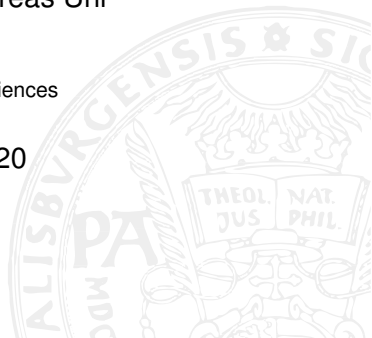


# Improved Liveness Detection in Dorsal Hand Vein Videos using Photoplethysmography

**Johannes Schuiki, Andreas Uhl**

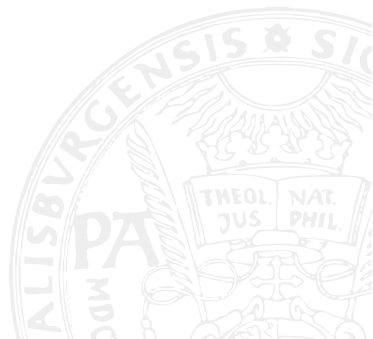
University of Salzburg  
Department of Computer Sciences

September 17, 2020

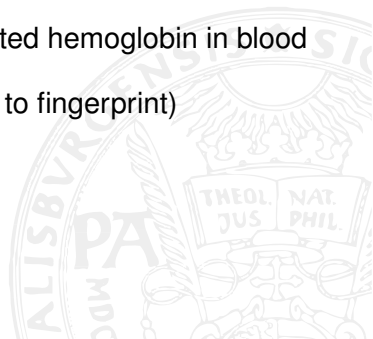


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- Pattern of blood vessels under the skin has become an emerging biometric trait due to uniqueness
- Captured through near-infrared (NIR) illumination
- NIR light is absorbed by oxygen-saturated hemoglobin in blood
- Can be used for authentication (similar to fingerprint)

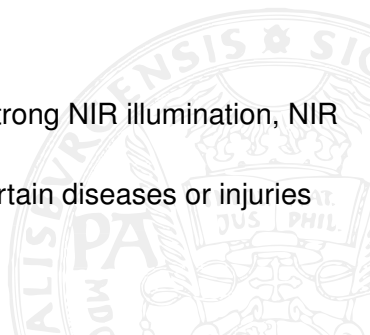


## Advantages

- Hard to forge since intrinsic trait and only visible through NIR illumination
- No abrasion as with fingerprints
- Invariant to sweat, sunscreen, etc.
- Non-intrusive

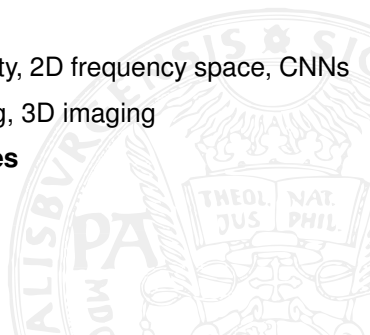
## Disadvantages

- Special hardware needed (Relatively strong NIR illumination, NIR imaging sensor, ... )
- Vein structure may be influenced by certain diseases or injuries
- In general low contrast and quality



# The Presentation Attack Problem

- Scenario: Access to previously captured hand vein data
- Presentation Attacks
  - Printed on paper [1]
  - Shown on smartphone display [2]
- Countermeasures
  - Still Images: Skin texture, image quality, 2D frequency space, CNNs
  - Additional sensors: capacitive sensing, 3D imaging
  - **Differences in adjacent video frames**



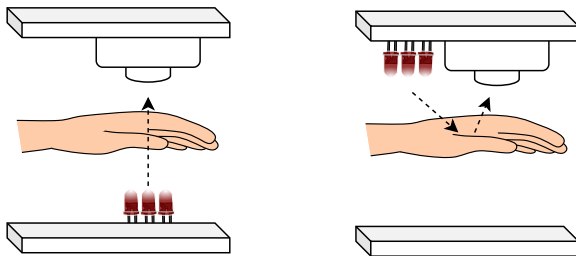
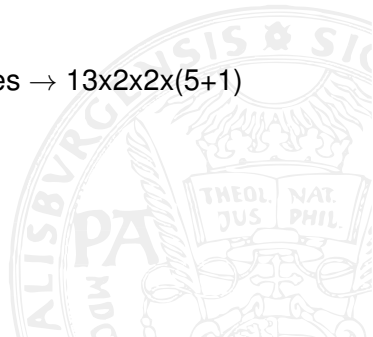


Figure: Modes of operation: Transillumination (left) and Reflected Light.



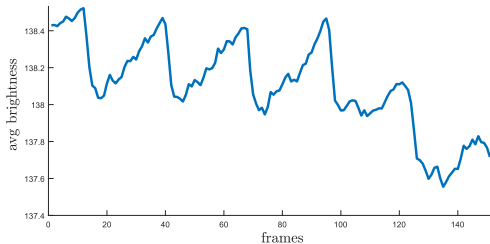
**Figure:** Example frames from NIR videos; top row: transillumination, bottom row: reflected light; left: real video frame, middle column: paper attack and right: smartphone attack, respectively. Captured with [3].

- 13 Persons à 2 Hands
- 2 Illumination Variants
- 5 Presentation Attack Scenarios
- Database in total: 312 Video Sequences  $\rightarrow 13 \times 2 \times 2 \times (5+1)$

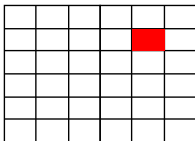




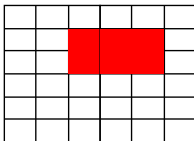
Related approaches build upon a common basis:



a) Pixel



b) ROI

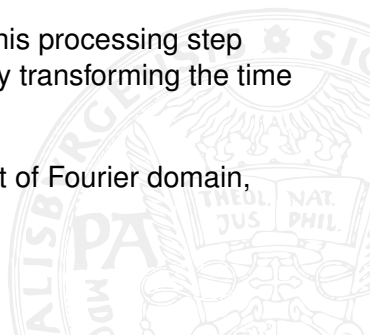


c) Entire Image

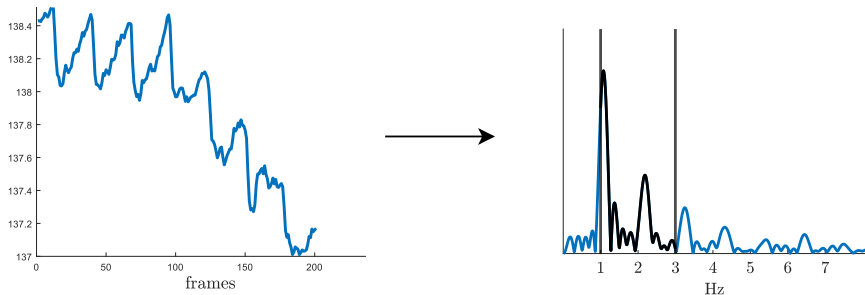


Figure: Idea of Average Pixel Illumination.

- Exception: [4] used "Eulerian motion magnification" → [5] used the described spoofing database to show that this method can still be fooled
- In [6], the peaks and valleys of the time series were used as classification criteria
- [7, 8, 9] reported the observation that this processing step contains information about heart rate by transforming the time series to Fourier domain
- Bok et al. [10] constructed classifier out of Fourier domain, although for finger vein videos



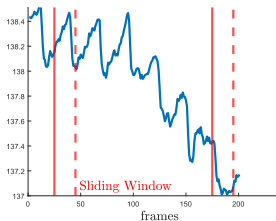
## Reference Method: Bok et al.



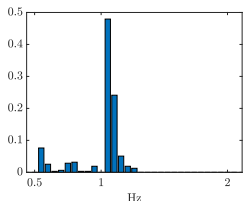
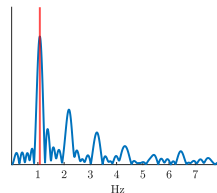
*Feature Vector*  $\in \mathbb{R}^{50}$

Figure: Feature Vector Construction Bok et al.

## Introduced Method 1



*PerWindow :*  
*Highpass Filtering*  
*Zero padding*  
*Fourier transform*

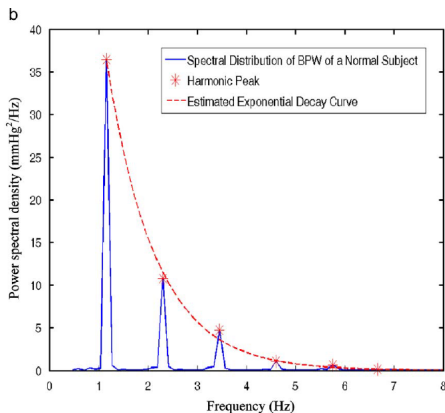


*Histogram Majority – voting*

*Feature Vector  $\in \mathbb{R}^{30}$*

**Figure:** Feature Vector Construction Method 1.

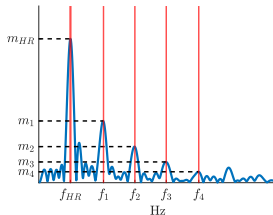
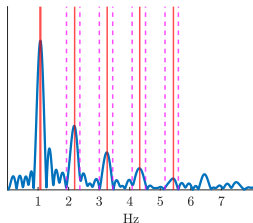
## Observation from Wei et al. [11]



**Figure:** Blood pressure measurement in Fourier space have harmonics that can be modelled through exponential decay. Figure taken from [11].

## Introduced Method 2

Like before: Fourier transform for every window



Use Means & Medians over all windows as Feature Vector.  
Note: Actually use Quotient w.r.t  $f_{HR}$  &  $m_{HR}$

$$\left( \frac{\overline{m_{HR}}}{f_{HR}}, \frac{\overline{m_1}}{f_1} \dots \frac{\overline{m_4}}{f_4}, Md(m_{HR}), Md(m_1) \dots Md(m_4), \right. \\ \left. \frac{\overline{f_{HR}}}{m_{HR}}, \frac{\overline{f_1}}{m_1} \dots \frac{\overline{f_4}}{m_4}, Md(f_{HR}), Md(f_1) \dots Md(f_4) \right)$$



Feature Vector  $\in \mathbb{R}^{20}$

Figure: Feature Vector Construction Method 2.

## Evaluation metrics according to ISO/IEC 30107-3:2017

- **Attack Presentation Classification Error Rate (APCER):**  
proportion of attack presentations incorrectly classified as bona fide presentations in a specific scenario
- **Bona Fide Presentation Classification Error Rate (BPCER):**  
proportion of bona fide presentations incorrectly classified as presentation attacks in a specific scenario

# Experimental Results II

RBF Kernel  $C=10$ ,  $\gamma = 0.001$

Spoof Method		Bok et al.		Method 1		Method 2	
		APCER	BPCER	APCER	BPCER	APCER	BPCER
Transill.	Paper	<b>10.31</b>	30.77	11.54	15.38	11.54	<b>3.85</b>
	Paper Mov.	<b>4.35</b>	37.50	26.92	11.54	11.54	<b>0.00</b>
	SP	<b>0.00</b>	89.42	3.85	7.69	<b>0.00</b>	<b>0.00</b>
	SP Mov.	77.36	89.42	<b>0.00</b>	19.23	19.23	<b>0.00</b>
	SP Zoom	59.43	17.31	<b>3.85</b>	19.23	11.54	<b>0.00</b>
Ref. Light	Paper	<b>0.00</b>	80.37	53.85	61.54	7.69	<b>23.08</b>
	Paper Mov.	49.06	5.61	<b>23.08</b>	<b>3.85</b>	42.31	7.69
	SP	63.55	6.54	7.69	<b>3.85</b>	<b>3.85</b>	<b>3.85</b>
	SP Mov.	23.08	6.54	3.85	<b>0.00</b>	<b>0.00</b>	3.85
	SP Zoom	15.24	7.48	3.85	<b>0.00</b>	<b>0.00</b>	3.85

**Table:** The table shows the SVM results with a RBF kernel, BoxConstraint of 10 and a  $\gamma$  value of 0.001 as proposed in [10]. Best results are highlighted **bold**.



# Experimental Results III

## Linear SVM Kernel

Spoof Method		Bok et al.		Method 1		Method 2	
		APCER	BPCER	APCER	BPCER	APCER	BPCER
Transill.	Paper	11.34	18.27	<b>7.69</b>	26.92	11.54	<b>0.00</b>
	Paper Mov.	<b>7.61</b>	16.35	26.92	19.23	11.54	<b>7.69</b>
	SP	13.21	54.81	7.69	26.92	<b>0.00</b>	<b>0.00</b>
	SP Mov.	17.92	57.69	11.54	11.54	<b>3.85</b>	<b>0.00</b>
	SP Zoom	27.36	40.38	7.69	11.54	<b>0.00</b>	<b>0.00</b>
Ref. Light	Paper	12.84	71.03	61.54	53.85	<b>0.00</b>	<b>3.85</b>
	Paper Mov.	25.47	<b>1.87</b>	15.38	3.85	<b>3.85</b>	7.69
	SP	46.73	6.54	7.69	19.23	<b>3.85</b>	<b>3.85</b>
	SP Mov.	28.85	6.54	3.85	<b>3.85</b>	<b>0.00</b>	<b>3.85</b>
	SP Zoom	18.10	5.61	<b>3.85</b>	<b>0.00</b>	<b>3.85</b>	3.85

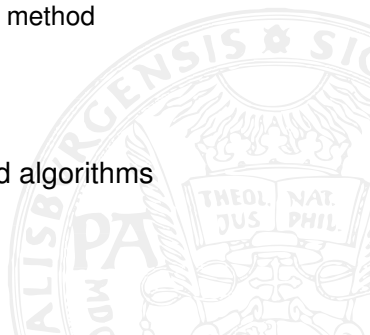
**Table:** The table contains results with a simple linear kernel. Best results are highlighted **bold**.

## Contribution

- Evaluated an existing Presentation Attack Detection Method for finger vein biometrics on a custom dorsal hand vein data set.
- Proposed two additional methods for PAD, employing spectral analysis of the average pixel illumination per frame.
  - Superior with respect to the reference method
  - High time consumption

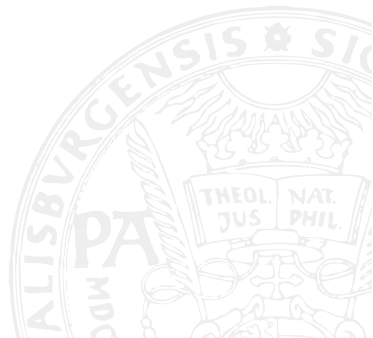
## Future Work

- Reduce computational cost of proposed algorithms
- Acquire more data samples



Thank you for your attention!

Thank You!  
Q & A



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