

POLITECNICO DI MILANO



REVERSE ENGINEERING OF VIDEO CONTENT FOR FORENSIC ANALYSIS

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Diffusion of multimedia sharing platforms



Motivations

- Diffusion of multimedia **sharing platforms**



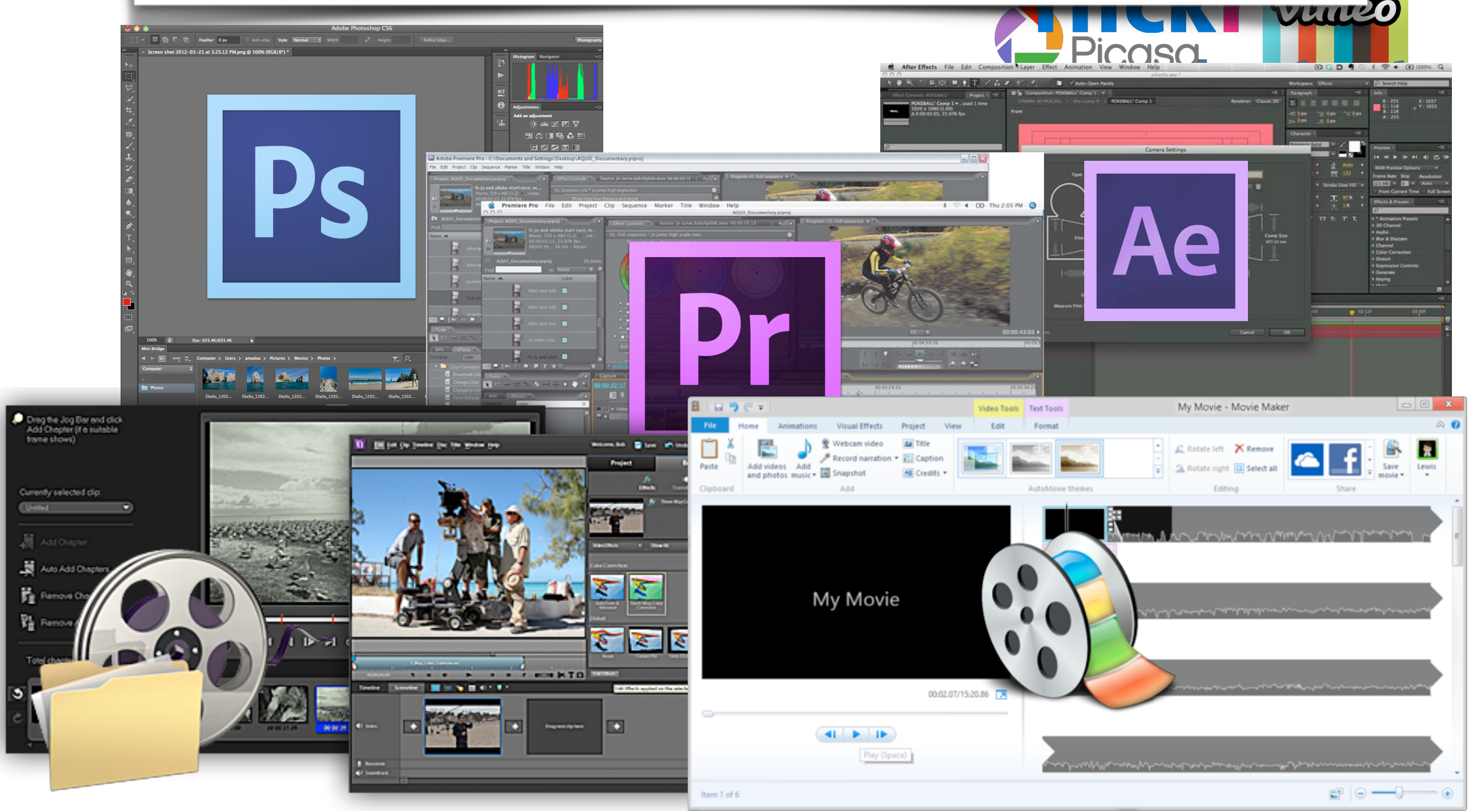
Motivations

- Diffusion of multimedia **sharing platforms**
 - Huge amount of user-generated content



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- F

Availability of user-friendly editing software



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 - Huge amount of user-generated content
- Availability of user-friendly **video-editing software**
 - Easy to tamper with videos



Motivations

- Diffusion of r
- Huge amo

We cannot trust what we see!



ftware



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Software



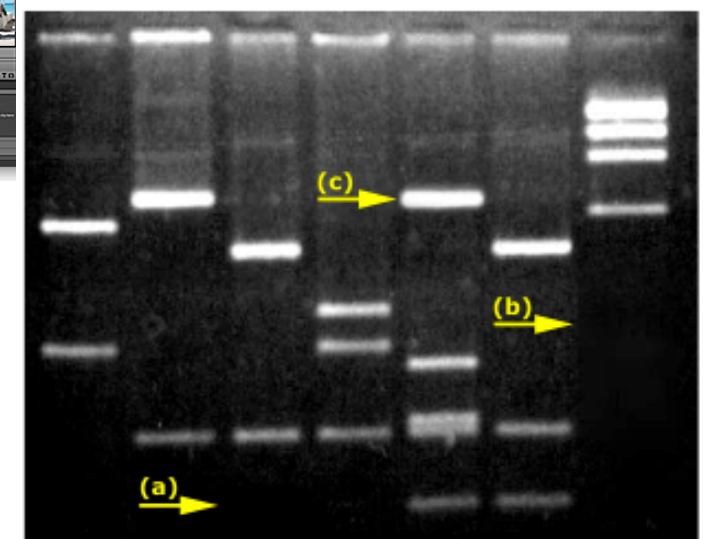
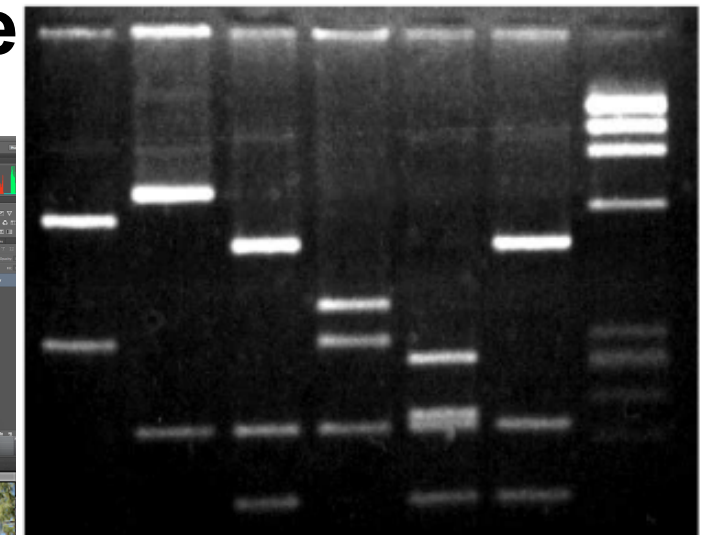
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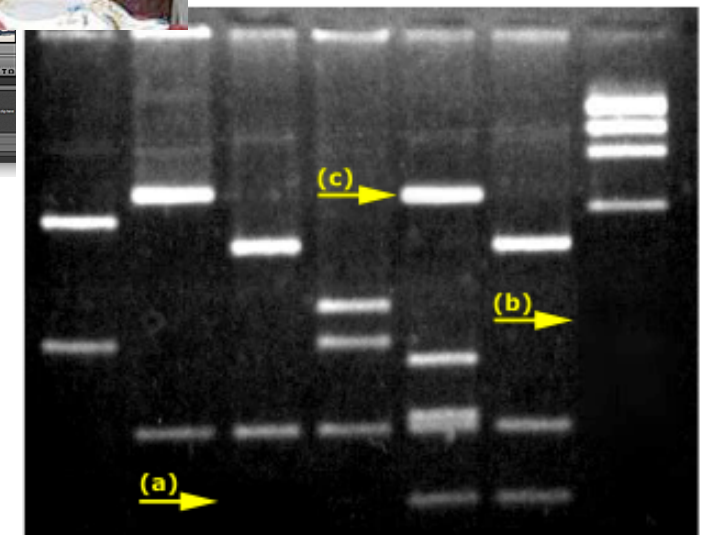
Software



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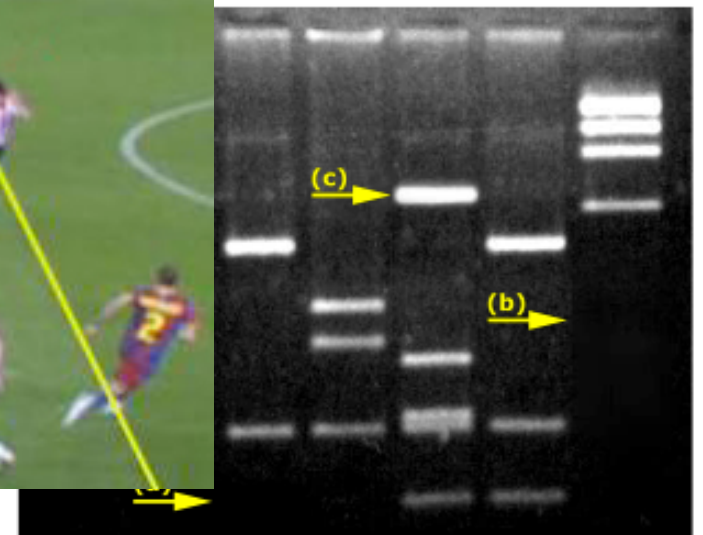
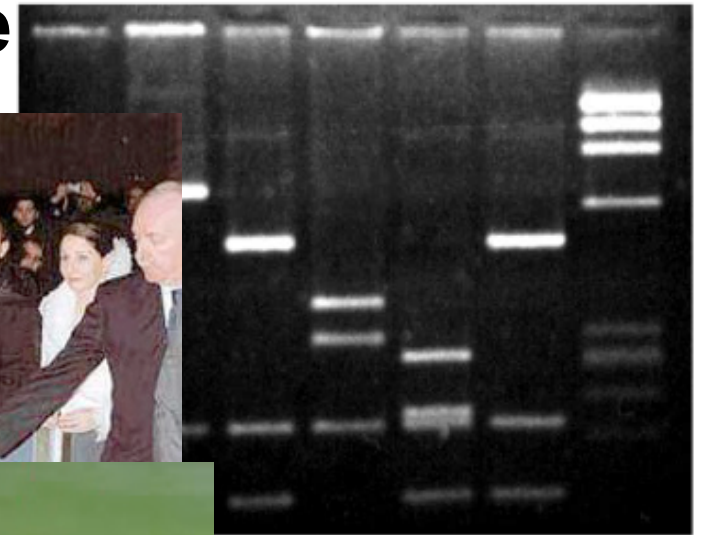
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 TC-SON 245043 85724-5043
 0**71486*03070**6**

Motivations

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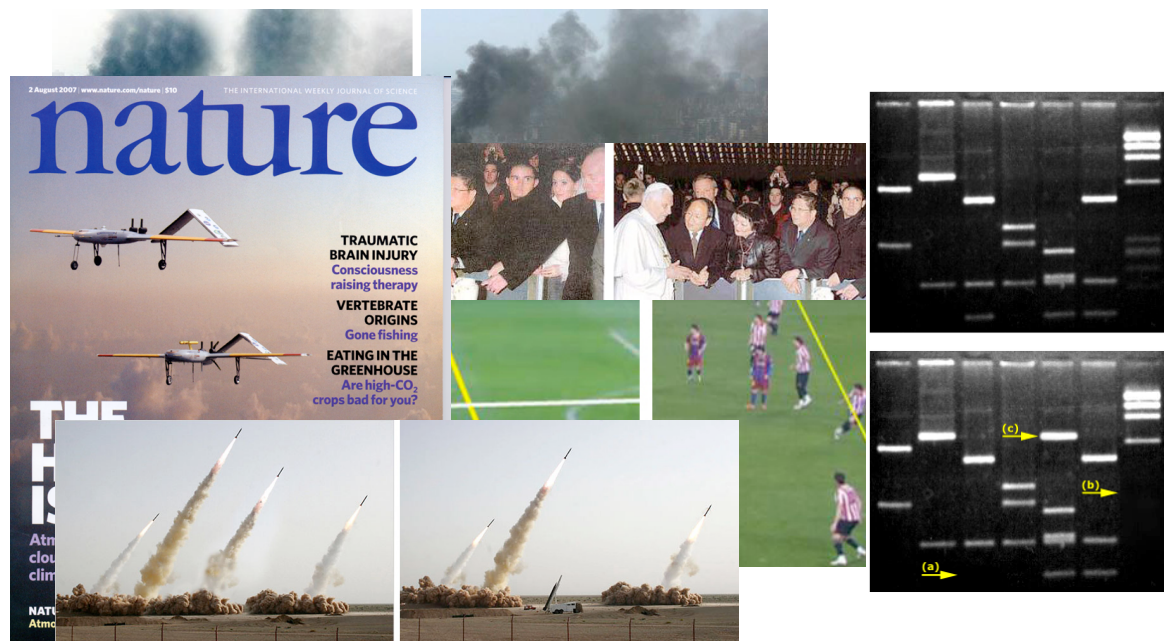


ftware



Motivations

- Diffusion of multimedia **sharing platforms**
 - Huge amount of user-generated content
- Availability of user-friendly **video-editing software**
 - Easy to tamper with videos
- **We cannot trust what we see!**



Motivations

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**TRACE BACK
FILE HISTORY**



Objectives

- **What can we do?**
 - To develop a series of **blind** algorithms and tools for **video forensic analyses** working in a real world scenario.

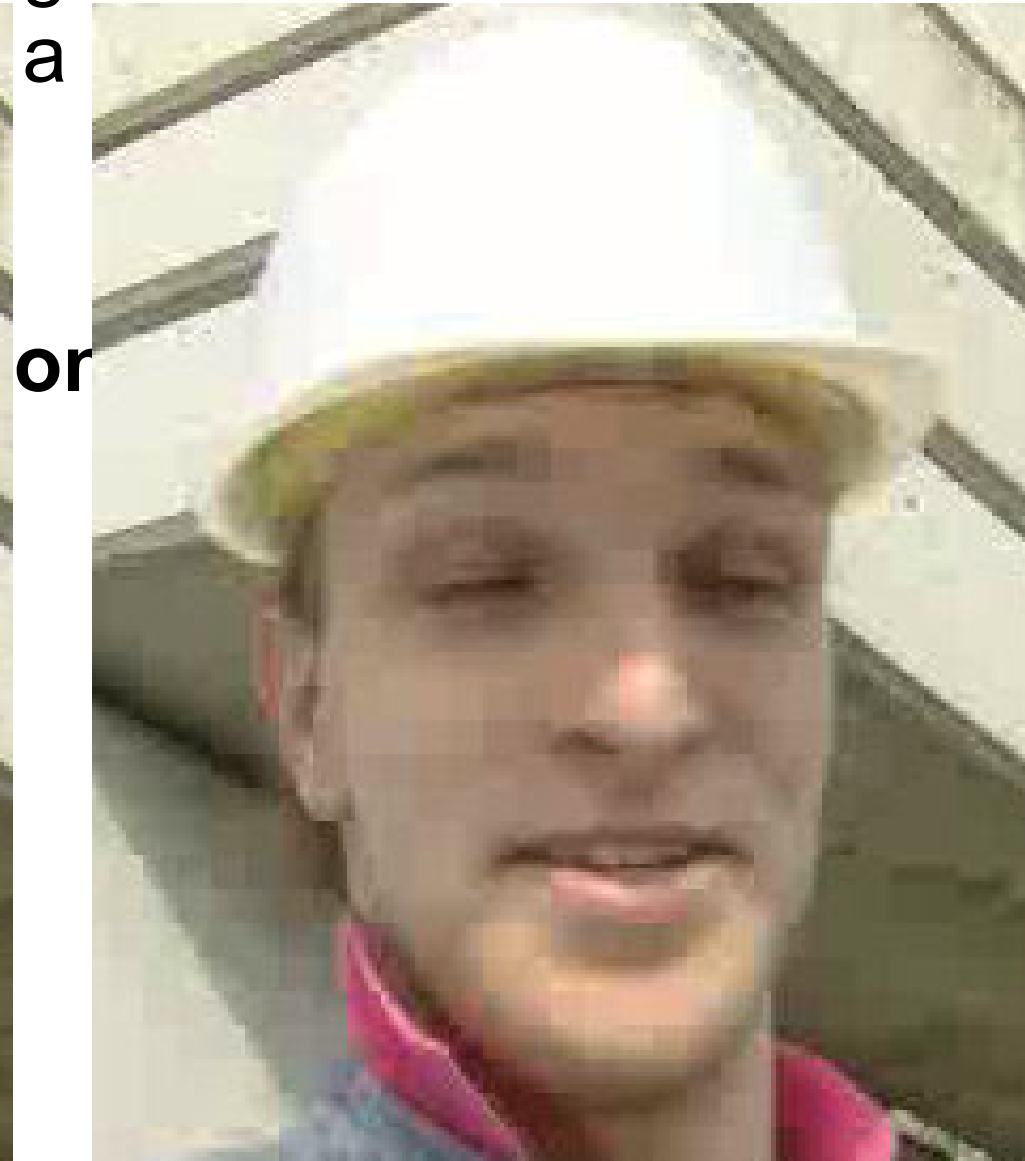
Objectives

- **What can we do?**
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- **How?**
 - Every **non-reversible operation** leaves peculiar footprints.
 - **Footprints** as an asset.

Objectives

- **What can we do?**
 - To develop a series of **blind** algorithms and tools for **video** forensics

- **How?**
 - Even
 - Focus



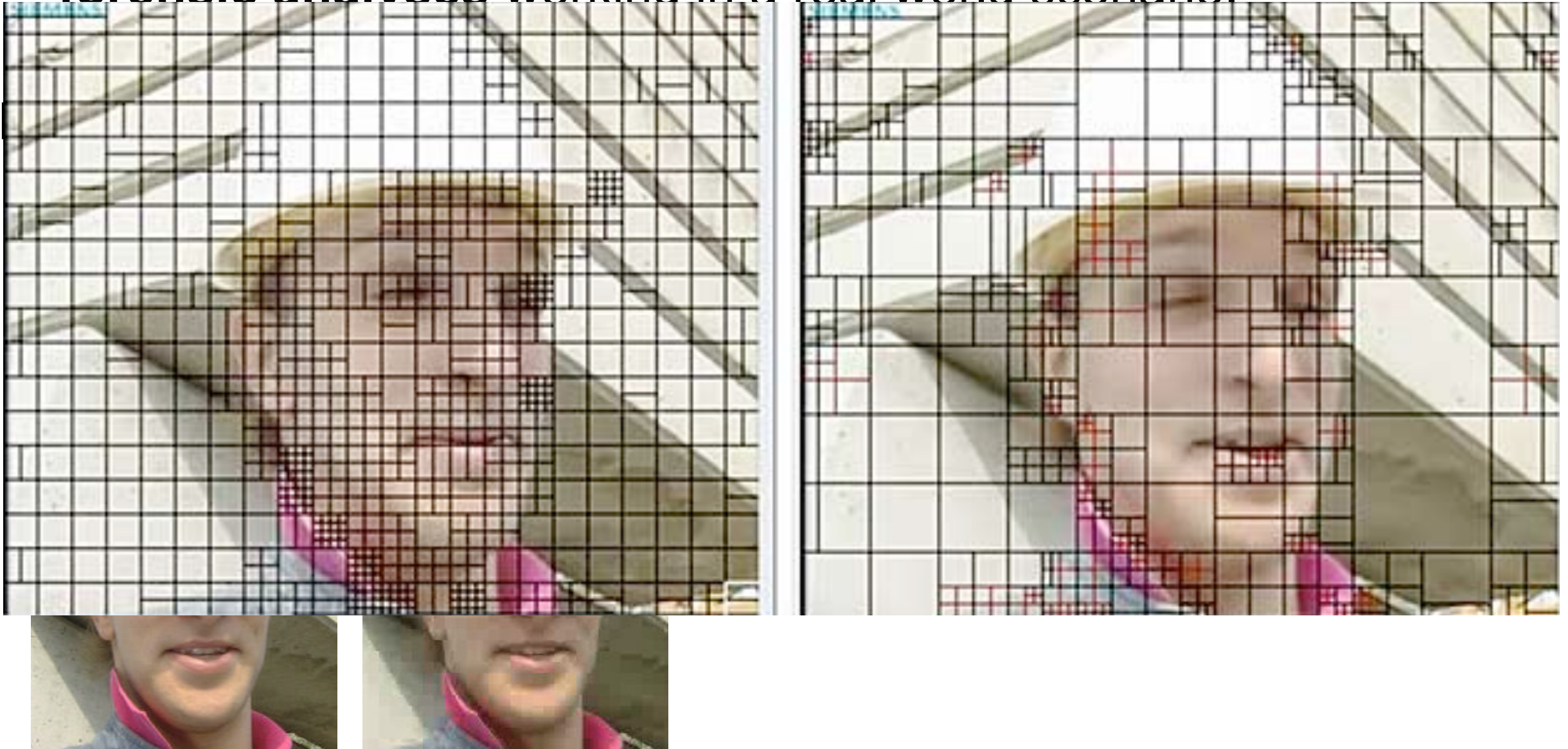
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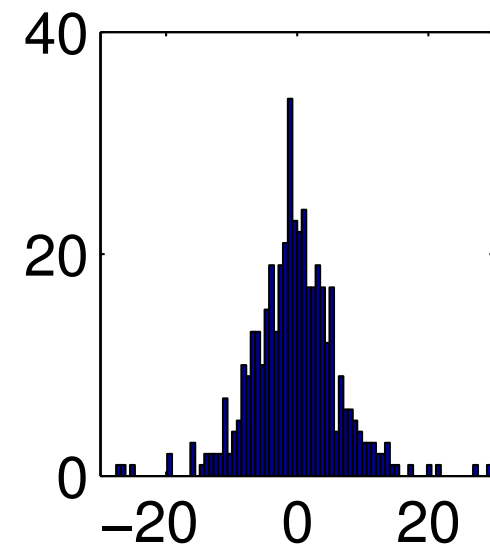
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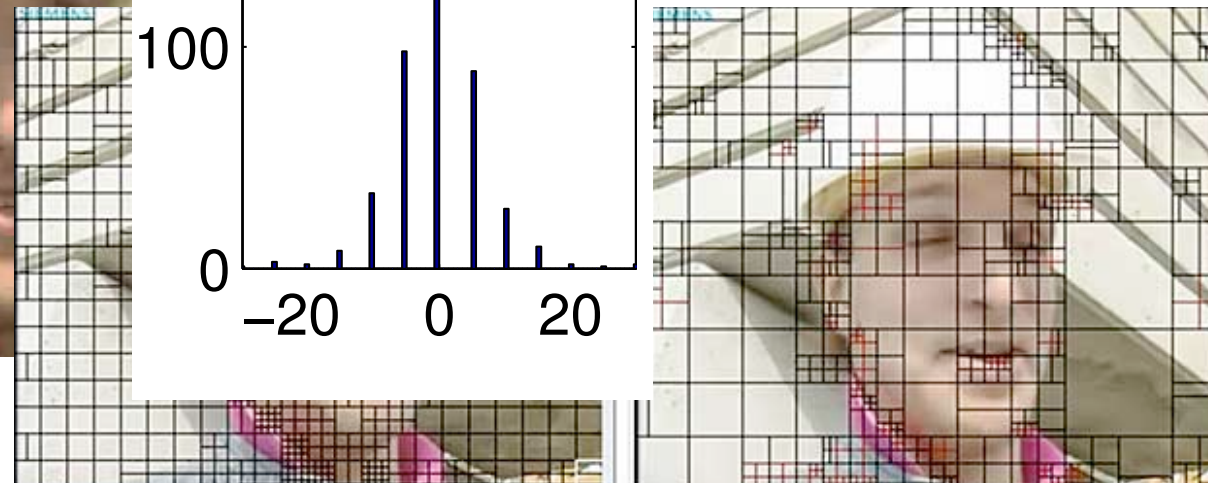
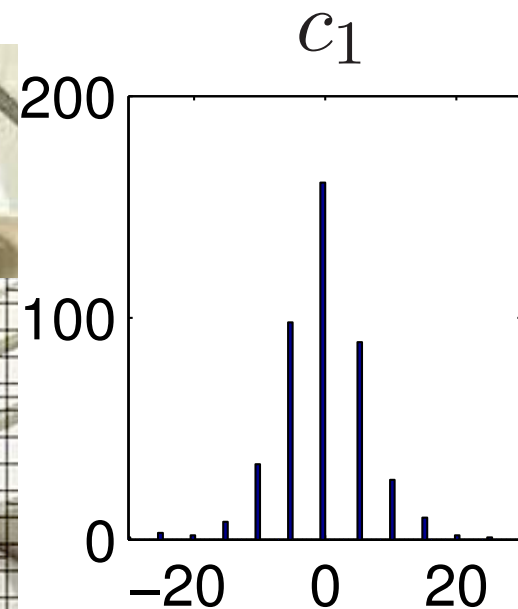
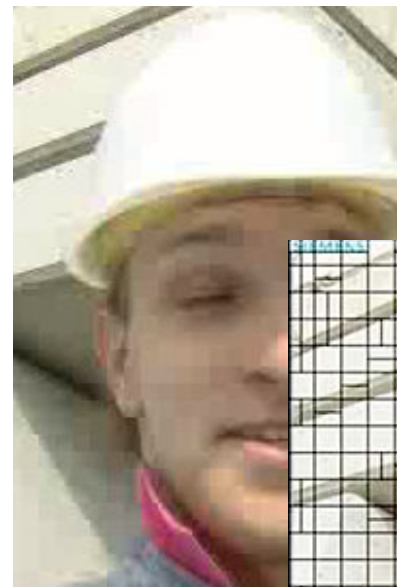
Objectives

- **What can we do?**
 - To develop a series of **biometric forensic analyses** work
- **How?**
 - Every **non-reversible** o
 - **Footprints** as an asset.

c_1 and tools for **video** world scenario.

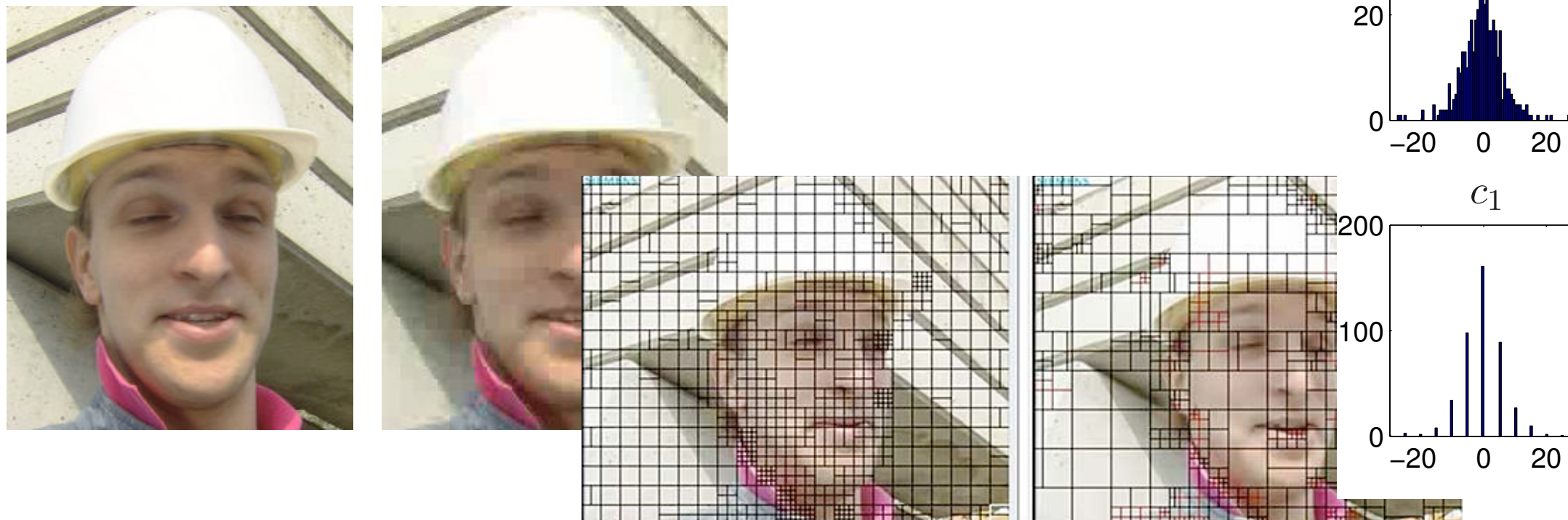


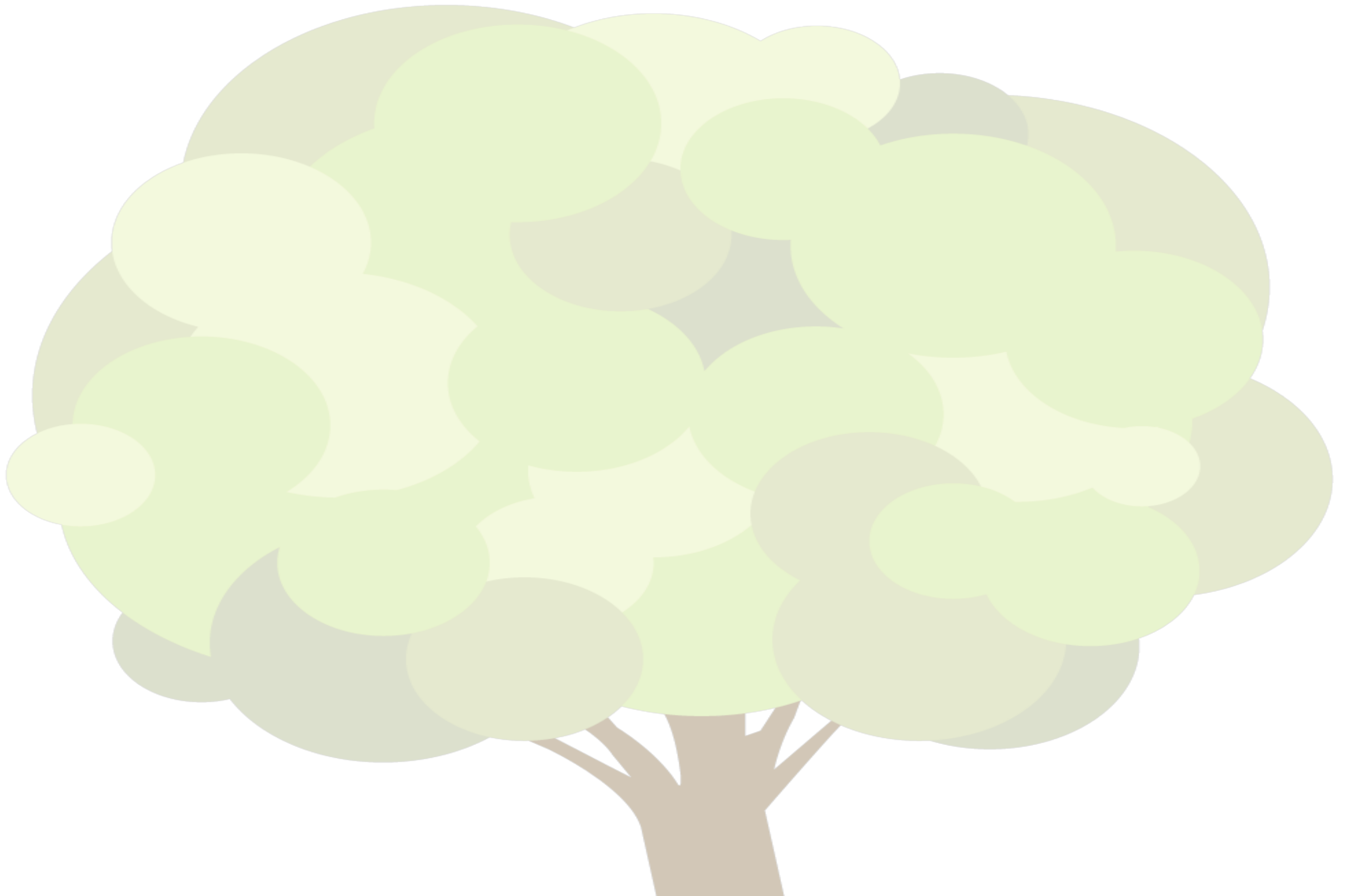
has peculiar footprints.

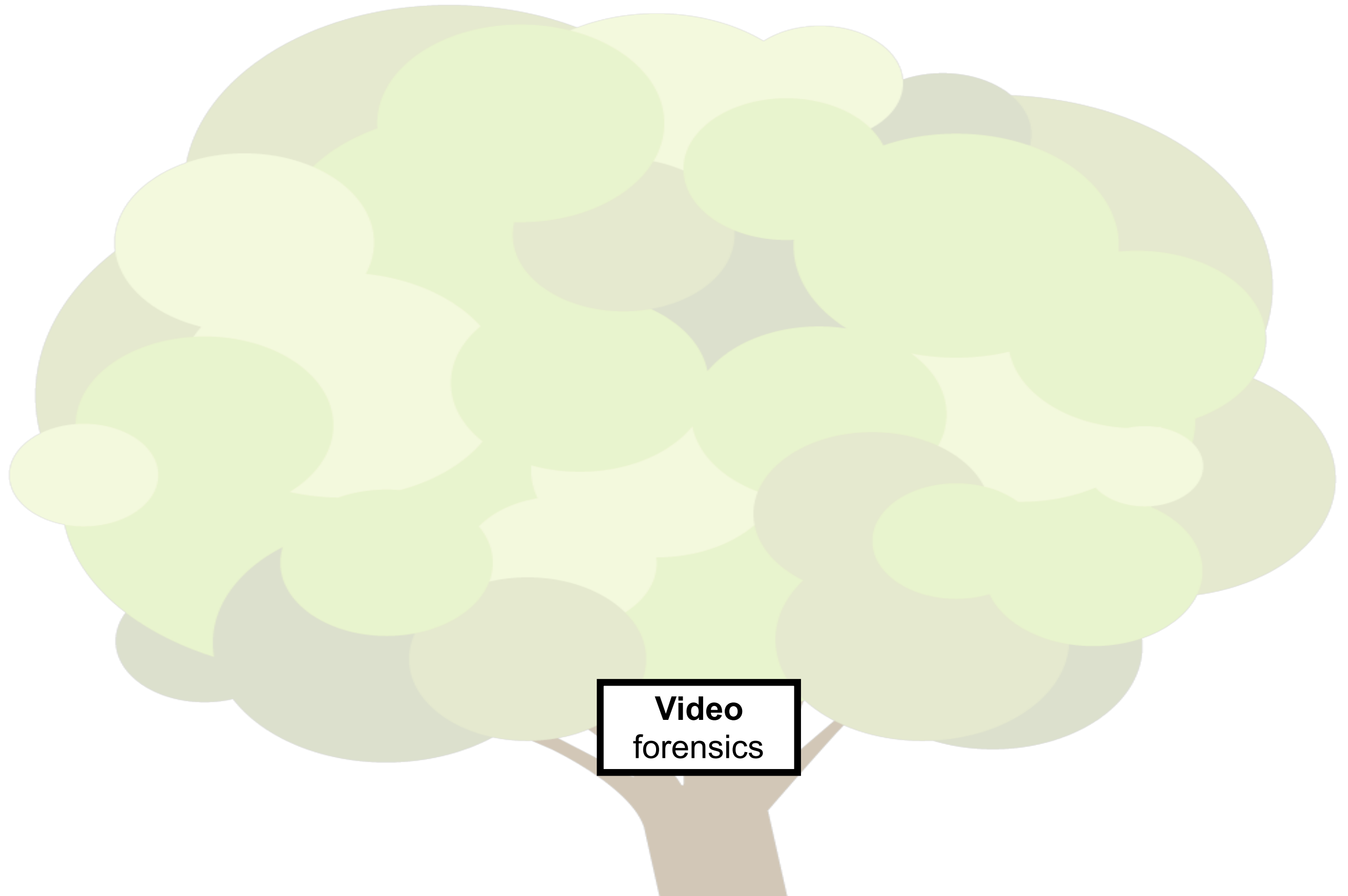


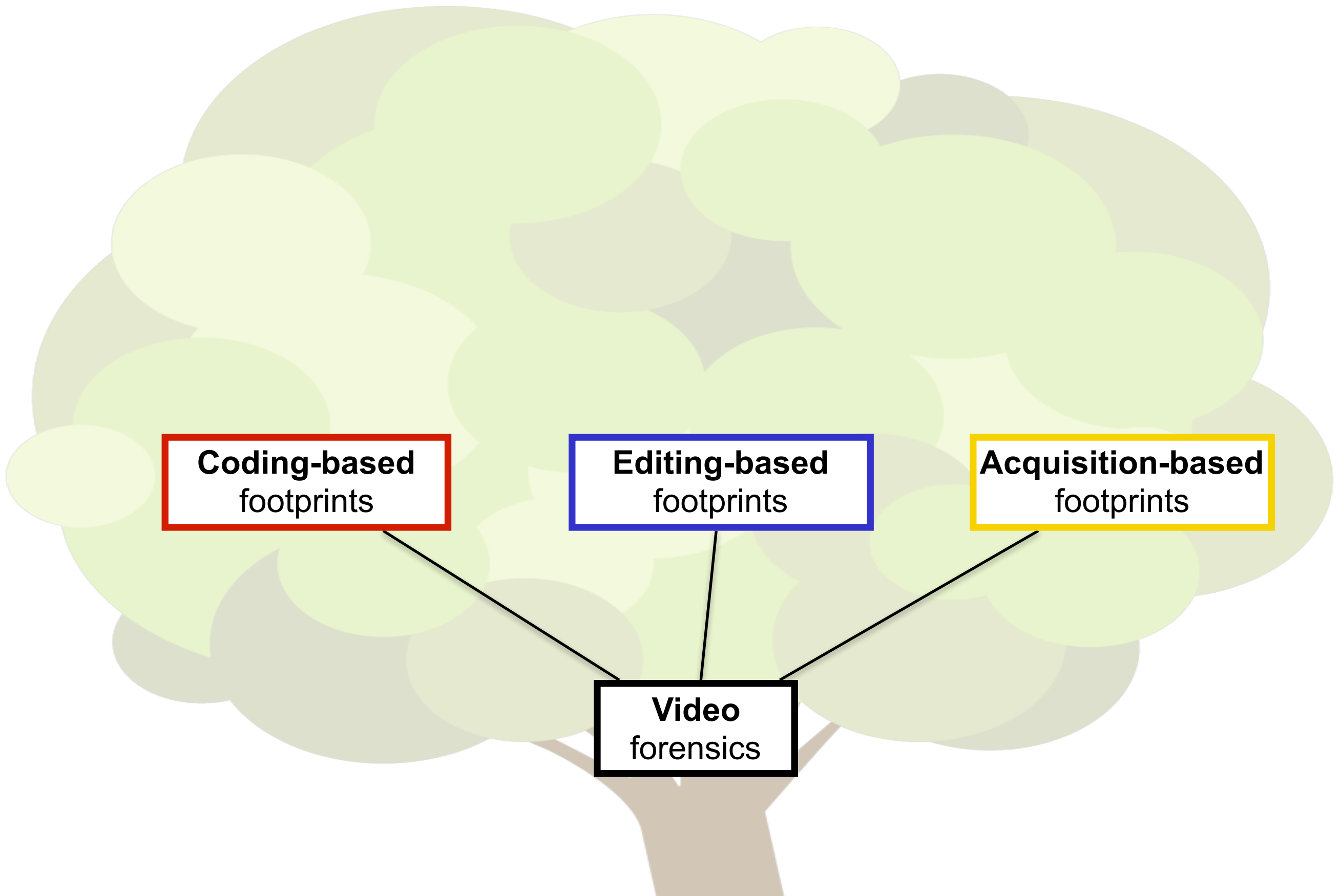
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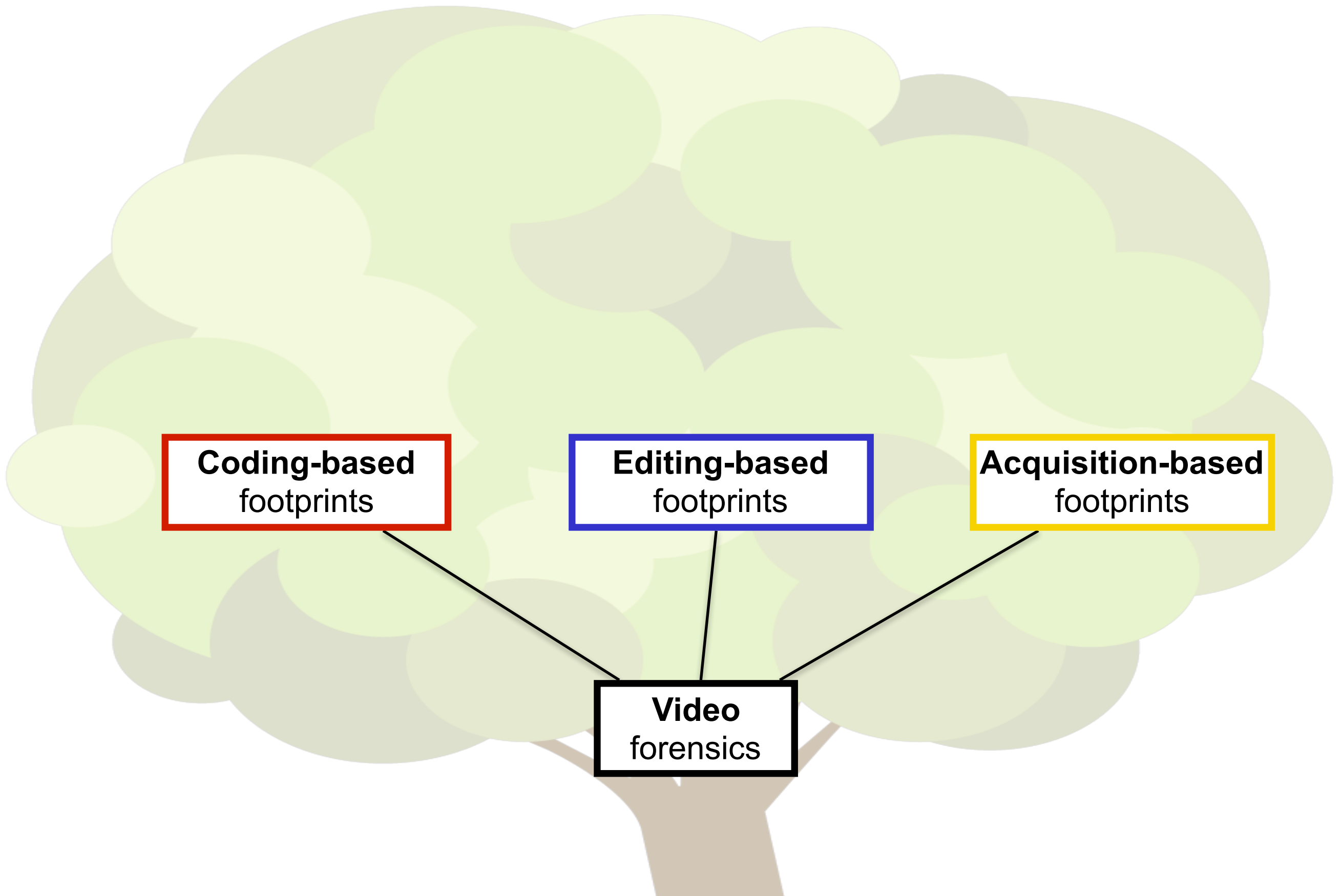
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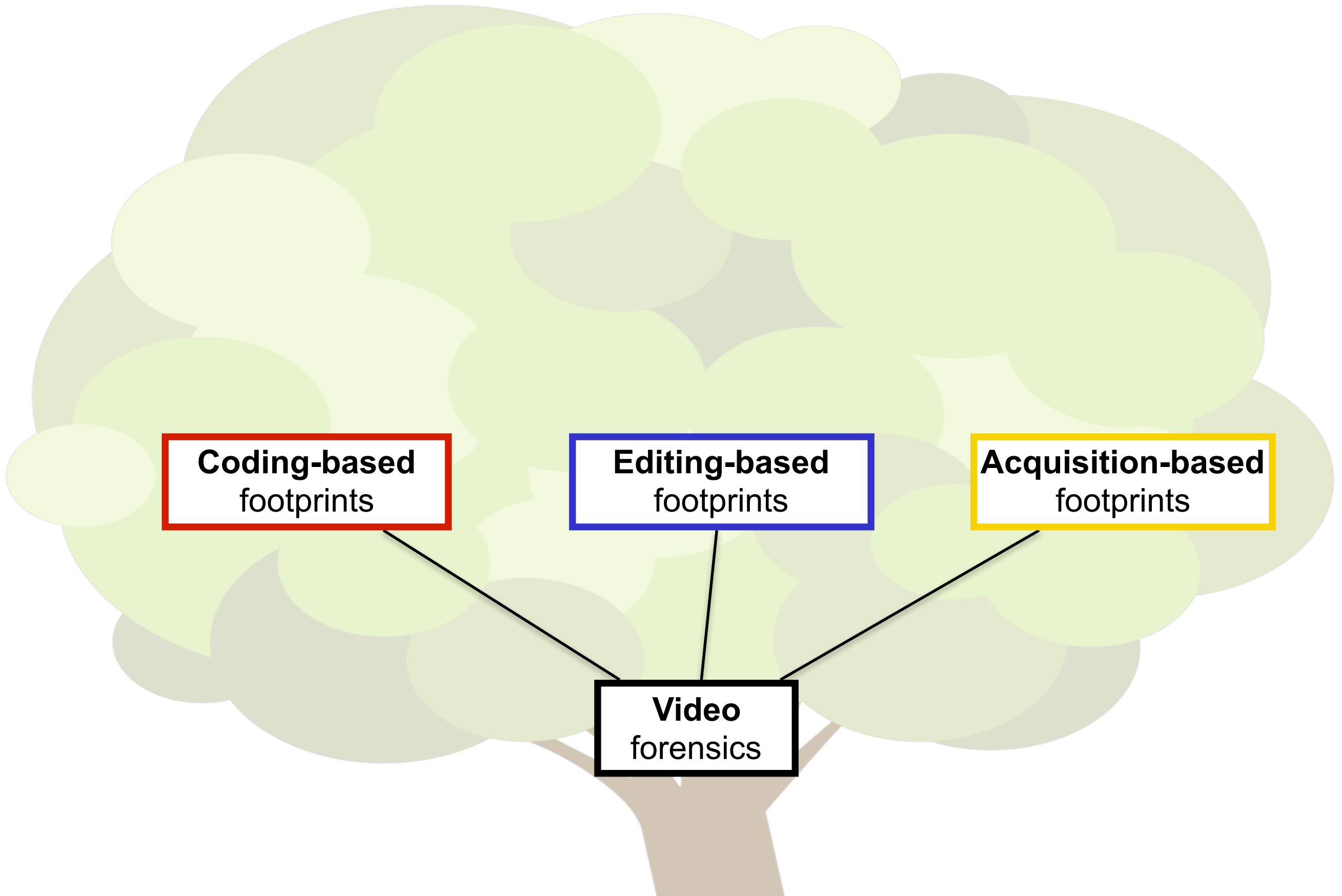


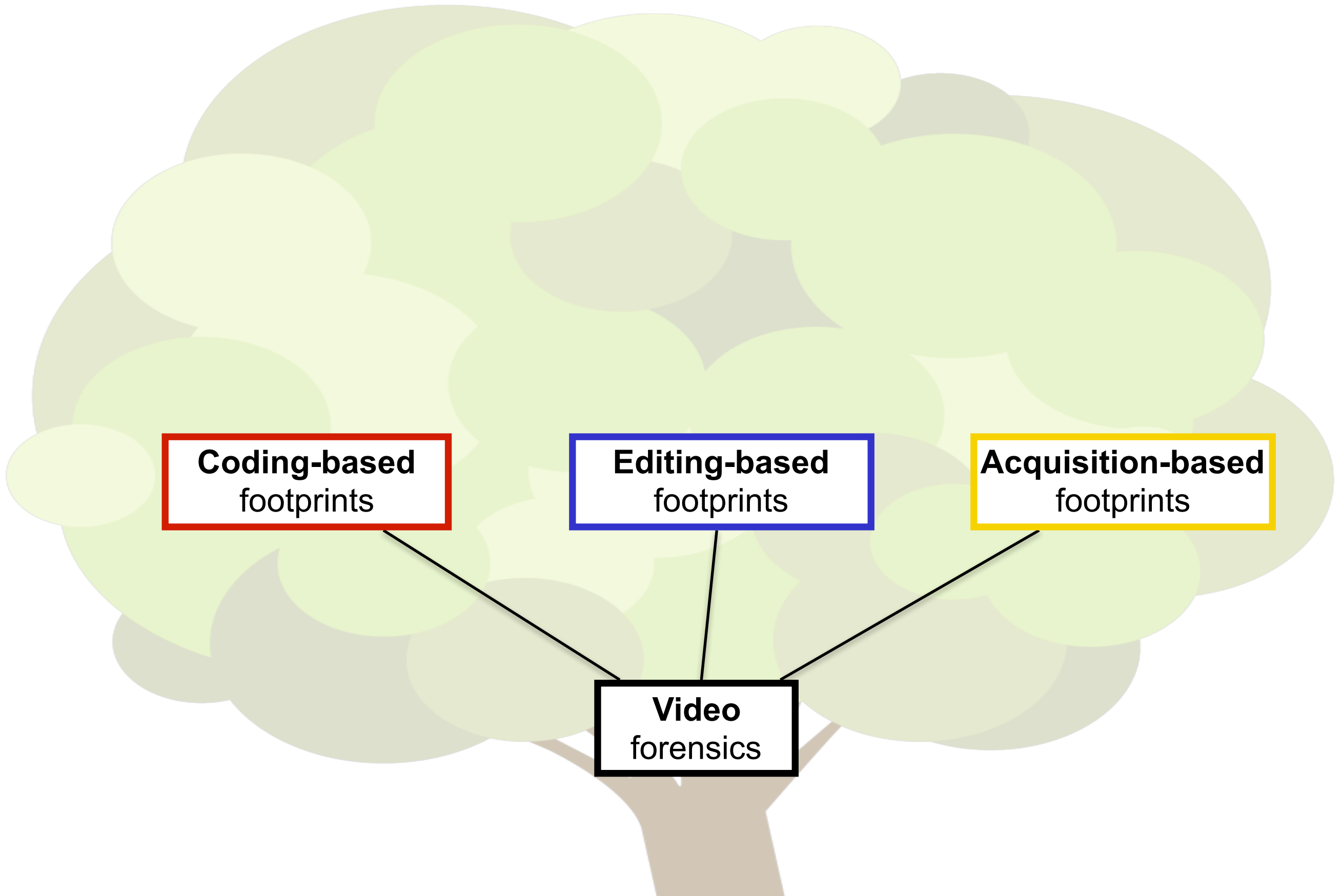


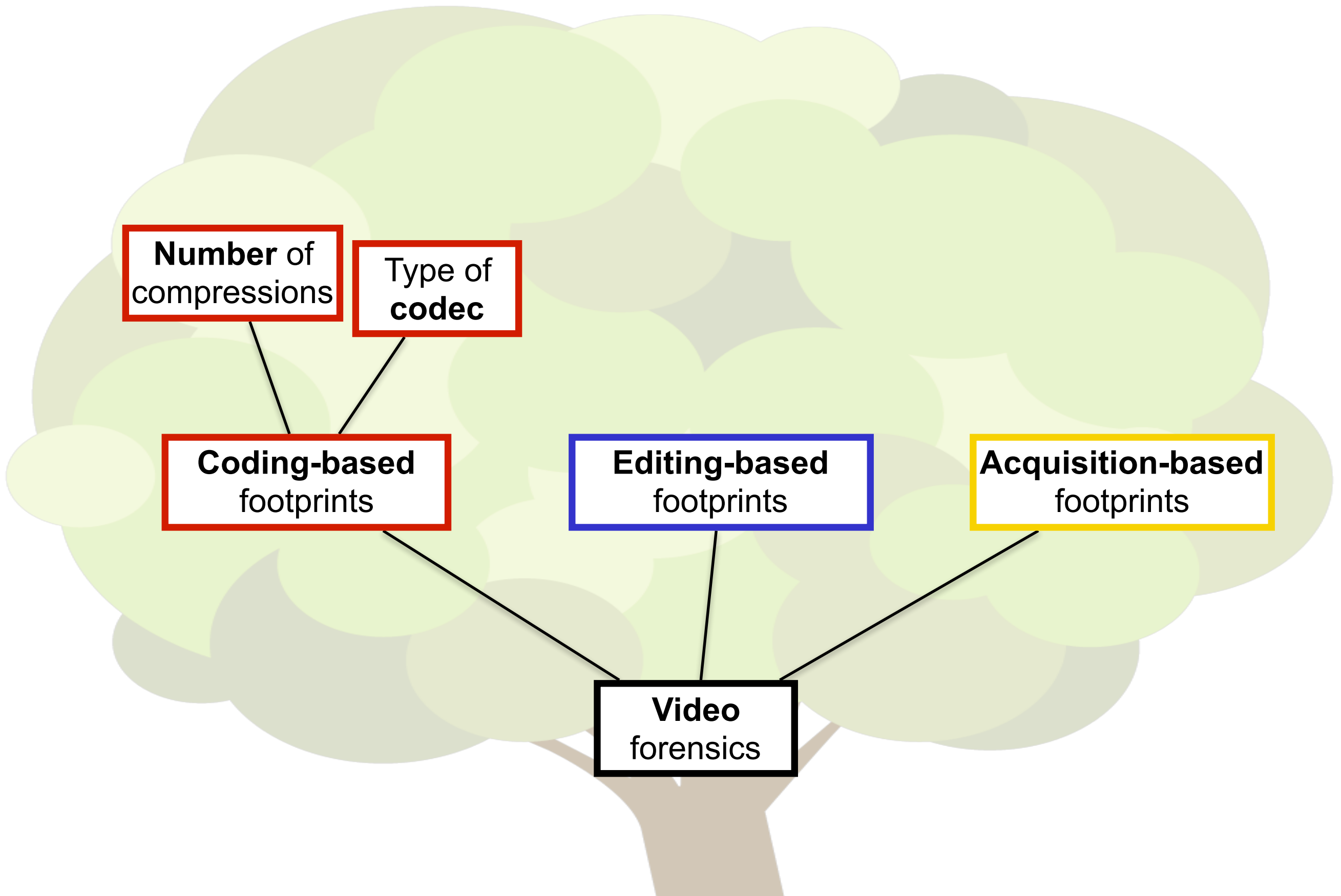


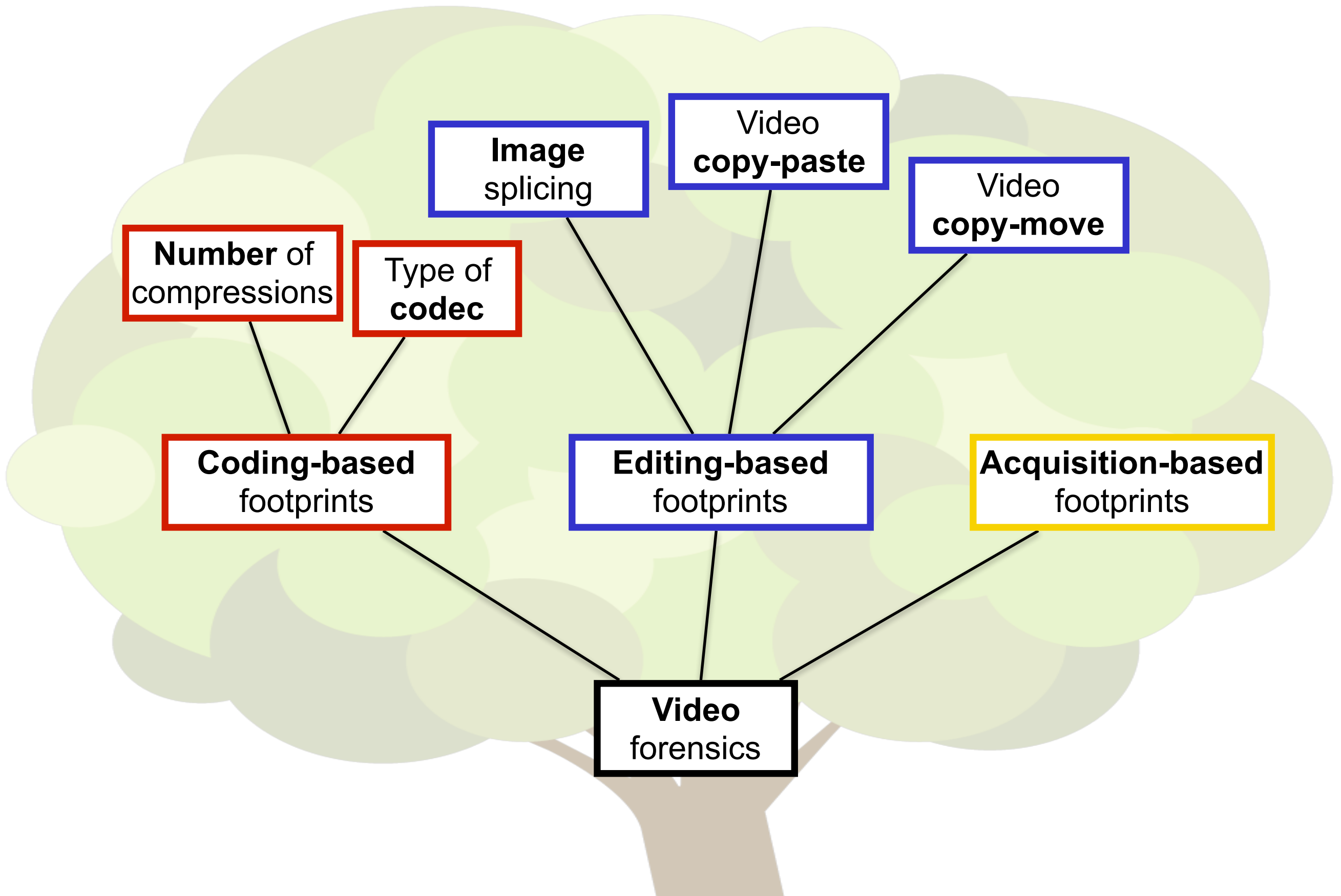


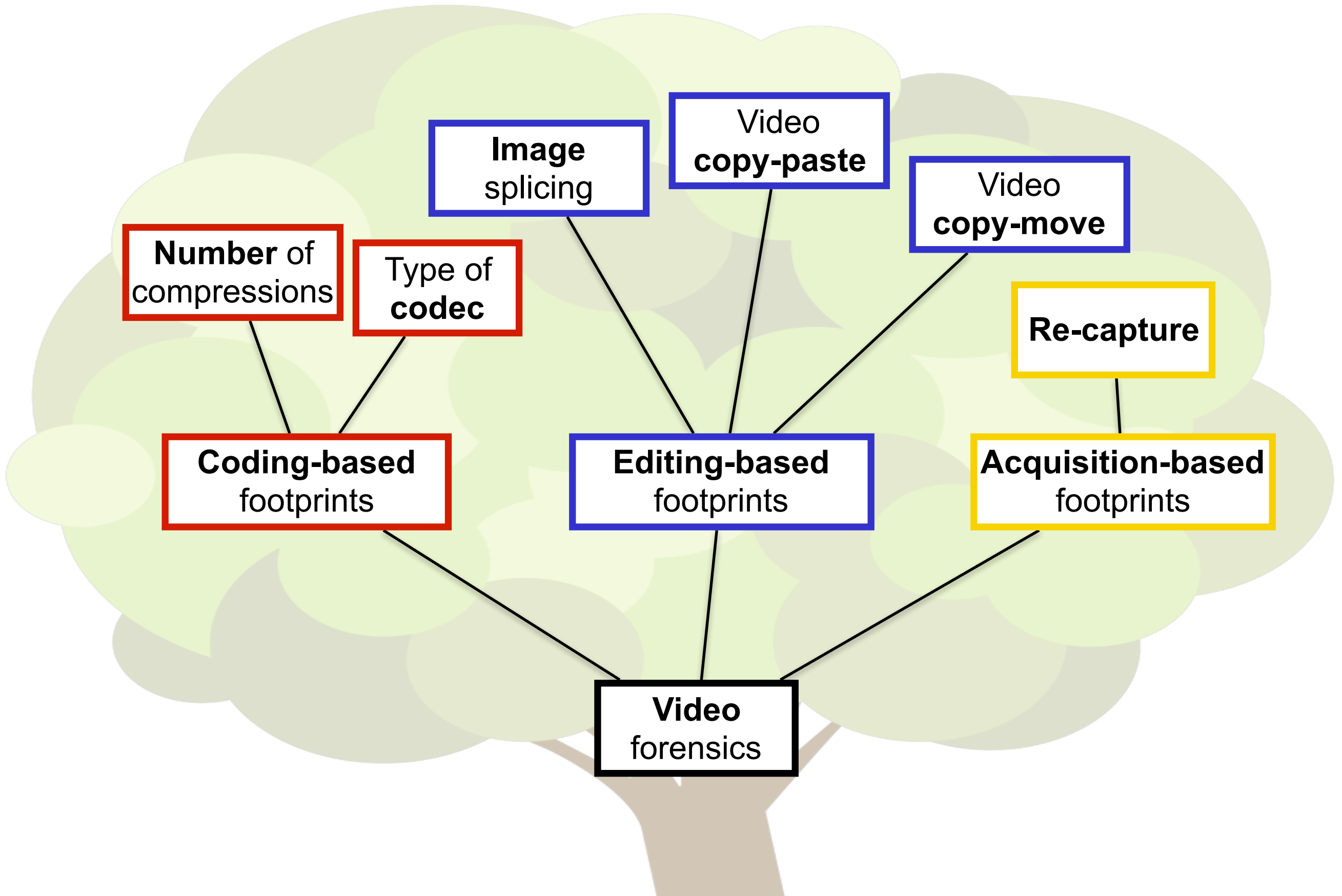




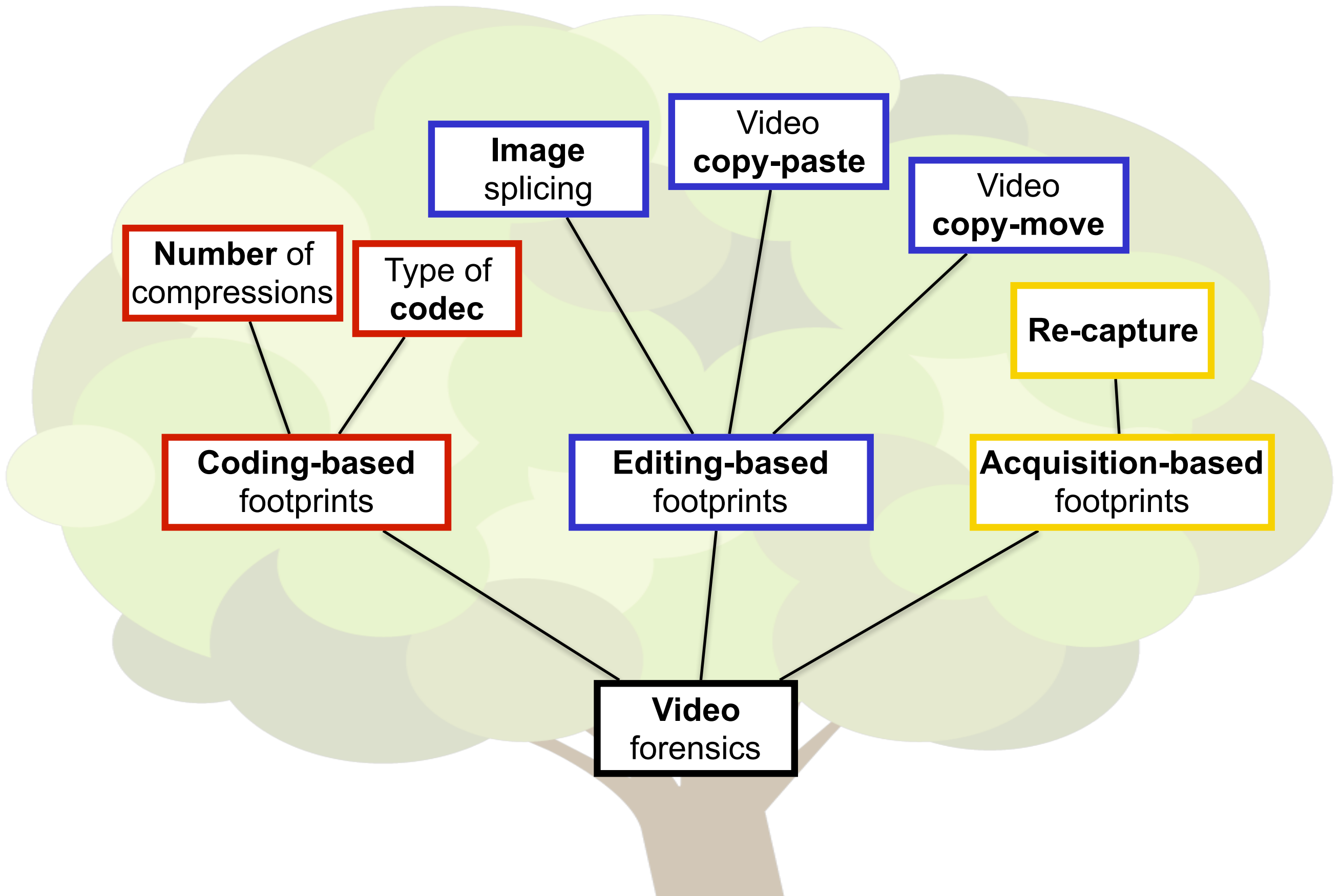








Work organisation



**Image
splicing**

**Number of
compressions**

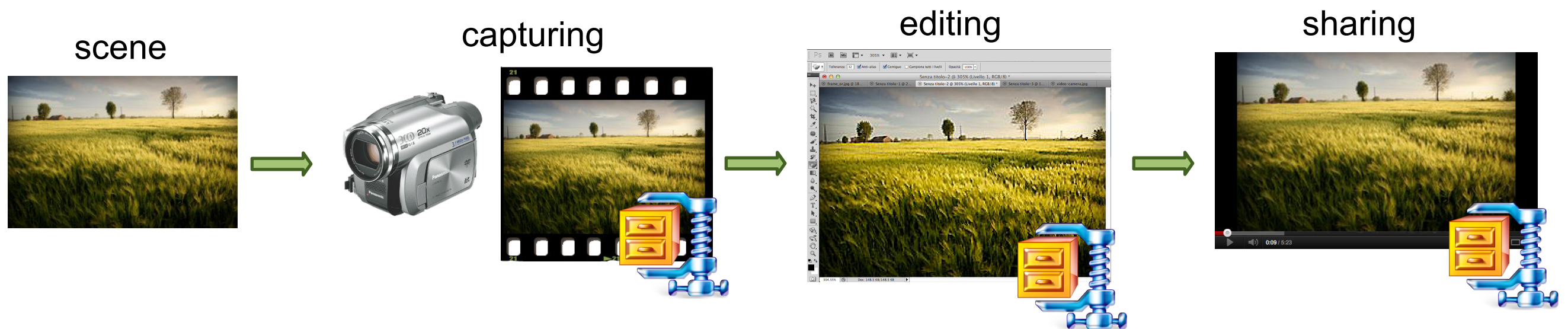
**Type of
codec**

**Coding-based
footprints**

[

Coding-based footprints

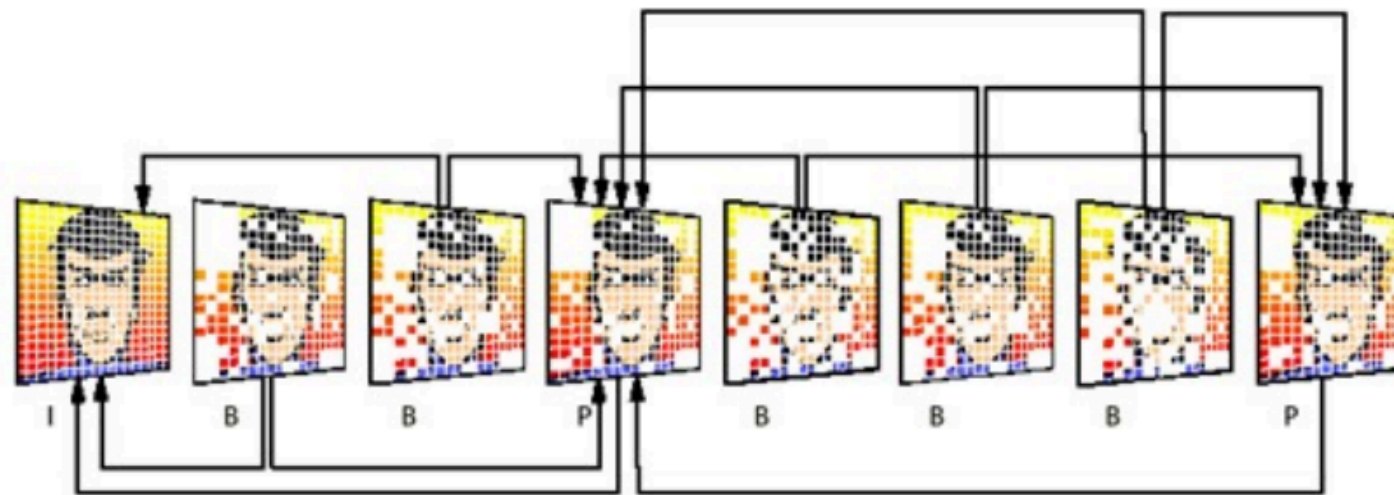
- Videos are often **encoded multiple times** during their life-time
 - Information about acquisition device
 - The number of compression steps is an indicator of a video reliability



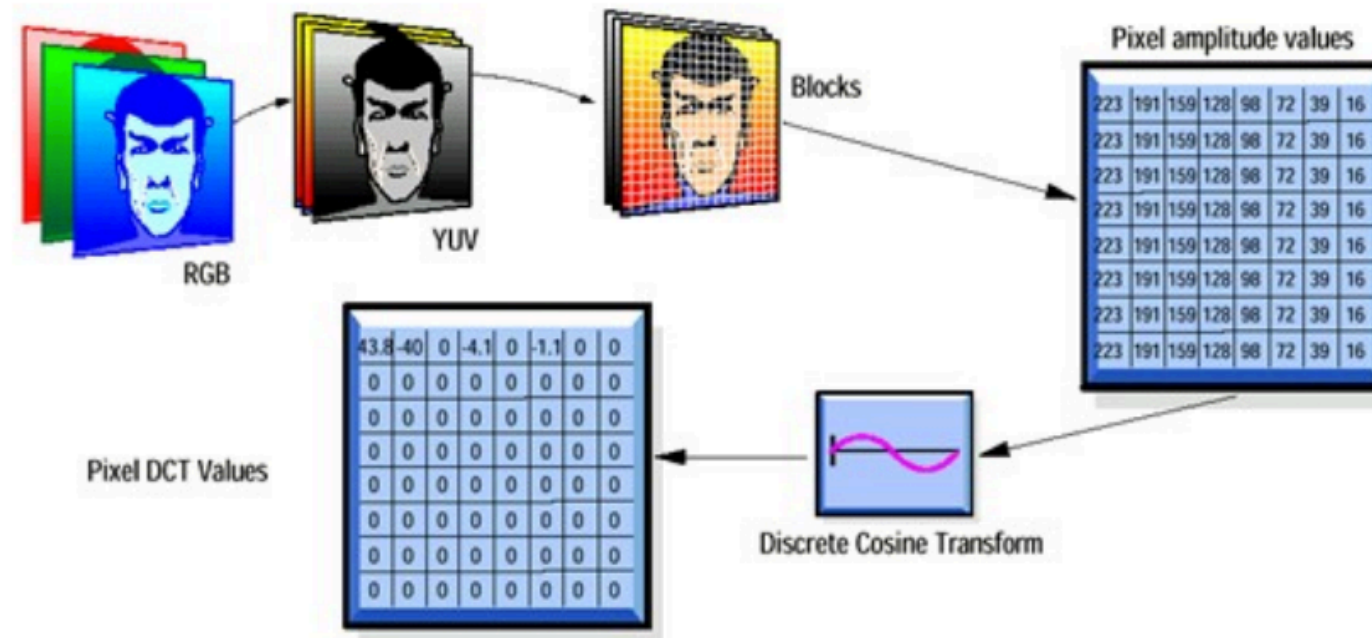
How many times has the video been compressed?

[Bestagini et al. MMSP 2012]

- Video coding:
 - Temporal redundancy



- Spatial redundancy



- **Benford's law:**

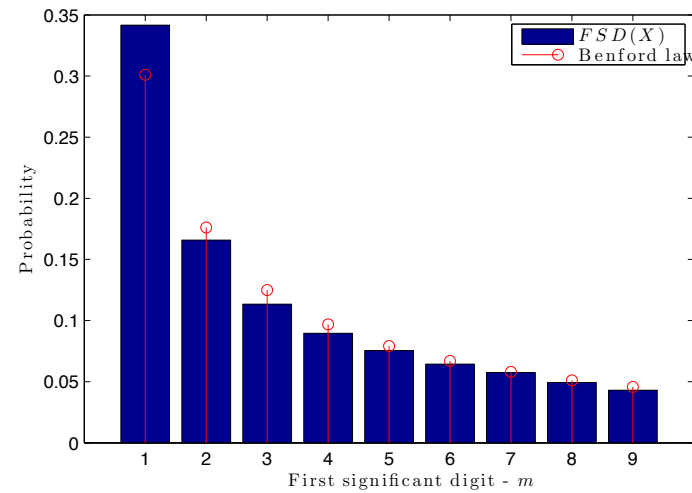
- The distribution of the first digit (FD) of a **single quantized** DCT coefficient approximatively follows Benford's law:

$$p(m) = K \log_{10} \left(1 + \frac{1}{\alpha + m^\beta} \right), \text{ with } m = 1, \dots, 9.$$

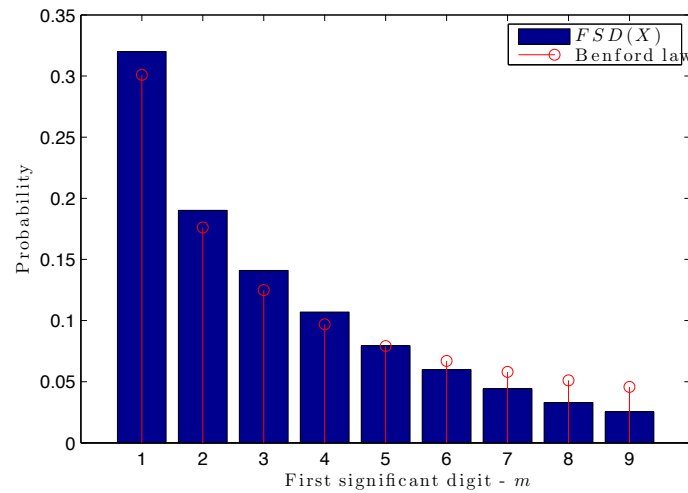
- When **multiple quantized**, this law does not hold!

- Single quantized:

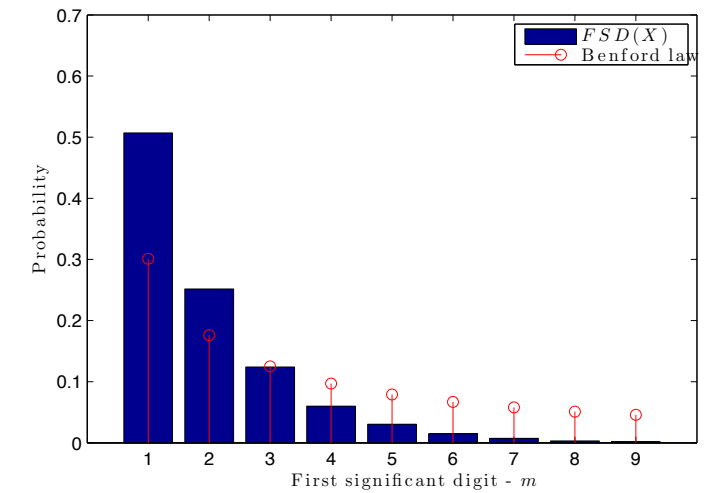
$$\Delta = 0.1$$



$$\Delta = 0.2$$

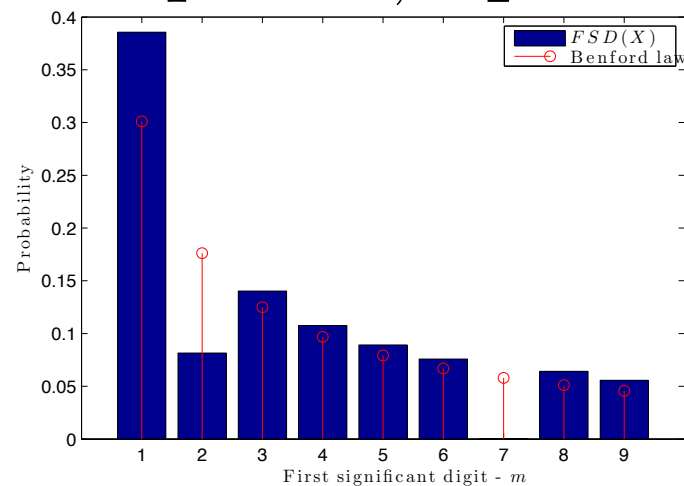


$$\Delta = 0.5$$

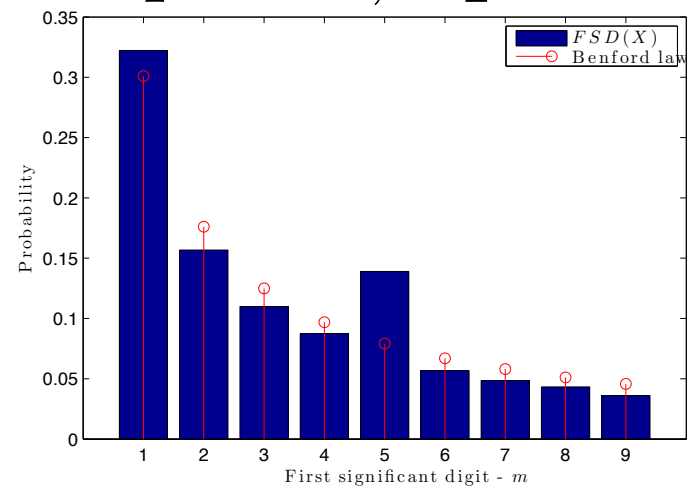


- Double quantized:

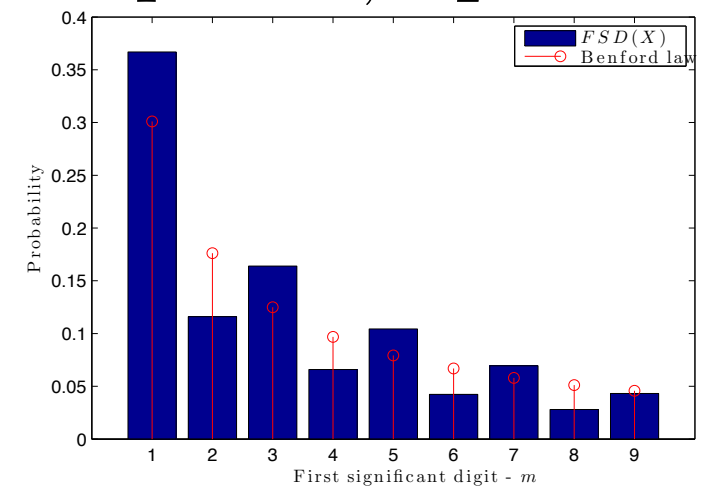
$$\Delta_1 = 0.1, \Delta_2 = 0.8$$



$$\Delta_1 = 0.1, \Delta_2 = 0.11$$

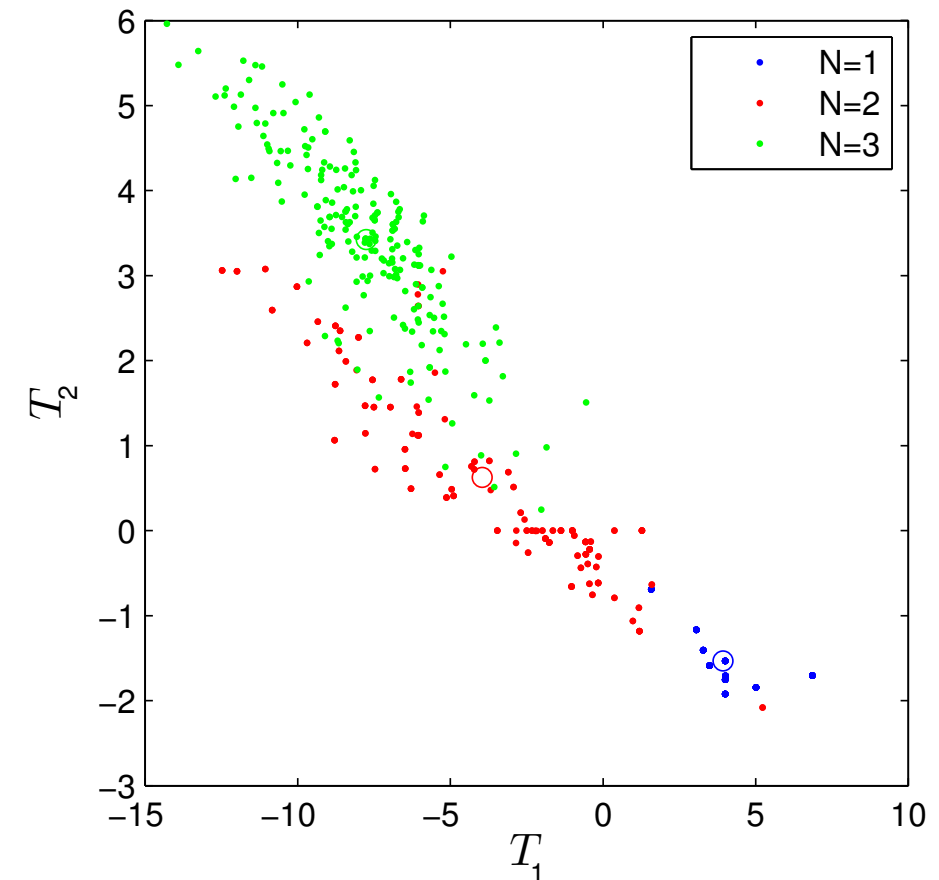


$$\Delta_1 = 0.1, \Delta_2 = 0.15$$



- **Method 1:**

- Compute FD histograms for a set of DCT frequencies.
- Train a set of SVMs.
- Combine SVMs outputs.



- **Results:**

- Up to three compressions successfully detected

N, N^*	1	2	3
1	100 %	0.00 %	0.00 %
2	0.00 %	73.89 %	26.11 %
3	0.00 %	22.22 %	77.78 %

Which codec has been used to encode a double-compressed video?

[Bestagini et al. ICASSP 2012]

[Bestagini et al. EUVIP 2013]

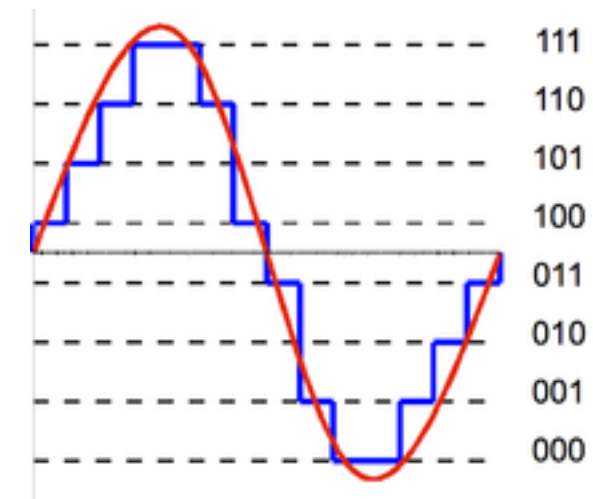
[Bestagini et al. TIP 2016]

- **Idempotency property:**

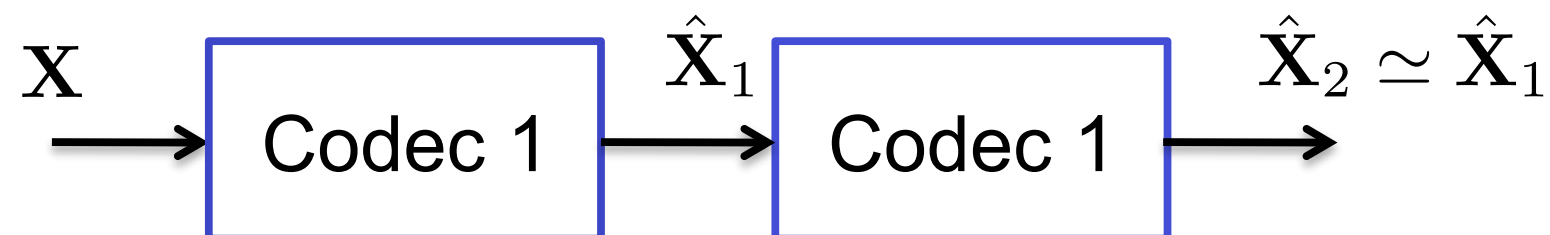
- if we re-quantize an already quantized signal with the same quantization step, the signal does not change

$$\hat{\mathbf{X}}_1 = Q_{\Delta_1}(\mathbf{X}) = \Delta_1 \left\lfloor \frac{\mathbf{X}}{\Delta_1} \right\rfloor$$

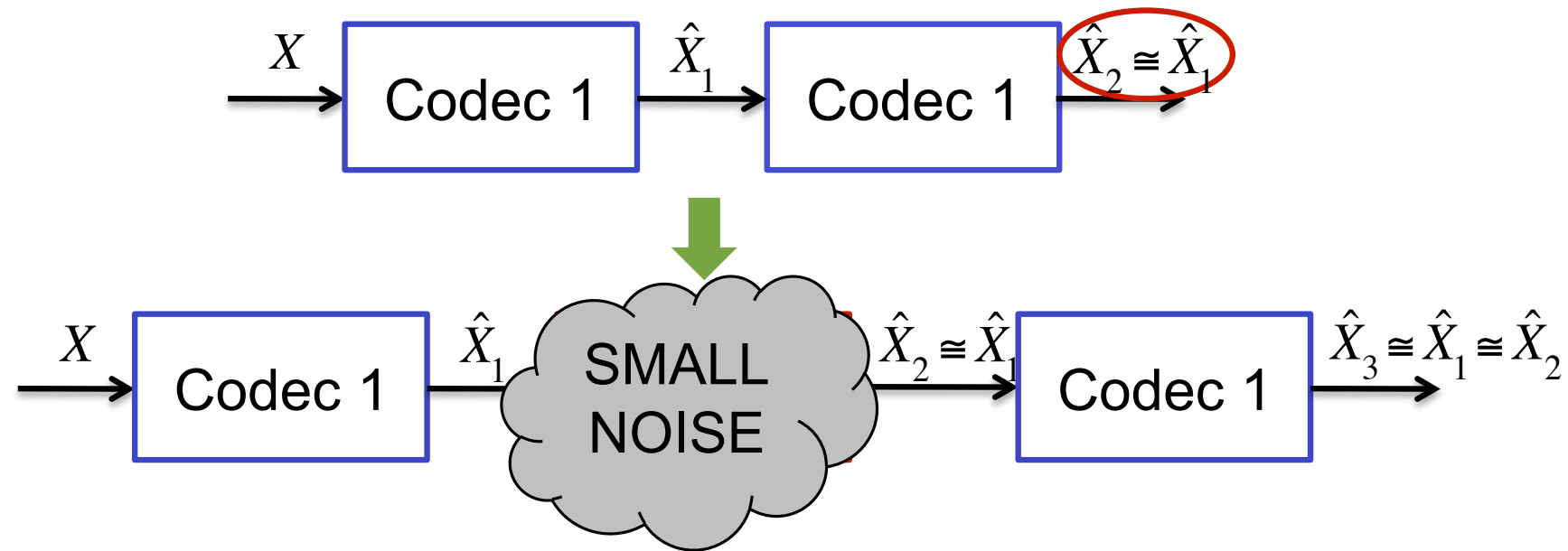
$$\hat{\mathbf{X}}_2 = Q_{\Delta_2}(\hat{\mathbf{X}}_1) = \Delta_1 \left\lfloor \frac{\Delta_1 \left\lfloor \frac{\mathbf{X}}{\Delta_1} \right\rfloor}{\Delta_1} \right\rfloor = \hat{\mathbf{X}}_1$$



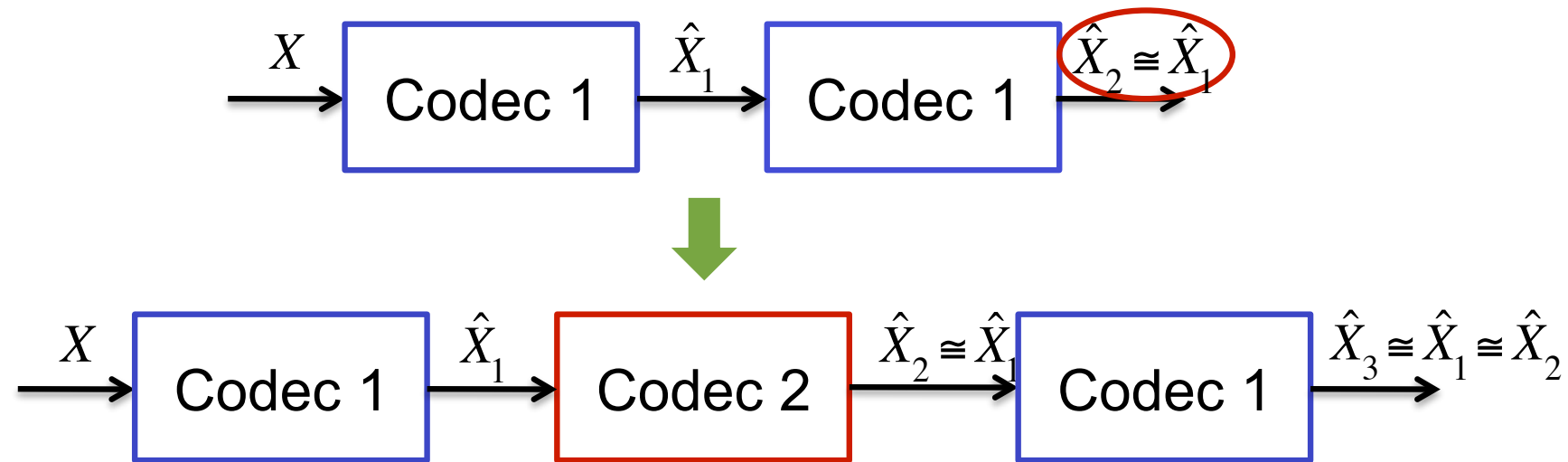
- This is partly true also for video codecs



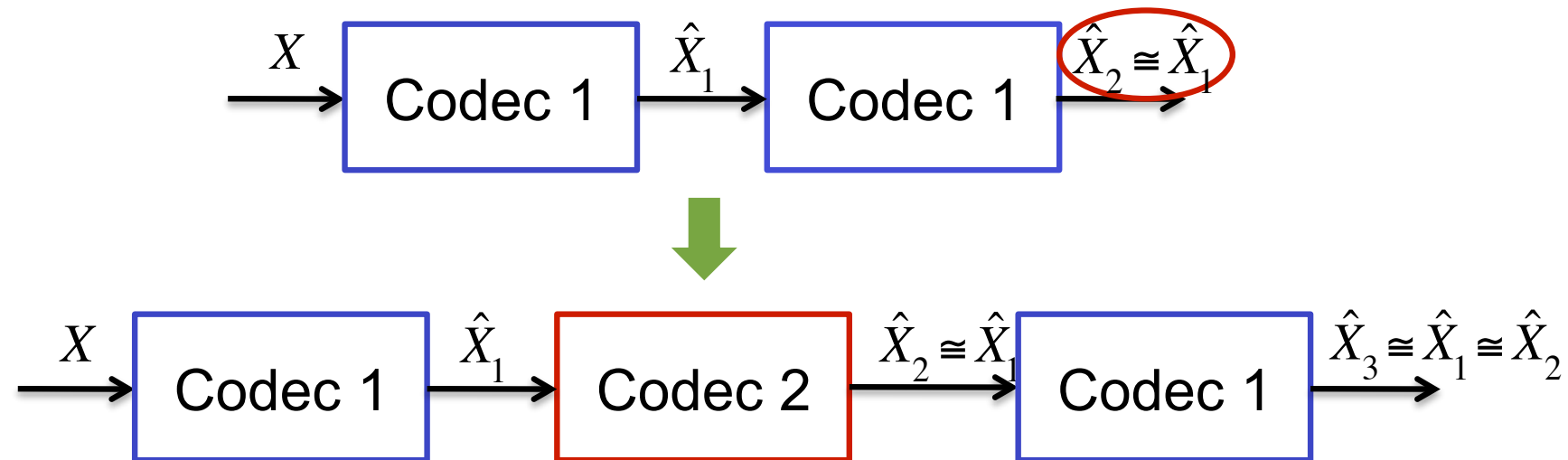
- Main idea:



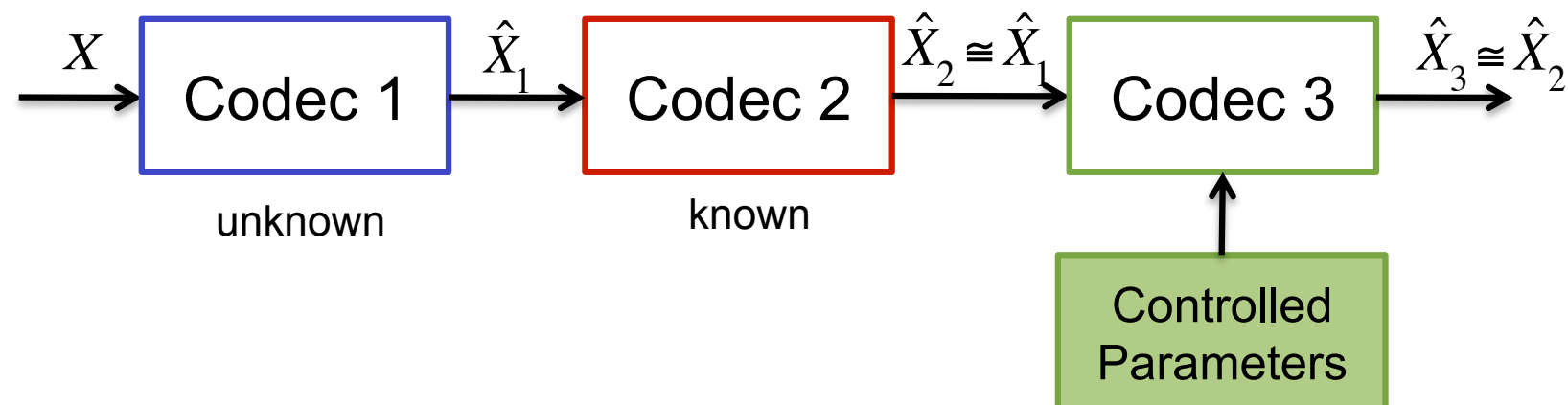
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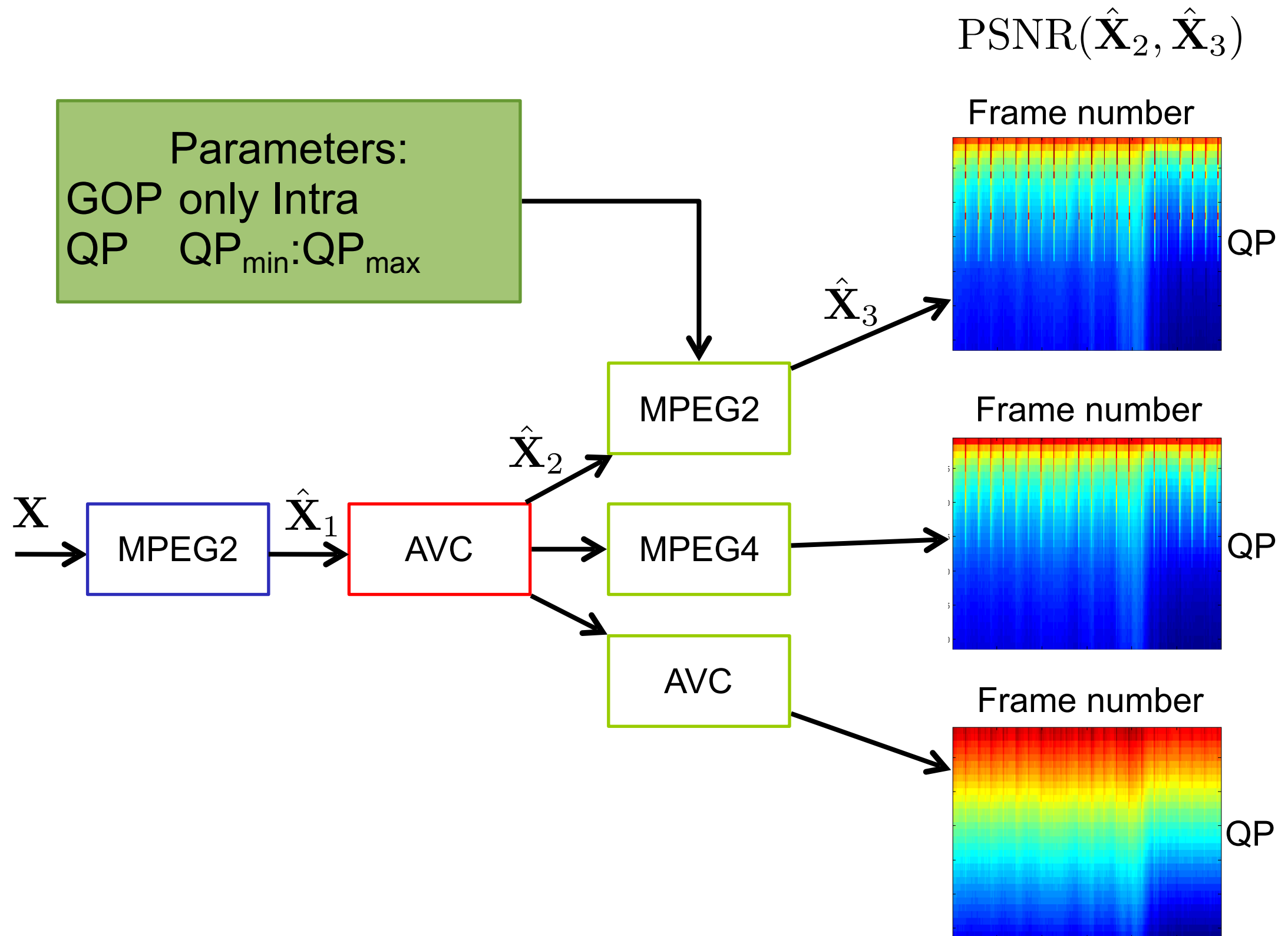


- Main idea:

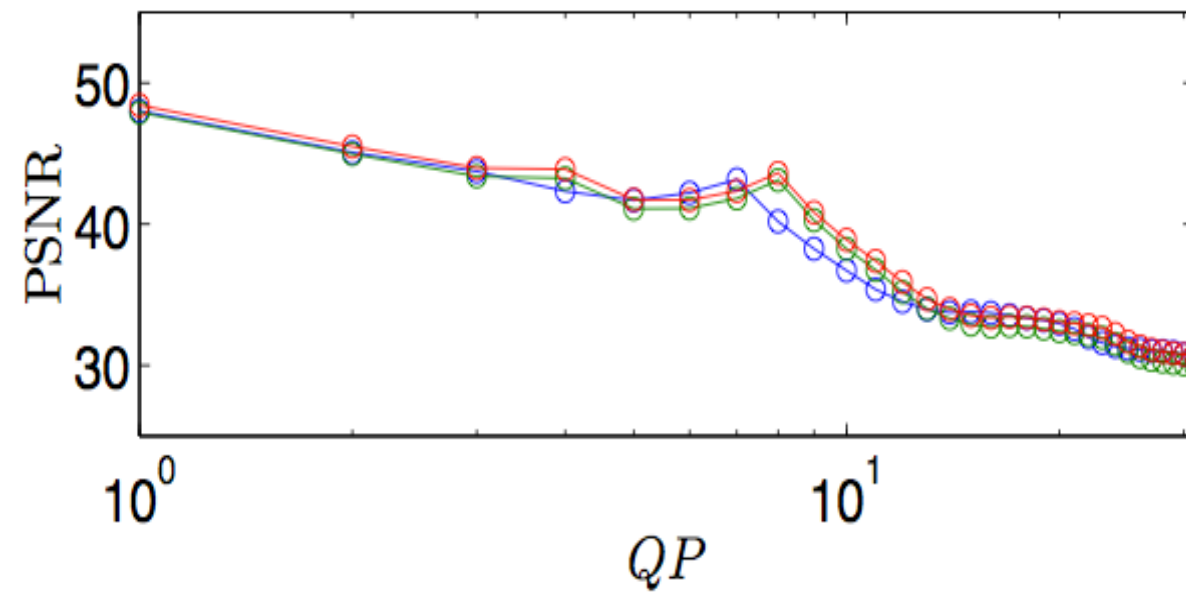
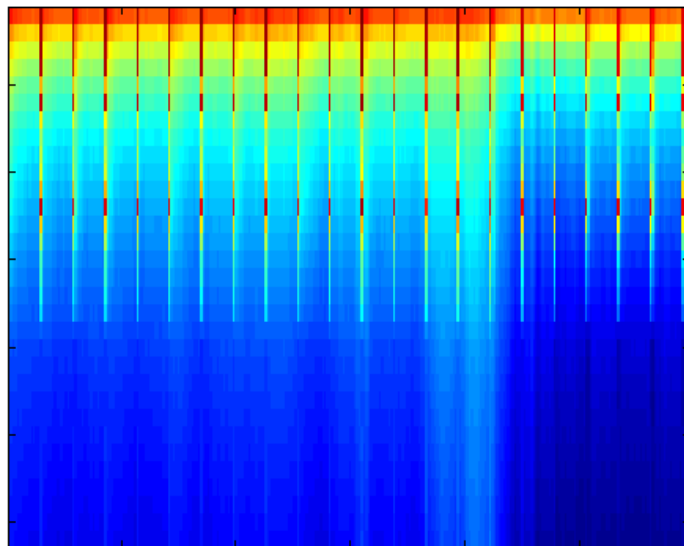


- Approach:

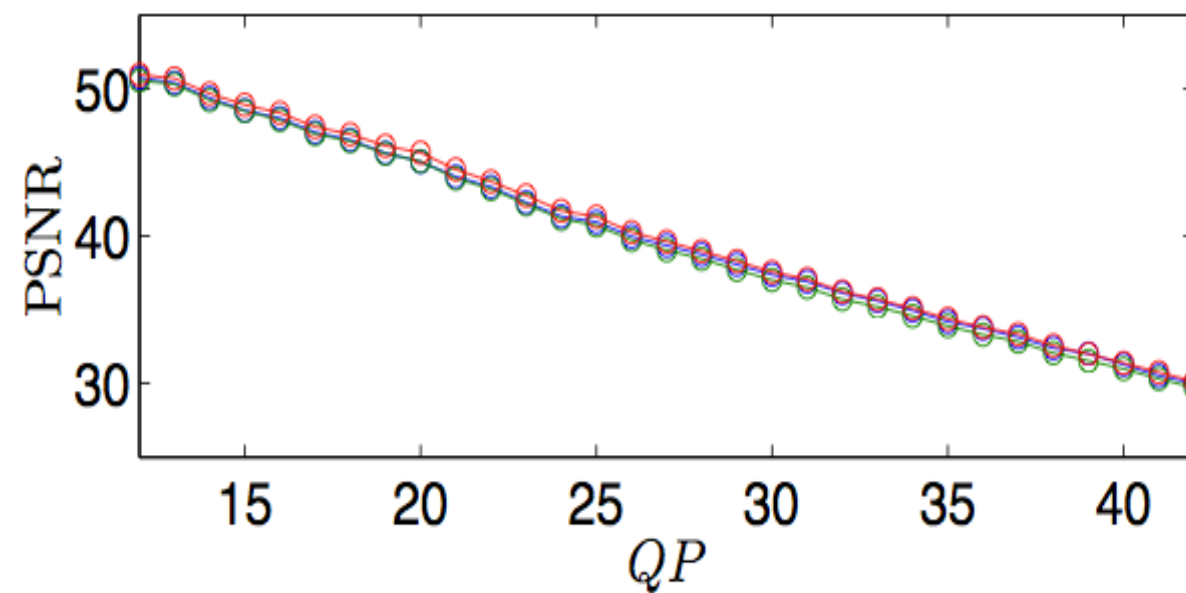
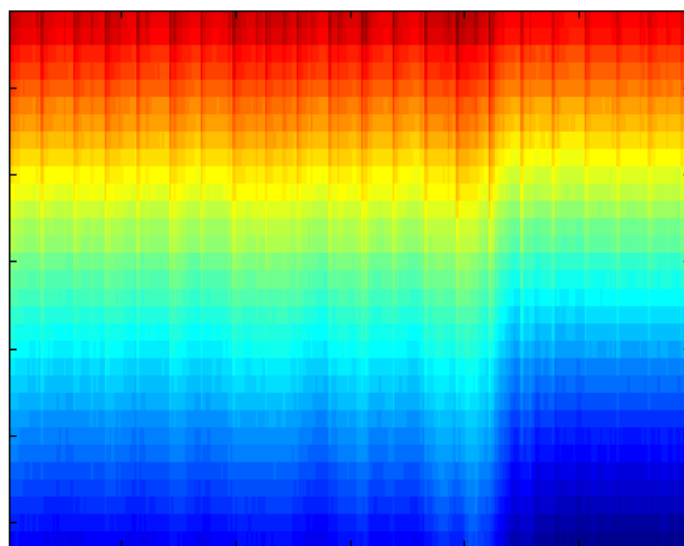




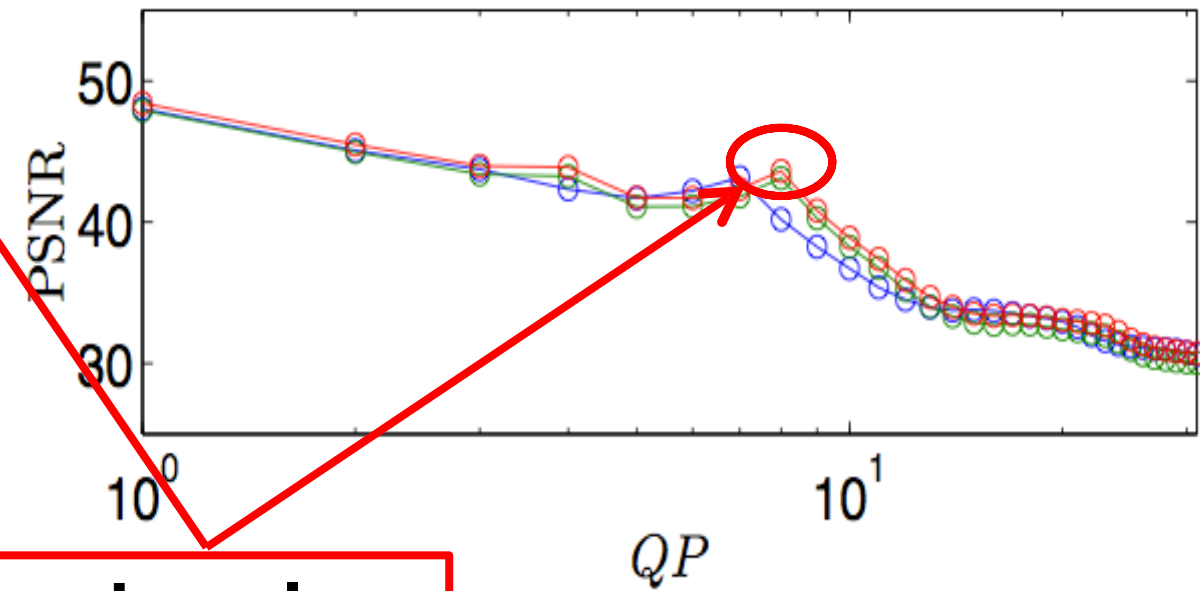
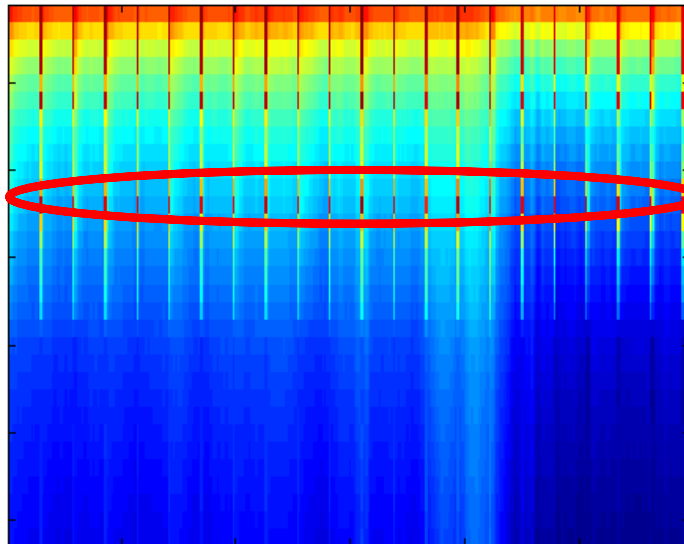
Correct



Incorrect

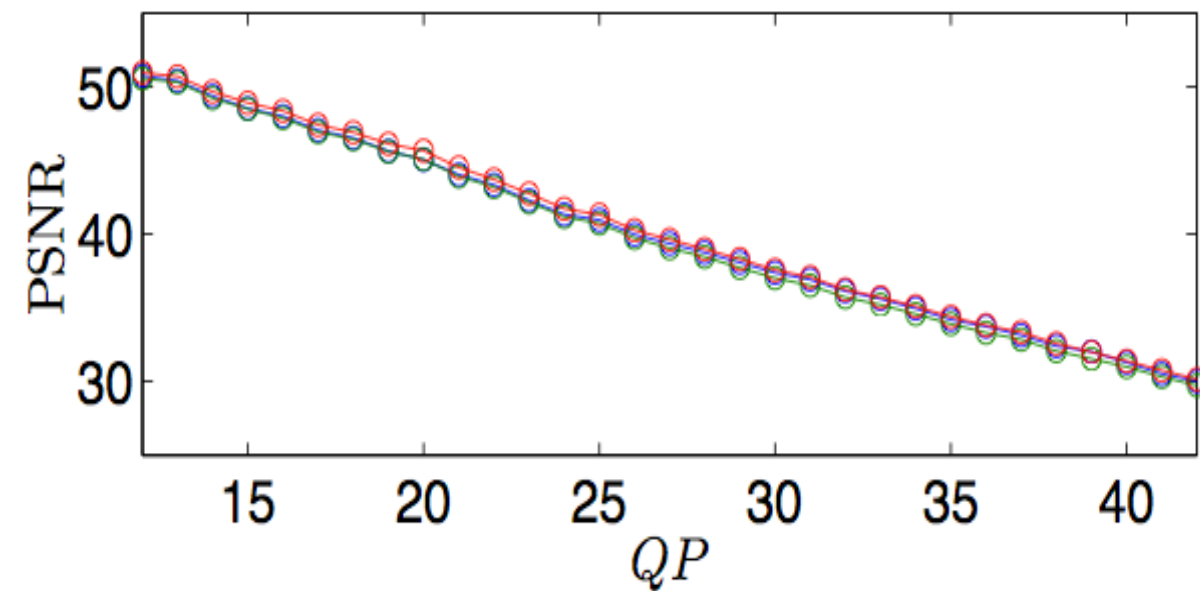
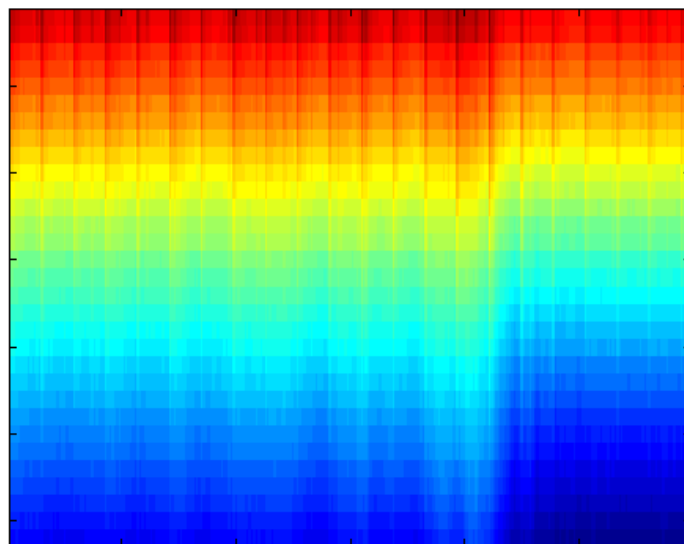


Correct



**Local
Maxima**

Incorrect



- Results:

c_1	\hat{c}_1											
	MPEG-2			MPEG-4			AVC			DIRAC		
MPEG-2	0.94	0.96	0.96	0.05	0.04	0.03	0	0	0.01	0.01	0	0
MPEG-4 (a)	0	0	0.04	0.93	0.92	0.76	0.02	0.02	0.2	0.06	0.06	0
MPEG-4 (b)	0.02	0.02	0.06	0.87	0.87	0.69	0.06	0.05	0.25	0.06	0.06	0
AVC (a)	0.01	0	0	0.14	0.24	0.06	0.79	0.68	0.94	0.06	0.08	0
AVC (b)	0	0.01	0	0.13	0.2	0.05	0.81	0.69	0.94	0.06	0.09	0.01
AVC (c)	0.06	0.06	0.15	0	0.03	0	0.92	0.87	0.81	0.02	0.05	0.04
DIRAC	0	0.02	0	0.09	0.12	0.01	0.13	0.12	0.21	0.78	0.74	0.78
	MPEG-2	MPEG-4	AVC	MPEG-2	MPEG-4	AVC	MPEG-2	MPEG-4	AVC	MPEG-2	MPEG-4	AVC

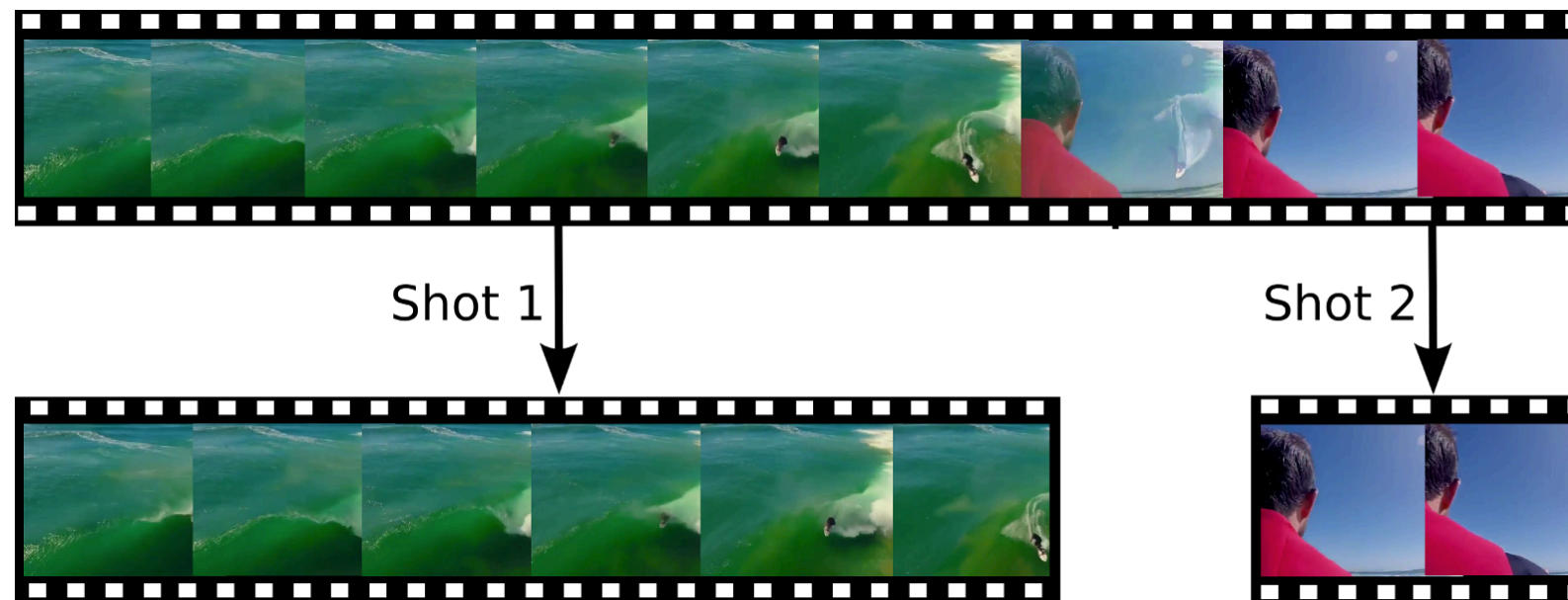
c_2

Are **codec** and **quality** coherent in time?

[Verde et al. ICIP 2018]

Problem

- Given a decoded video sequence, detect whether it is a compilation from **multiple video shots**.

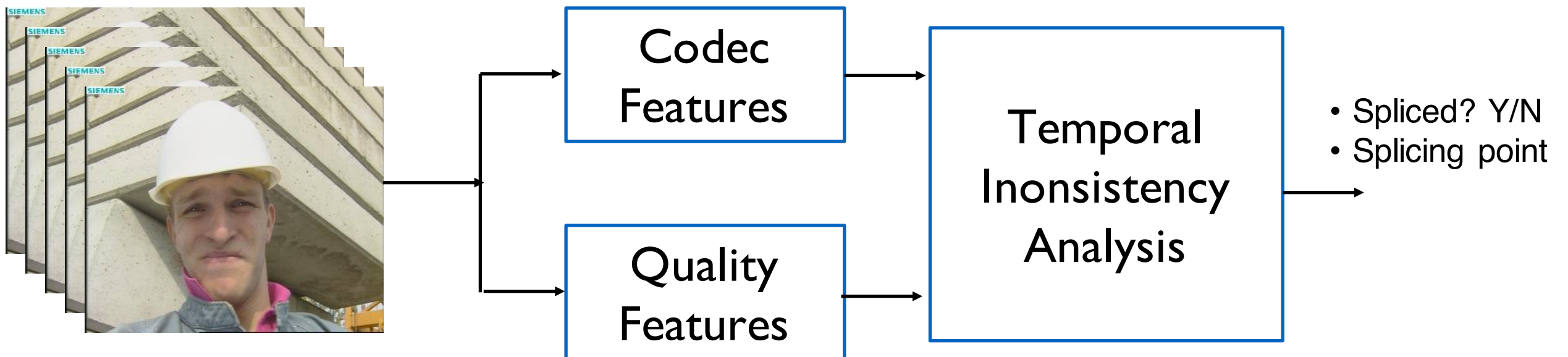


Assumptions

- Shots are seldom originally encoded with the exact same **codec** or **parameters** due to different sources and used software.

Main Pipeline

- Compute a frame-wise indicator of the used **codec**
- Compute a frame-wise indicator of the video **quality**
- Check **inconsistency** of these indicators frame-by-frame



Feature Extraction

Codec Features

- A CNN is trained to classify 4 different codecs (MPEG2, MPEG4, H264, H265)
- Feature vector is $\mathbf{f}_C^p(n) = [f_{H264}^p(n), f_{H265}^p(n), f_{MPEG2}^p(n), f_{MPEG4}^p(n)]$

Quality Features

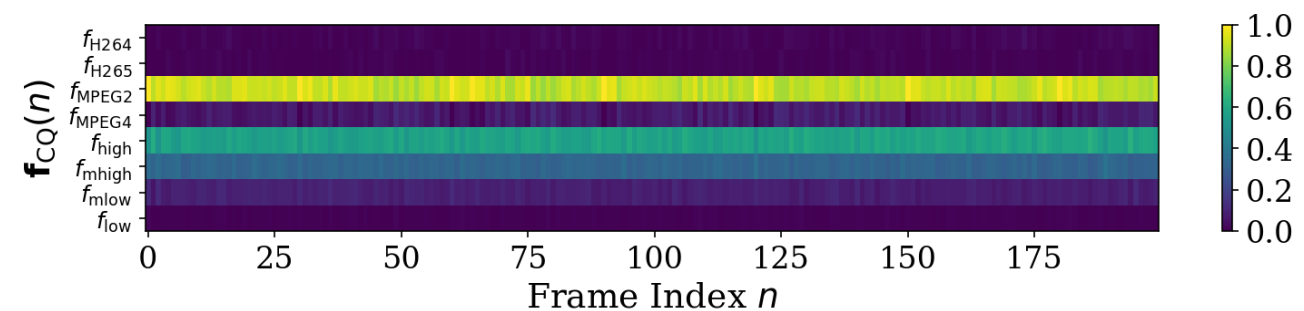
- A CNN is trained to classify 4 different qualities (H264 with QP=5, 10, 15, 20)
- Feature vector is $\mathbf{f}_Q^p(n) = [f_{low}^p(n), f_{m-low}^p(n), f_{m-high}^p(n), f_{high}^p(n)]$

Temporal inconsistency analysis

Feature Merge

- Feature vectors are concatenated into a single one

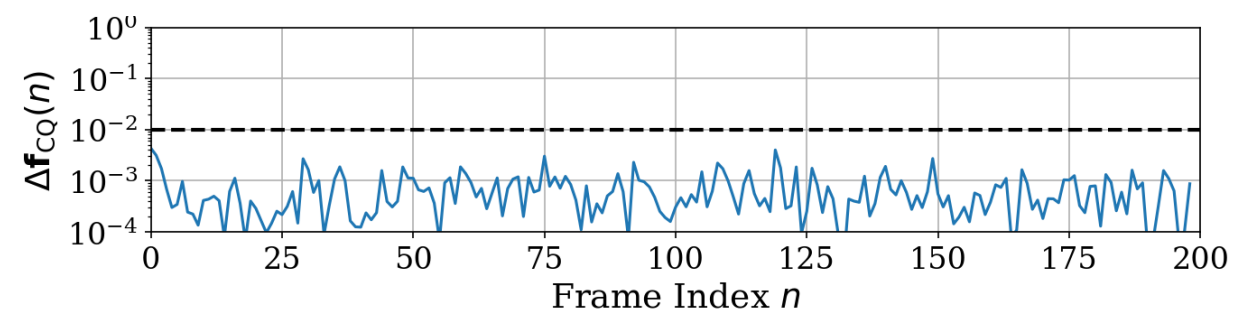
$$\mathbf{f}_{CQ}(n) = [\mathbf{f}_C(n), \mathbf{f}_Q(n)]$$



Time Analysis

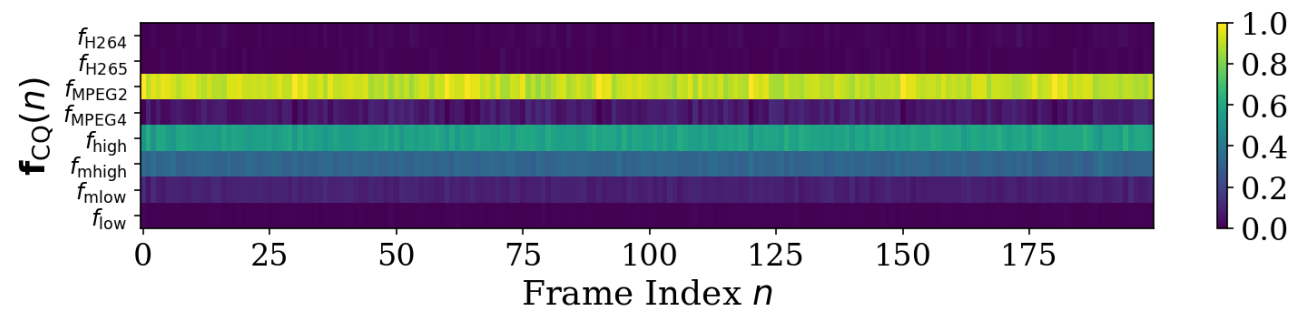
- Compute MSE between feature pairs

$$\Delta \mathbf{f}_{CQ}(n) = \text{MSE}(\mathbf{f}_{CQ}(n), \mathbf{f}_{CQ}(n+1))$$

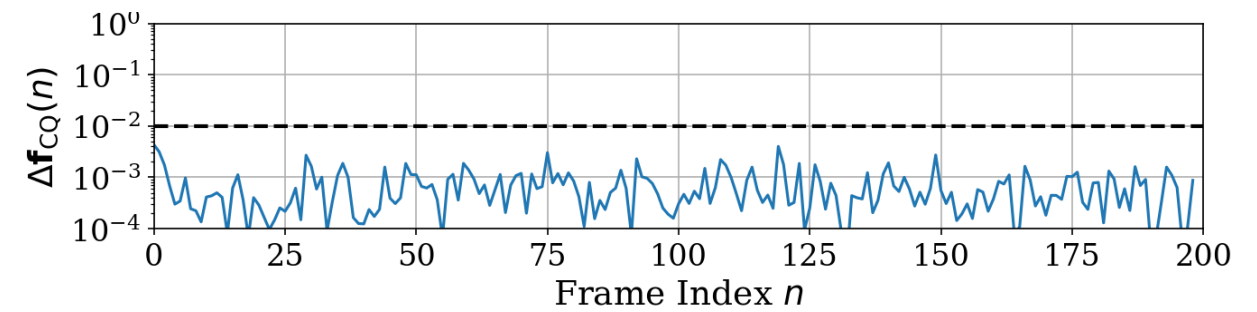


Threshold MSE

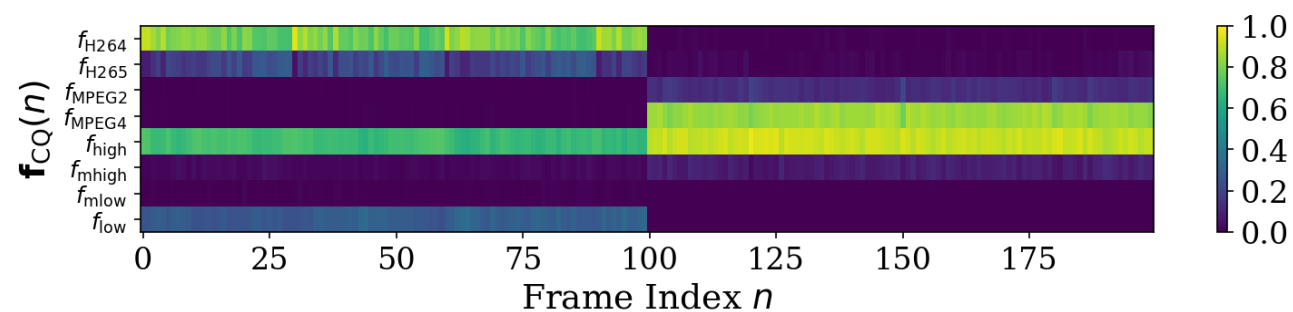
Visual example



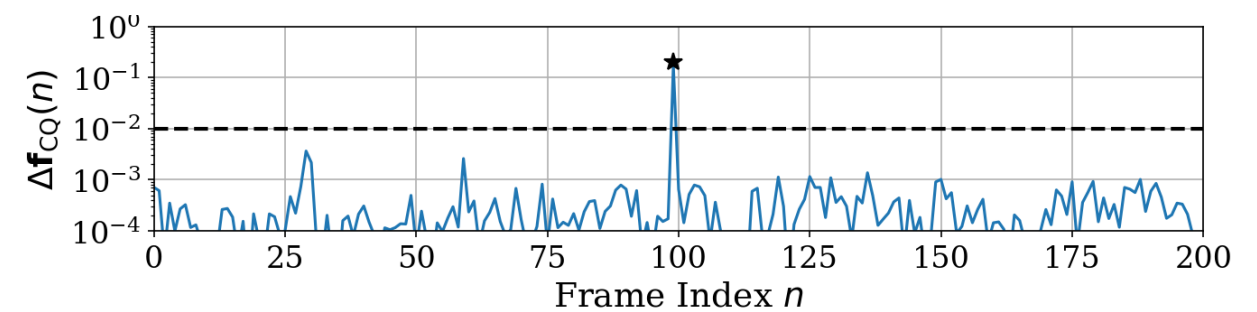
(a) Pristine



(a) Pristine



(b) Composition



(b) Composition

Challenging Example



Challenging Example



Video Codec Identification Results

H264	0.76	0.20	0.02	0.02
H265	0.12	0.85	0.01	0.02
MPEG2	0.02	0.01	0.84	0.13
MPEG4	0.05	0.03	0.03	0.89
	H264	H265	MPEG2	MPEG4

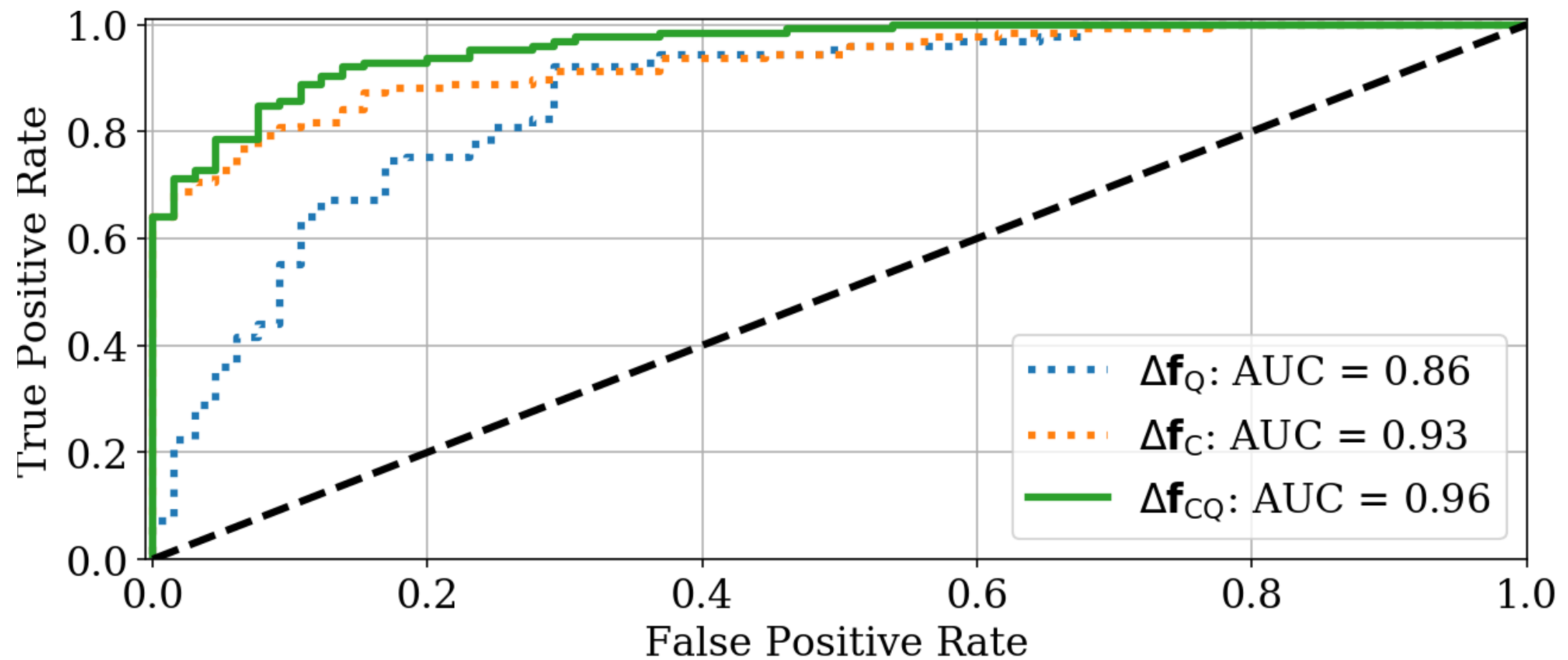
Predicted

Video Quality Identification Results

high	0.84	0.12	0.04	0.00
m-high	0.09	0.83	0.08	0.00
m-low	0.03	0.07	0.85	0.04
low	0.00	0.00	0.04	0.96
	high	m-high	m-low	low

Predicted

Splicing Detection Results



**Image
splicing**

**Number of
compressions**

**Type of
codec**

**Coding-based
footprints**

E

**Image
splicing**

**Video
copy-paste**

**Video
copy-move**

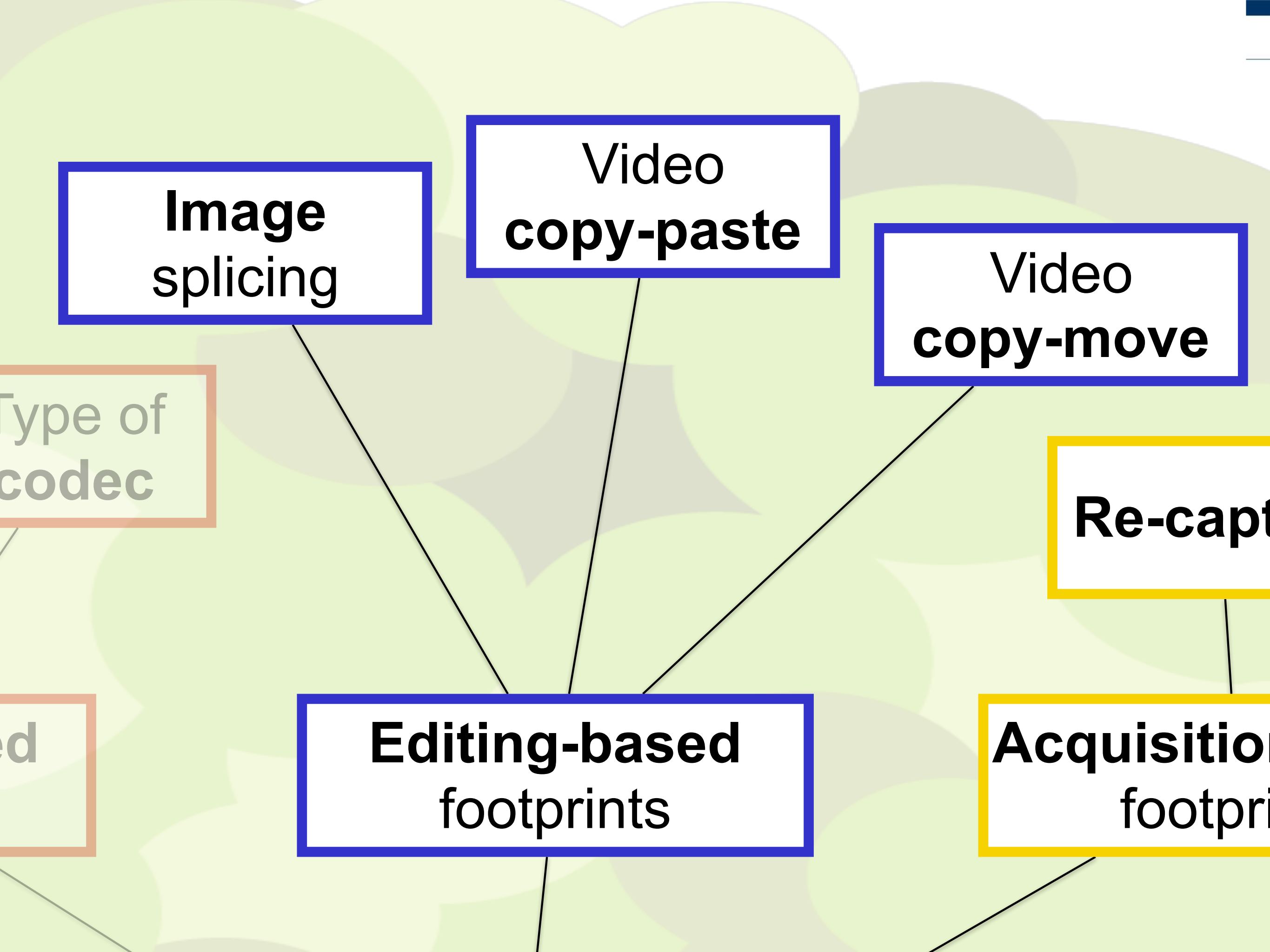
Type of
codec

Re-capt

ed

**Editing-based
footprints**

**Acquisition
footprints**











Forged



Original



Forged



Original



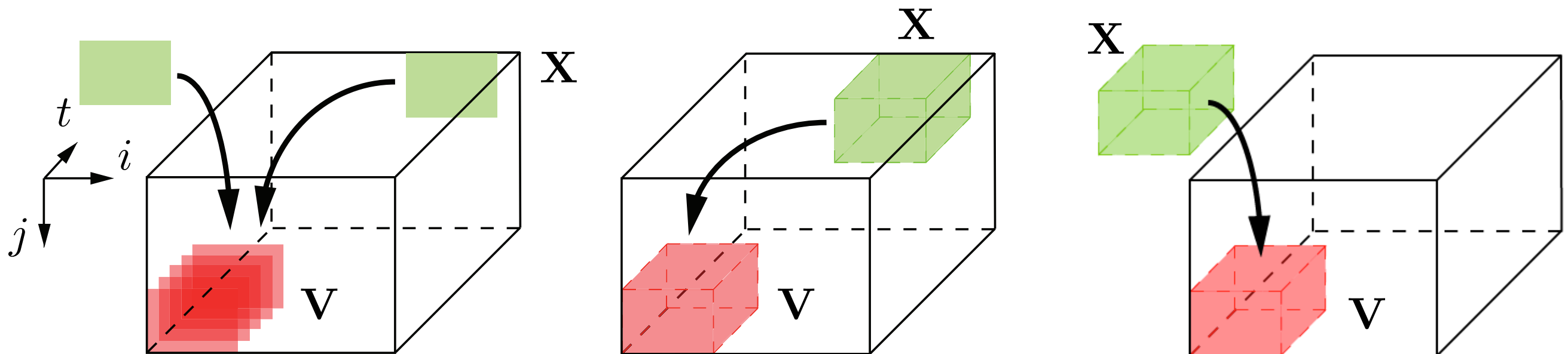
Is the video **forged**?

[Bestagini et al. **ICASSP 2013**]

[Bestagini et al. **MMSP 2013**]

Editing-based fingerprints

- Video **forgeries** can be operated through different kinds of editing attacks
- We considered:
 1. insertion of a **still image**
 2. insertion of a portion of **video from the same source**
 3. insertion of a portion of **video from a different source**



Editing-based footprints: image copy-paste

- **Problem:**

- An **image is inserted** and repeated in time

- **Method:**

- Exploit characteristic residual between adjacent frames



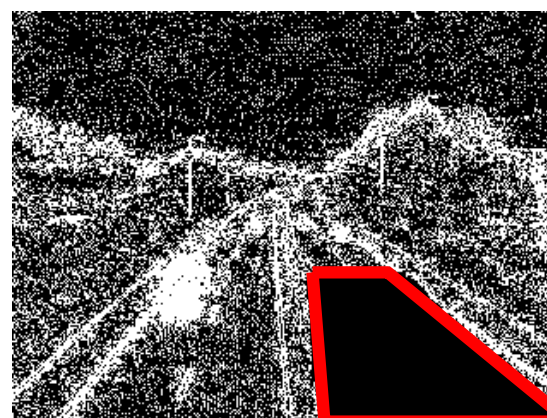
Original



Tampered



(a) Original residual



(b) Tampered residual



(c) Tampered and compressed residual

Editing-based footprints: image copy-paste

ALGORITHM:

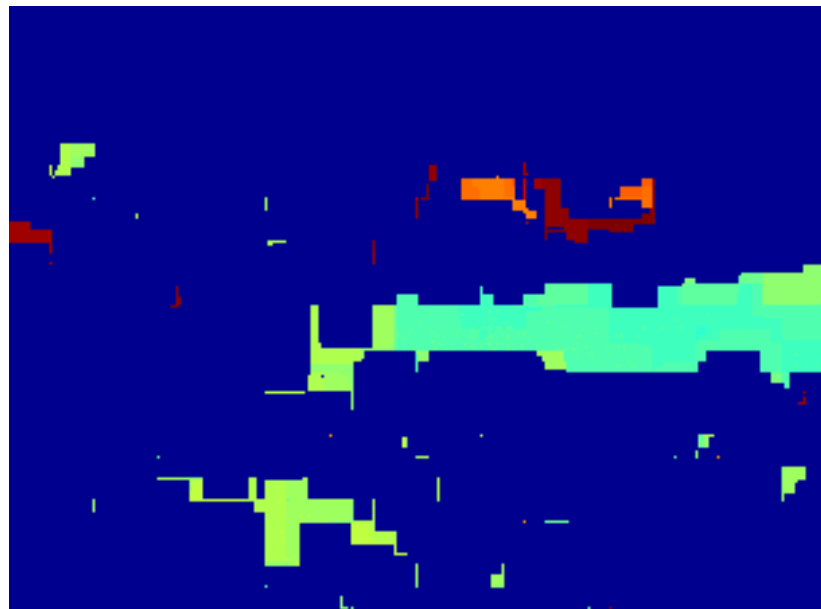
- Define the residual $r_{i,j}^t = x_{i,j}^t - x_{i,j}^{t+1}$
 - Zero for possibly tampered pixels
- Define the residual mask as $m_{i,j}^t = \begin{cases} 1 & \text{if } r_{i,j}^t = 0, \\ 0 & \text{otherwise,} \end{cases}$
 - One for possibly tampered pixels
- Apply an erosion with a Structuring Element $\mathbf{H}^{di,dj,dt}$ and obtain $\mathbf{E} = \{e_{i,j}^t\} = \mathbf{M} \ominus \mathbf{H}^{di,dj,dt}$,
 - Remove small areas
- Compute the feature vector $F_{i,j} = [f_{i,j}^1, f_{i,j}^2]$
 - $f_{i,j}^1$ cardinality of the longest set of ones in (i,j)
 - $f_{i,j}^2$ starting t value of the longest set of ones
- Search the longest set of ones starting from the same t



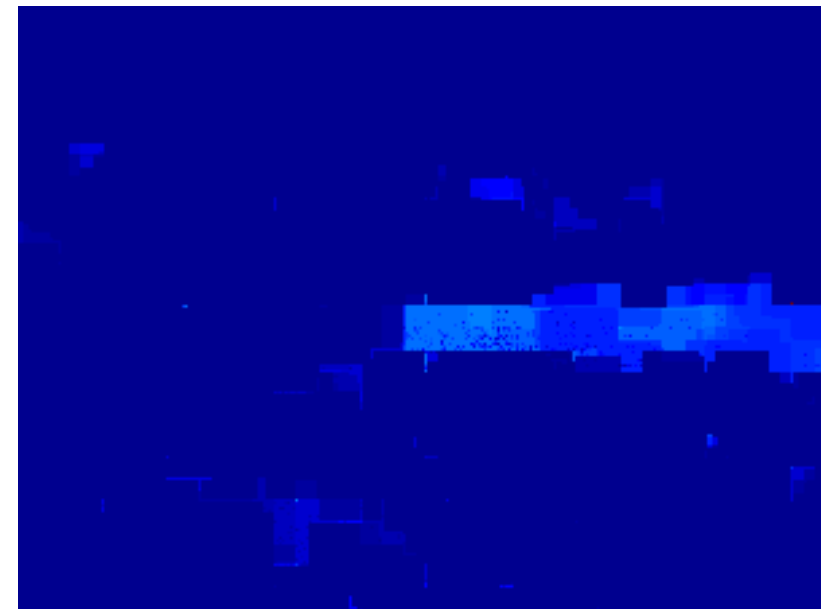
(a) Tampered frame



(b) Detected mask



(c) $f_{i,j}^1$



(d) $f_{i,j}^2$

Editing-based footprints: image copy-paste

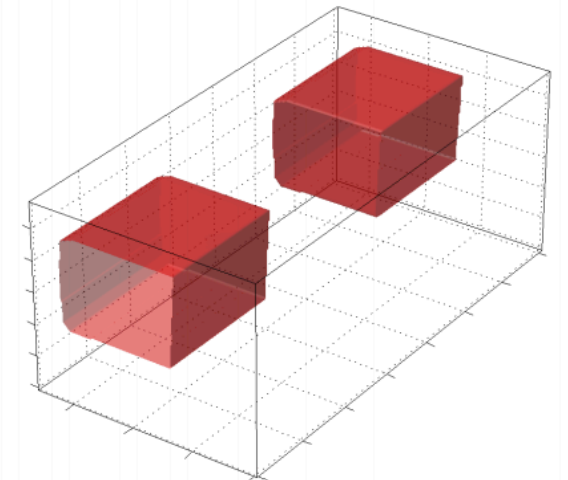
- **Problem:**

- A **video is inserted** from the same sequence



- **Method:**

- Implementation of an automatic correlation analysis to detect local duplication



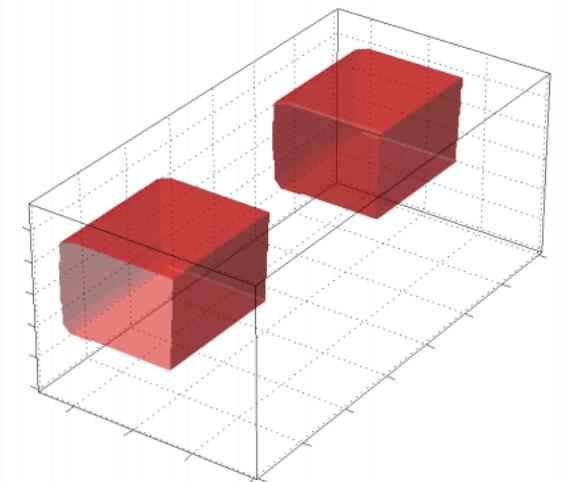
Editing-based footprints: image copy-paste

- **Problem:**
 - A **video is inserted** from the same sequence



Method:

- Implementation of an automatic correlation analysis to detect local duplication



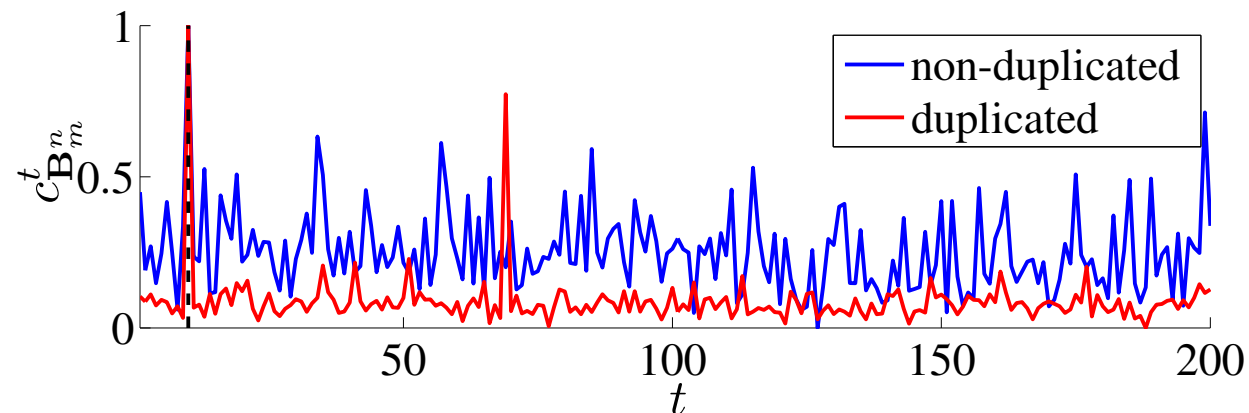
Editing-based footprints: image copy-paste

Algorithm:

- Compute the residual $\mathbf{R} = \{r_{i,j}^t\}$
- Divide the residual into non-overlapping 3D blocks \mathbf{B}_m^n
- Compute the phase correlation

$$\mathbf{C}_{i,j}^t(\mathbf{B}_m^n) = \mathcal{F}^{-1} \left(\frac{\mathcal{F}(\mathbf{B}_m^n) \mathcal{F}(\mathbf{R})^*}{|\mathcal{F}(\mathbf{B}_m^n) \mathcal{F}(\mathbf{R})^*|} \right)$$

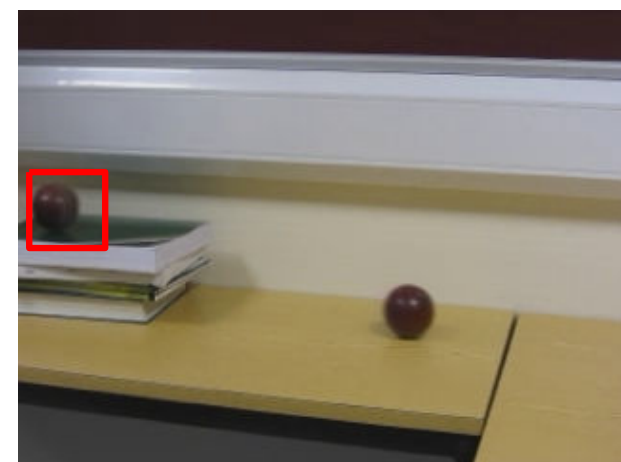
- Compute the maximum correlation value for each time position $c_{\mathbf{B}_m^n}^t = \max_{i,j} (|\mathbf{C}_{i,j}^t(\mathbf{B}_m^n)|)$



- Search for peaks indicating duplication by thresholding the max-mean ratio

$$p_{\mathbf{B}_m^n} = \frac{\max(c_{\mathbf{B}_m^n}^t)}{\frac{1}{(T-1)} \sum_t c_{\mathbf{B}_m^n}^t}$$

- Check if the detected duplicated block is similar to its original version (MSE)



original

forged

detected duplication

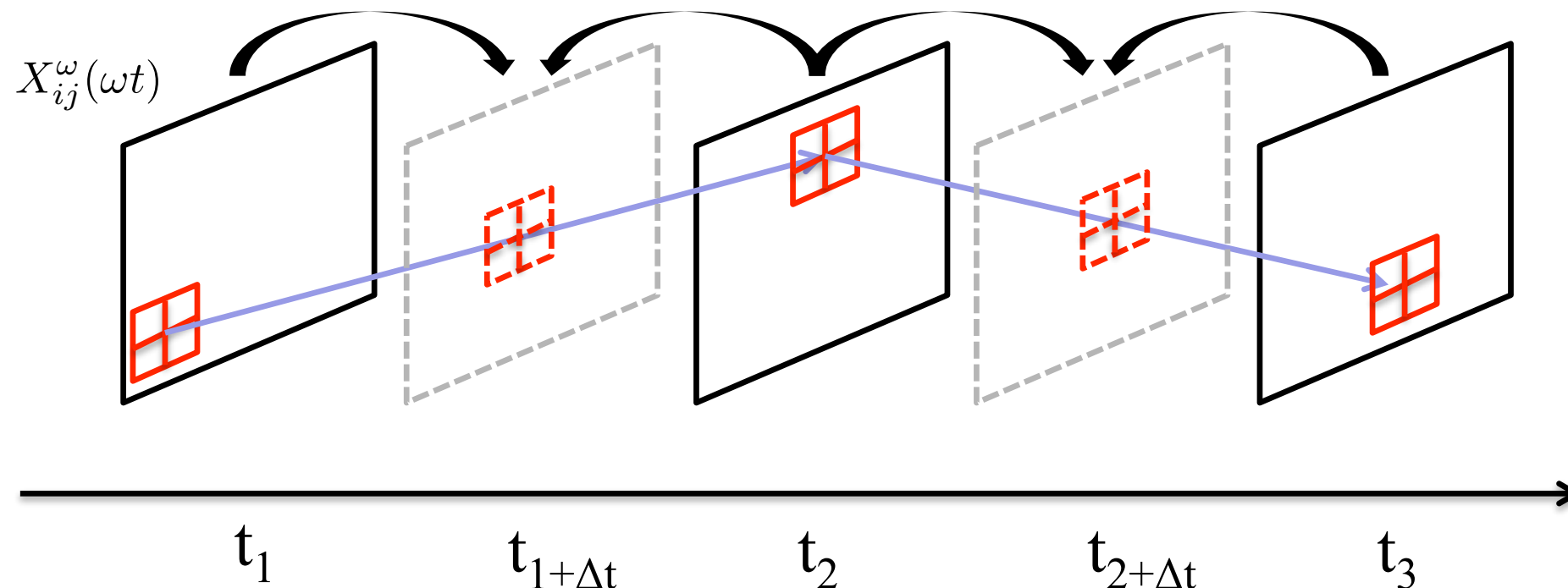
Editing-based footprints: image copy-paste

- **Problem:**

- A **video is inserted** from the same a different sequence at with different frame-rate

- **Method:**

- Search for traces left by frame-rate equalisation
- Up-sampling and down-sampling leave a characteristic pixel correlation in time



Editing-based footprints: image copy-paste

Algorithm:

- Estimate each frame from their neighbors
 - Compute motion vectors
 - Average frames

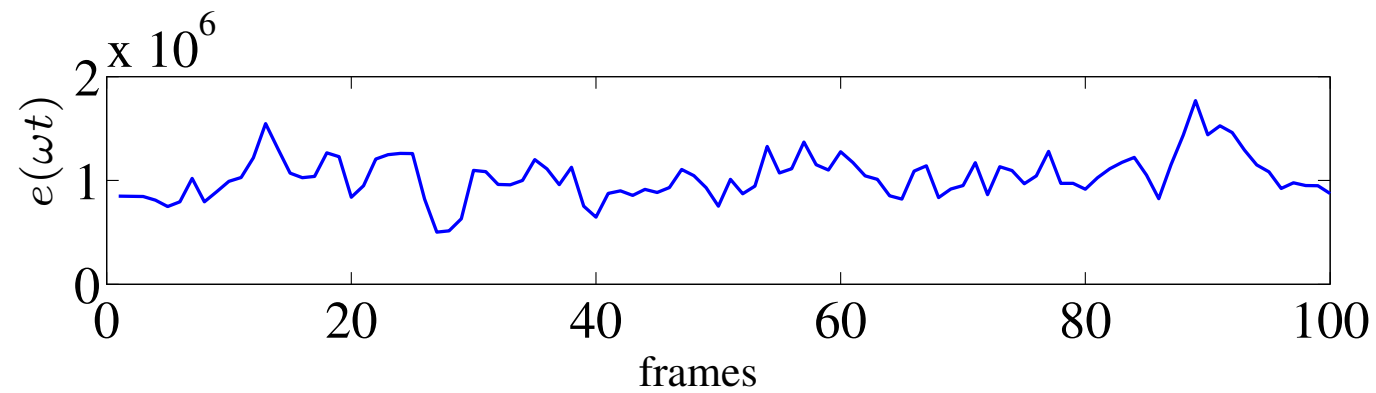
- Compute the prediction error
 - Original frames → high error
 - Predicted frames → low error

$$e_{ij}(\omega t) = X_{ij}^{\omega}(\omega t) - \sum_{k=-K}^K \boxed{h_k^*} \cdot X_{\boxed{m_{t,i,j} n_{t,i,j}}}^{\omega}(\omega t + \omega k)$$

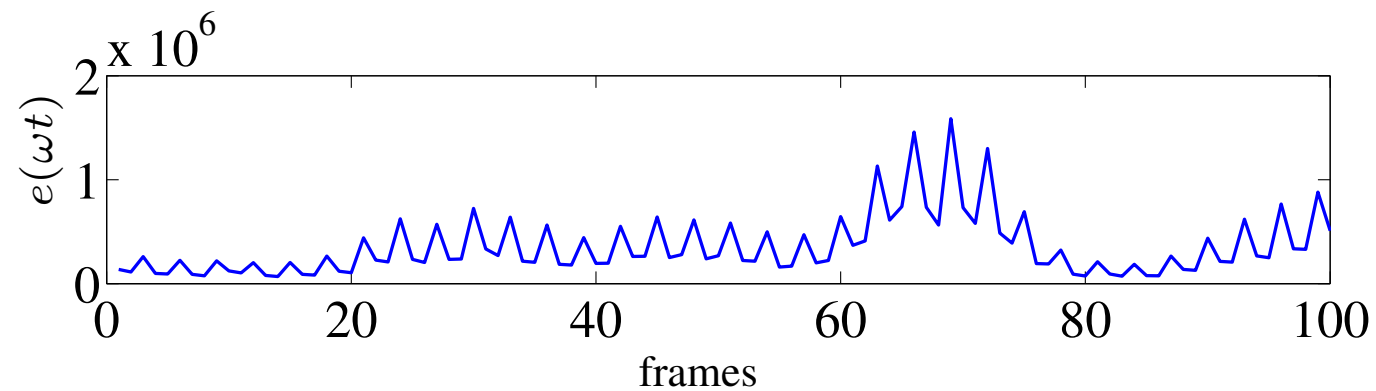
Analysis filter
Estimated MVs

$$e(\omega t) = \sum_{ij} |e_{ij}(\omega t)|^2$$

- Estimate error periodicity (spectral analysis)
 - If non-periodic → not interpolated
 - If periodic → interpolated

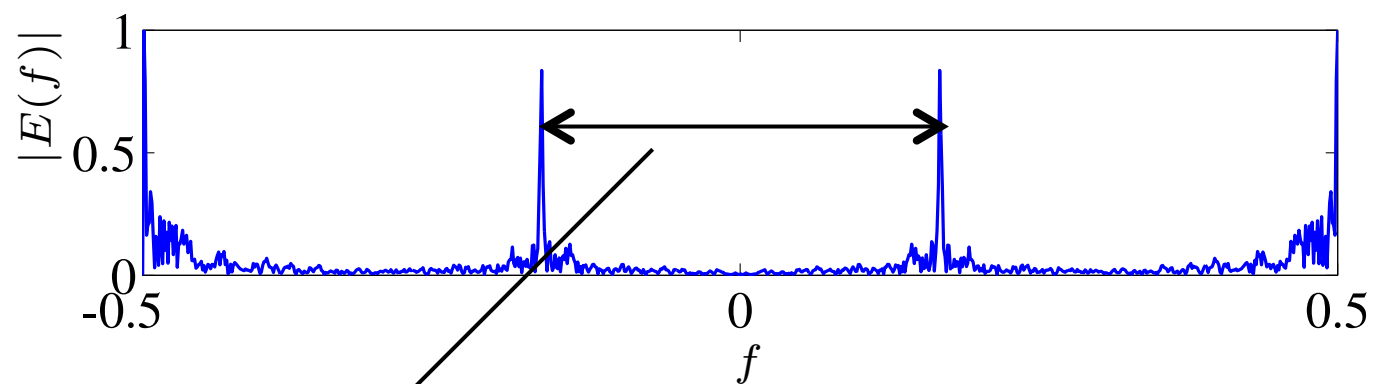


Foreman (original)



Foreman (interpolated)

Fps: 30 \rightarrow 90



Foreman (interpolated)
Spectral analysis

$$\Delta f = 0.5 - |\omega - 0.5|,$$

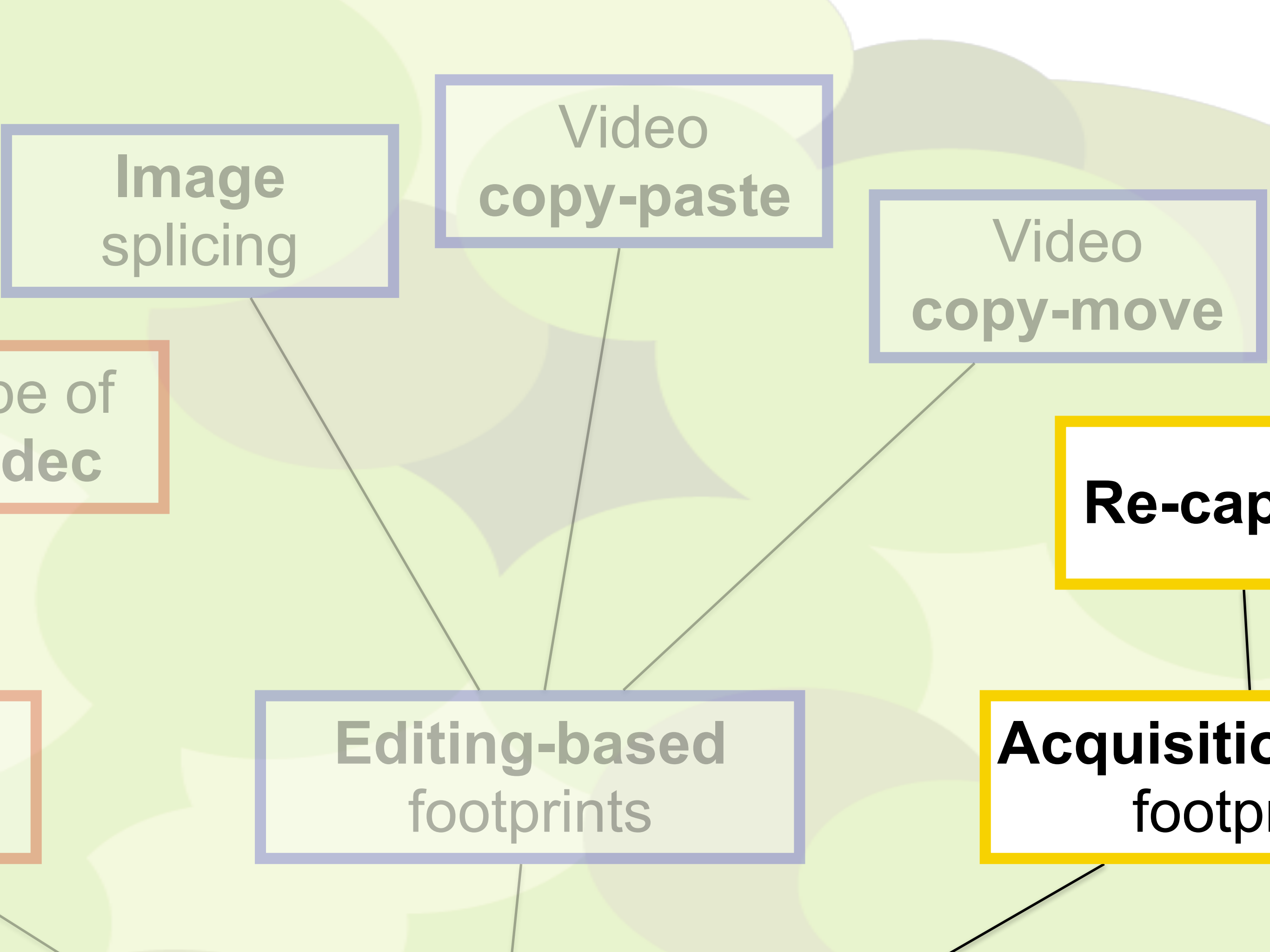


Image
splicing

Video
copy-paste

Video
copy-move

Re-capture
footprint

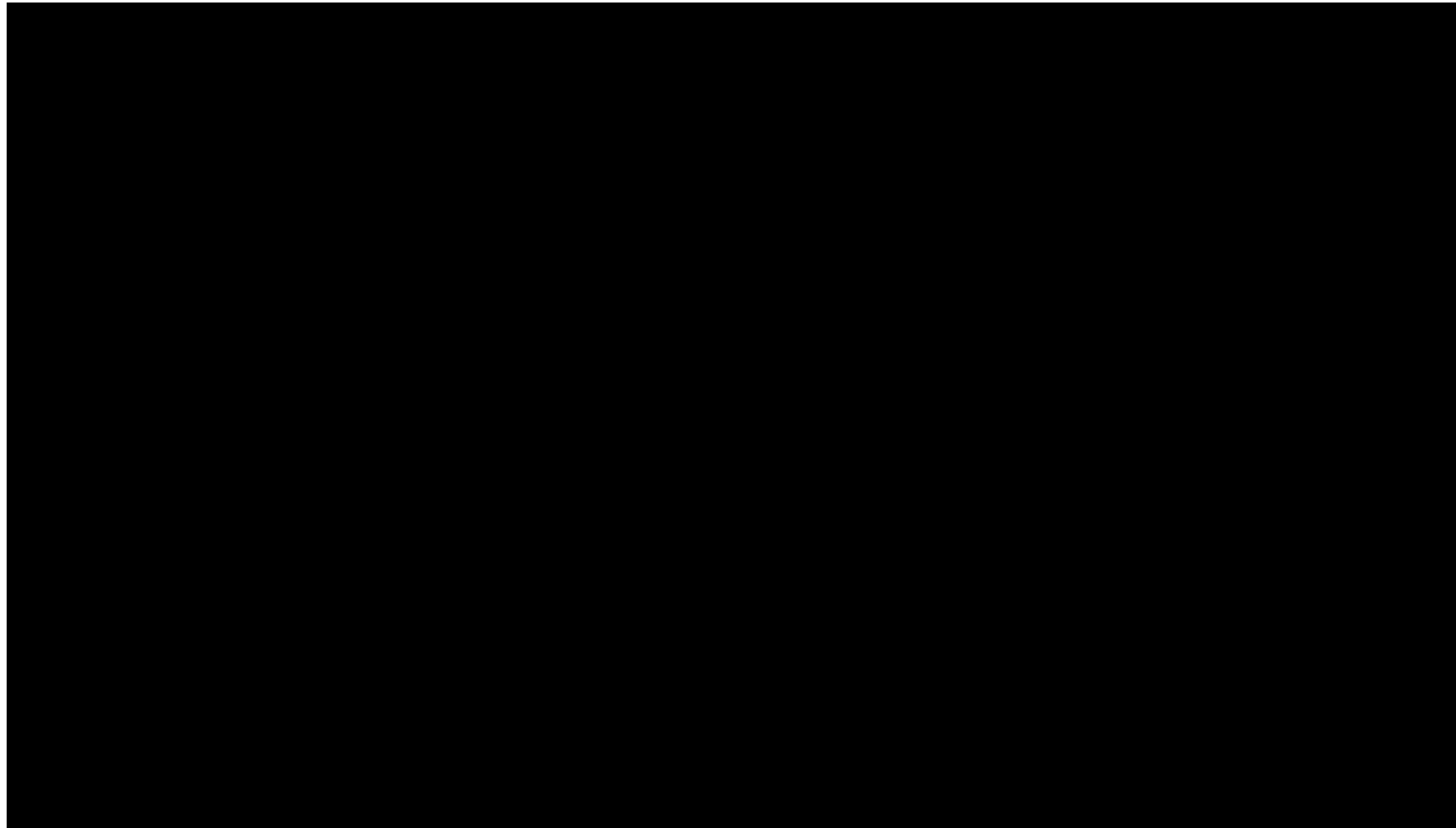
Re-capture
footprint

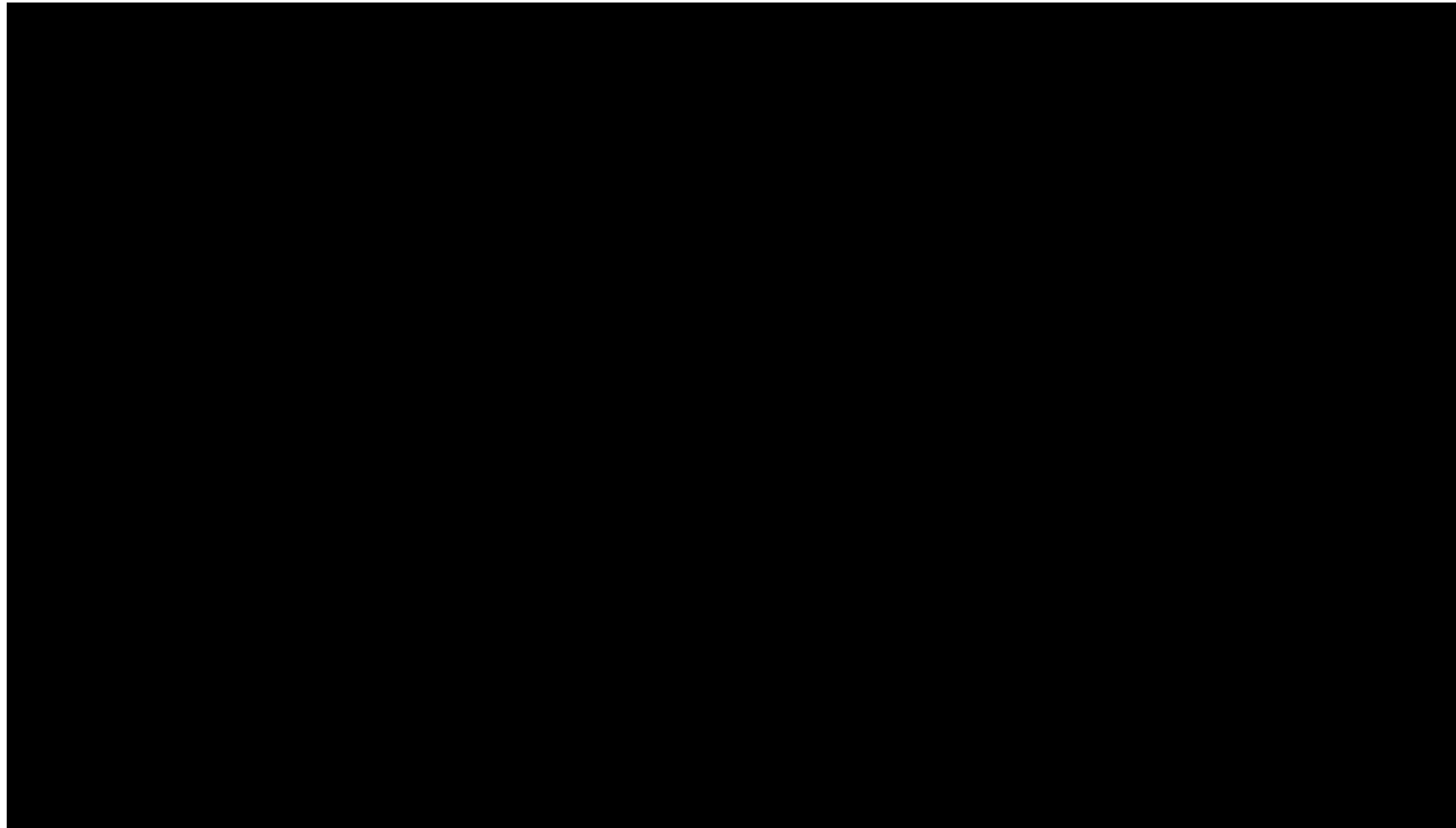
Editing-based
footprints

Acquisition
footprint

Re-capture

**Acquisition-based
footprints**



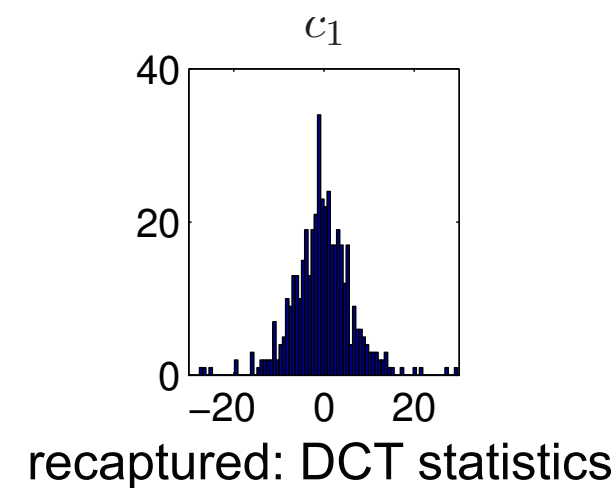
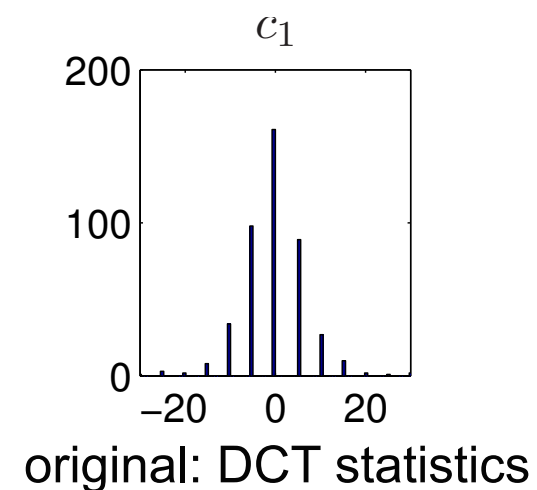


Acquisition-based footprints

- Re-acquisition is a powerful **anti-forensic** tool



- Re-acquired videos are visually **similar to the originals**
- Many footprints are **masked**
 - Detectors can be-fooled



Is the video **recaptured**?

[Bestagini et al. ICIP 2013]

Acquisition-based footprints: re-capture

- **Setup:**

- A video is **re-captured** from a **LCD** monitor

- **Ghosting as filtering:**

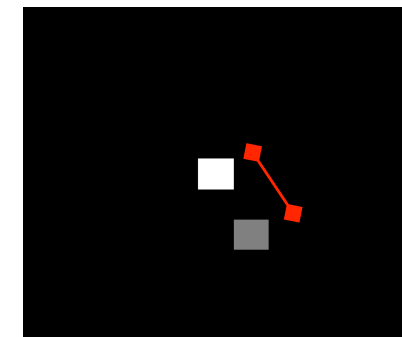
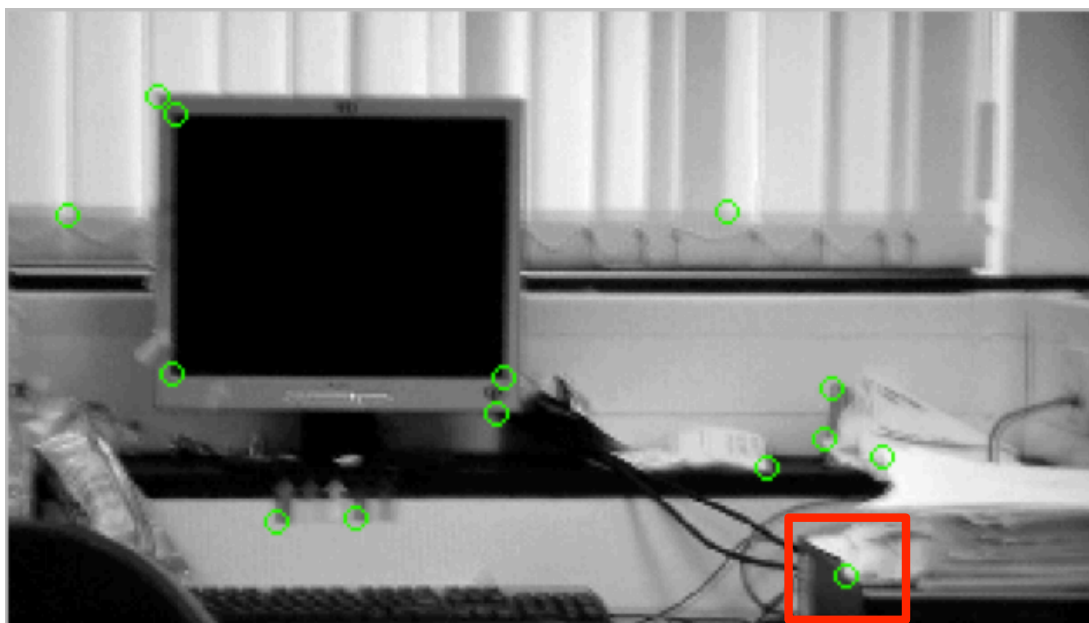
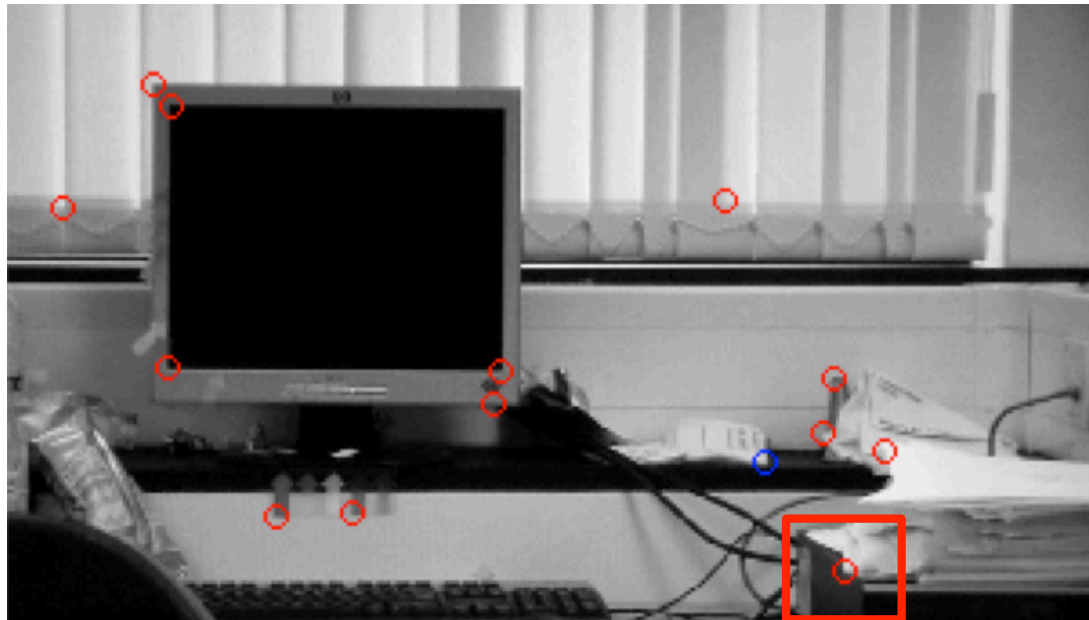
$$\begin{array}{ccc}
 X_1 = \text{Orig. scene} & X_2 = \text{Orig. scene} & Y_1 = \text{Recap. scene} \\
 \alpha \begin{array}{c} \blacksquare \\ \square \\ \blacksquare \end{array} + (1-\alpha) \begin{array}{c} \blacksquare \\ \square \\ \blacksquare \end{array} & = & \begin{array}{c} \blacksquare \\ \square \\ \blacksquare \end{array}
 \end{array}$$

Approximating the motion as a translation between adjacent frames:

$$\begin{array}{ccc}
 X_1 = \text{Orig. scene} & H(\alpha) = \text{Ghost. Filter} & Y_1 = \text{Recap. scene} \\
 \begin{array}{c} \blacksquare \\ \square \\ \blacksquare \end{array} * \begin{array}{c} \blacksquare \\ \square \\ \blacksquare \end{array} & = & \begin{array}{c} \blacksquare \\ \square \\ \blacksquare \end{array}
 \end{array}$$

Acquisition-based footprints: re-capture

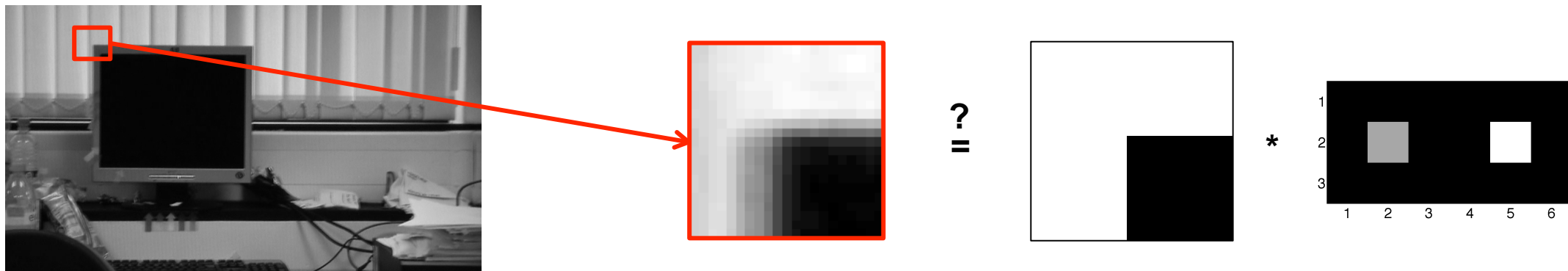
- Filter shape is derived from motion estimation



Acquisition-based fingerprints: re-capture

- **Method:**

- Minimize a cost function to detect whether key-points underwent “ghosting filtering”



- **Results:**

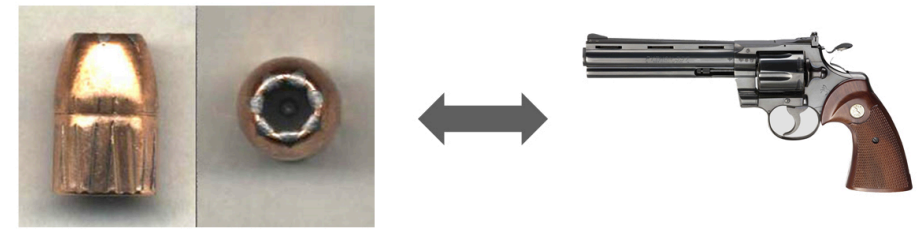
- Detection accuracy over 91%

From **which camera** does the video come from?

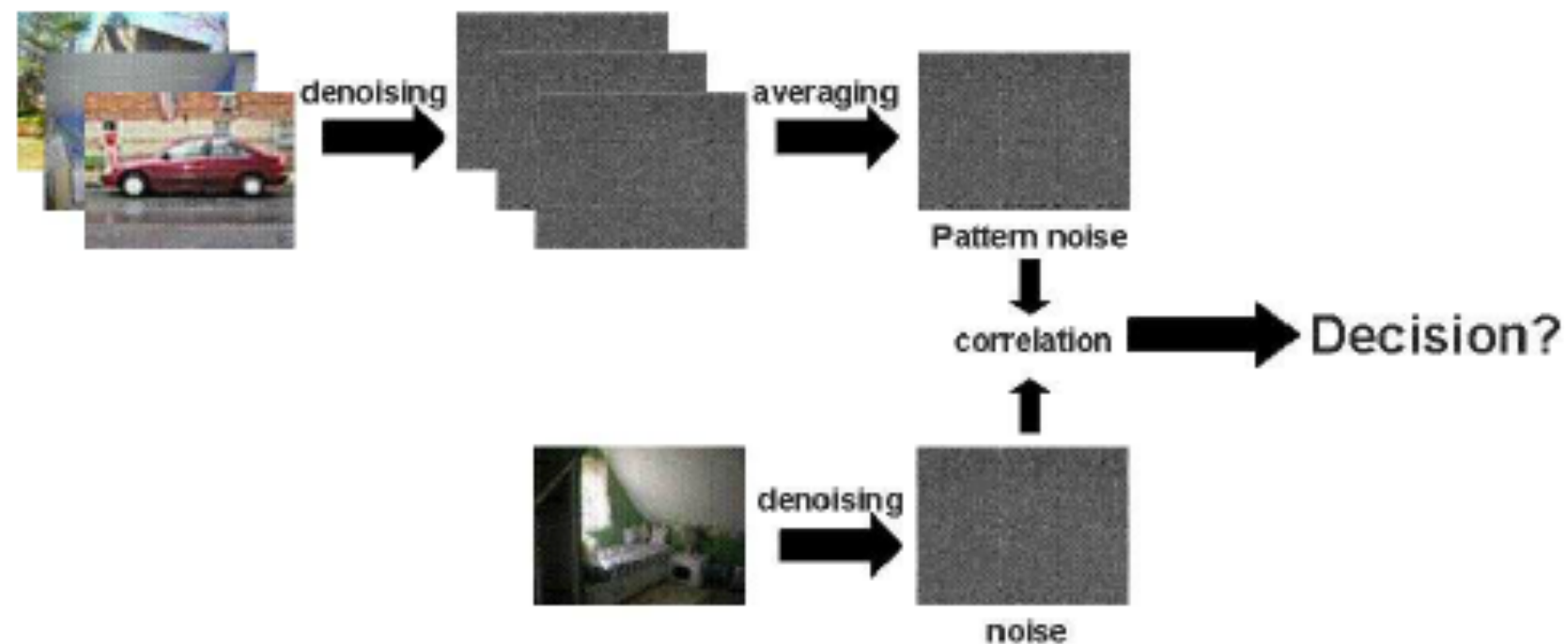
[Mandelli et al. **EUSIPCO 2018**]

[Mandelli et al. **to be submitted**]

- **Photo Response Non Uniformity:**
 - It enables linking images to devices



- **How to:**
 - Extract noise pattern from images
 - Compute correlation



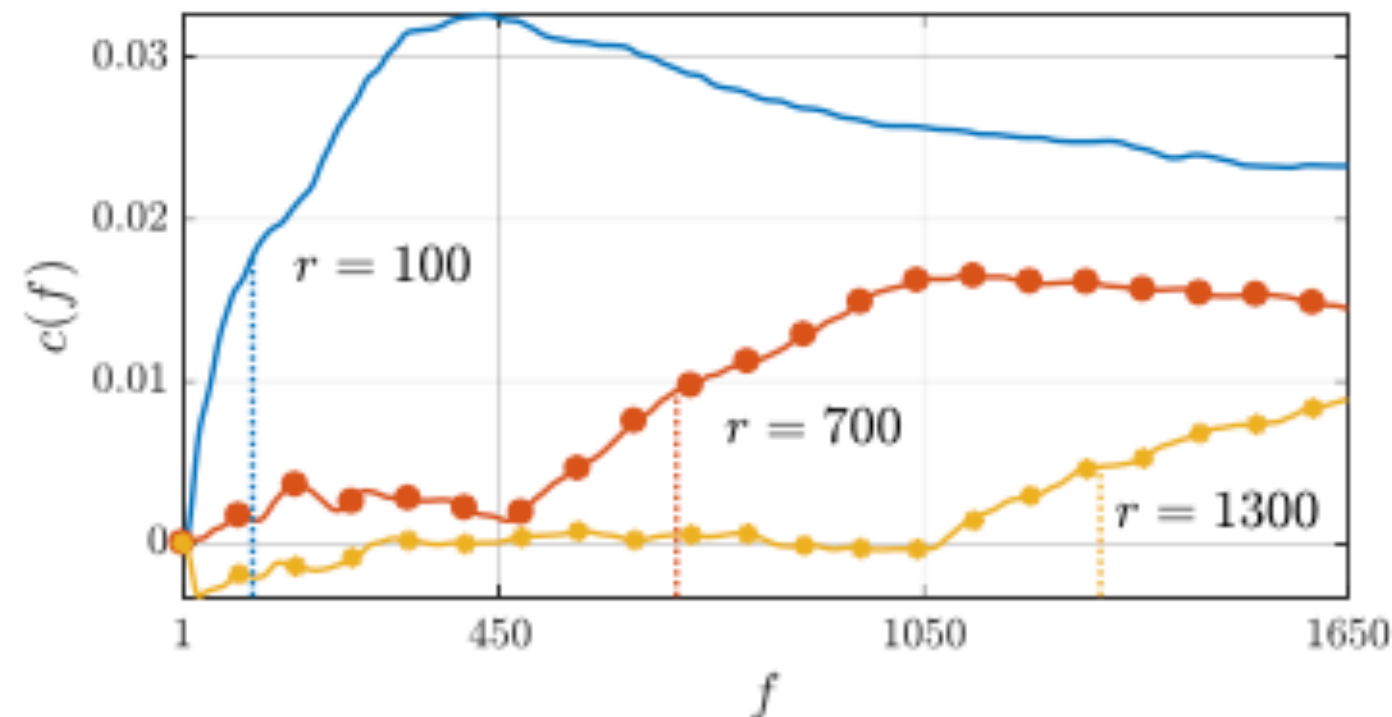
- **Application:**

- Video compilation detection and segmentation



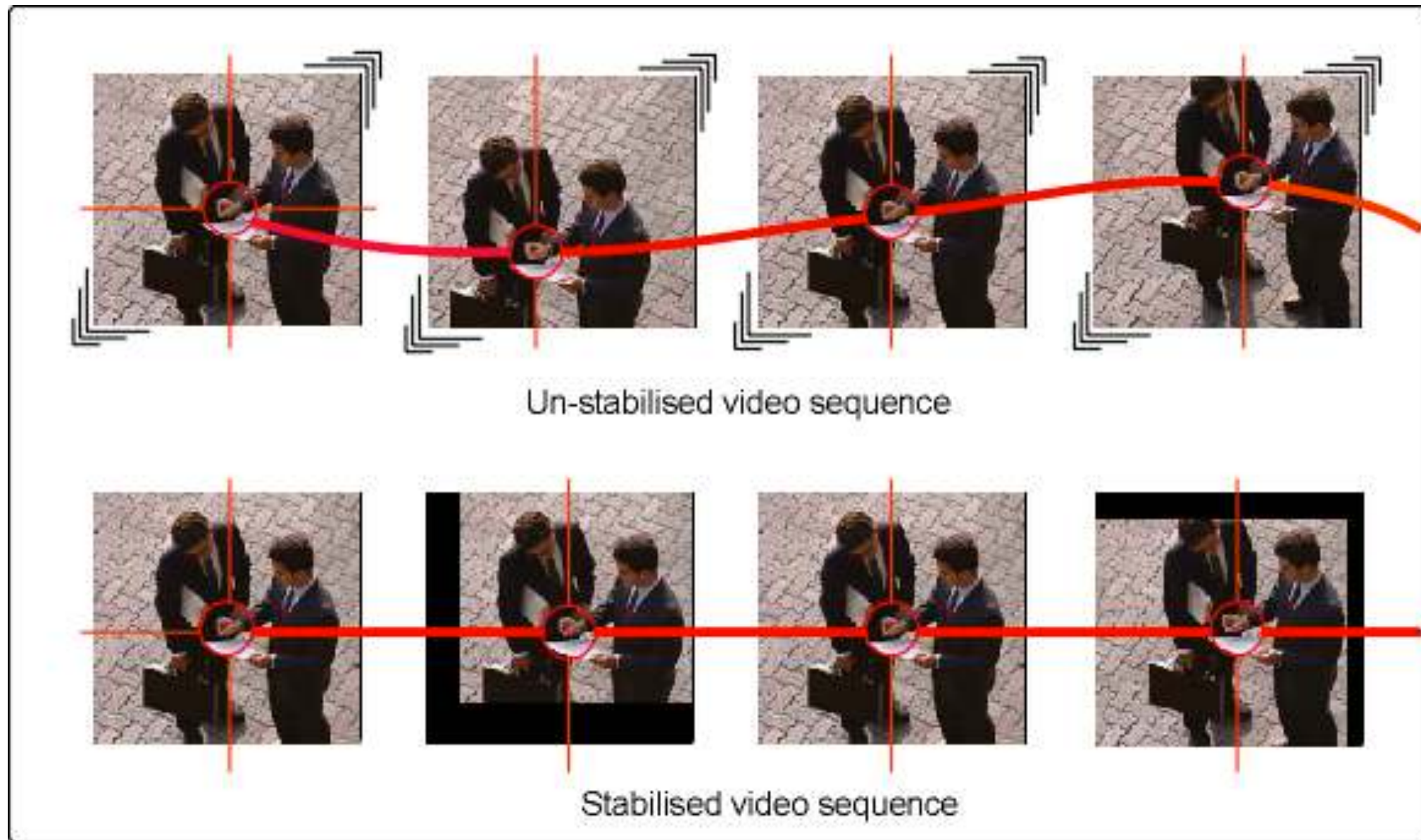
- **Method:**

- Compute cumulative correlation $c(f) = \rho(\mathbf{W}_r, \overline{\mathbf{W}}(f)\mathbf{I}_r)$



- **Challenges:**

- Aggressive coding
- Digital video stabilizaion

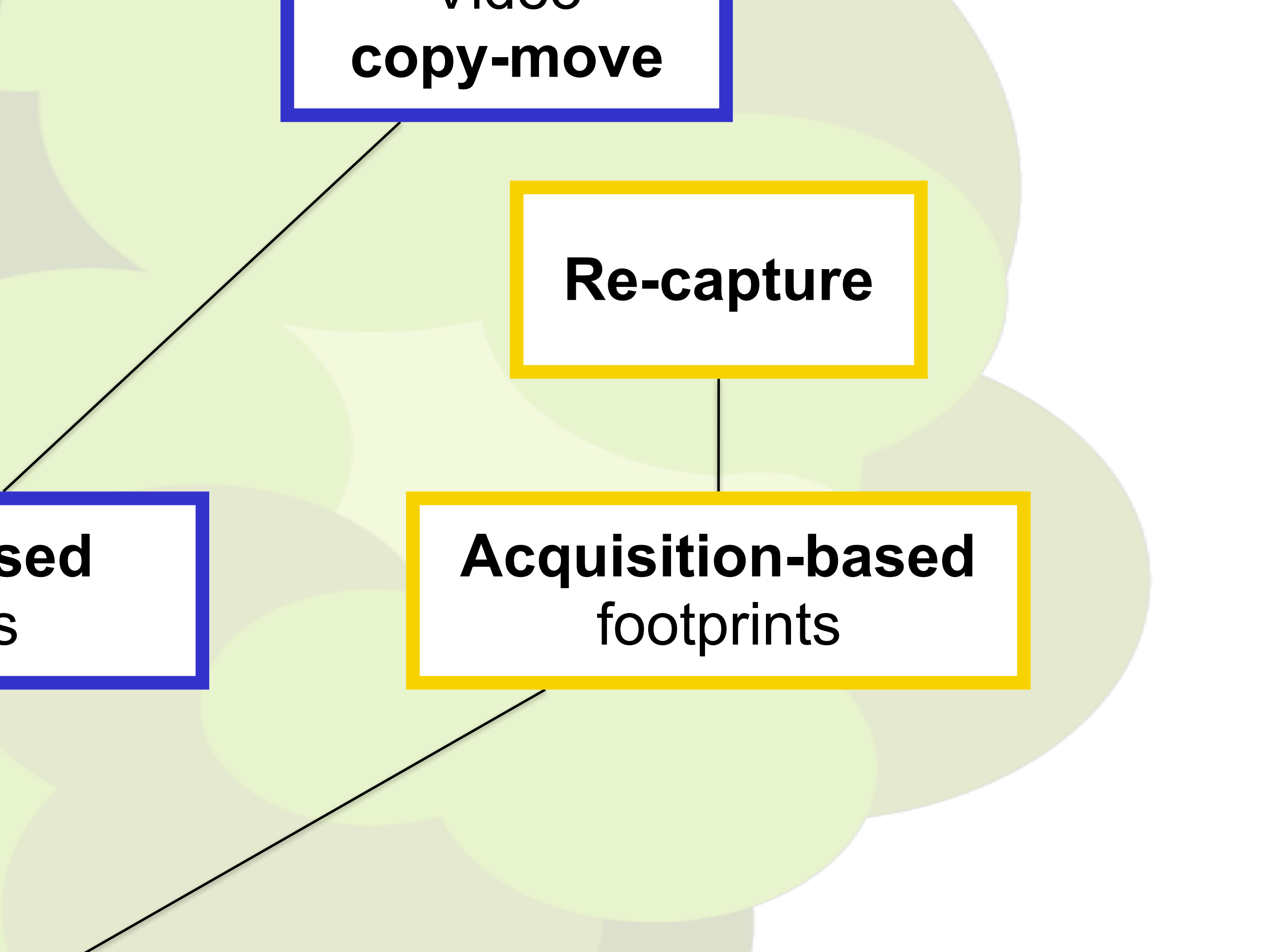


copy-move

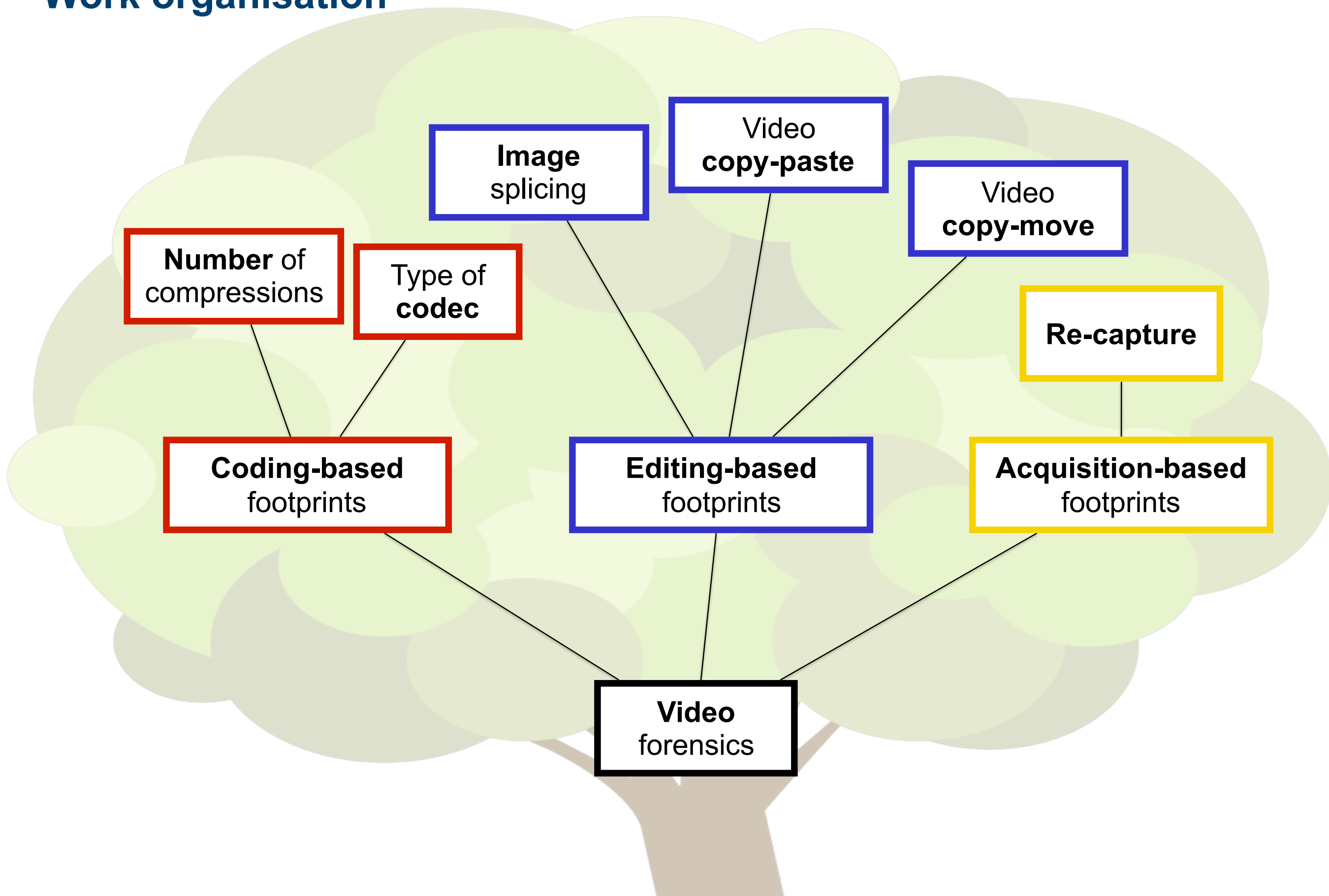
Re-capture

**Acquisition-based
footprints**

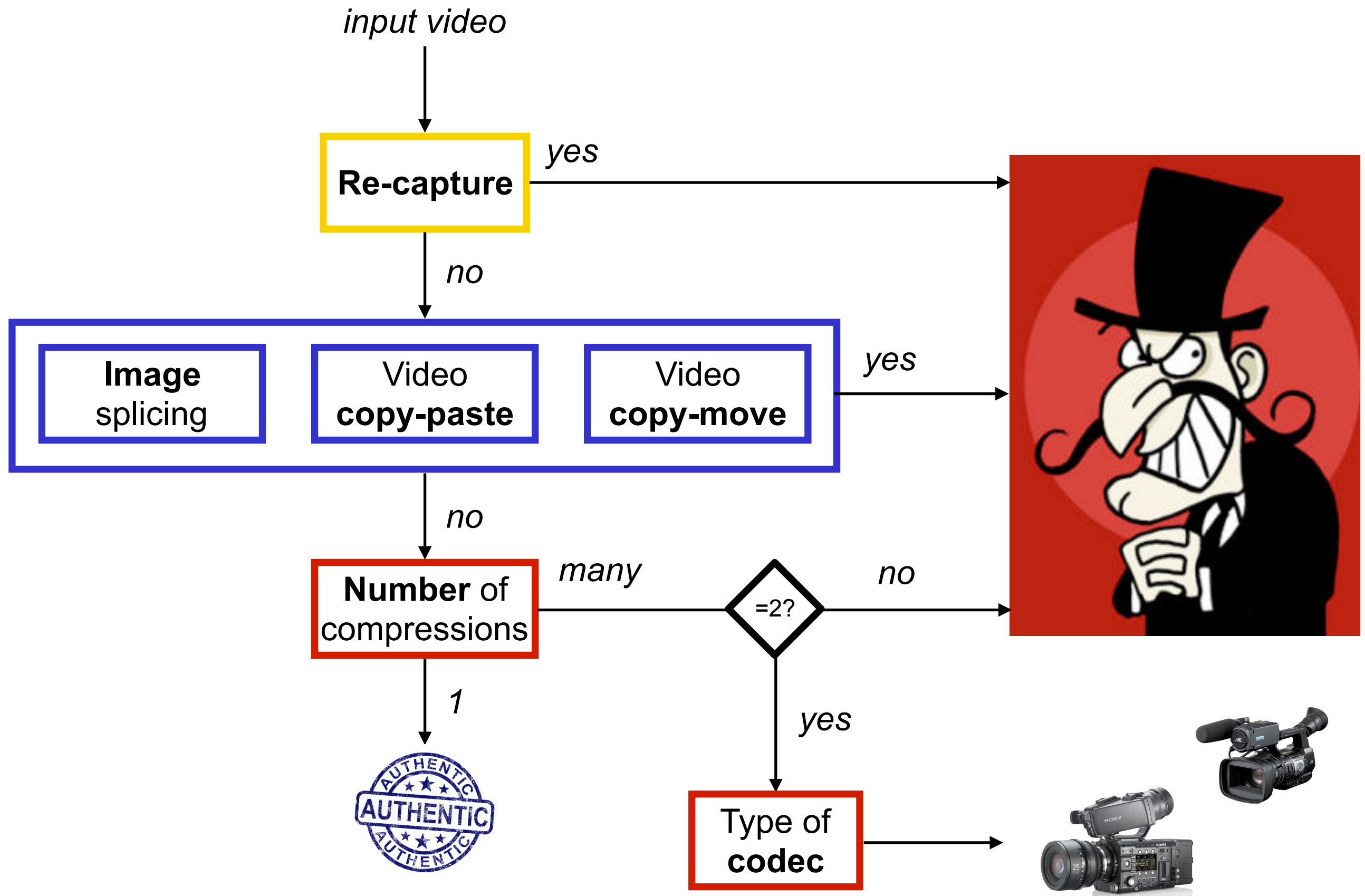
**sed
s**

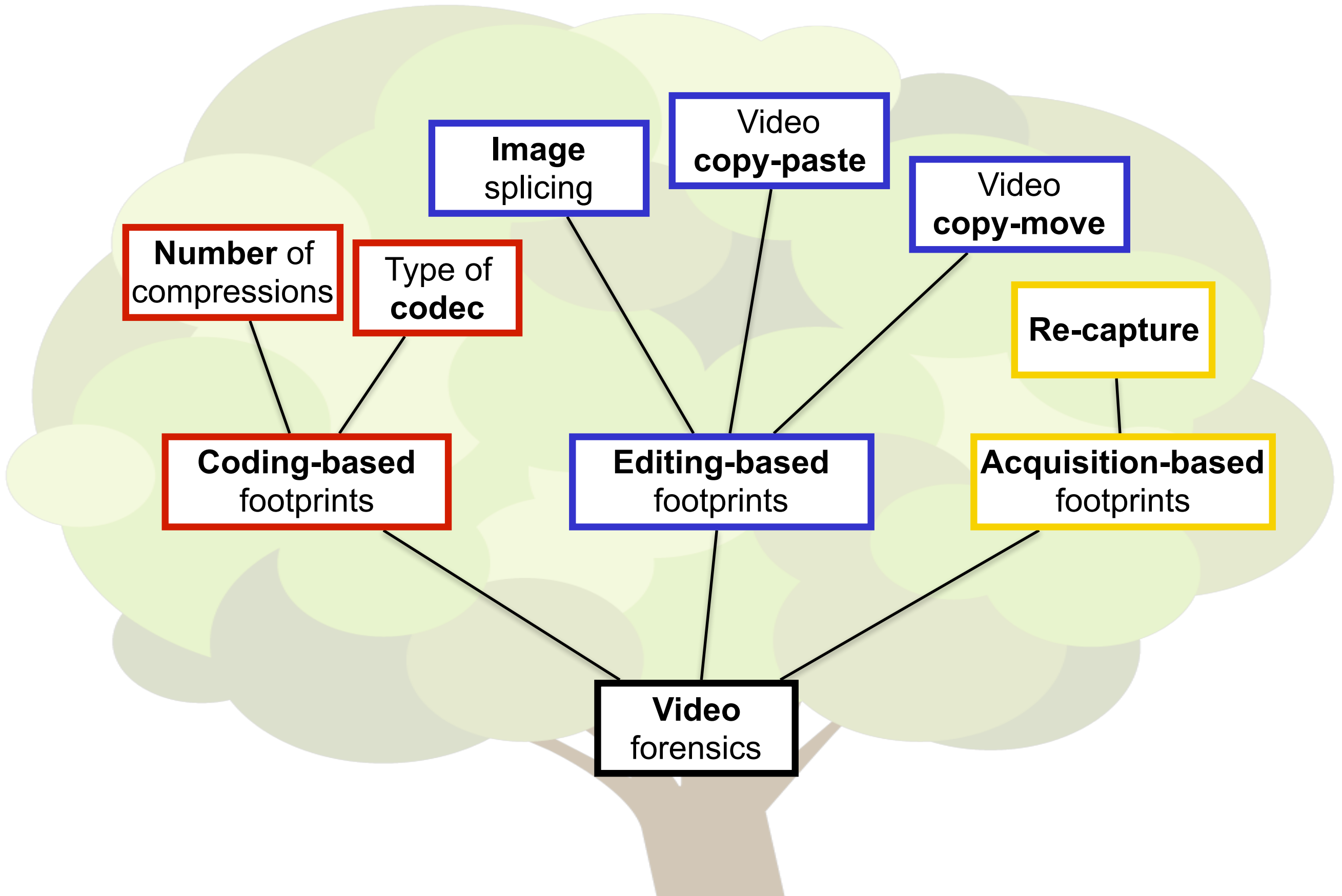


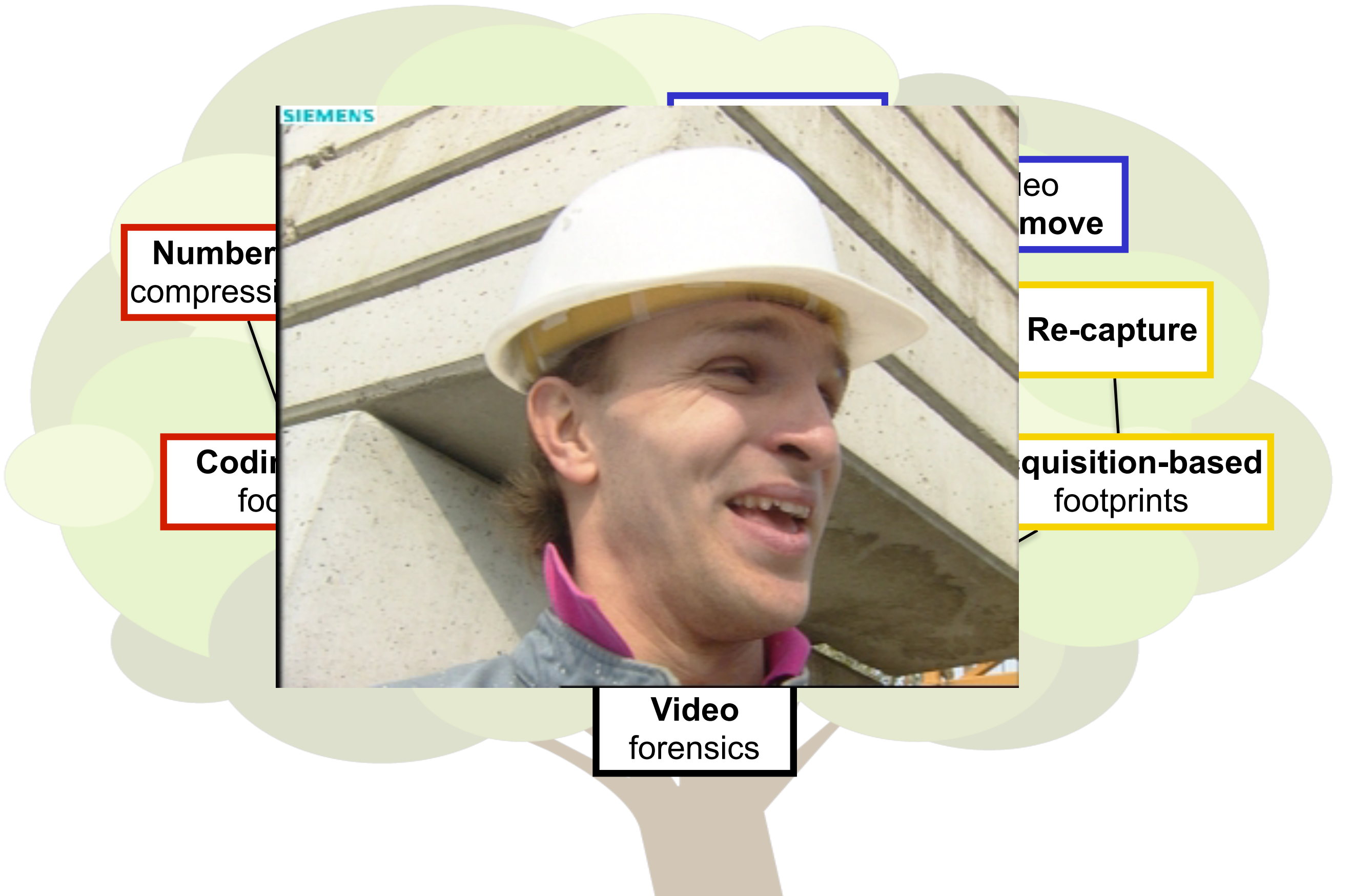
Work organisation



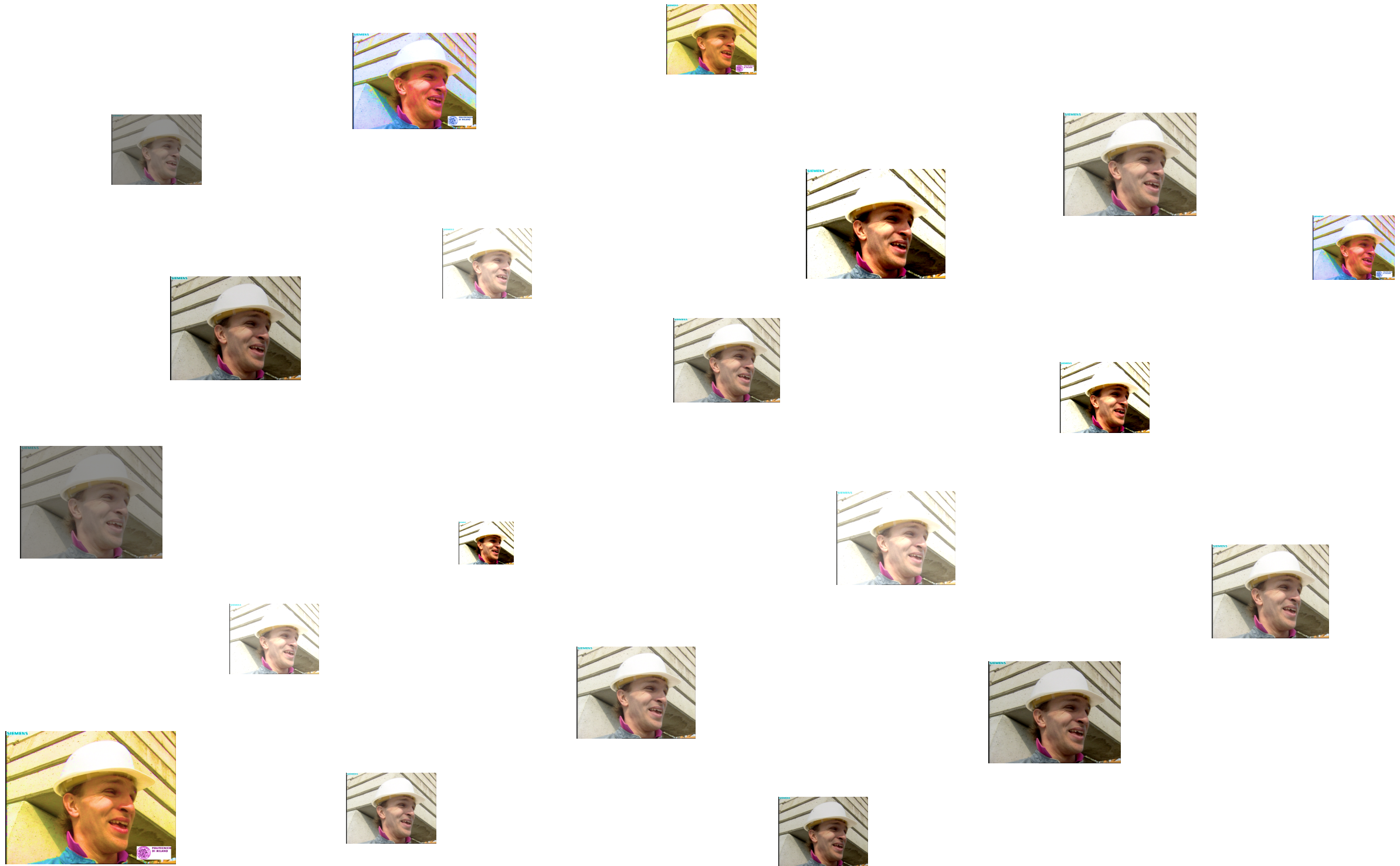
Put everything together



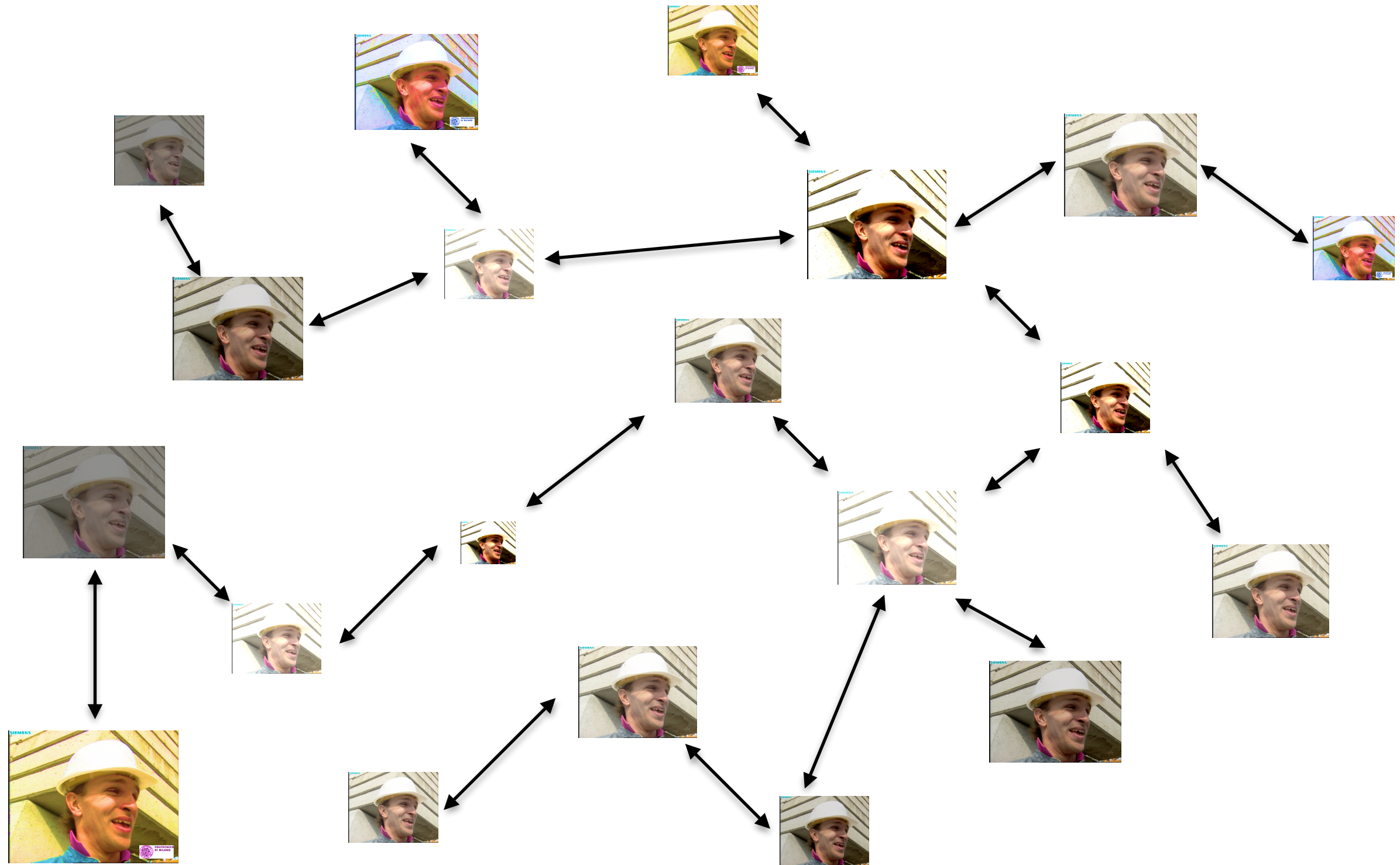




From one to many



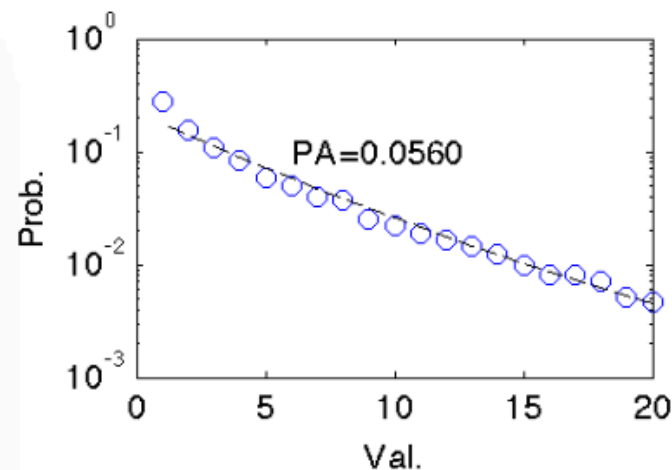
From one to many



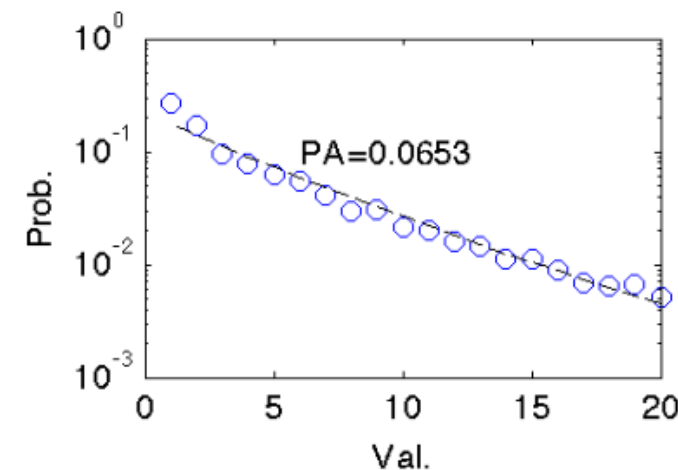


Déjà Vu

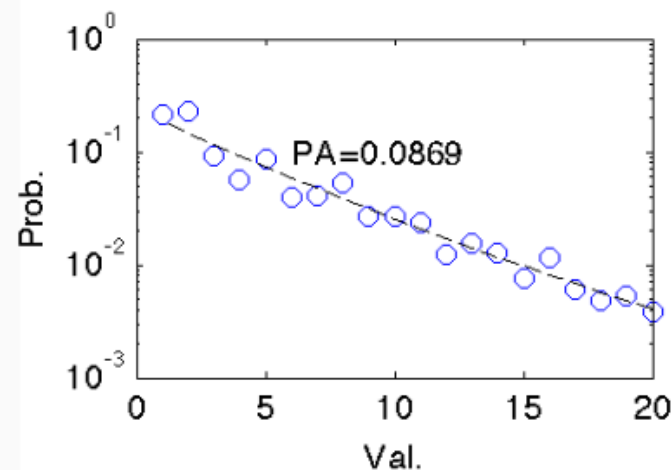
- Which video has been more processed?
 - Extend Benford's law to **base-N** first digits
 - **Fit** logarithmic curve
 - Check goodness of fit (**processing age**)
 - The better, the younger!



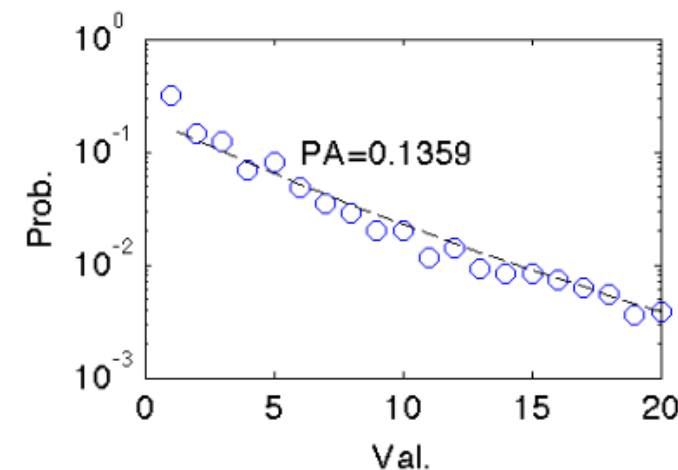
(a) $N_c = 1$ QP=21.



(b) $N_c = 2$ QP=25.

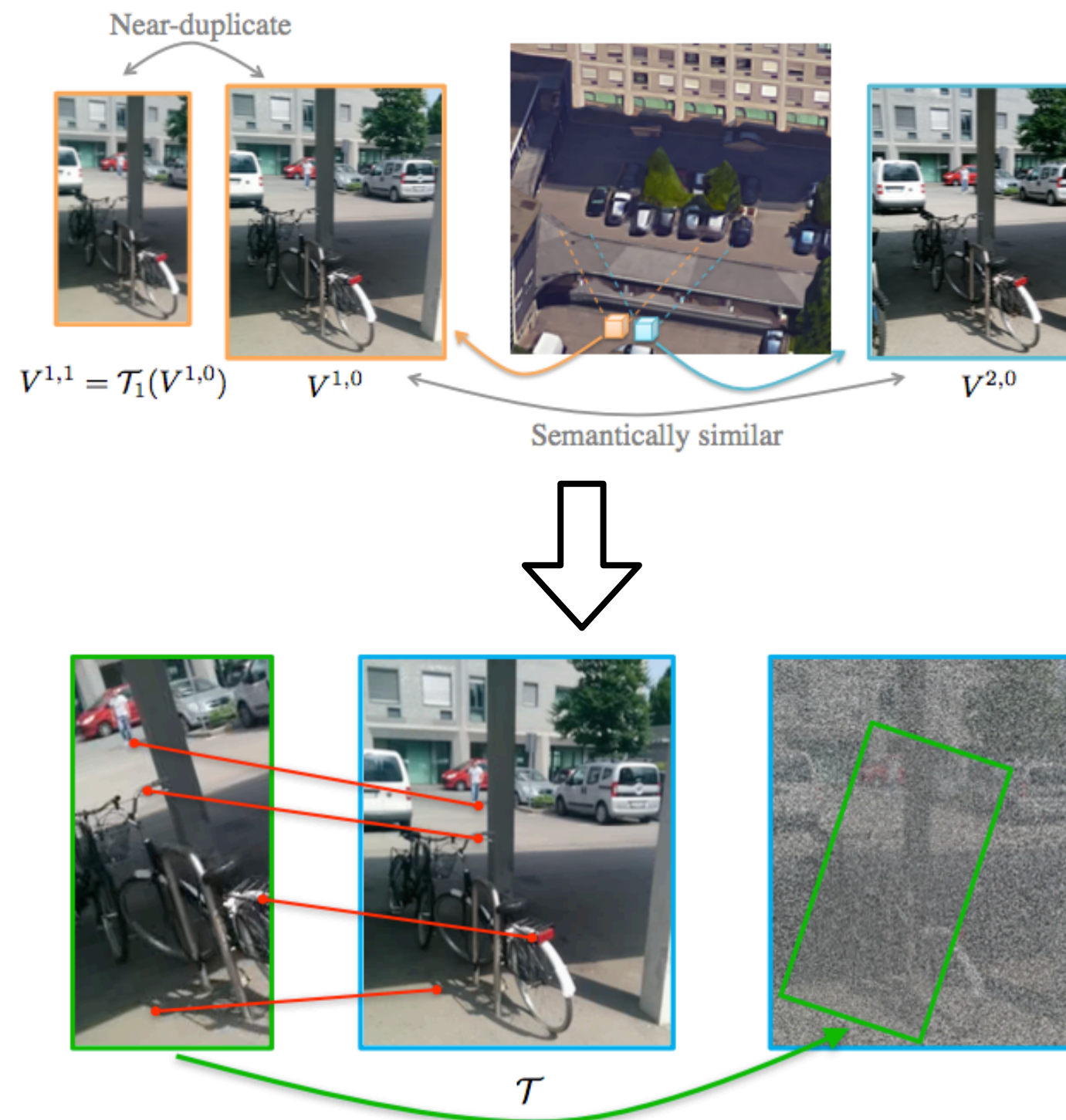


(c) $N_c = 3$ QP=28.



(d) $N_c = 4$ QP=24.

- Which views are the redundant?
 - match video **PRNUs** to detect those from same device



- **Who is my parent?**
- It is possible that we are analysing a short shot (**child**) of a longer sequence (**parent**)
 - e.g., a VIP speech



- Can we find other (partially overlapping) **child** sequences to reconstruct the **parent**?

- Download a set of videos related to the topic under analysis:

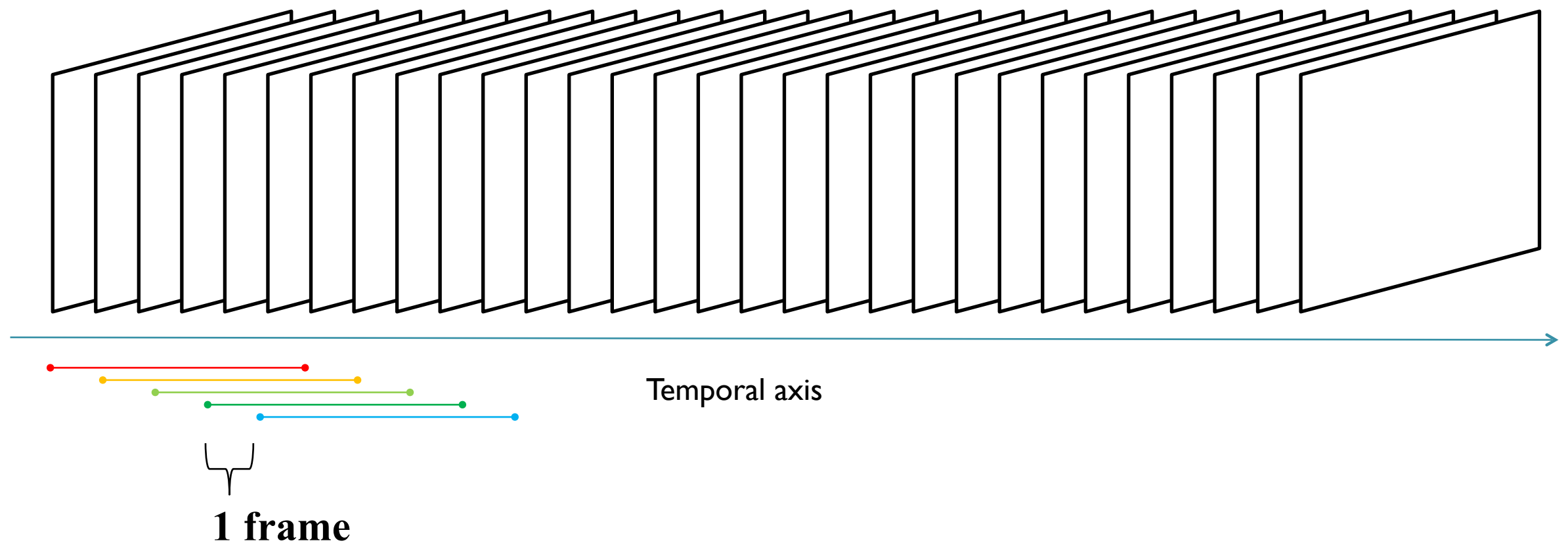


- Download a set of videos related to the topic under analysis:



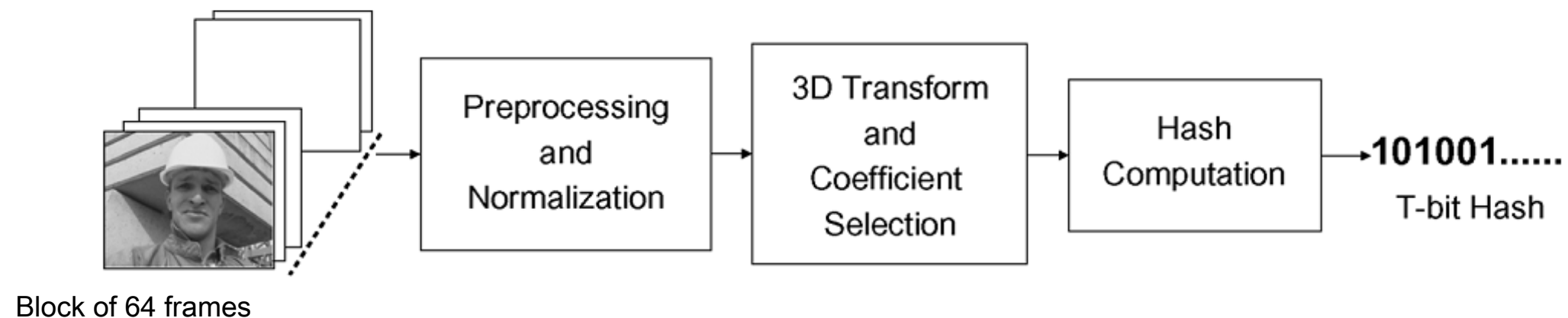
Parent reconstruction

- Analyse each pair of sequences exploiting a **robust hash** algorithm
 - A sequence is split in overlapping time segments of 64 frames each



Parent reconstruction

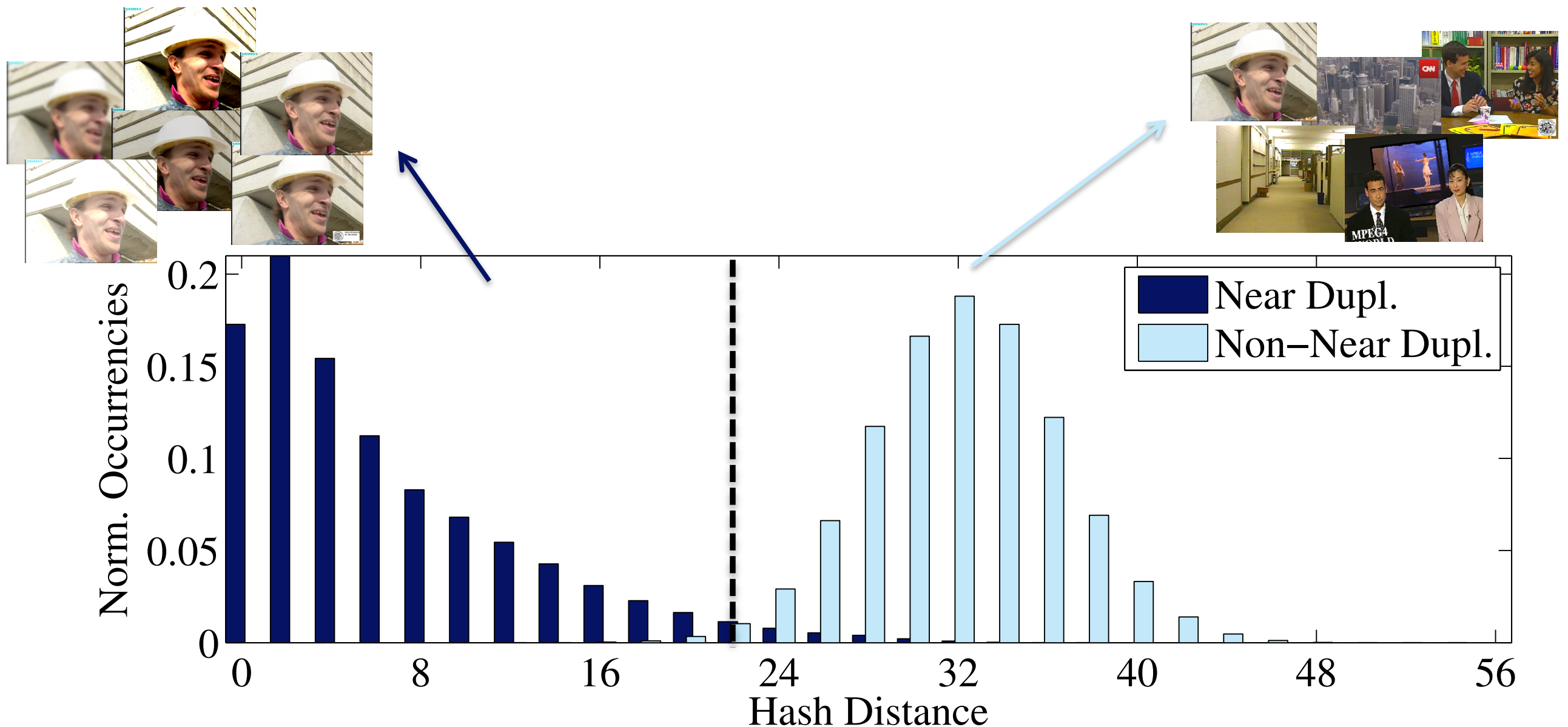
- Analyse each pair of sequences exploiting a **robust hash** algorithm
 - each block is described by a binary hash



- Every frame in the block is **spatially resized** to 32x32 pixels
 - The block now measures 32x32x64 pixels
- **3D DCT** is applied to the block
- **64 DCT coefficients** are selected
- This 64 DCT coefficients are **binarized** according to their median value
 - 32 are set to zero, 32 are set to 1
- The **hash** is this 64 binary string

Parent reconstruction

- Analyse each pair of sequences exploiting a **robust hash** algorithm
 - Hashes of different blocks are compared by computing **hamming distance**

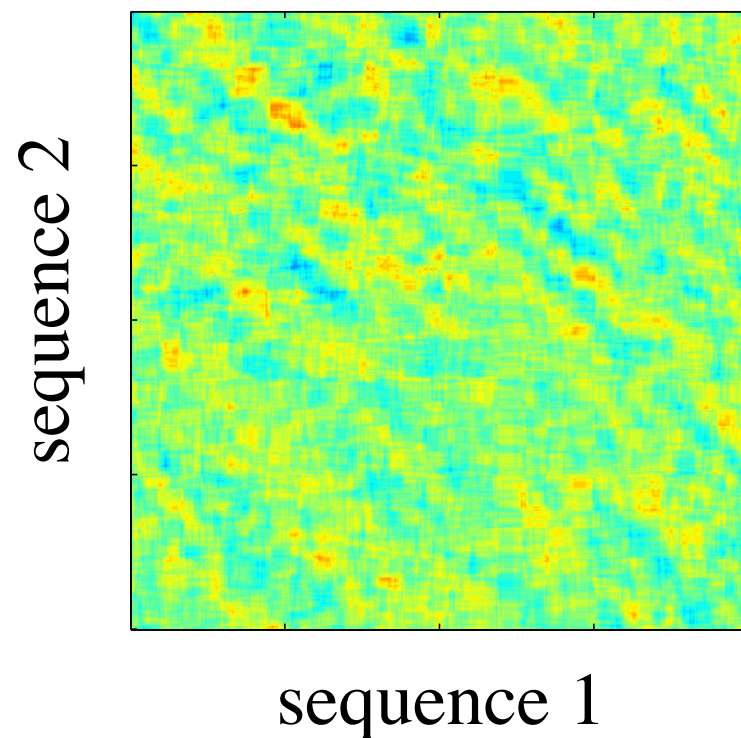


Parent reconstruction

- Compute the **distance between every block** of *sequence 1* and every block of *sequence 2*

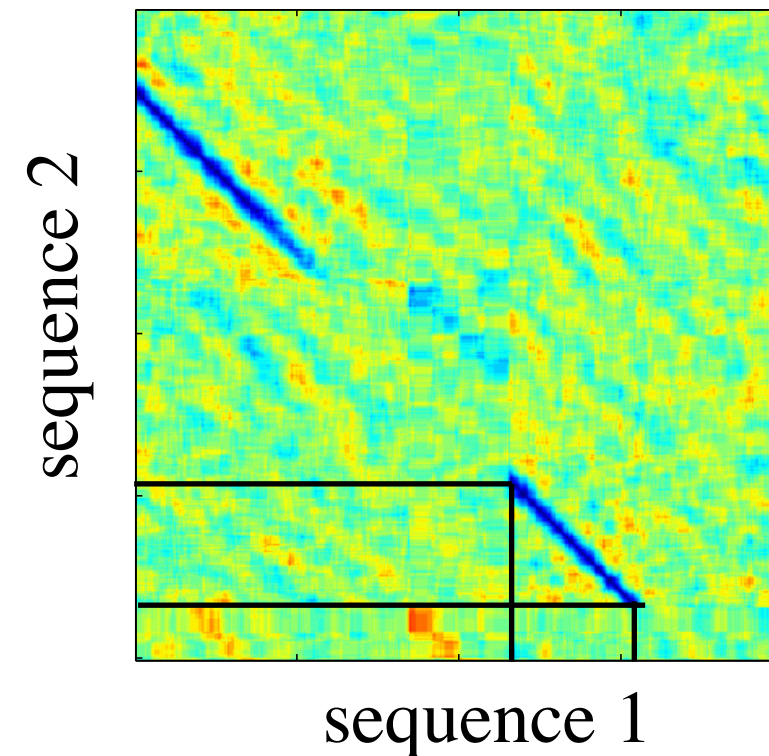
- **Non-near duplicates**

- High distance
- No regular patterns



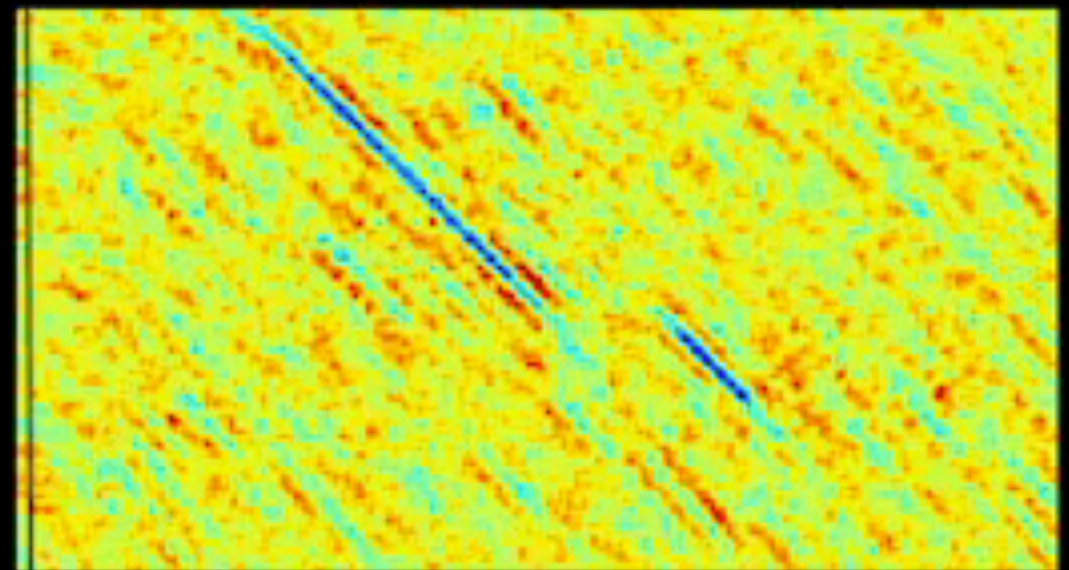
- **Near duplicates**

- Low distances = matching
- Start and end points used for alignment



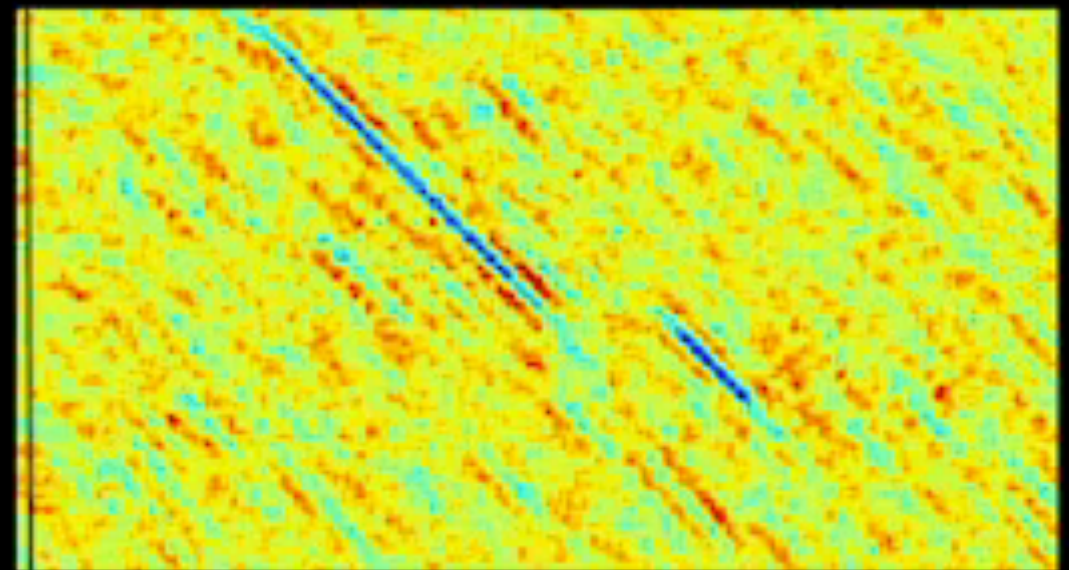
Parent reconstruction

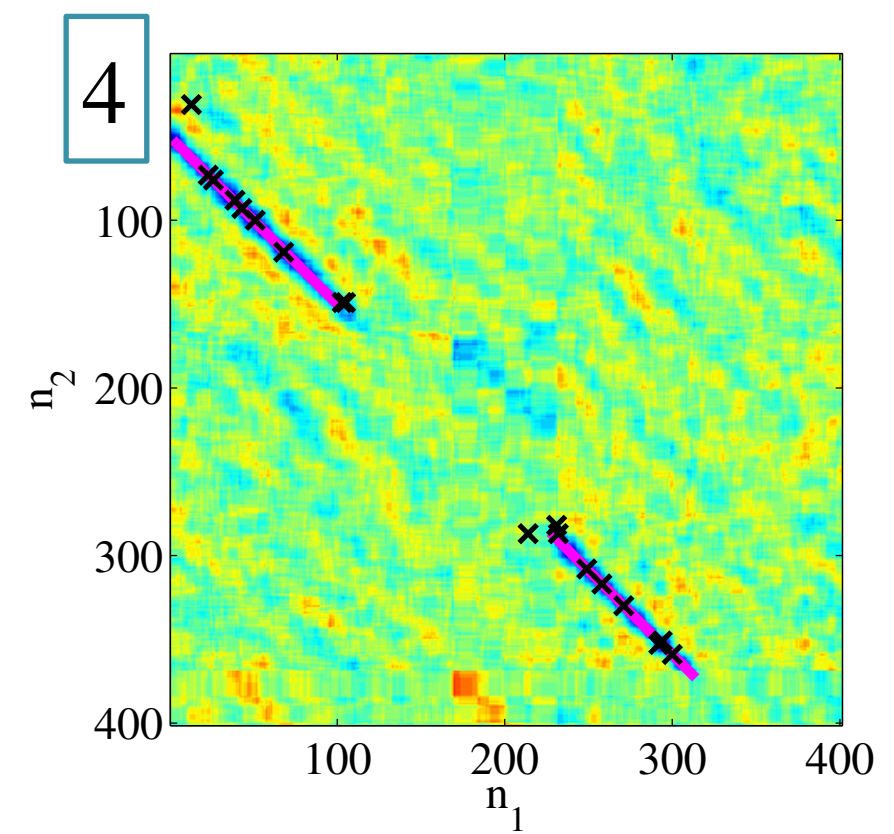
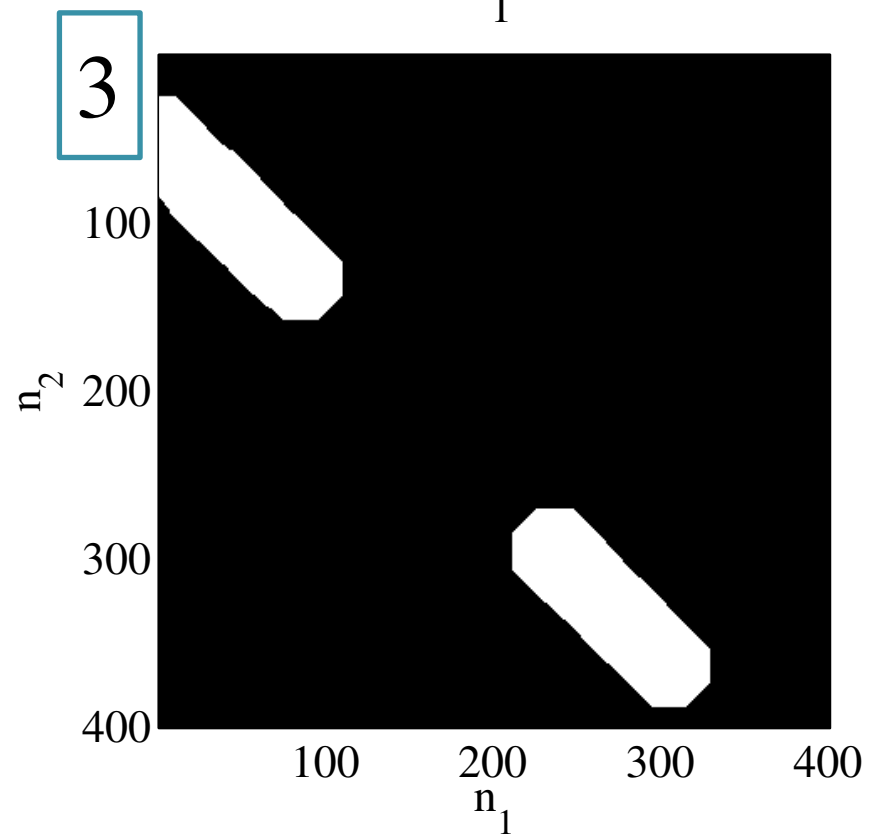
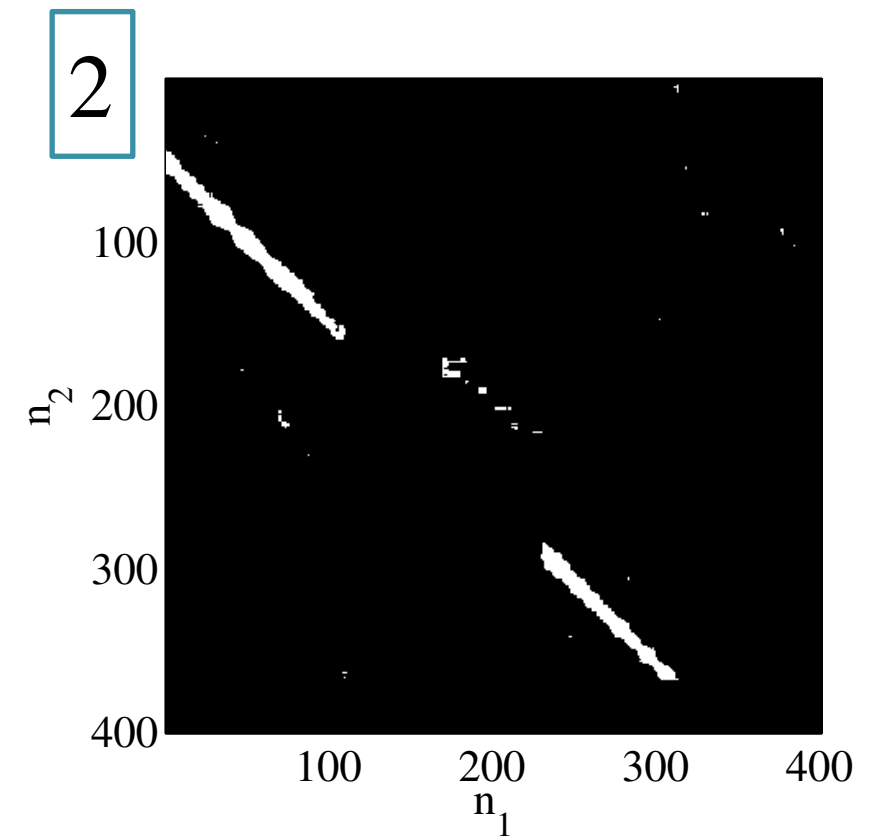
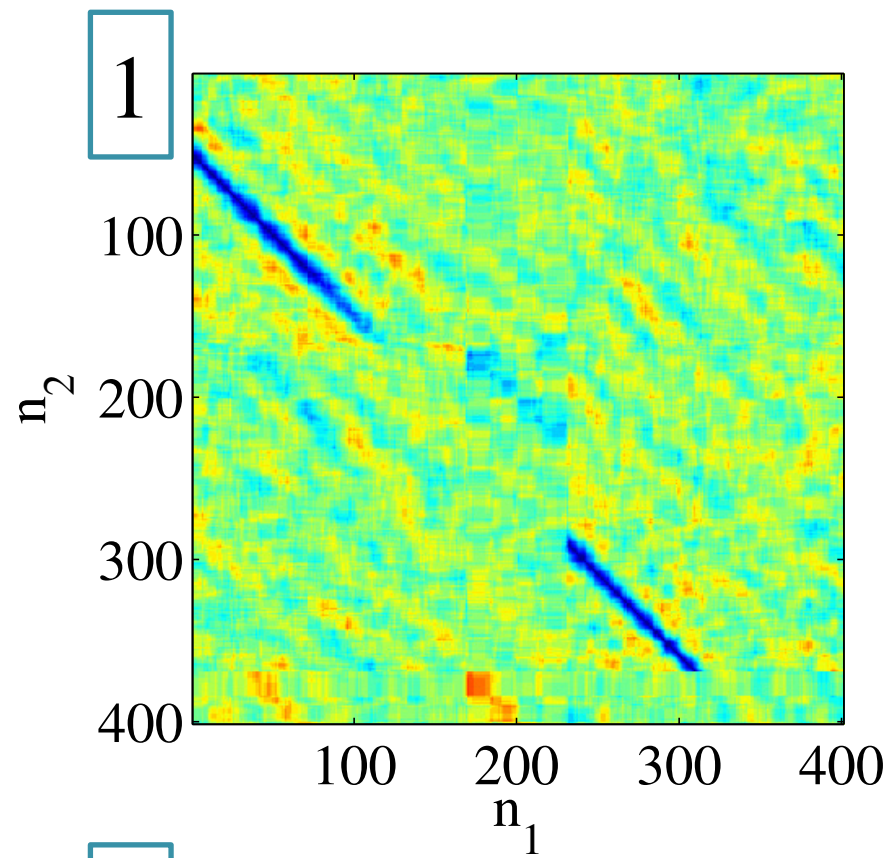
- Analyse each pair of sequences exploiting a robust hash algorithm:



Parent reconstruction

- Analyse each pair of sequences exploiting a robust hash algorithm:





Parent reconstruction

- Segment each sequence according to the matching shots:



Parent reconstruction

- Segment each sequence according to the matching shots:



- Reconstruct the most part of the **parent** sequence for the analysis:

Reconstruction of parent



from child 1

- Reconstruct the most part of the **parent** sequence for the analysis:

Reconstruction of parent



from child 1

- Being able to **reconstruct the parent** from the children enables to shed very interesting insights on the way content is reused:
 1. Analyse the **context** from which a child sequence was taken
 2. **Reconstruct** sequences no longer available online in their totality
 3. Establish causal **relationship** between children

Conclusions

- **Remarks**

- Forensics vs. **Anti-forensics**
- **Single video** analysis is just part of the problem
- **Multiple video** analysis paves the way to the development of novel applications

- **Open questions**

- Merge results from **content-** and **context-aware** detectors
 - Do metadata match the video content?
- Deal with **big data**
 - Time-consuming algorithms need optimisation
- **Deep learning**
 - Still under-investigated in video forensics (space-time?)
 - Training data hardly available...

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**Thank you for
the attention!**

Any questions?

