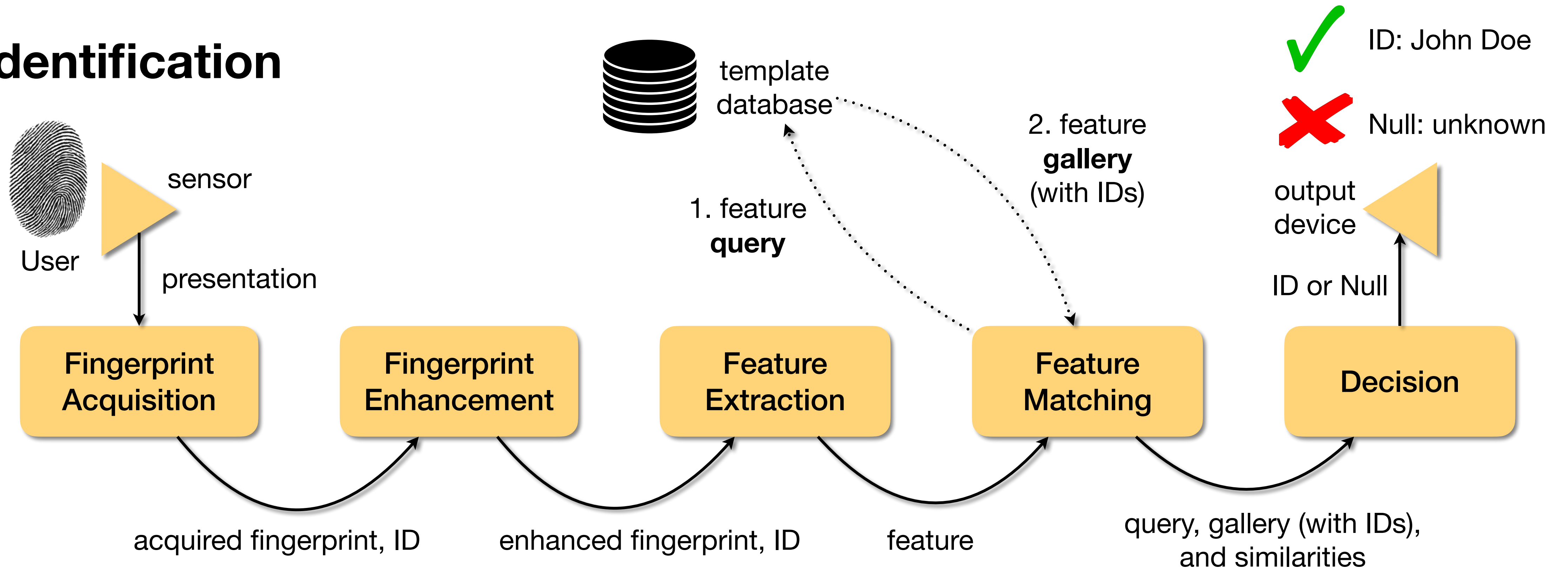
# Fingerprint Recognition

## Verification

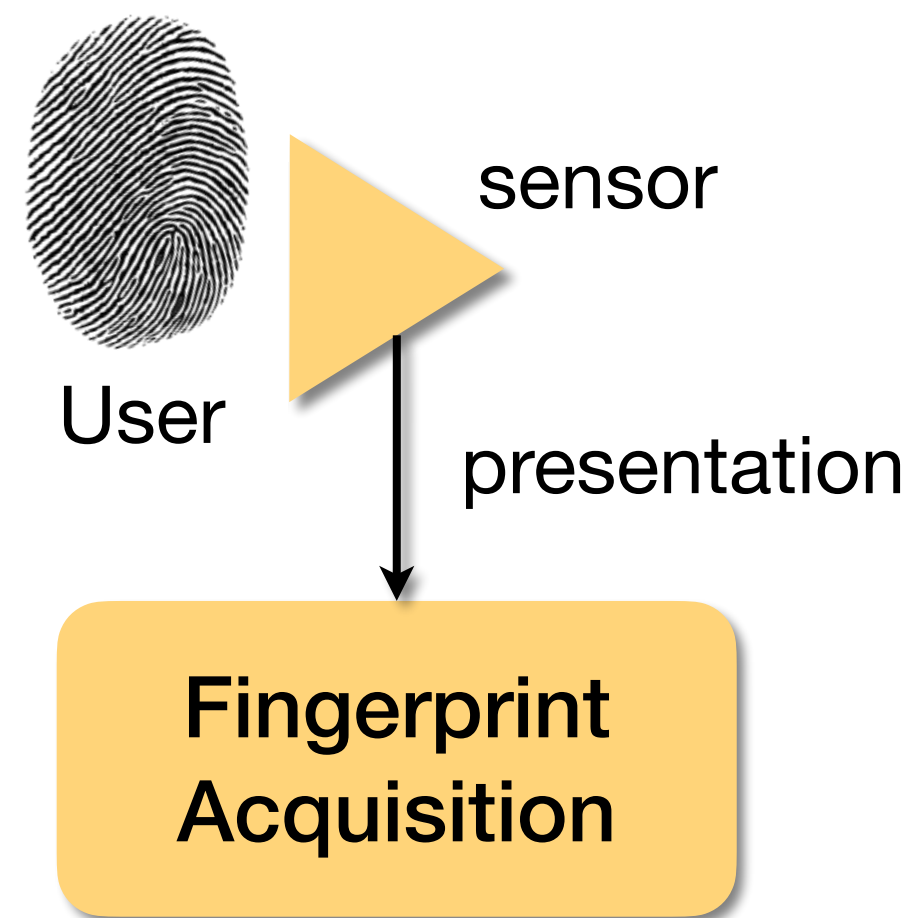


# Fingerprint Recognition

## 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

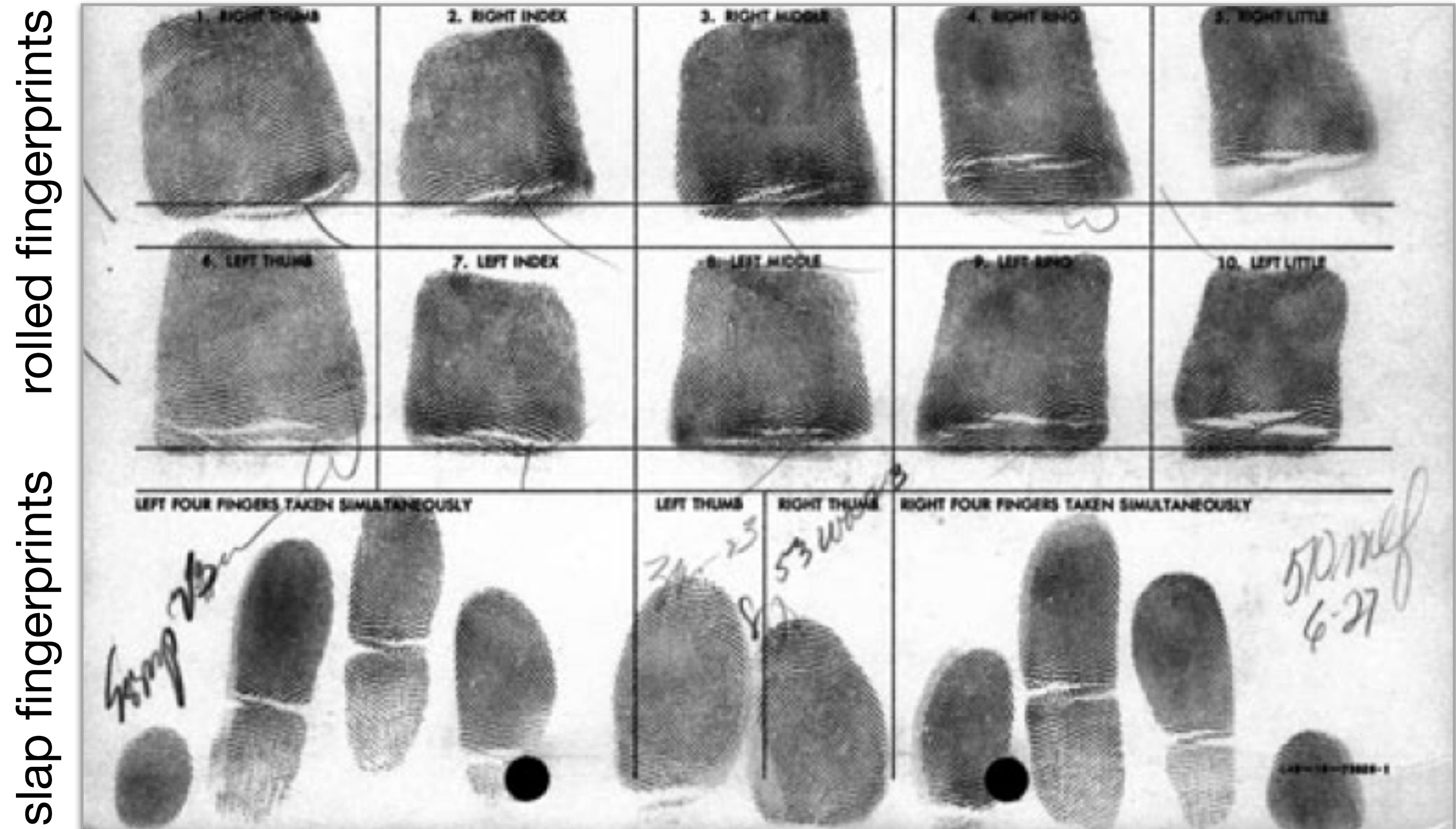


latent fingerprint

# 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.



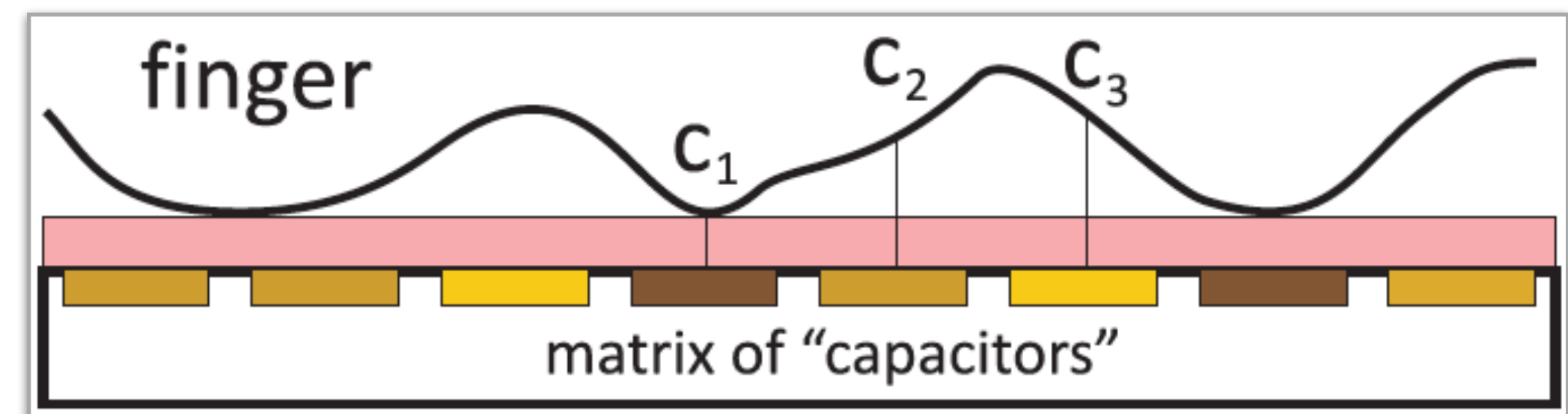
Source: Dr. Adam Czajka

# 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

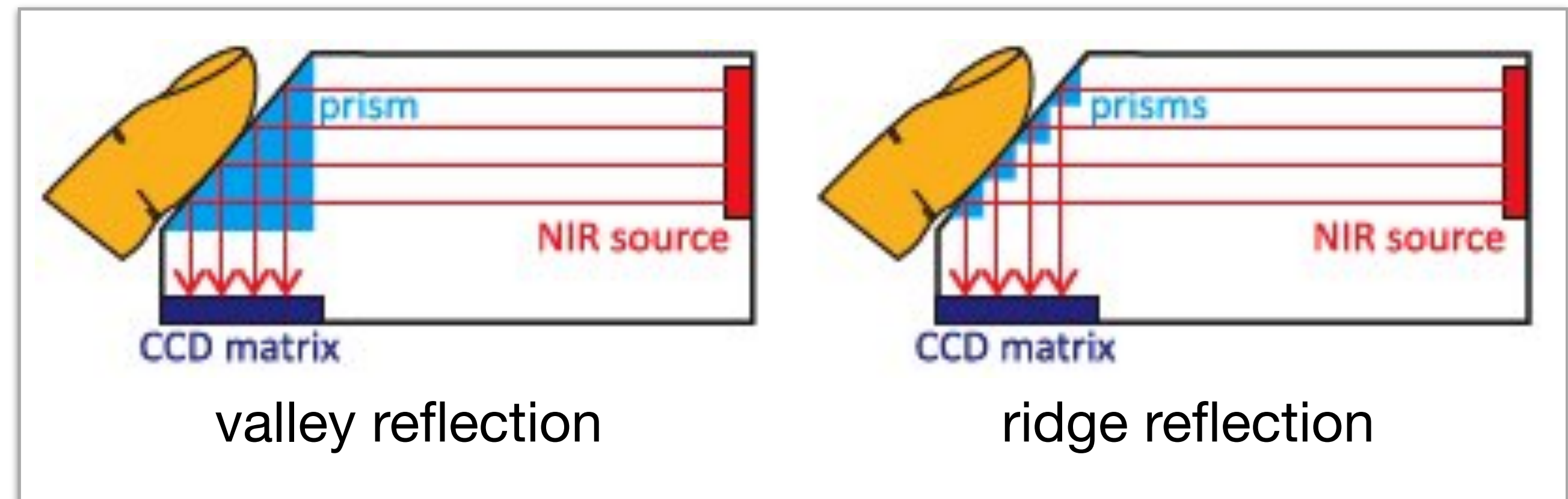
## On-line Acquisition

### 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)

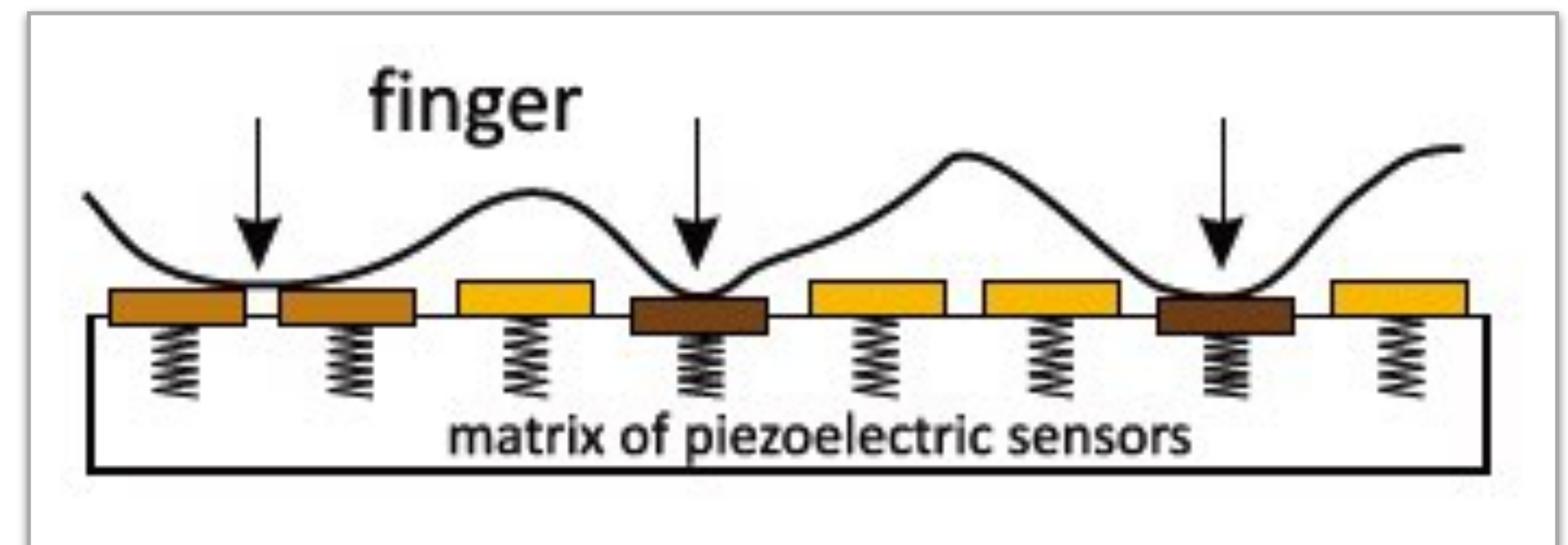
Also known as piezoelectric.

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



# 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

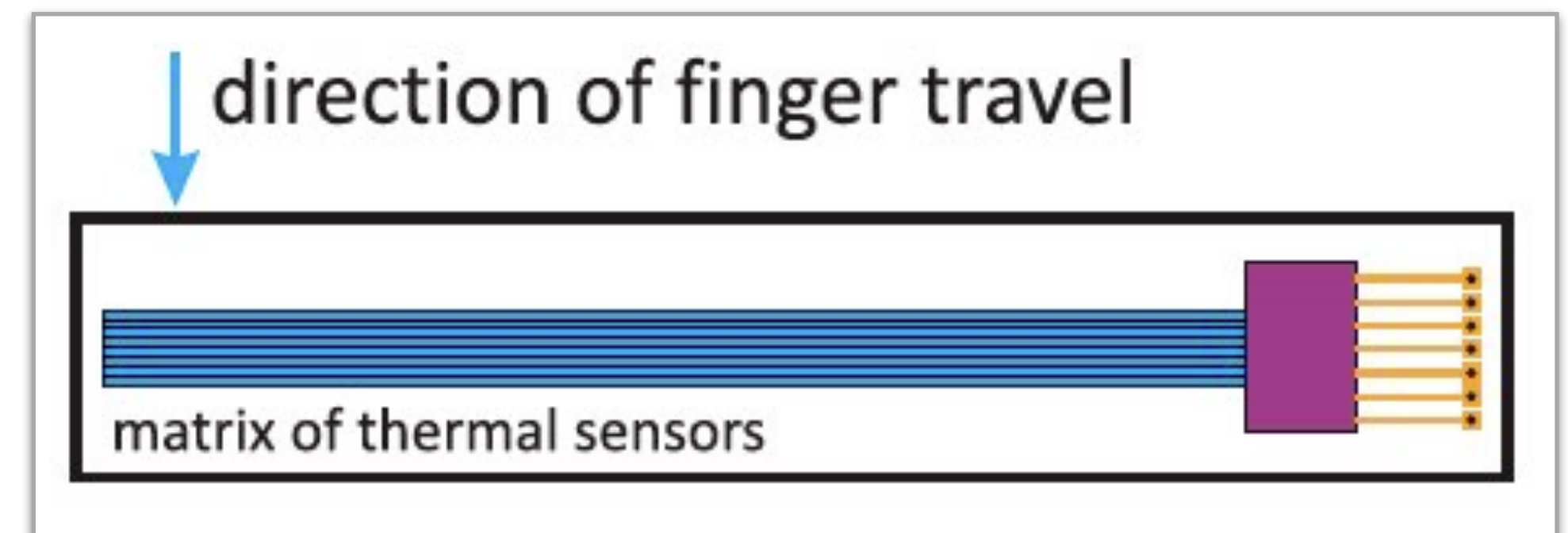
### Thermal sensors (4/6)

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.



finger sweep

discrete collection

fingerprint reconstruction

Source: Atmel FingerChip

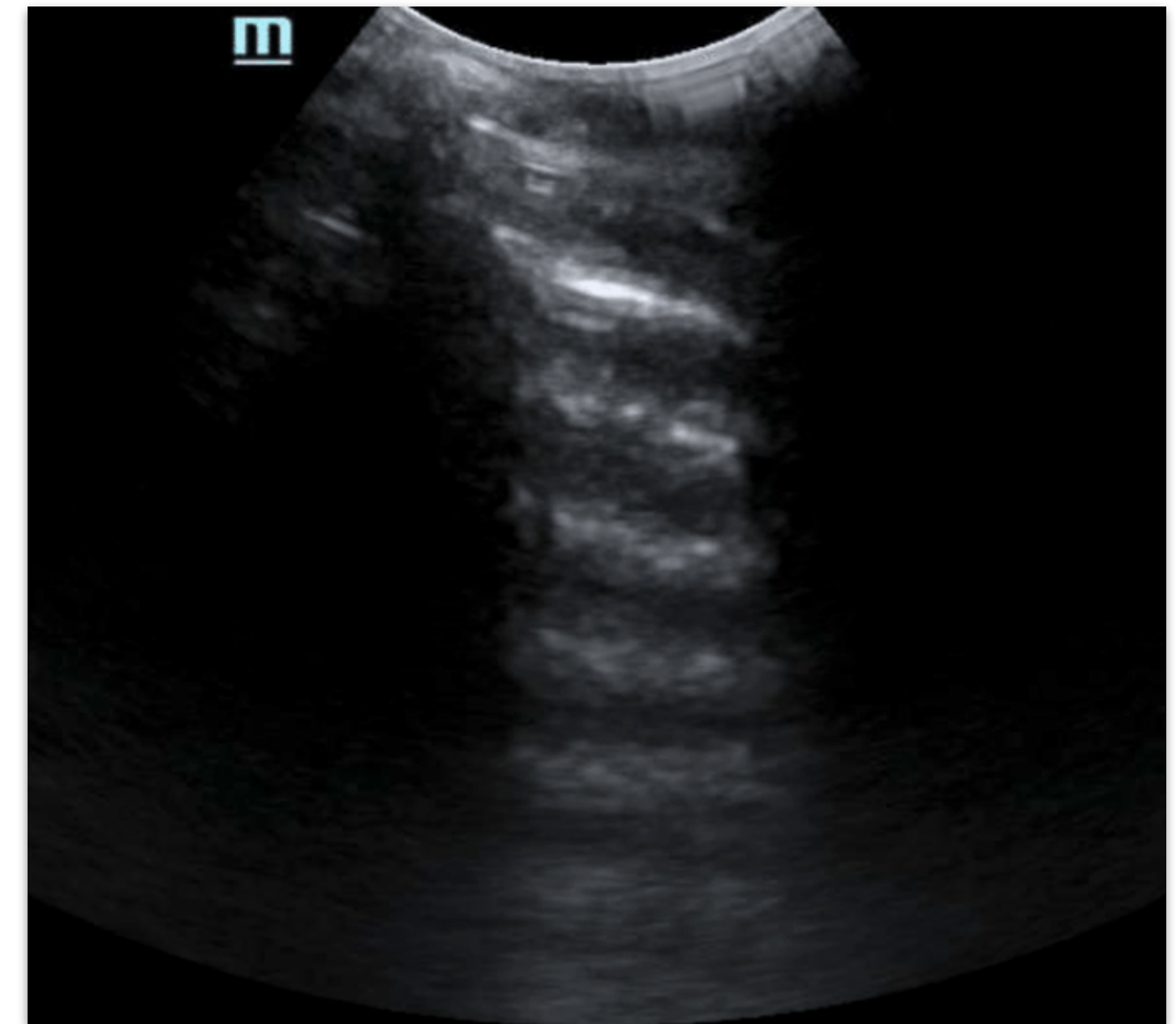
# Acquisition

## 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

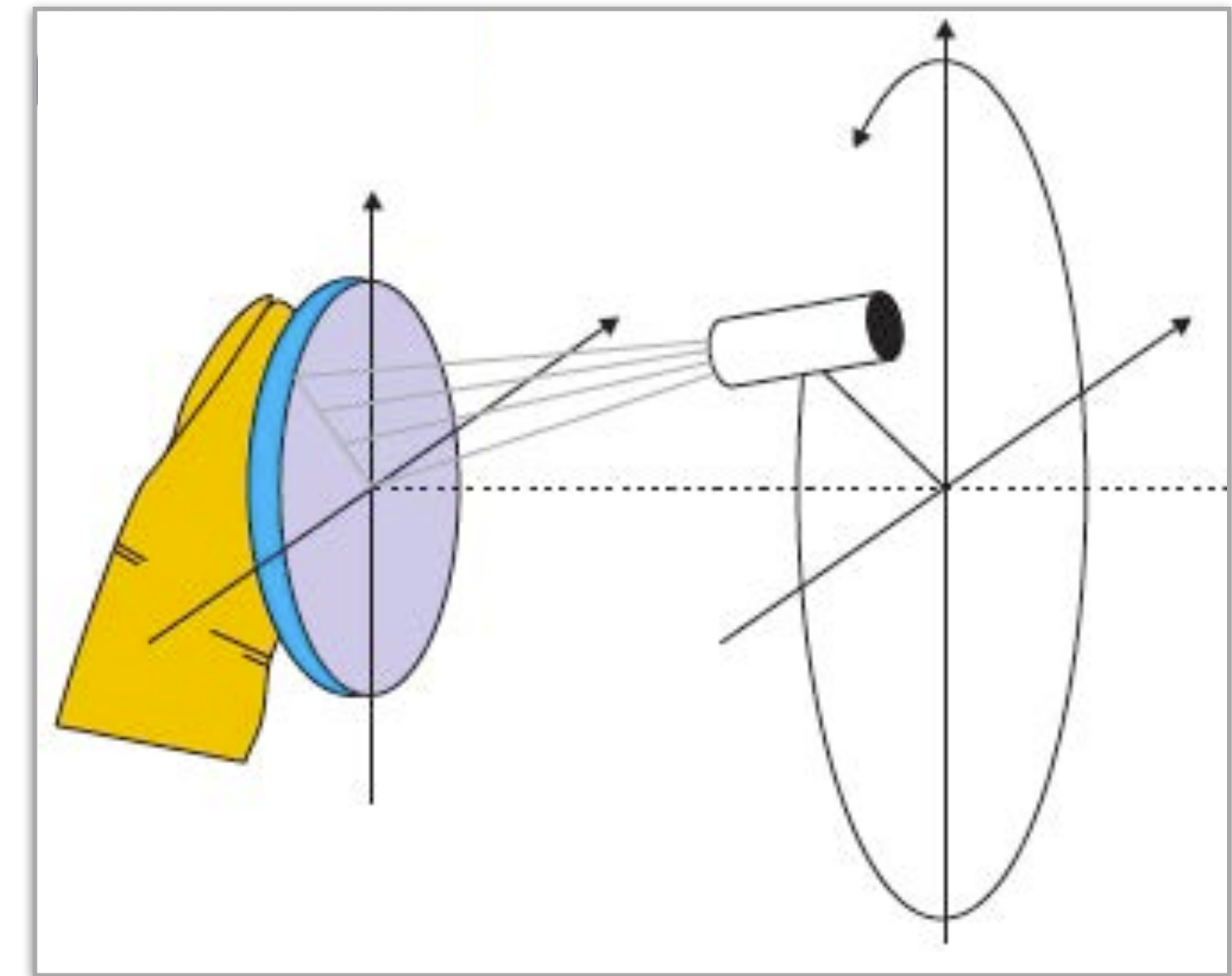
## On-line Acquisition

### Ultrasound sensors (5/6)

Example: Optel

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

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



Source: Dr. Adam Czajka

# Acquisition

## On-line Acquisition

### Ultrasound sensors (5/6)

Example: Optel  
Device and sample.



Source: [www.optel.com.pl](http://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)

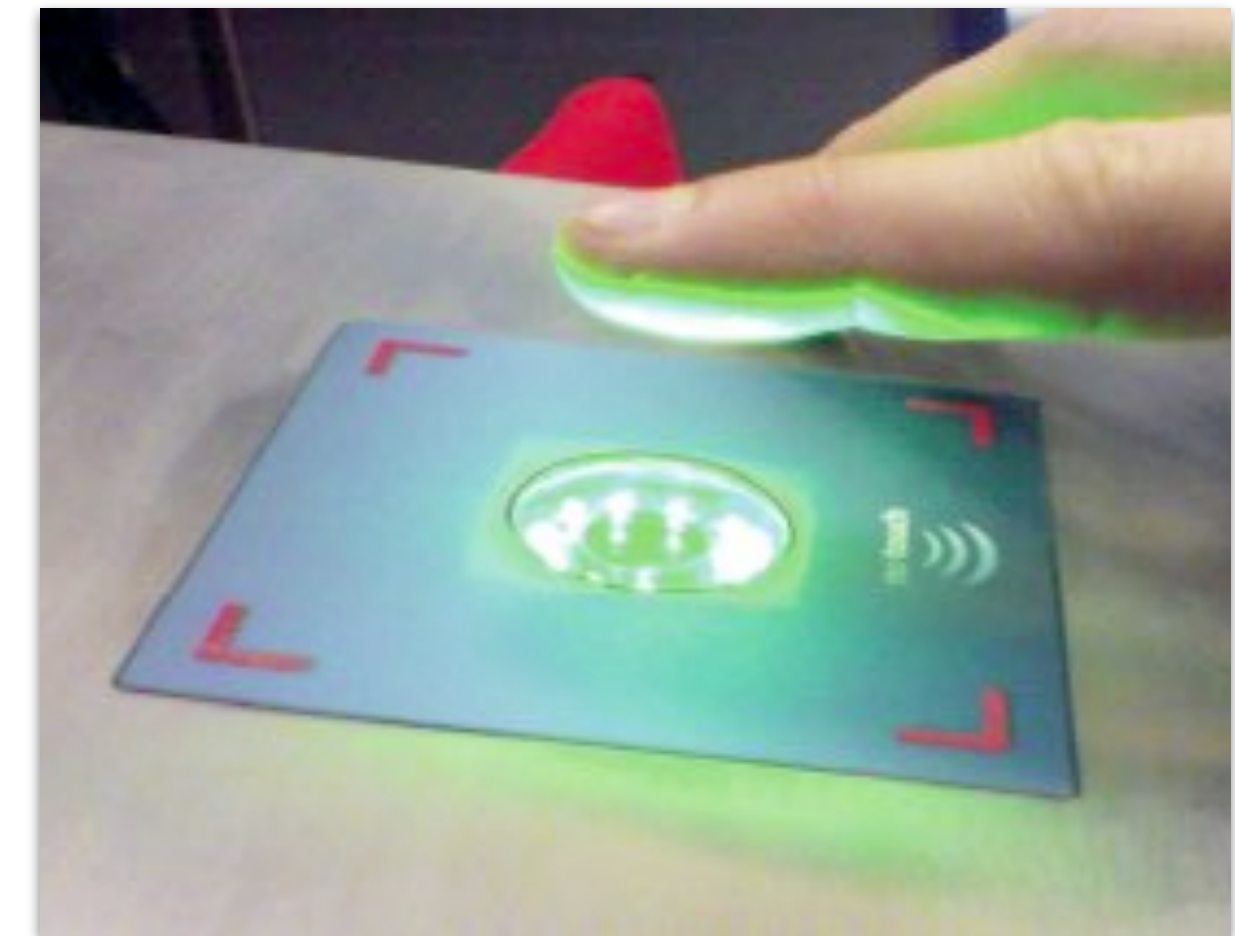
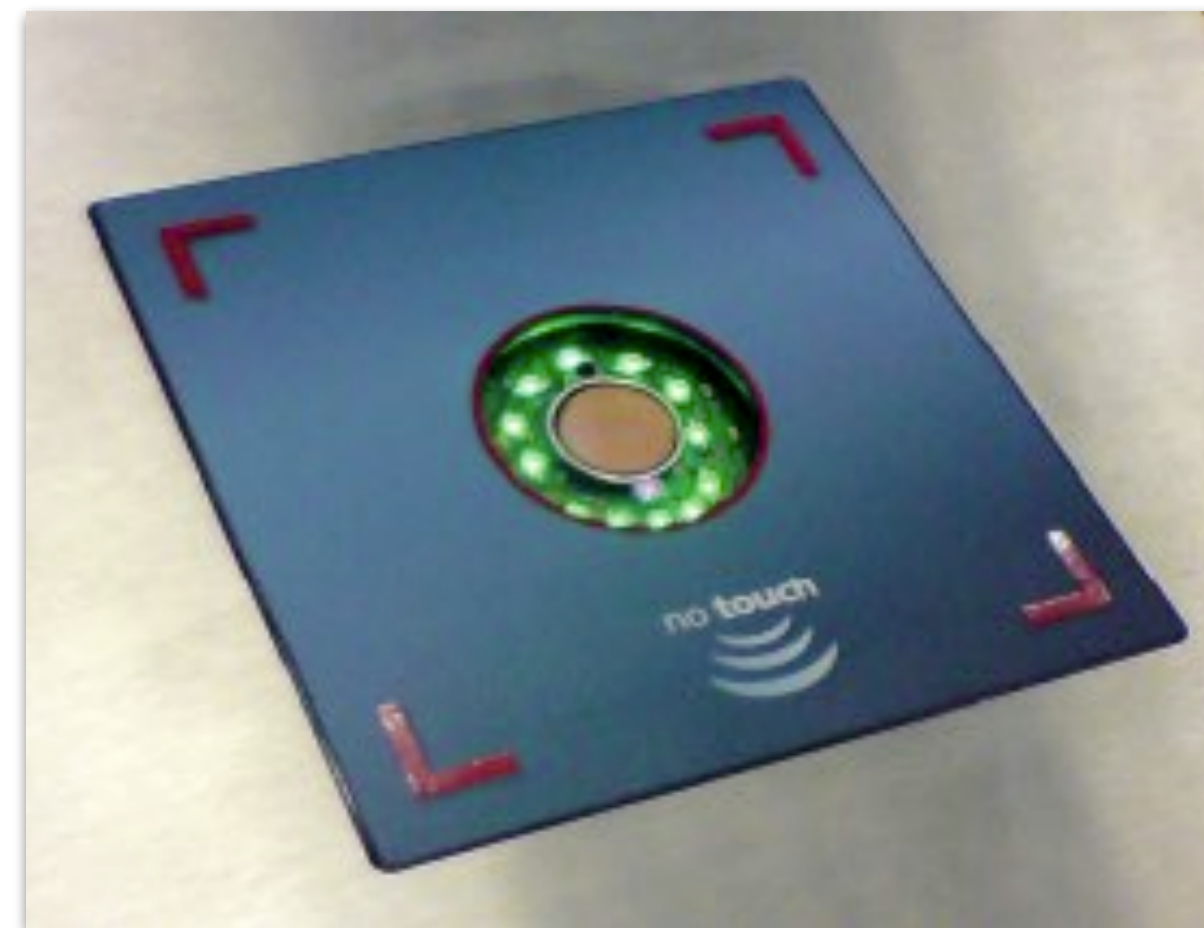
3D imaging with CCD sensor.



# 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

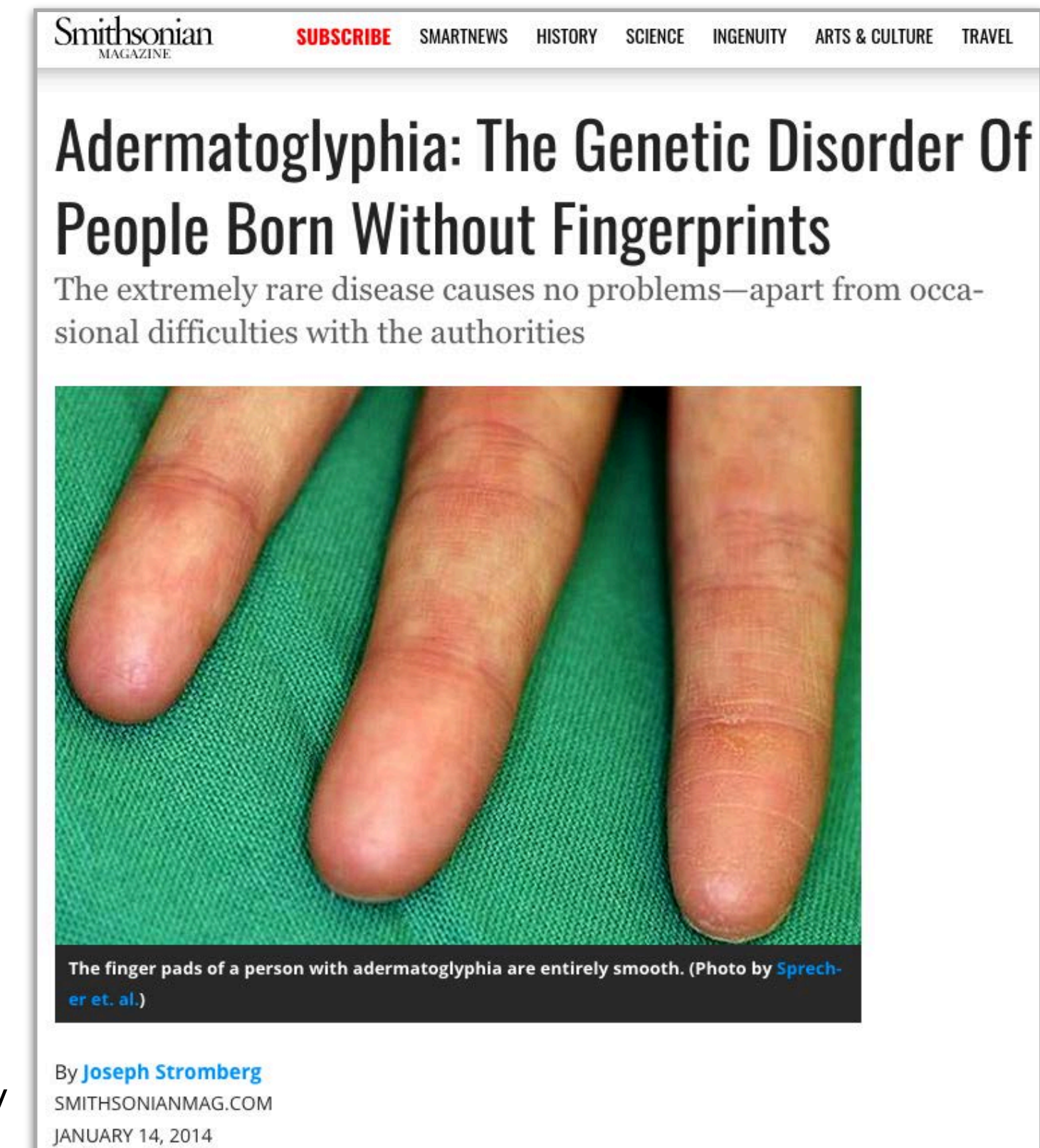
# Acquisition

## 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?

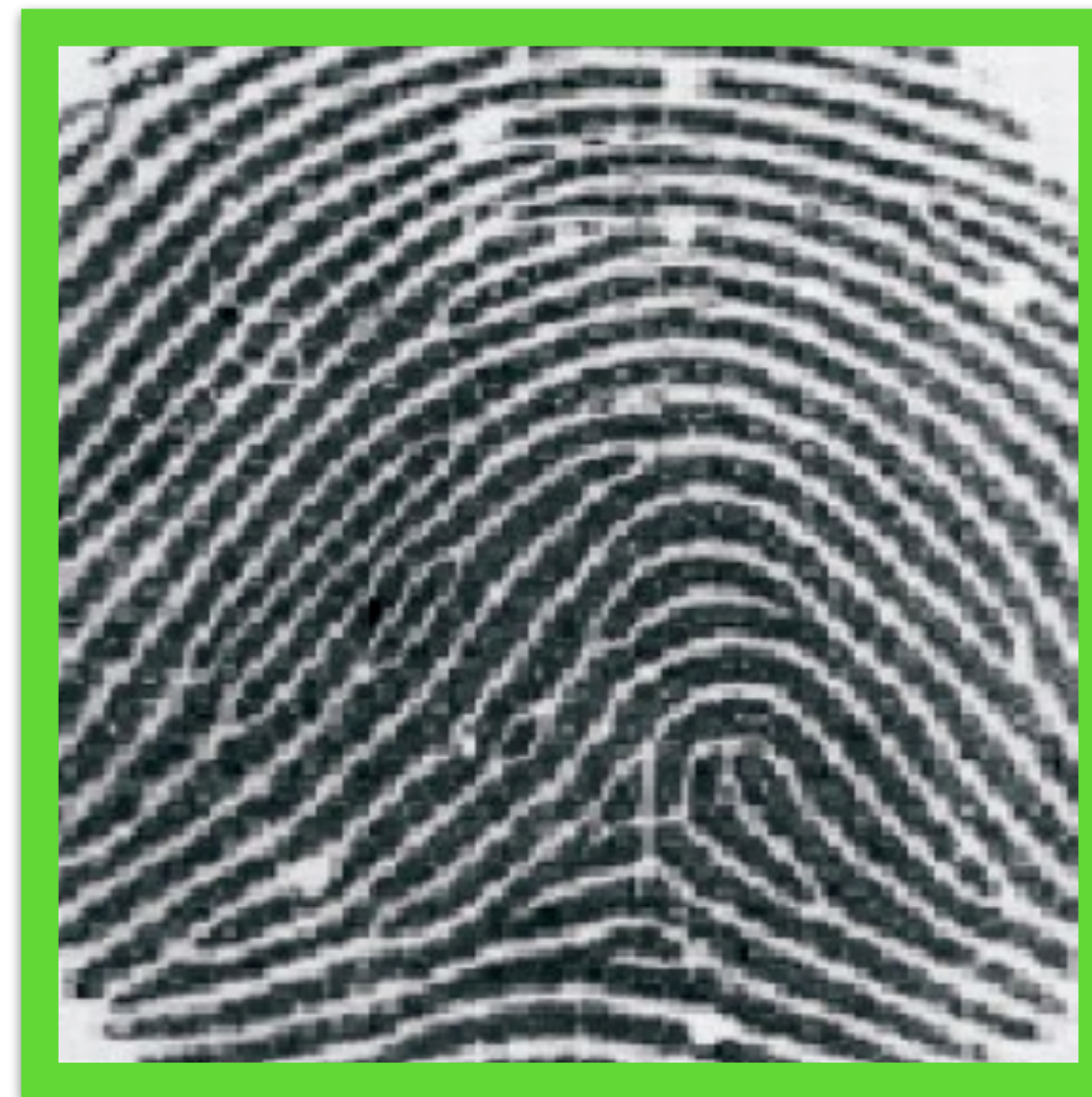
From capacitive sensor



# 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



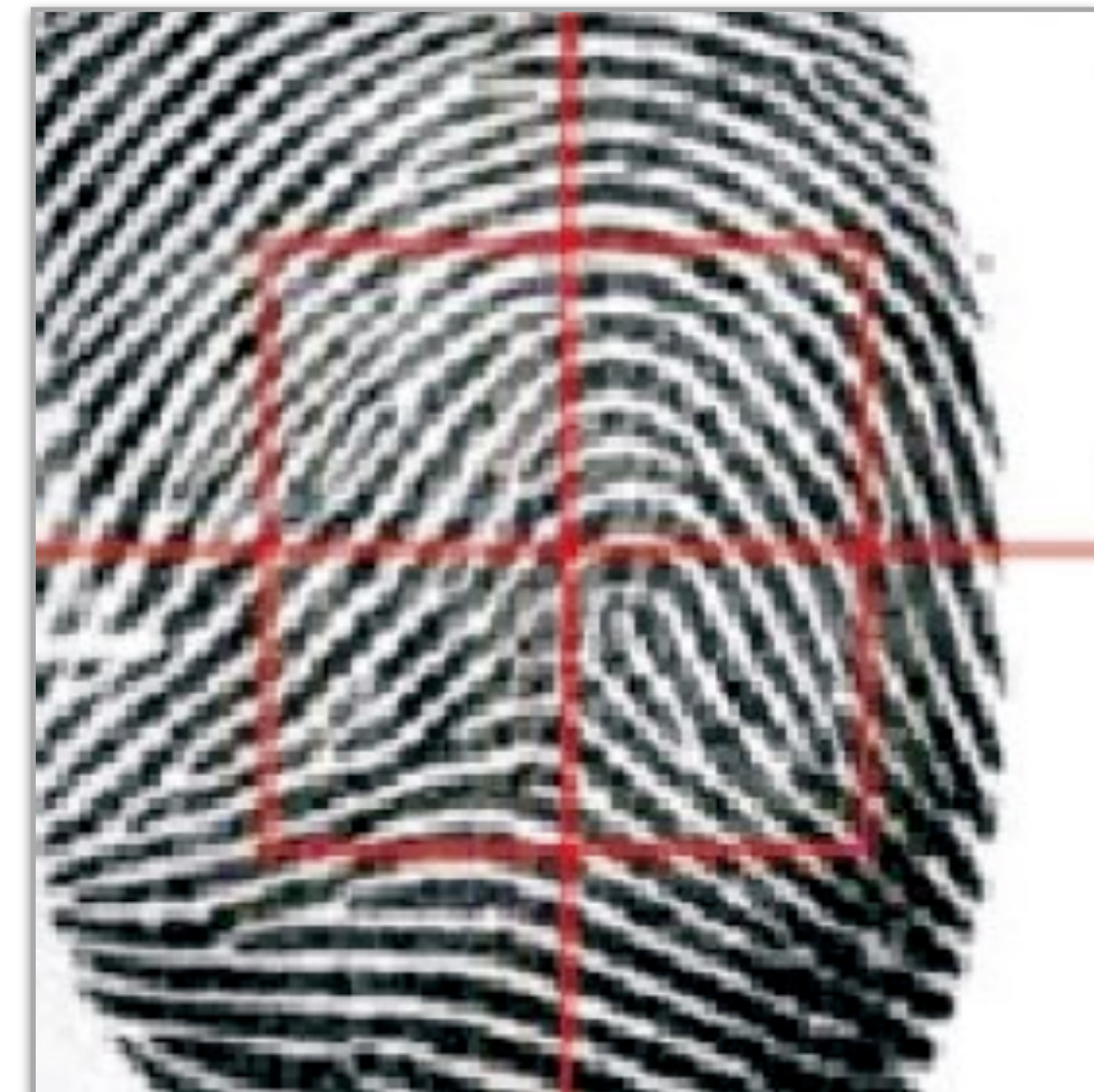
authentic



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

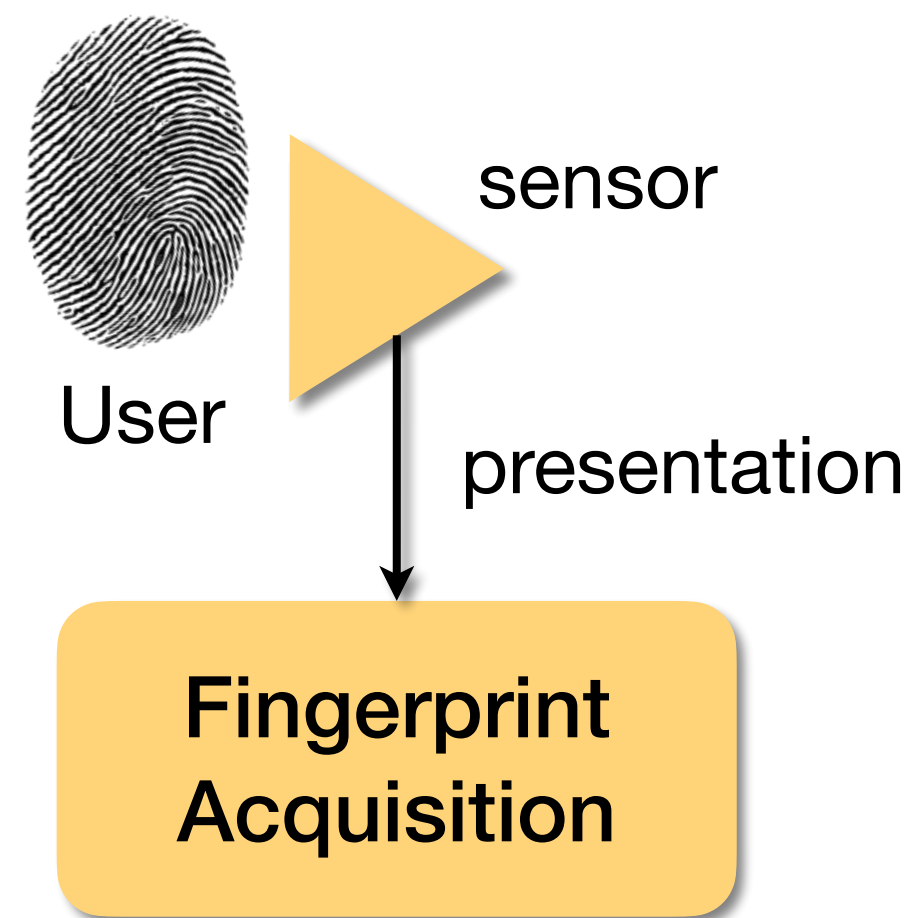


authentic

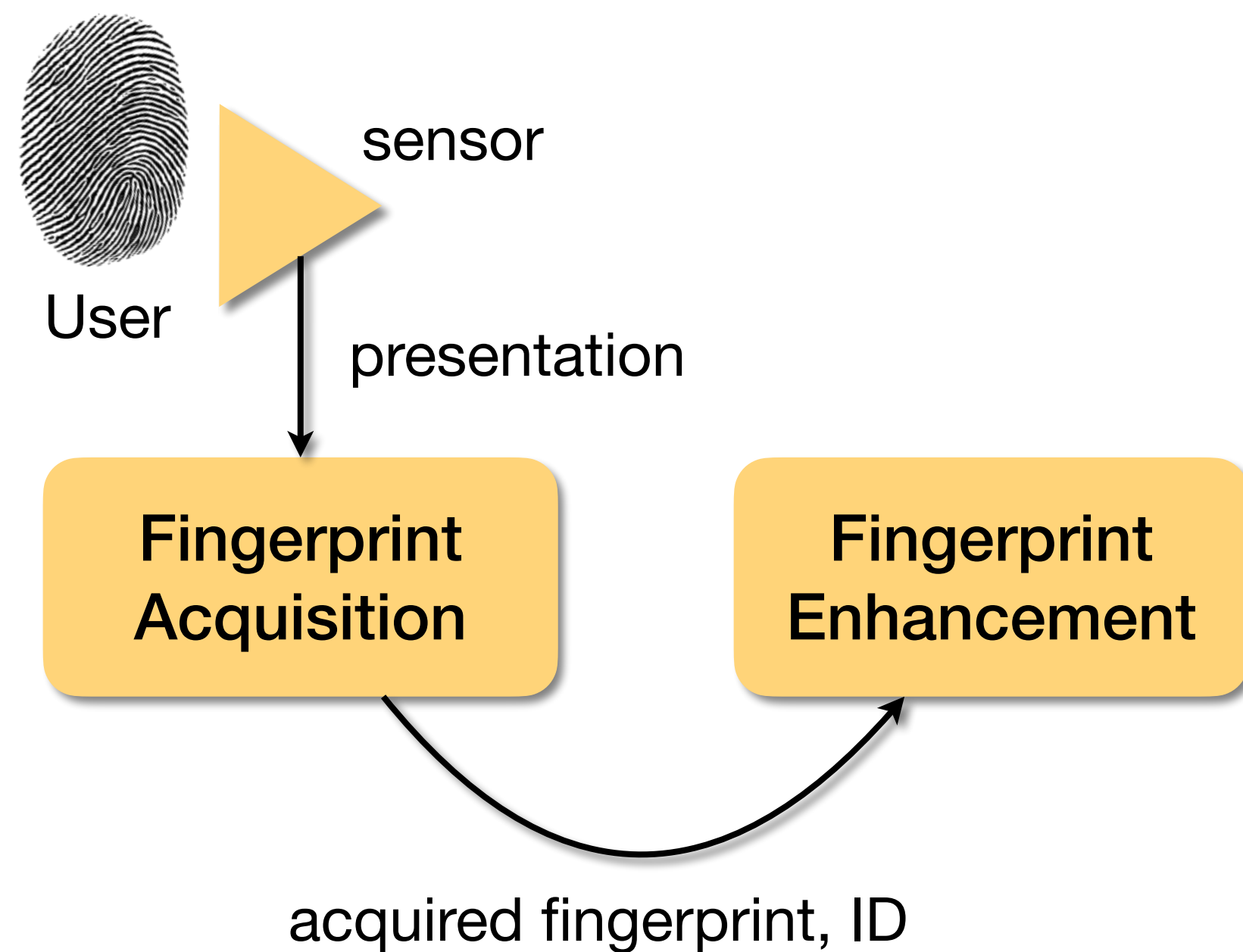




# Fingerprint Recognition



# 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



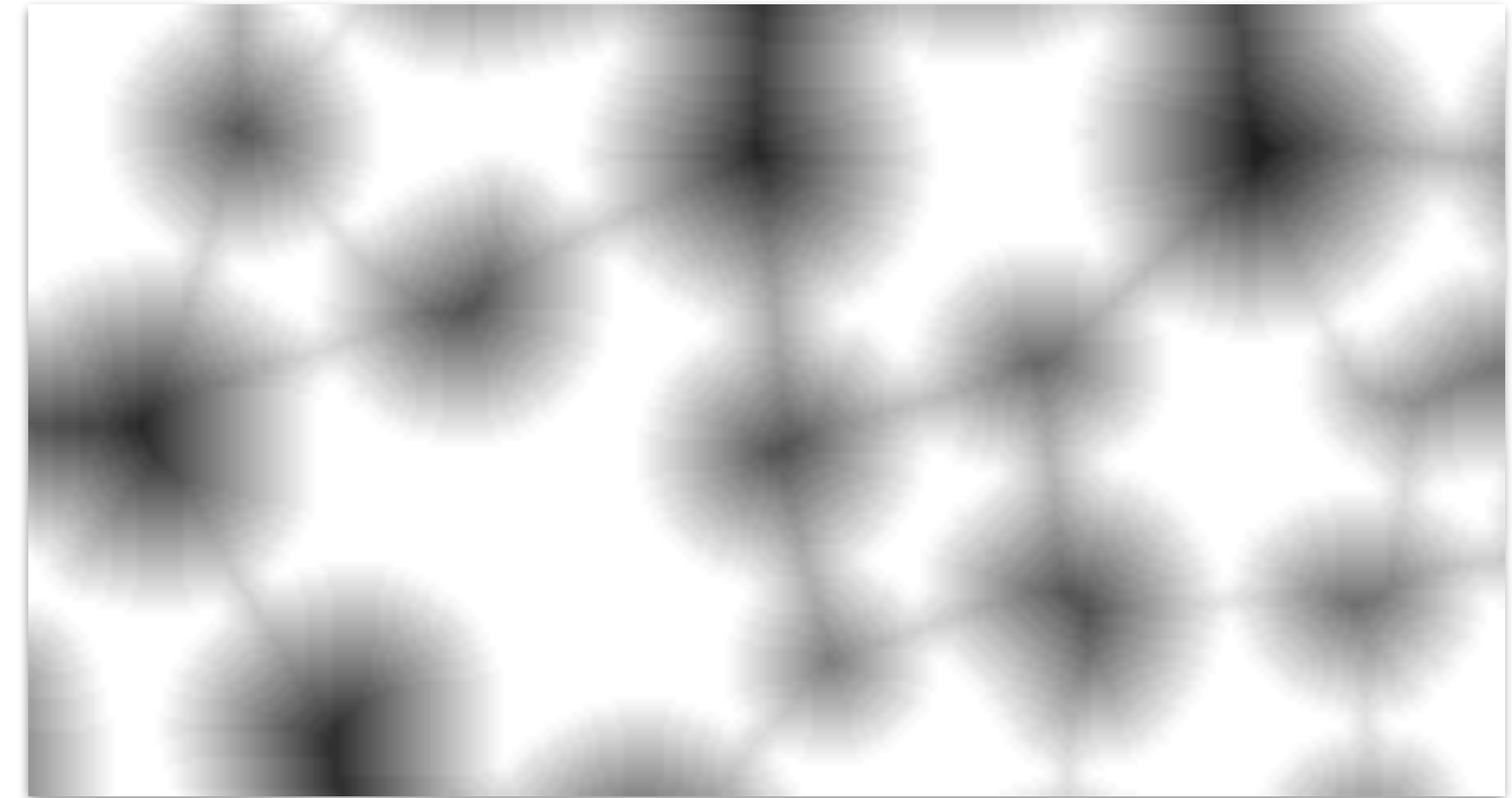
wet

# Enhancement

## Image Processing Solutions

### Tasks

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

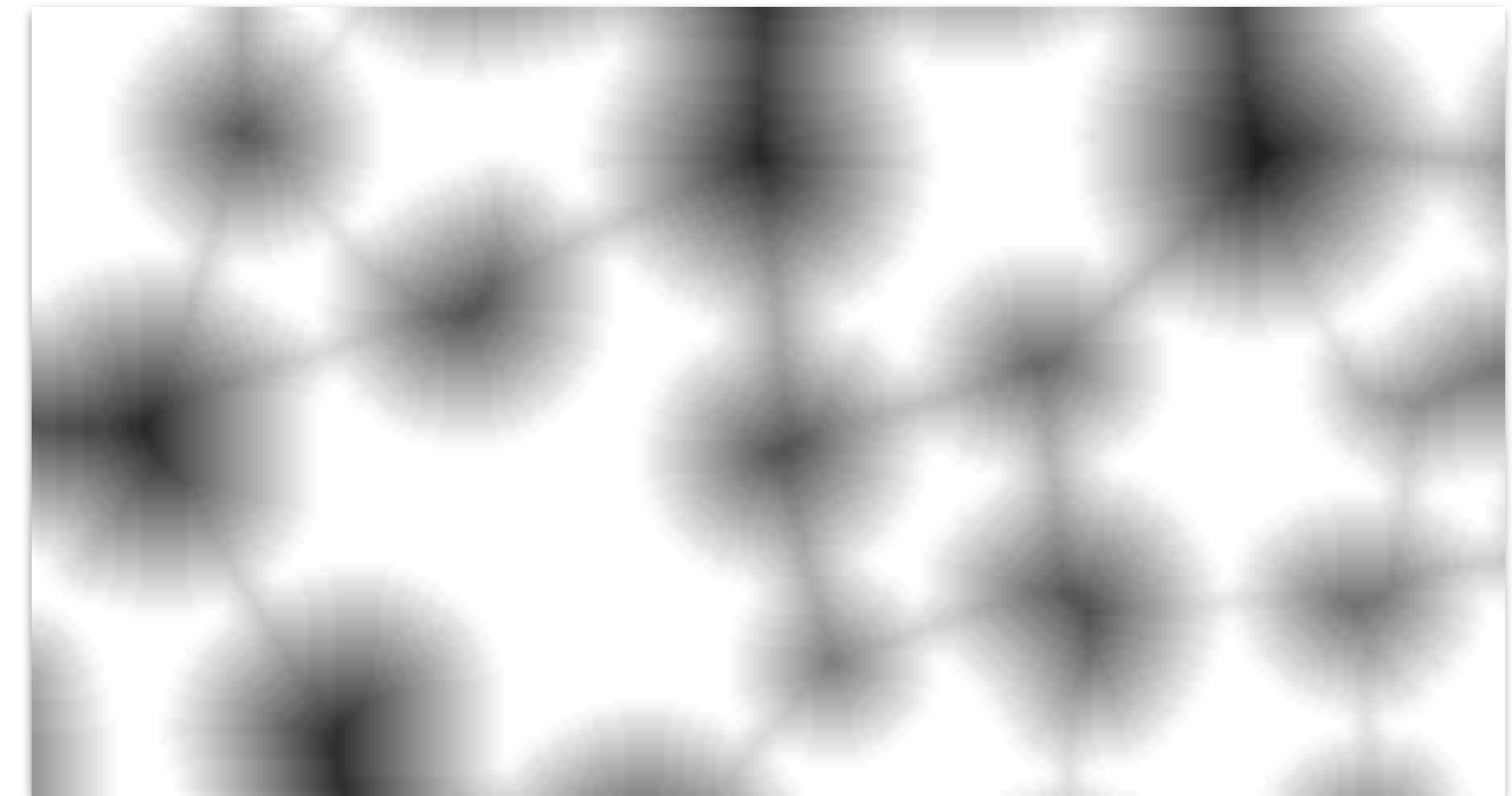


# Enhancement

## Image Processing Solutions

### Tasks

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



# Enhancement

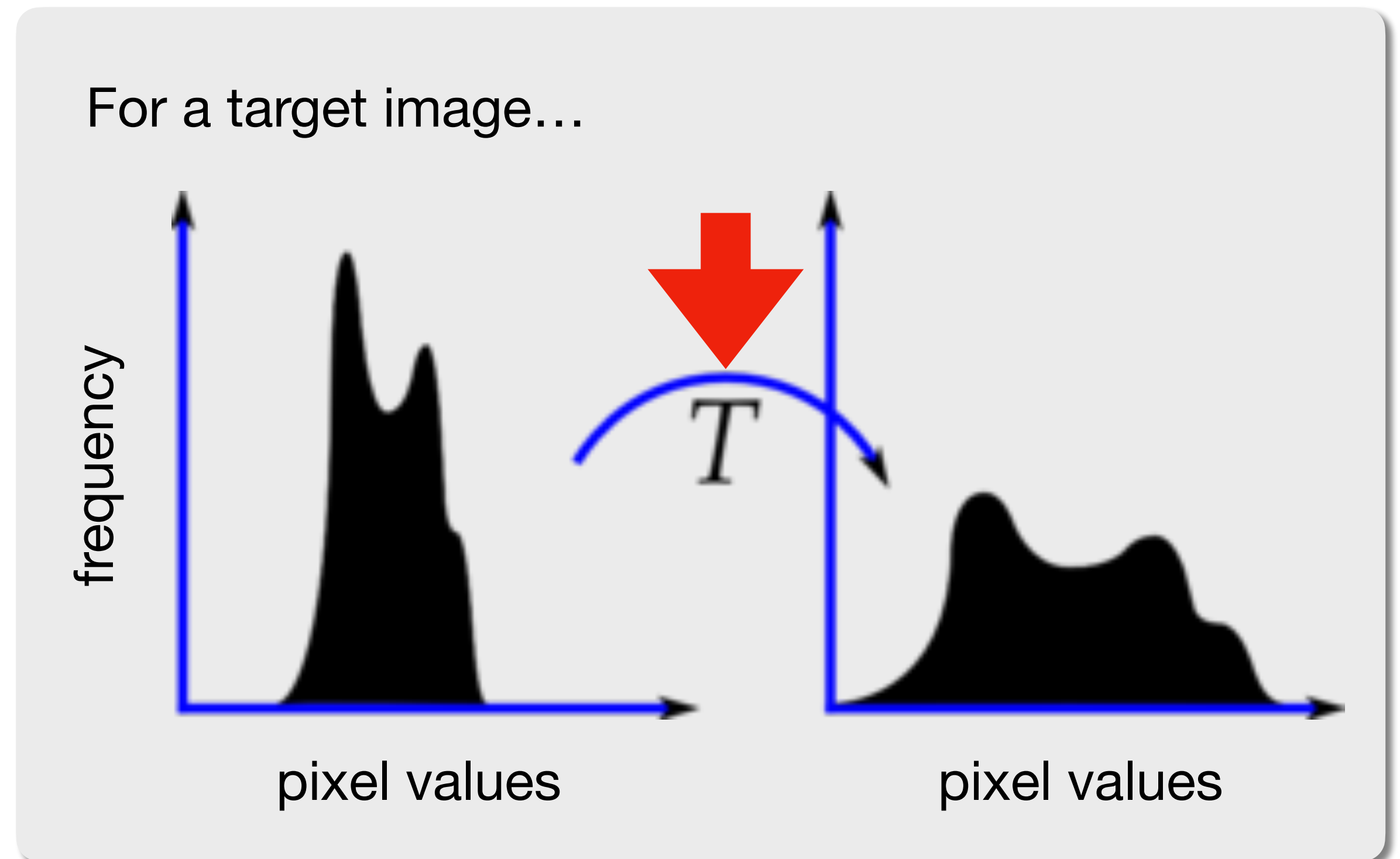
## Image Contrast

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.





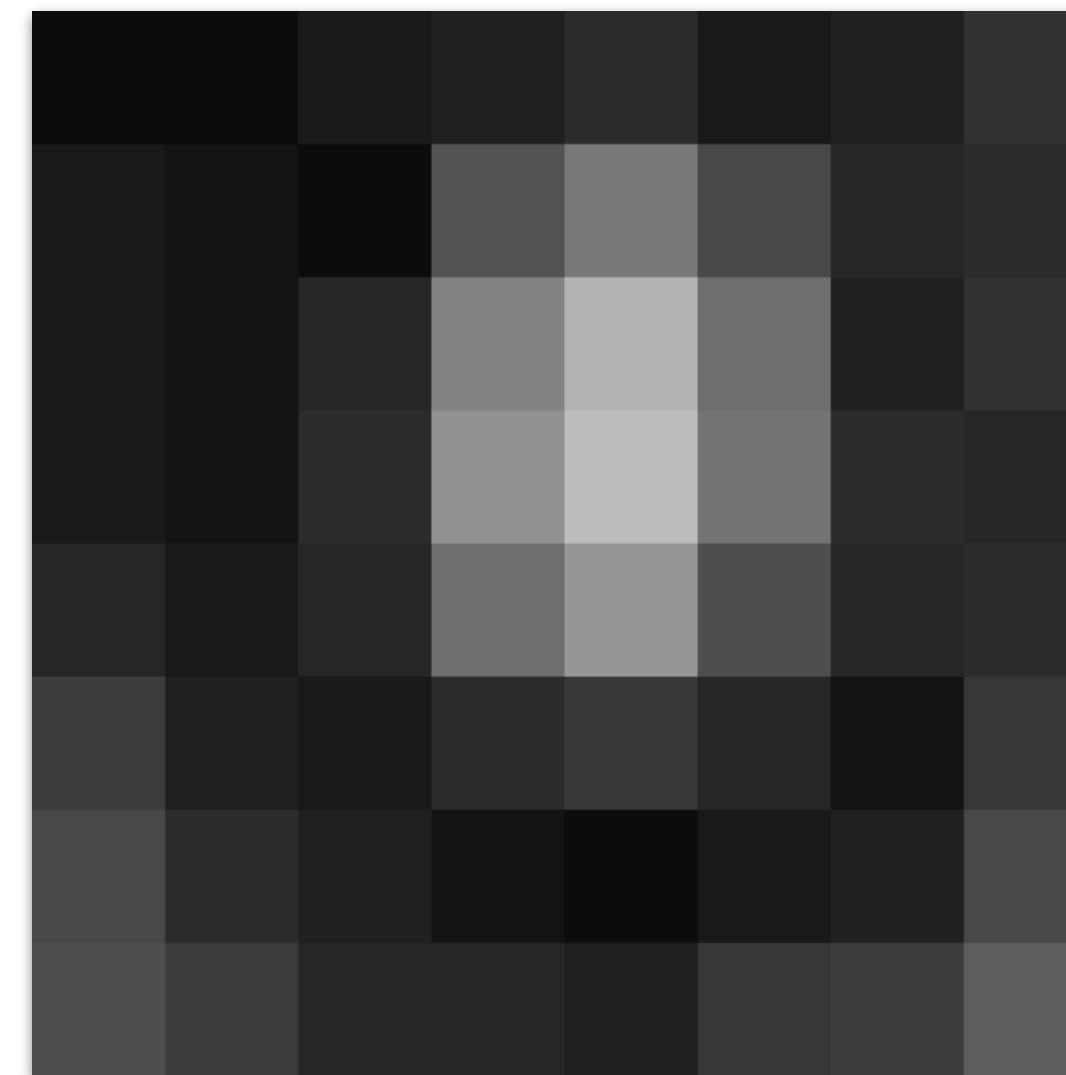
# Enhancement

## 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](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](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
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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](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](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](https://en.wikipedia.org/wiki/Histogram_equalization)

# Enhancement

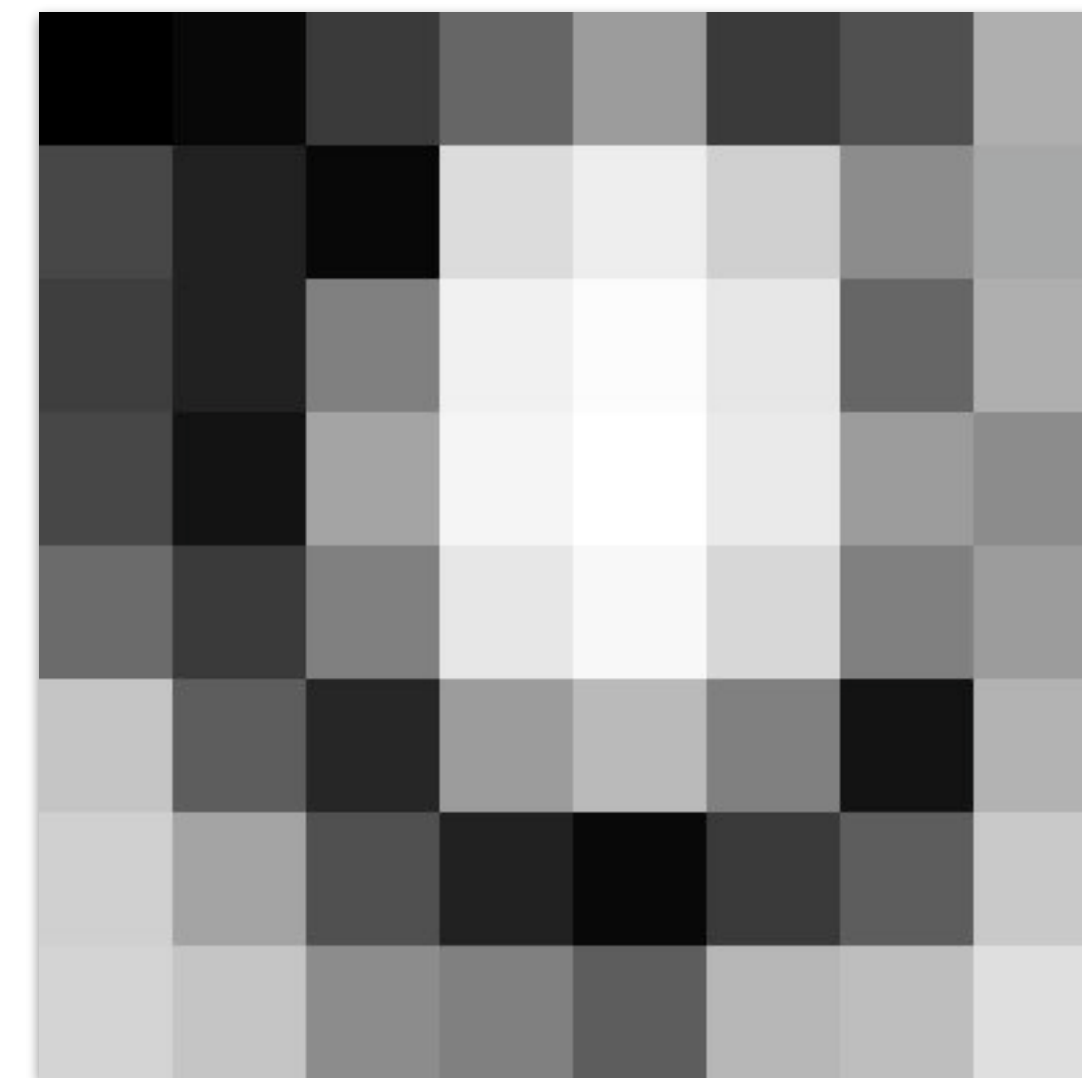
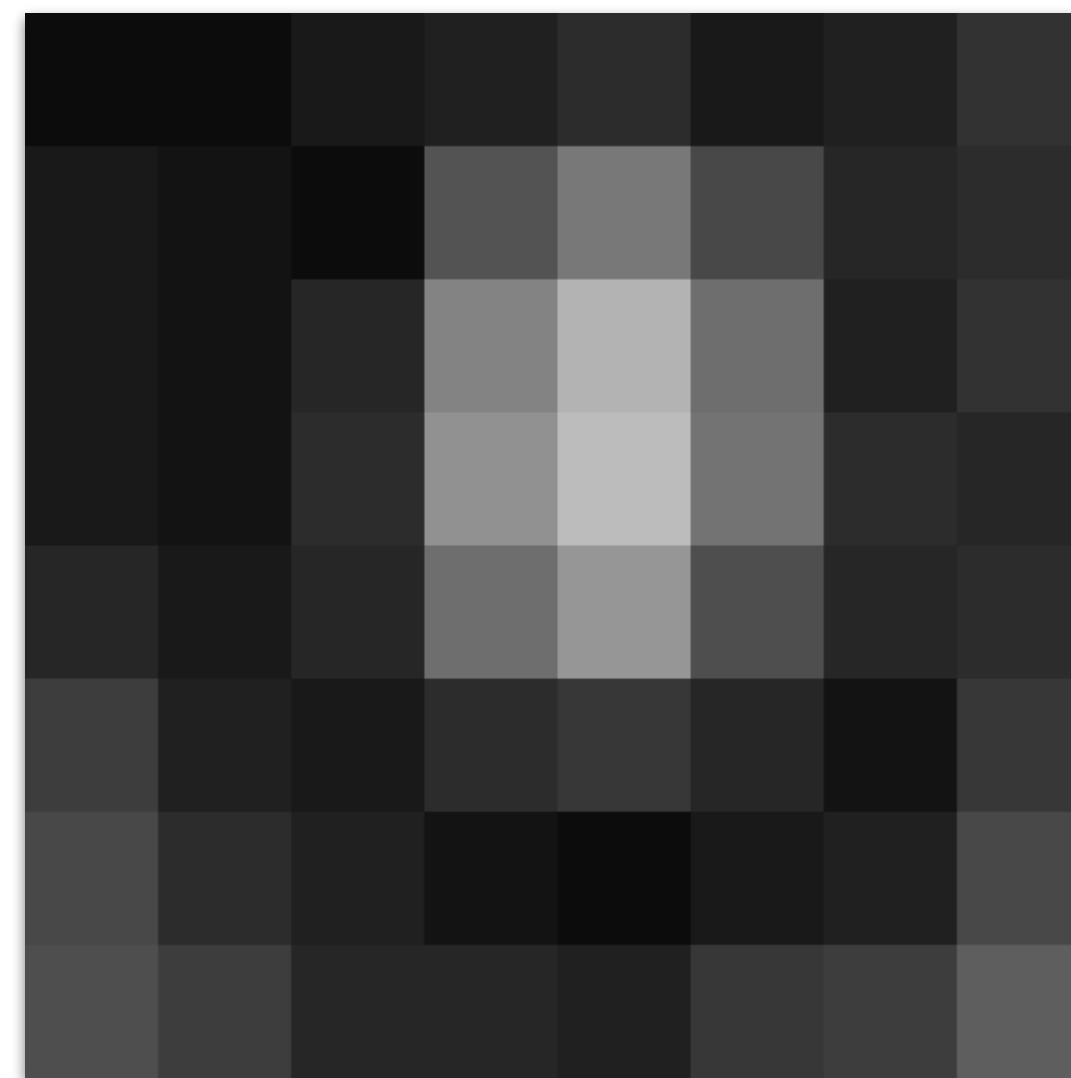
## Color Histogram Equalization

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](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

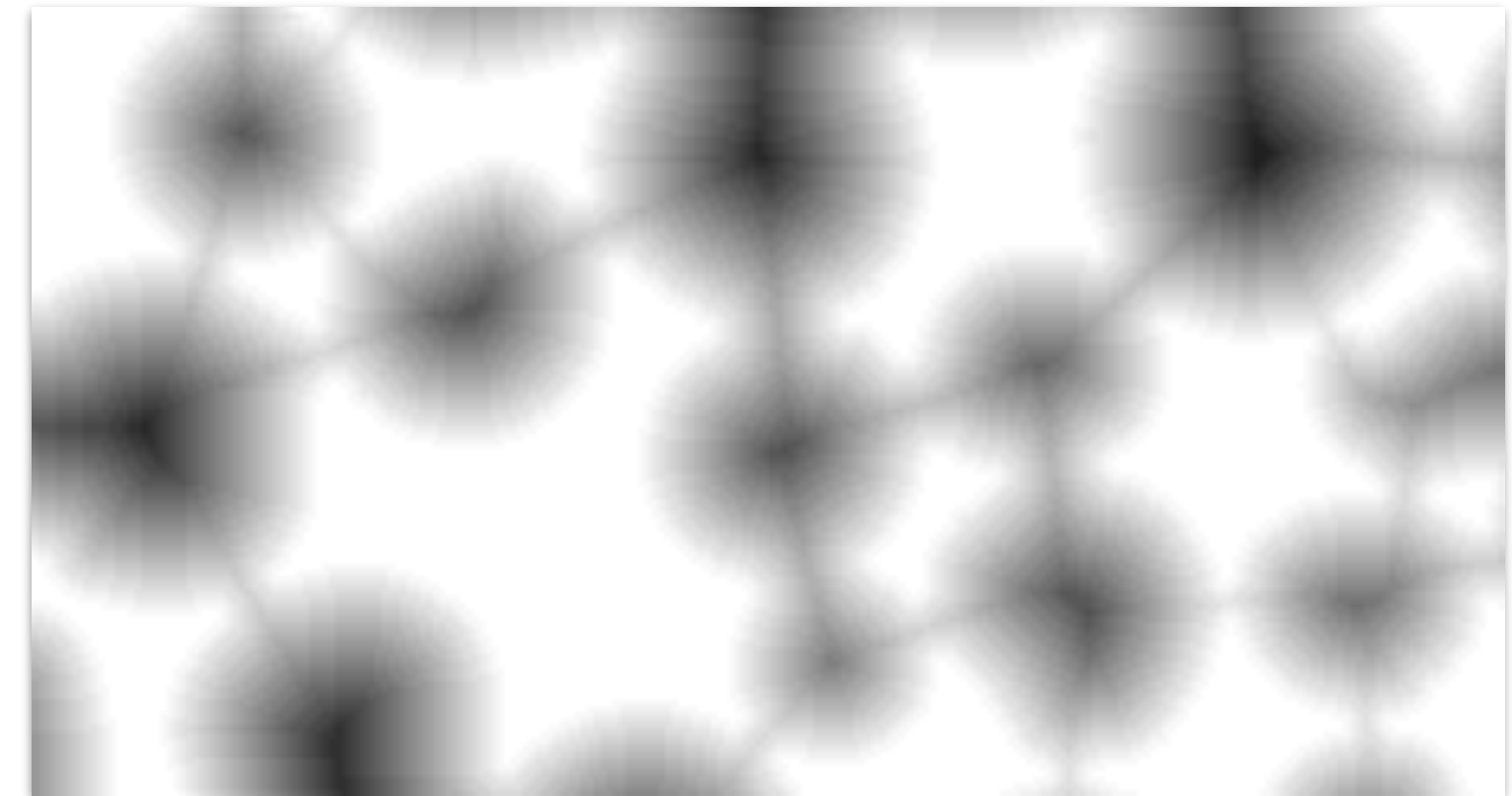
### Tasks

Enhancement of image contrast.

**Enhancement of ridges and valleys.**

Content segmentation.

Others.



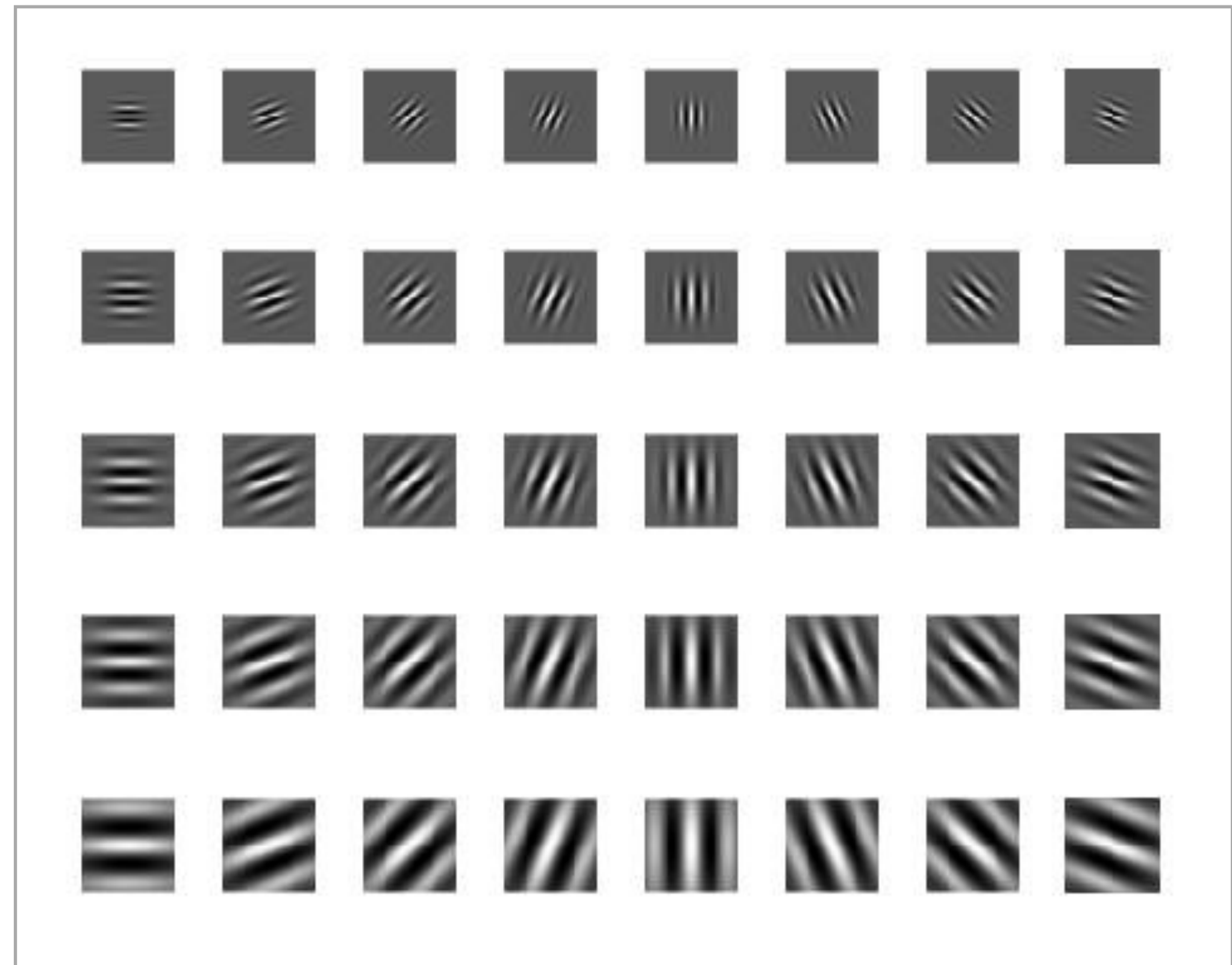
# Enhancement

## 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.



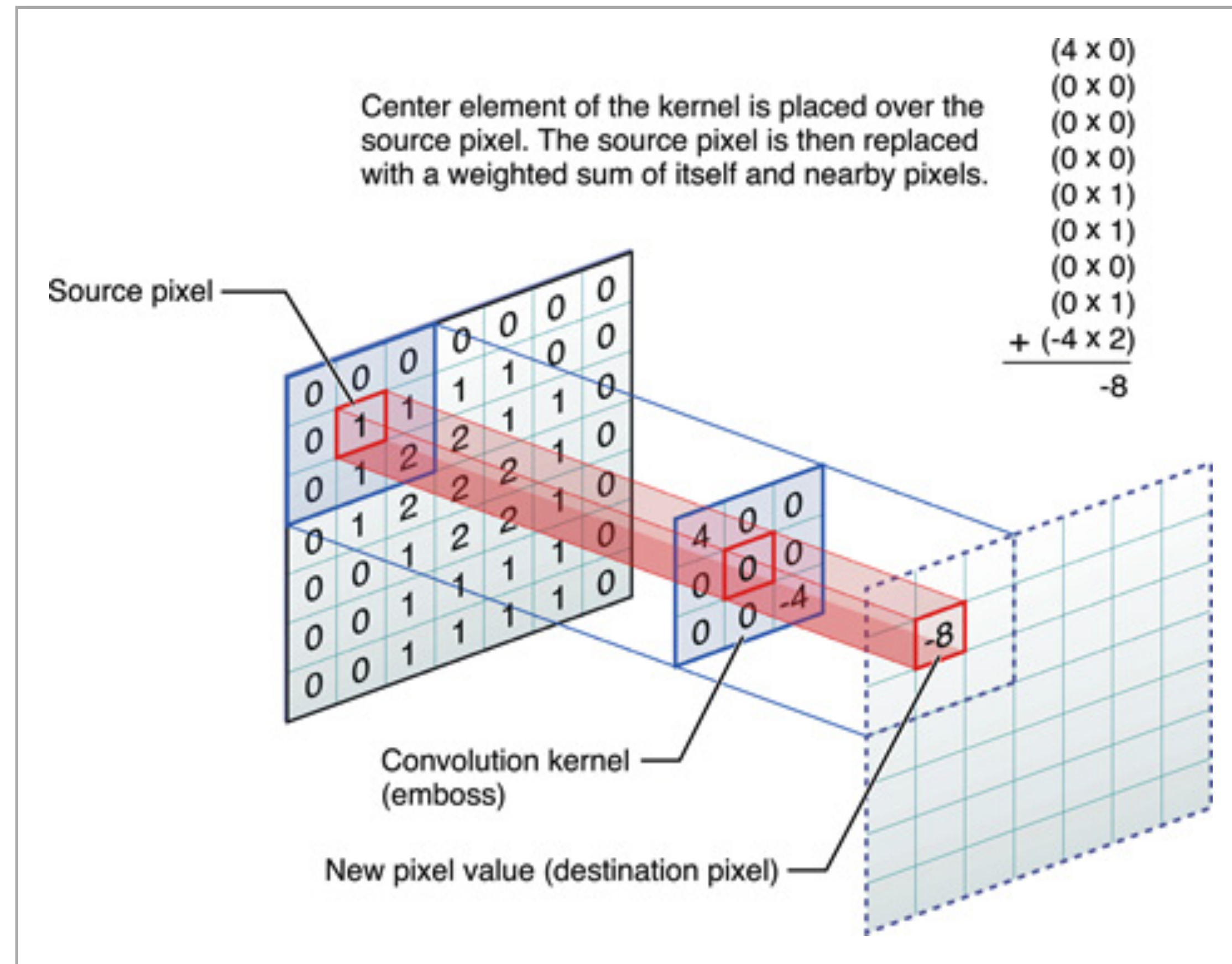
# Enhancement

## Ridges and Valleys

Example:

Image filtering with Gabor filters.

Gabor filters may be applied to an image through convolutions.



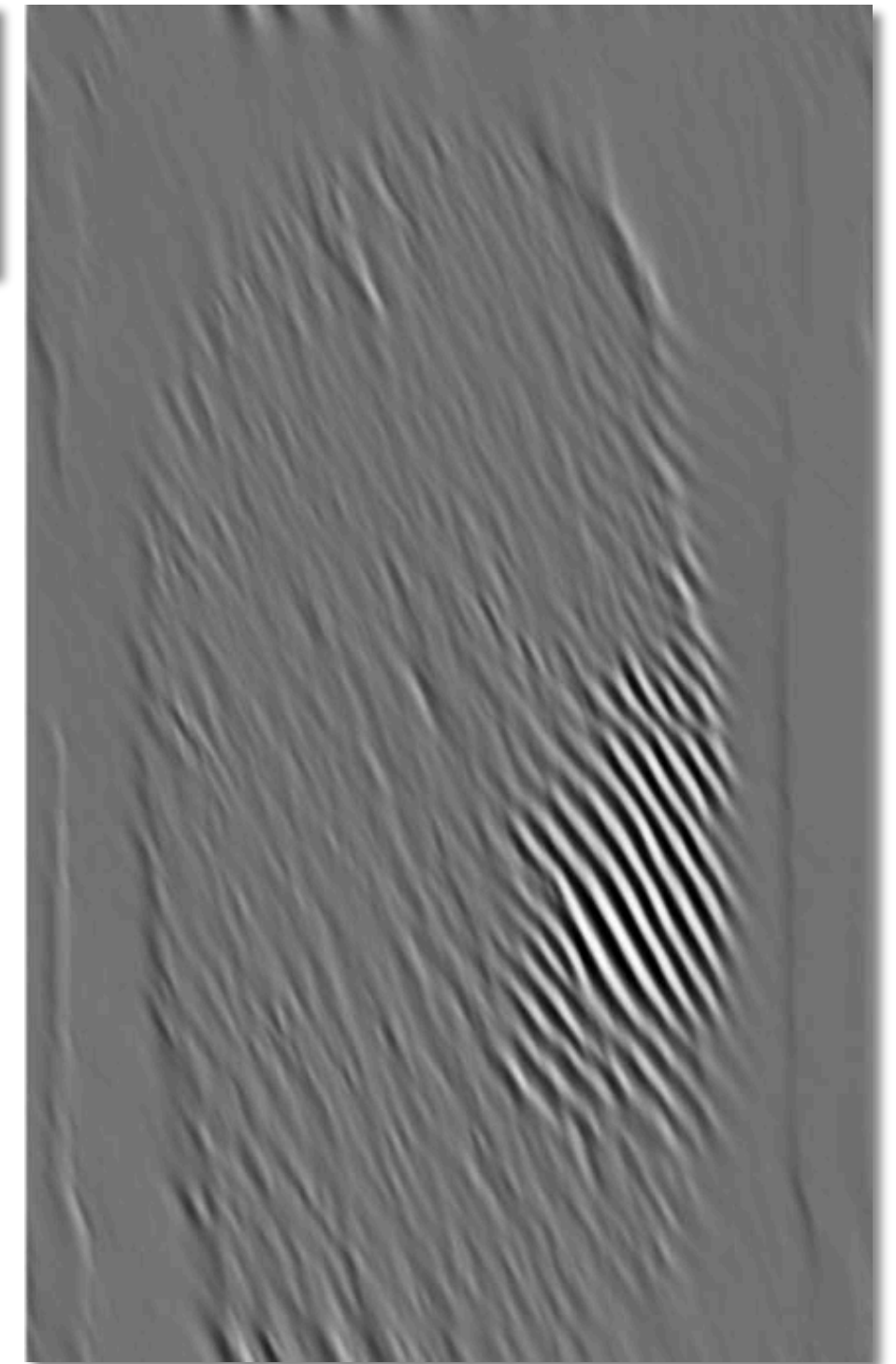
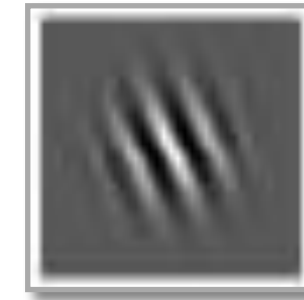
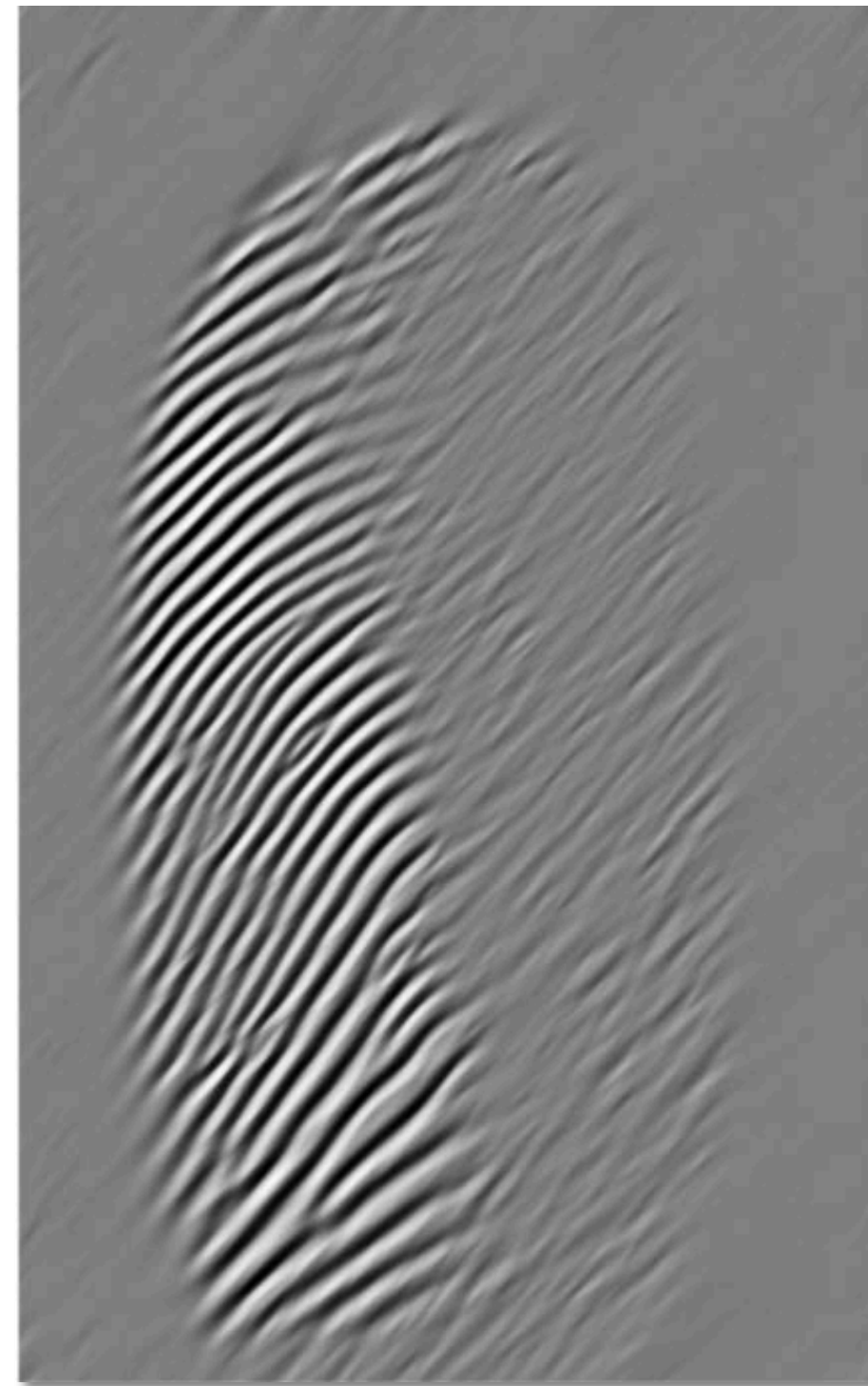
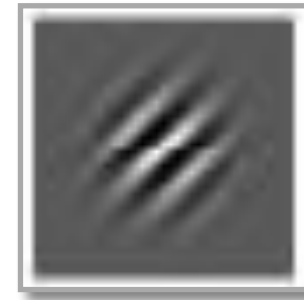
Source: <https://developer.apple.com/library/archive/documentation/Performance/Conceptual/vimage/ConvolutionOperations/ConvolutionOperations.html>

# 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

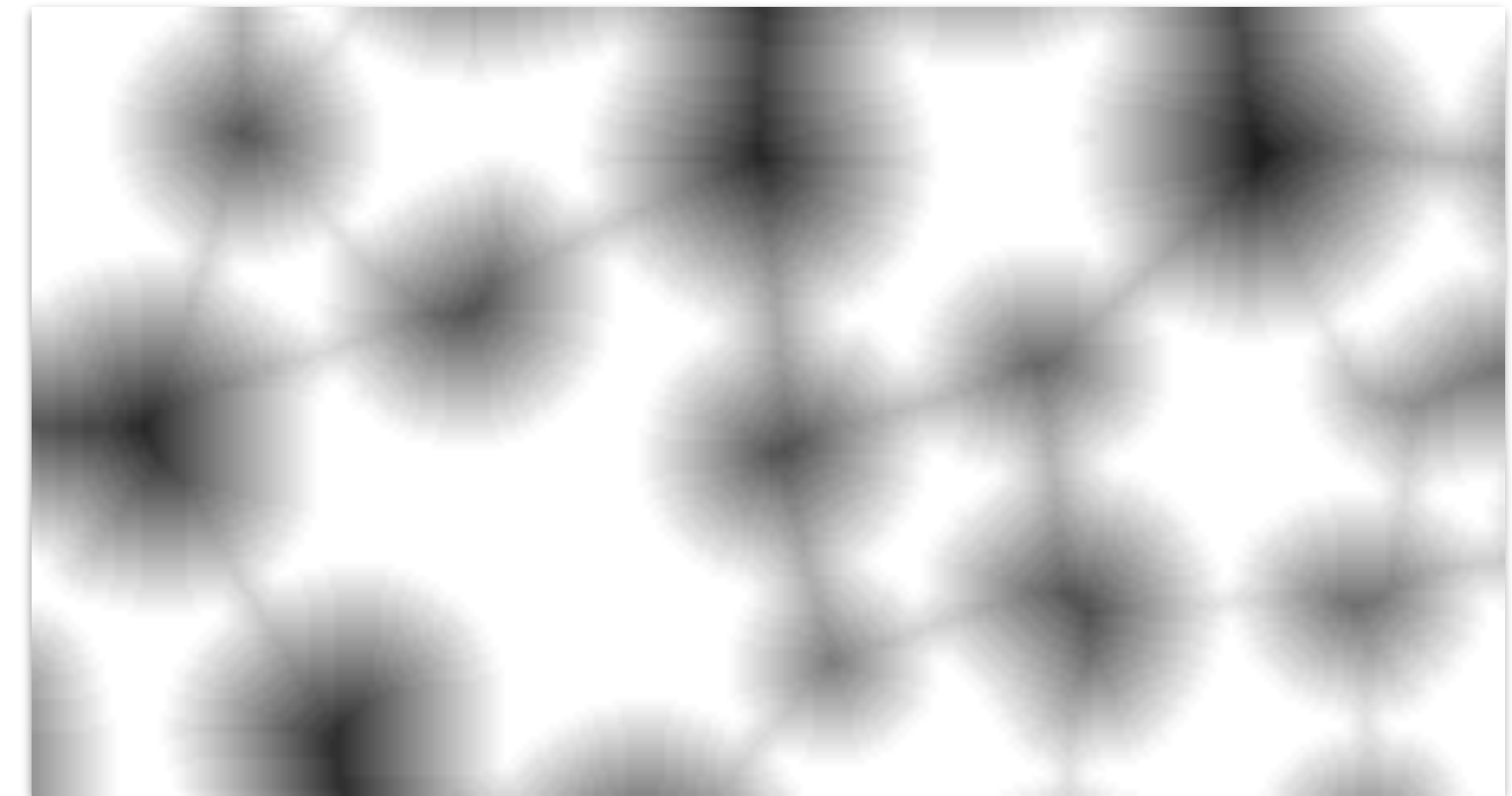
### Tasks

Enhancement of image contrast.

Enhancement of ridges and valleys.

**Content segmentation.**

Others.



# Enhancement

## Segmentation

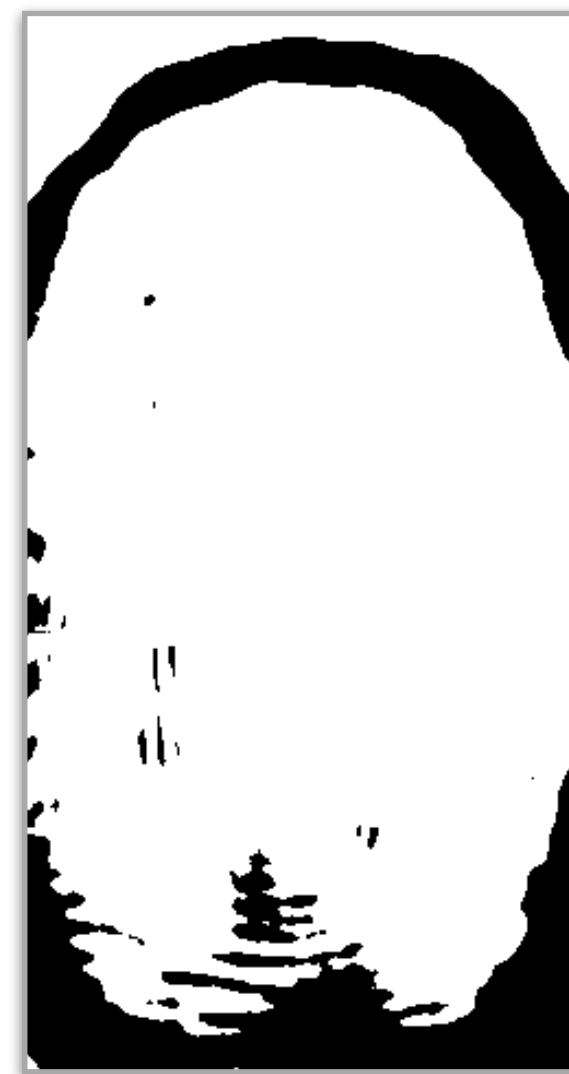
Example: blurring, thresholding, and morphological operations.



before



blur



threshold



open



after

# Enhancement

## Image Processing Solutions

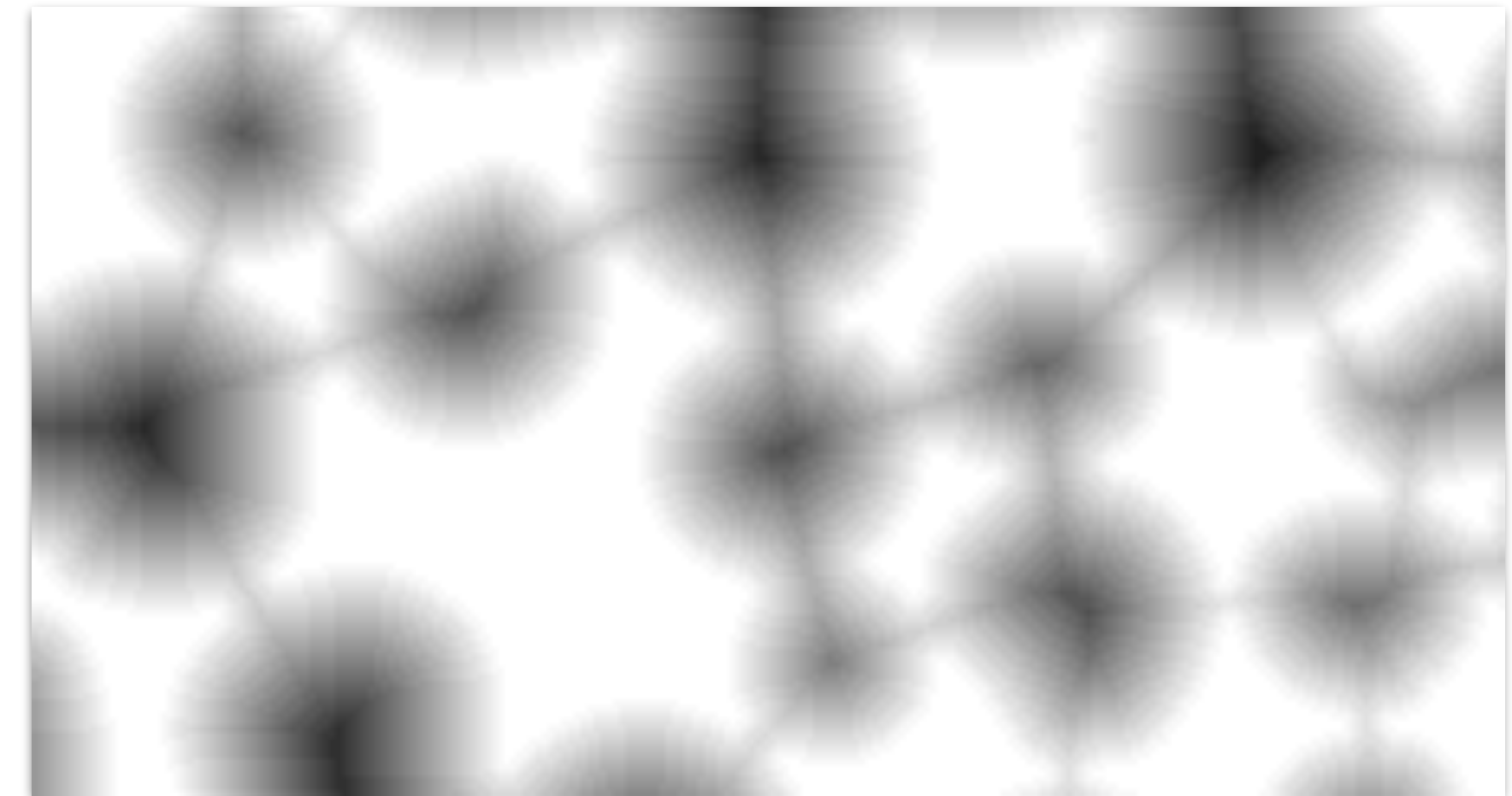
### Tasks

Enhancement of image contrast.

Enhancement of ridges and valleys.

Content segmentation.

**Others.**





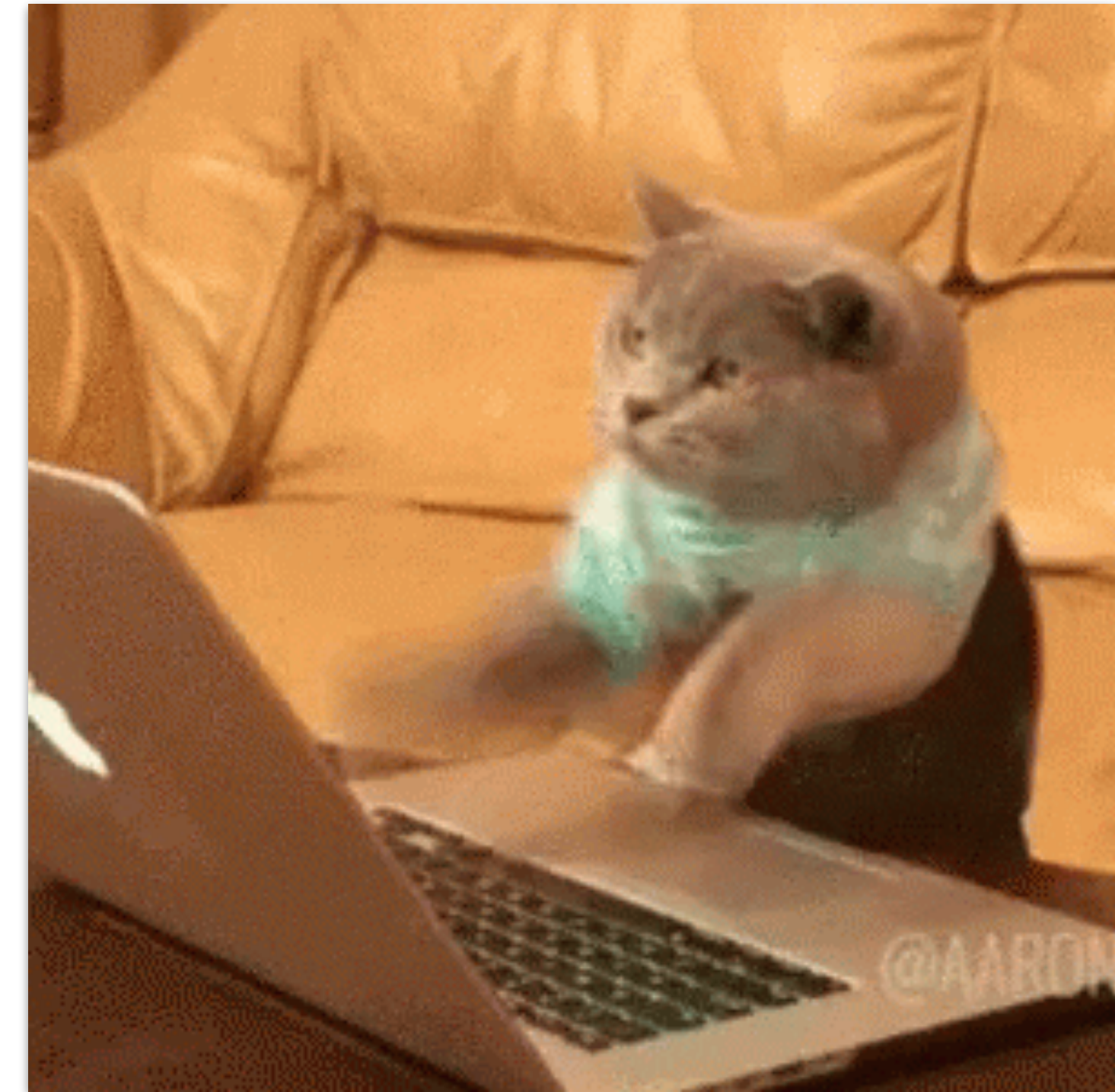
# Enhancement

## 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.



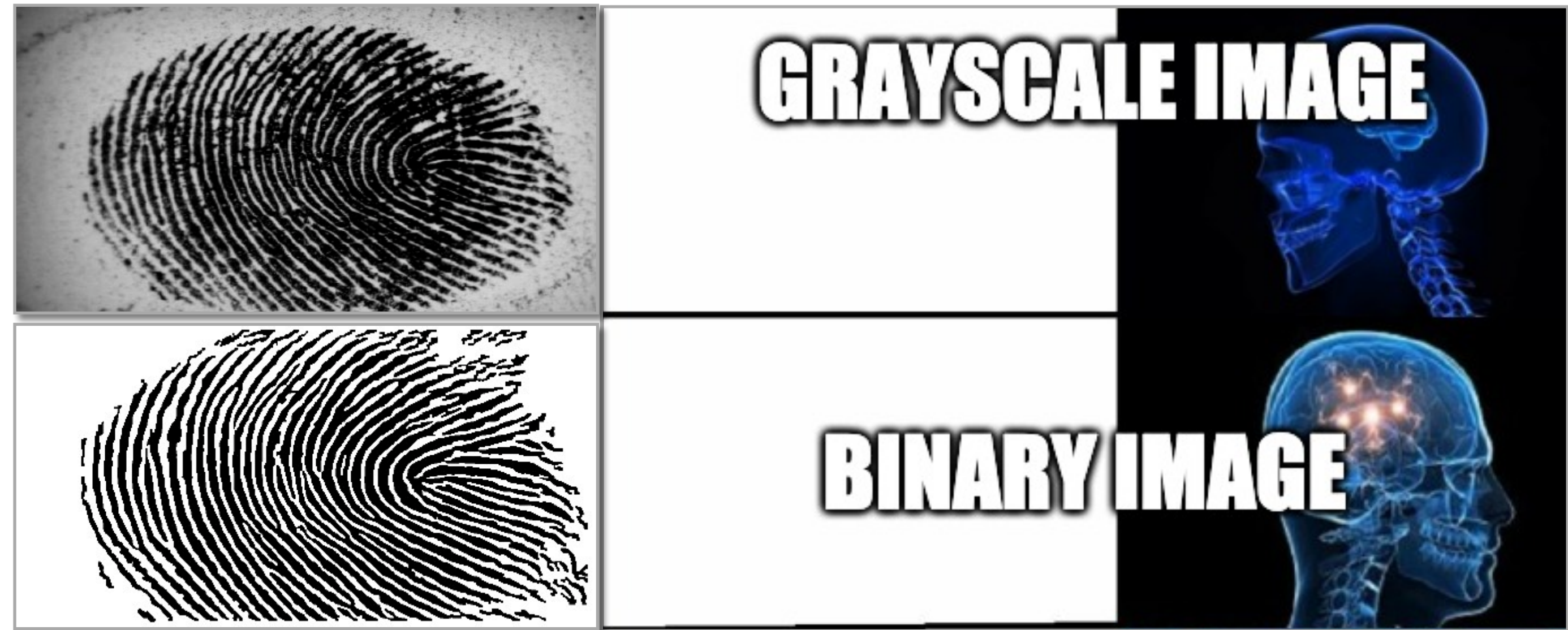
# Enhancement

**Other Strategies**  
Start from...



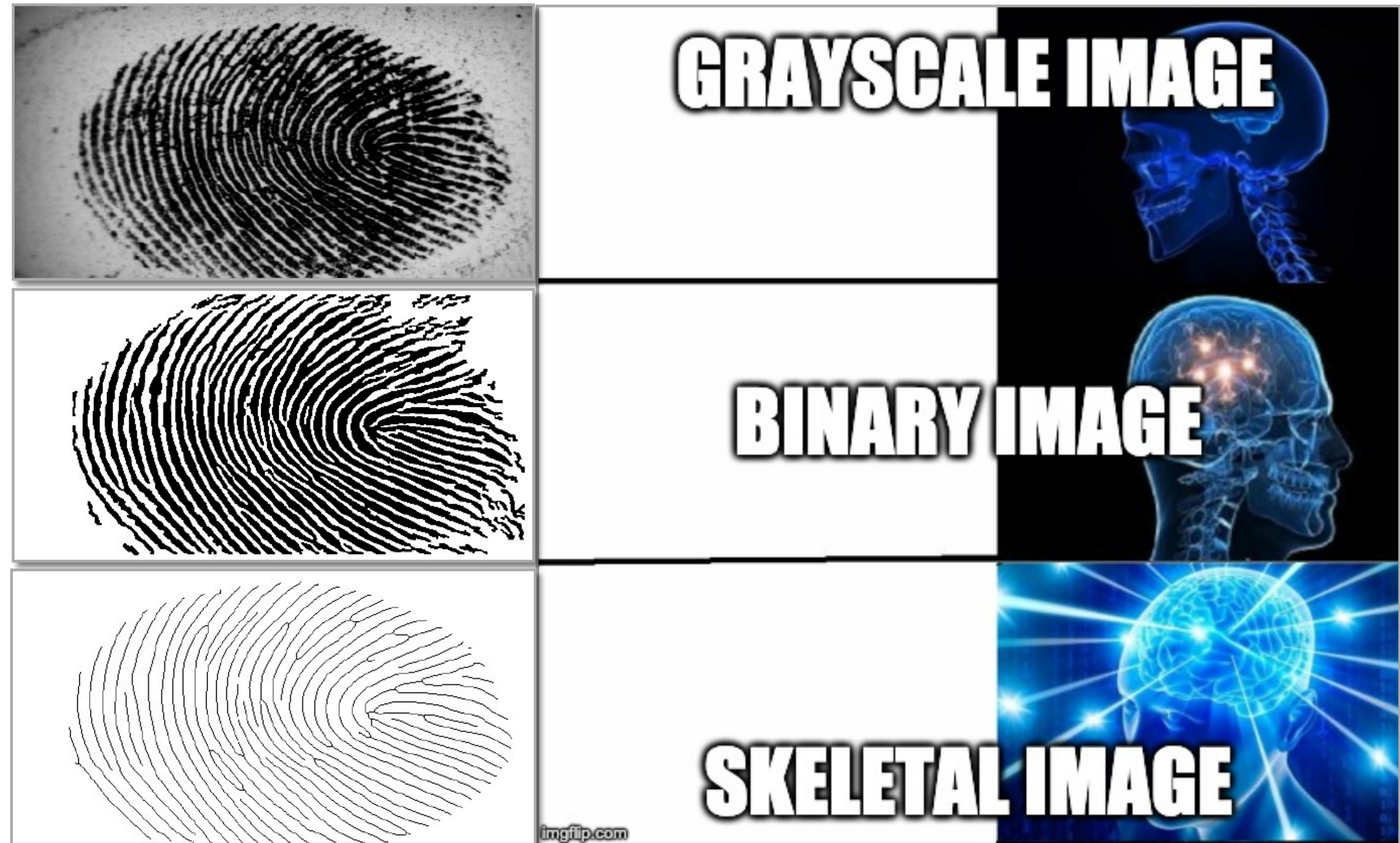
# Enhancement

**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.



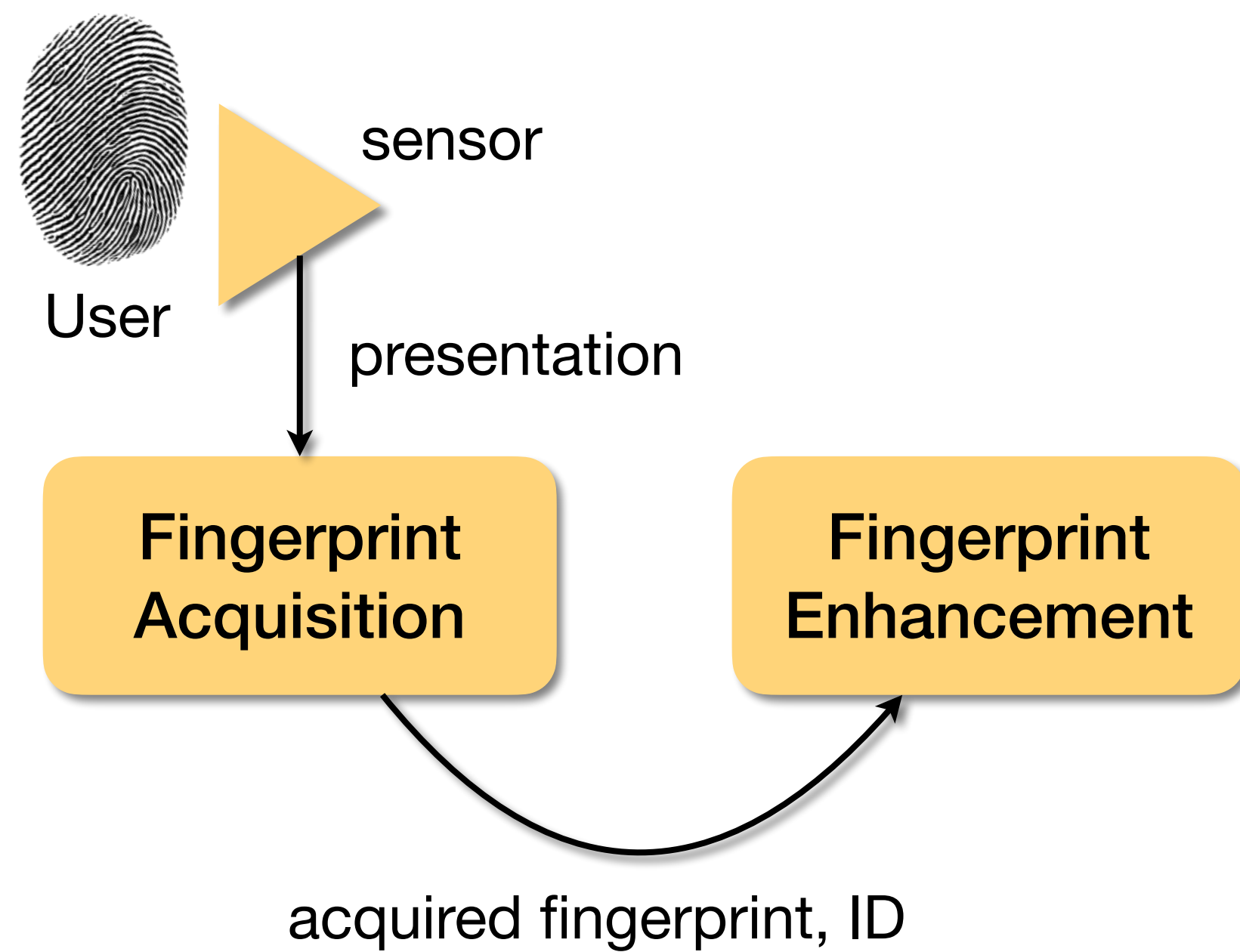
Source: Dr. Adam Czajka

# Enhancement

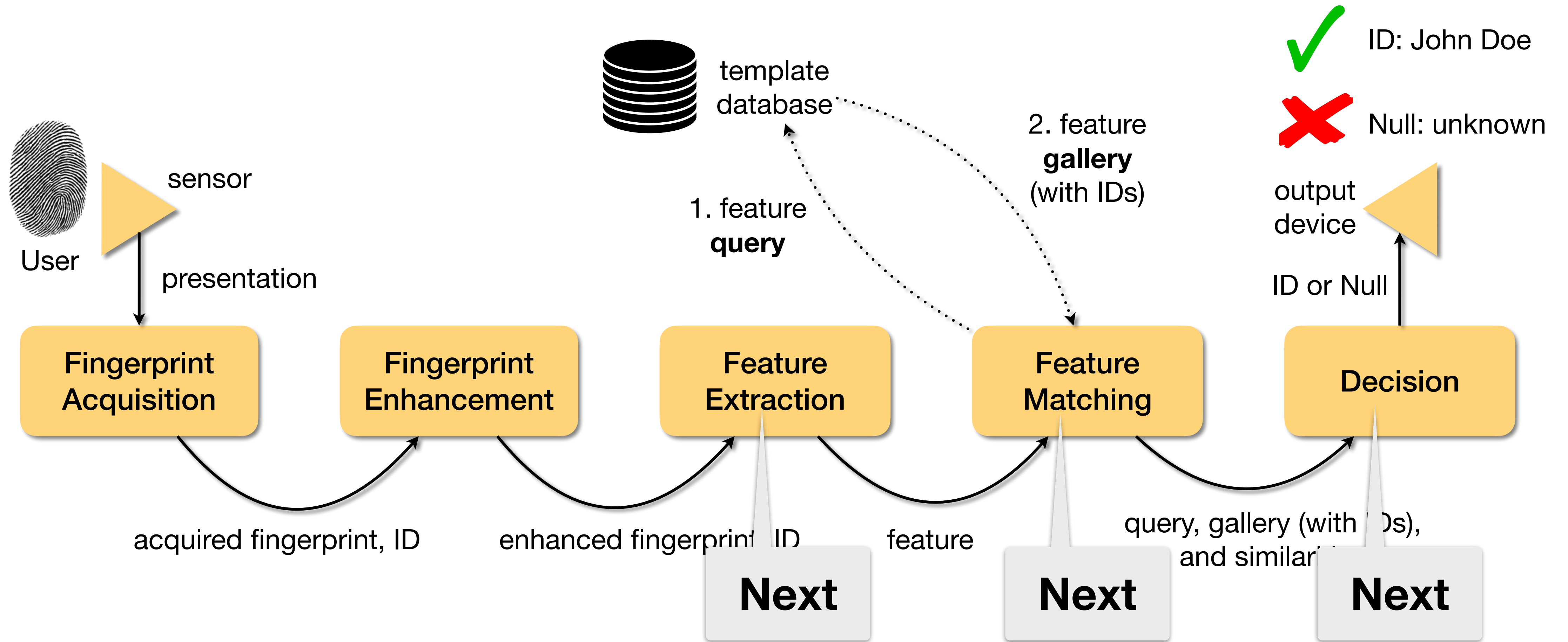


Source: Dr. Adam Czajka

# 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 <https://sakai.luc.edu/x/HAZC1P>.

