MICHIGAN STATE UNIVERSITY



Generalizing Fingerprint Spoof Detector: Learning a One-Class Classifier

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Spoofing TouchID



Successful spoof attack on iPhone 5S by German Hacking Club

Spoofs are a specific type of presentation attack Figure retrieved from, https://www.ccc.de/de/updates/2013/ccc-breaks-apple-touchid

Spoof Detection Systems



Goal: automatically detect and flag spoofs prior to authentication or search

Challenge: Large Variety of Spoofs



Two-class (Spoof vs. Live) Classifier



Material

Problems:

- Two-class classifiers prone to overfit to specific materials
- Fail to detect spoofs of "unseen" materials; studies report 3-fold increase in error

Solution: One-Class Classifier



Goal:

- Only train with live fingerprint samples
- Generalize to ALL
 spoofs

Challenges:

- Need large *number* of *diverse* live training samples
- Grayscale live and spoof images are
 very similar

Training: 3 DCGANs only on Live Impressions



Idea: Discriminators separate *real* live from *synthesized* live; can be used to distinguish live and spoof fingerprints

Testing: Discriminate between Live and Spoof



Fused discriminators distinguish live and spoof fingerprints

RaspiReader







FTIR image (1900 ppi) 🗡 direct image (1900 ppi)

Open-source, multi-image, highresolution fingerprint reader

Source: <u>github.com/engelsjo/RaspiReader</u> DIY video: <u>bit.do/RaspiReader</u>

RaspiReader Dataset



Example: RaspiReader Live Impression



RaspiReader Live Data

- •6,000 unique fingers
- 11,880 impressions
- 3 collection locations (MSU, Clarkson, JHUAPL)

RaspiReader Spoof Data

- 12 materials
- •5,531 impressions

Larger and more diverse dataset than LiveDet

Example: RaspiReader Spoof Impression

Preprocessing: ROI Extraction

(1) Obtain ROI Mask



(2) Crop ROI from RaspiReader images



Remove background noise which both live and spoof images share

Training and Testing Protocol

Live Dataset	Training Location*	Testing Location	# Trai Impres	ining #` ssions Im	Validation pressions	# Testing Impressions					
	CU & JHUAPL	MSU 8,3		30	500	3,050					
-	* Location refers to the site where data was collected										
Spoof Partition Set_1	Training Materials	Testing Materials		# Training Impressions	# Validation Impressions	# Testing Impressions					
	Dragonskin, Ecoflex, Crayola, Paper, Body Latex, Monster Latex	Gelatin, Woodglue, Pigmented, Gold, Transparency		2,851	134	2,312					
Spoof	Training Materials	Testing Mate	erials	# Training Impressions	# Validation 5 Impressions	# Testing Impressions					
Partition Set ₂	Gelatin, Woodglue, Pigmented, Gold, Transparency	Dragonskin, E Crayola, Pape Latex, Monste	coflex, r, Body r Latex	2,195	117	2,985					

Experimental Results

	Algorithm	Gelatin I	Pigmented	Playdoh	Woodglue	Transparency	Gold
Testing Partition Set ₁	OCSVM [1]	0.0%	2.0%	0.3%	0.0%	0.0%	0.9%
	CNN [2]	67.7%	29.4 %	6.0%	55.7%	34.0%	11.8%
	Proposed	74.5%	22.3%	96.3 %	85.2 %	94%	34.2%
	Algorithm	Dragonski	n Ecoflex	Monste Latex	er Crayo	a Body Latex	Paper
Testing	Algorithm OCSVM [1]	Dragonski 0.0%	n Ecoflex 0.2%	Monste Latex 0.3%	er Crayol 15.0%	a Body Latex	Paper 33.8%
Testing Partition	Algorithm OCSVM [1] CNN [2]	Dragonski 0.0% 49%	 n Ecoflex 0.2% 39.3% 	Monste Latex 0.3% 54.3%	er Crayol 15.0% 5 78.1%	Ia Body Latex 6 20.6% 6 12.1%	Paper 33.8% 46.1%

 [1] Yaohui Ding, and Arun Ross. "An ensemble of one-class svms for fingerprint spoof detection across different fabrication materials." IEEE WIFS (2016)

• [2] Joshua J. Engelsma, Kai Cao, and Anil K. Jain. "Raspireader: Open source fingerprint reader." IEEE TPAMI (2018).

True Detection Rate (TDR) @ False Detection Rate (FDR) = 0.2%

Experimental Analysis



2D tSNE plot of live and spoof features

Outperforms two-class CNN in 7 / 12 testing materials



(playdoh) (gold finger) GAN successes; CNN failures



(ecoflex) (live finger) GAN failures (thin clear spoofs)

On-going Research

- Fuse two-class CNN and proposed one-class classifier
- Benchmark one-class classifier on public datasets (LiveDet)
- Additional evaluation of one-class classifier on "unseen" materials

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Questions?

