Weakly Supervised Training for Hologram Verification in Identity Documents

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Code and data: https://github.com/EPITAResearchLab/pouliquen.24.icdar

Introduction

Context: remote verification of ID documents to comply with Know Your Customer (KYC) regulation

Task: Integrity verification of Optically variable devices (OVDs) or "holograms" used to secure ID docs.

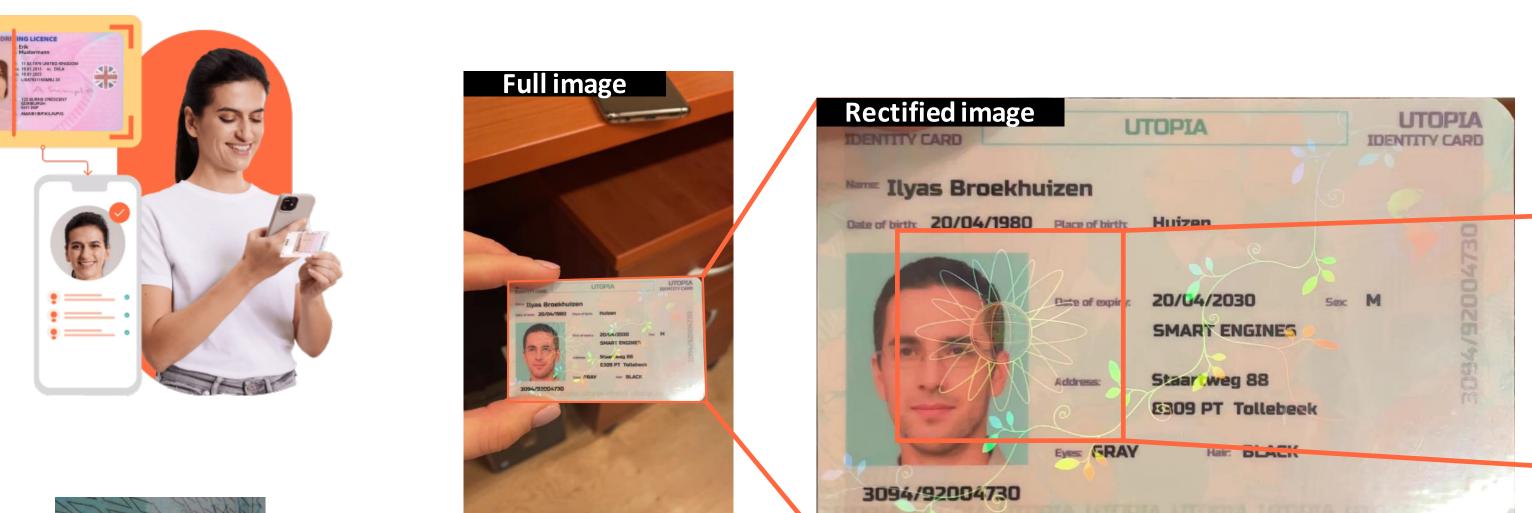
Challenges: Weak signals, changing

background (face picture)

How to model and verify appearance

and **behaviour**?

Data: MIDV Datasets





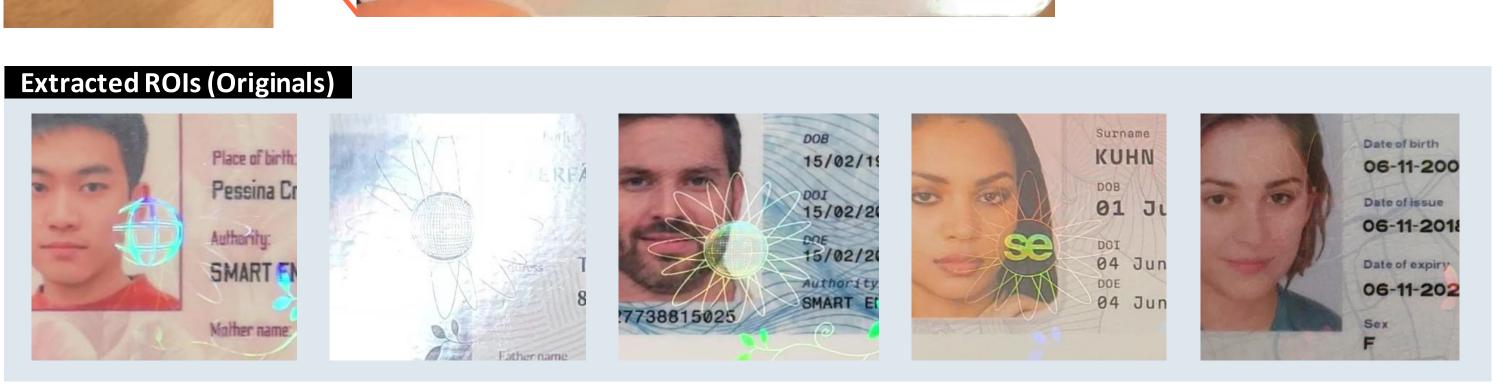


Fig 1. Extracted ROIs from the MIDV-Holo dataset

Method

Input: Sequence of cropped images **Output:** Authenticity indicator

Main contributions:

- Learn hologram representation with contrastive loss
- Robust to static attacks: checks holographic behaviour
- Weakly supervised: video level annotation

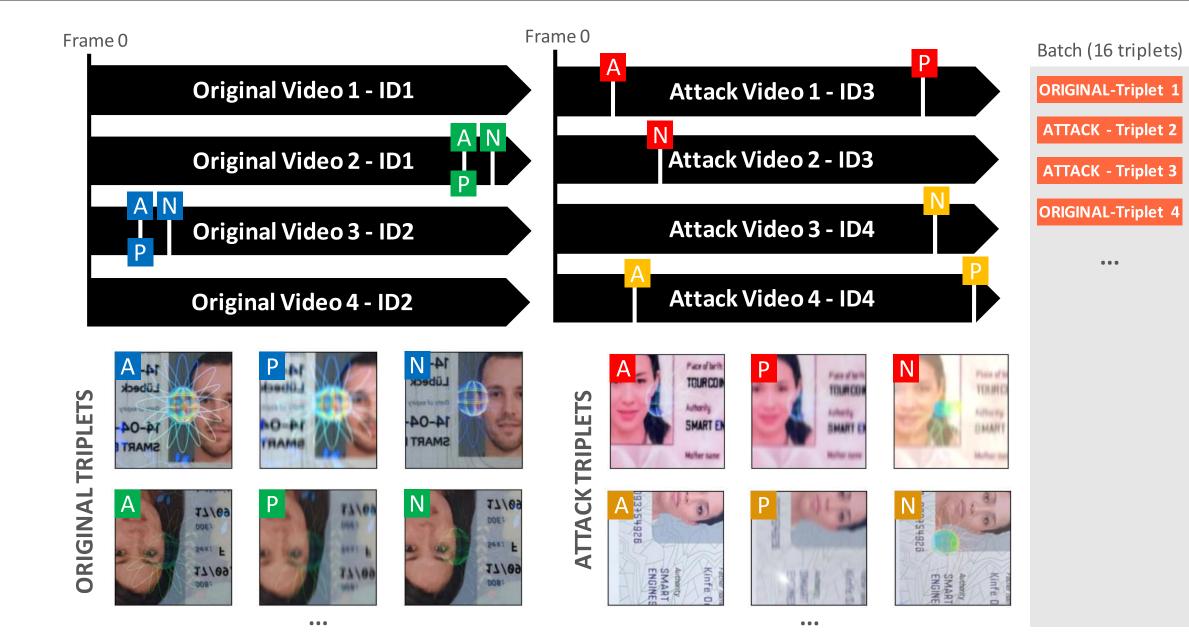


Fig 3. Frame sampling overview

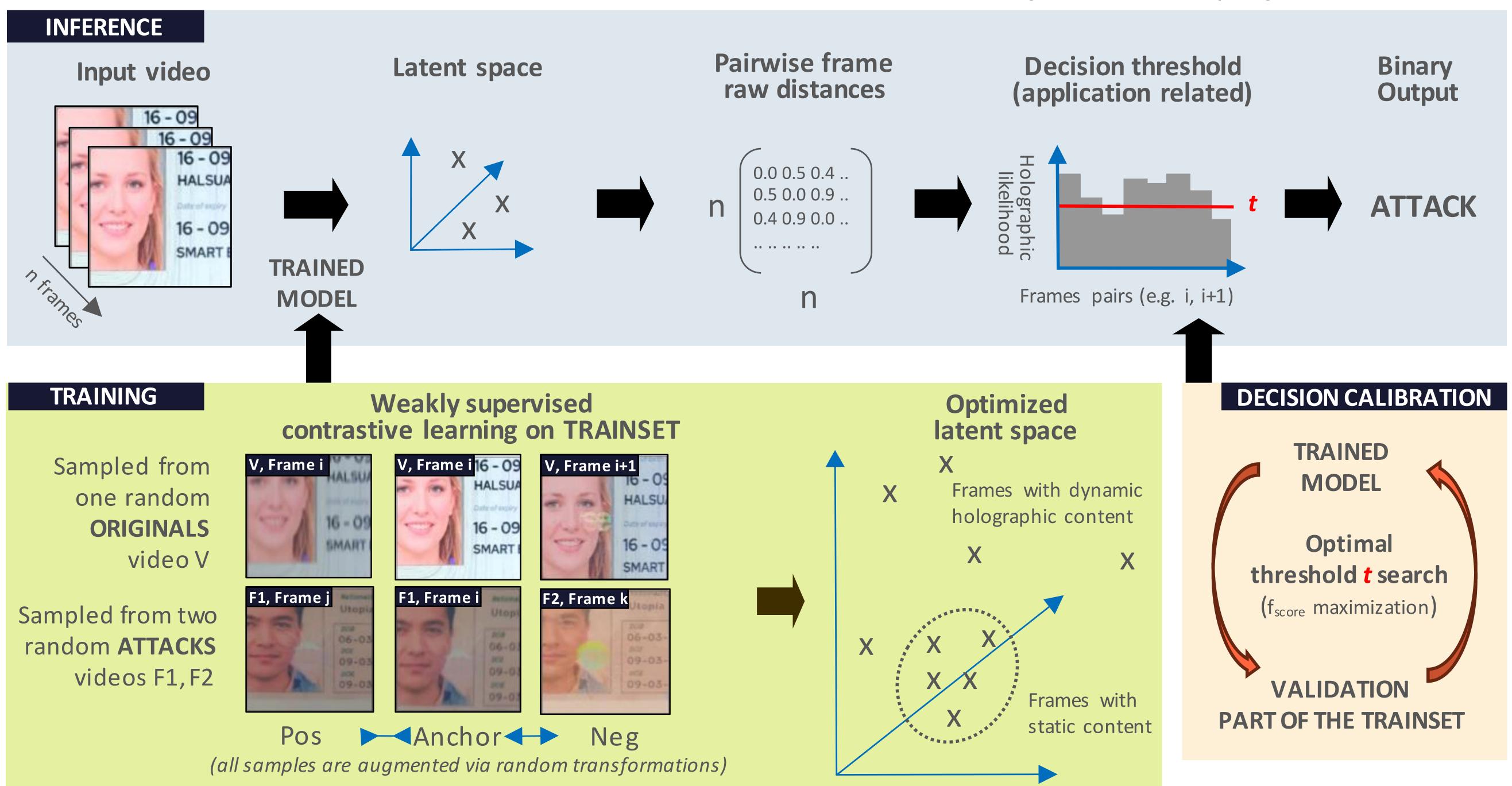


Fig 2. Proposed approach overview: 1) Hologram representation that discards the background 2) Control dispersion of the representations

Experiments

- Trained on 1 dataset (MIDV-Holo "Vanilla")
- Focus on the face picture region
- Tested on 3 datasets
- Comparison to a reproduced MIDV-Holo ROI
- Evaluated on attack detection

${\bf Test\ dataset} \rightarrow $	MIDV-HOLO "Vanilla"	MIDV-HOLO "Photo repl."	MIDV 2020 "Clips"
	(120 mixed vids)	(20 attack vids)	(1k attack vids)
$\boxed{\hspace{1.5cm} Metric \to }$	F_{score} (%)	Recall (%)	Recall (%)
$oxed{ ext{Method}}\downarrow$, , ,	, ,
MIDV-Holo ROI	80 ± 3	63 ± 10	92 ± 2
OUR - $mobilevit_{xxs}$	90 ± 2	87 ± 14	93 ± 6