





# rtCaptcha: A Real-Time CAPTCHA Based Liveness Detection System



<u>Erkam Uzun</u>, Simon Pak Ho Chung, Irfan Essa and Wenke Lee Department of Computer Science Georgia Institute of Technology, USA Background Cloud Services Attacks Defense Methods Threat Sec. of Current System System System System Conclusion

**Face Authentication Systems** 

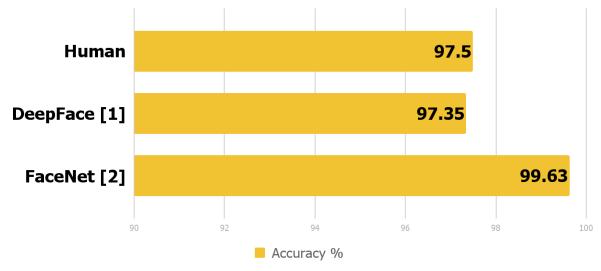
**Background** 





# **Deep Learning Outperforms**

#### Face recognition performance on LFW dataset





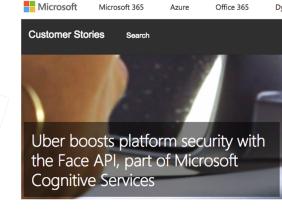
Background Cloud Services Attacks Defense Methods Threat Sec. of Current Proposed User Sec. of Proposed Conclusion



## **Deployed by Major Companies**





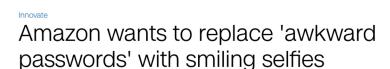


#### **Face Verification Cloud Services**

- Microsoft Cognitive Services [3]
- Amazon Rekognition [4]
- Face++ [5]
- Kairos Human Analytics [6]







GADGETS

by Ivana Kottasova @ivanakottasova

cm tech



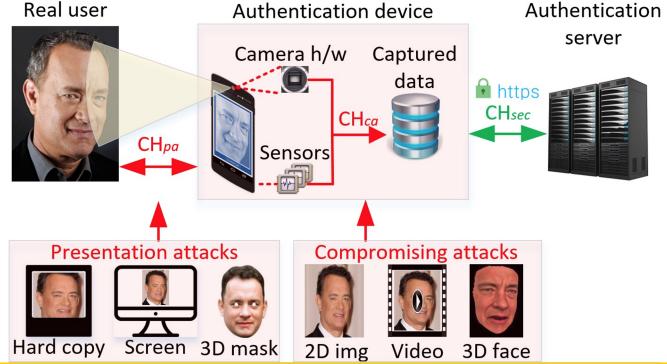




ackground Cloud Services Attacks Defense Threat Threat Sec. of Current Proposed User Sec. of Proposed Conclusion

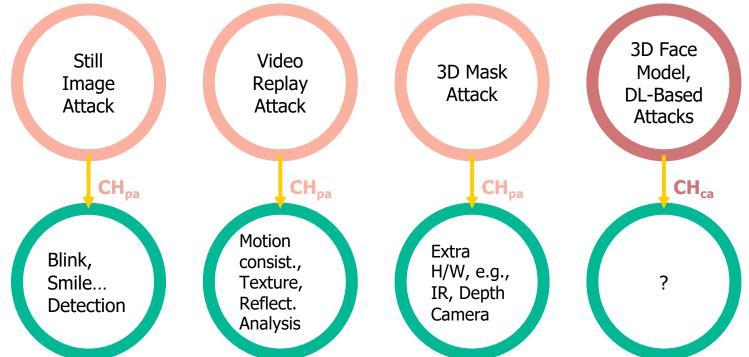


#### **Attack Channels of Biometric Authentication**





## **Adversarial Models vs Defense Systems**







# Threat Model

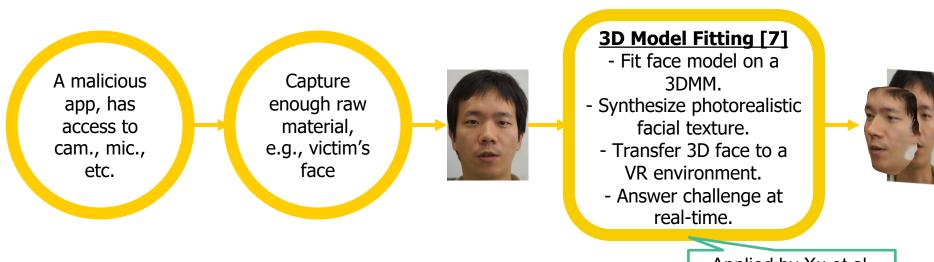
#### **Automated compromising attacks.**

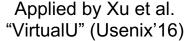
- Camera, microphone and device kernel are compromised.
- No form of attestation.
- Known client-server protocol.
- State-of-the art synthesizers and Captcha breaking tools.
- Authentication server is NOT compromised.





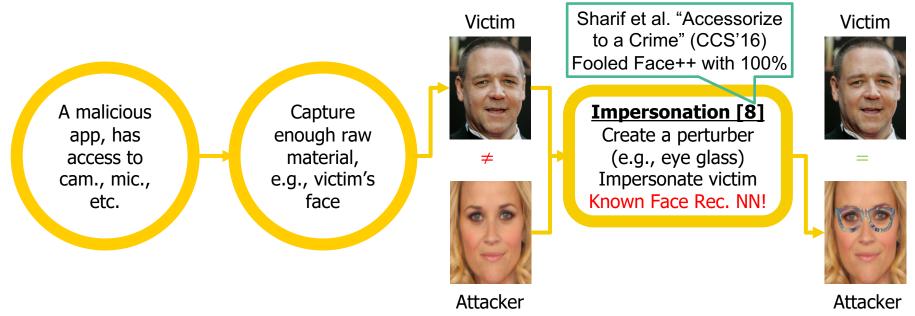
#### **Compromising Attack: Example-1**







#### **Compromising Attack: Example-2**





Face Authentication Face Spoofing Methods Face Spoofing Results Challenge Spoofing Voice Authentication Voice Spoofing Methods Voice Spoofing Results

# Security of Industry Leading Solutions (Face Authentication)

Do we need sophisticated attacks?



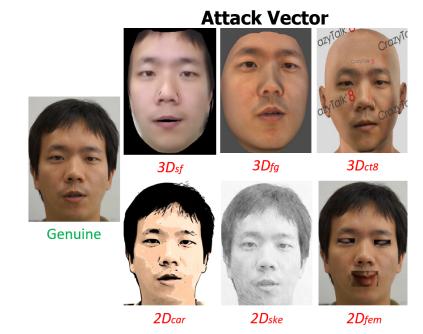
#### **Security of Cloud Systems**

#### **Face Verification Cloud Services**

- Microsoft Cognitive Services
- **Amazon Rekognition**
- Face++
- Kairos Human Analytics

#### **Database**

- First 10 subjects of CASIA Face Anti-Spoofing Database [9].
- Six attack images are generated for each subject.





Face Spoofing Results Challenge Spoofing

Voice Authentication

Voice Spoofing Methods Voice Spoofing Results



#### **Security of Cloud Systems (cont'd)**

Cognitive Service	Baseline/	Conf. (%)	Spoofed/Overall Confidence (%)						
	TP	TN	$3D_{sf}$	$3D_{fg}$	$3D_{ct8}$	2D <sub>car</sub>	2D <sub>ske</sub>	2D <sub>fem</sub>	
MS Cognitive	100/78	100/65	100/70	100/75	100/70	100/82	100/84	100/86	
Amazon	100/97	100/82	100/89	80/77	90/67	70/84	60/84	90/89	
Face++	100/87	100/83	100/86	100/71	100/72	90/77	70/80	70/75	
Kairos	100/80	80/58	100/75	100/78	100/73	100/91	100/83	100/80	

















#### **Security of Cloud Systems (cont'd)**







MS Cognitive Service

Face Authentication Face Spoofing Methods Face Spoofing Results Challenge Spoofing Voice Authentication Voice Spoofing Methods Voice Spoofing Results

3

# Security of Industry Leading Solutions (Speaker Authentication)

Do they also vulnerable to spoof?

ace Spoofing Methods

Face Spoofing Results

Challenge Spoofing

oice Authentication

Voice Spoofing Methods

Voice Spoofing Results



# Security of Cloud Systems (cont'd)

#### **Speaker Verification Cloud Services**

Microsoft Cognitive Services

#### **Database**

- $V_{dnn}^{1-7}$ : Contain 7 different DL-based synthesized version of genuine samples from two subjects, both female and male [10].
- $V_{asv}^{1}$  to  $V_{asv}^{10}$ : Contain genuine samples and their voice converted (7) and synthesized (3) versions of randomly selected 8 subjects from ASV Spoofing Challenge database [11].

#### **Methodology**

- 30 seconds of genuine samples are enrolled for each subject. Hence, a group with 10 people in MS Cognitive Service is created.
- Randomly selected different samples for genuine and spoofed voices are tested.





Face Authentication

# **Security of Cloud Systems (cont'd)**

Test Sample	Detected as Original (%)	Test Sample	Detected as Original (%)	Test Sample	Detected as Original (%)
Original	97.0	$V_{asv}^4$	60.0	$V_{asv}^9$	71.3
$V_{dnn}^{1-7}$	100	$V_{asv}^5$	77.5	$V_{asv}^{10}$	91.3
$V_{asv}^1$	81.3	$V_{asv}^6$	77.5		
$V_{asv}^2$	28.8	$V_{asv}^7$	50.0		
$V_{asv}^3$	47.5	$V_{asv}^8$	33.8		



Background Cloud Services Attacks Defense Threat Threat Sec. of Current Proposed User Sec. of Proposed Conclusion

# **Proposed System**

#### Fundamental Problem of Existing Schemes

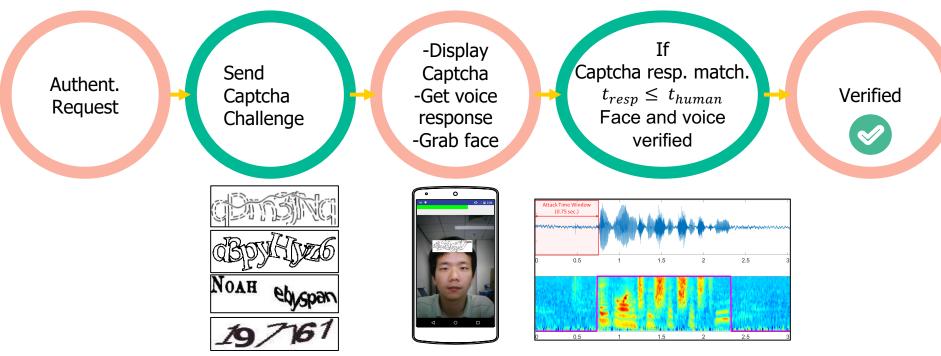
- Predictable challenges.
- Security relies on audio/face analysis, which has endless improvement in adversarial settings.

#### Real-Time Captcha (rtCaptcha)

- Randomized challenges.
- Security relies on an existing liveness detection mechanism.
- Captcha provides two types of randomness:1) Challenge semantic, 2) Captcha scheme



# **System Overview**





#### **Challenges**

- Plaintext Numeric and Phrases
- Numeric Captchas reCaptcha, Ebay, Yandex
- Animated Phrase Captchas reCaptcha
- Blink/Smile



Challenge	Accuracy (%) (1 trial)	Accuracy (%) (2 trials)	Response Time (seconds)		
Plain-text	90.3	100	0.77		
Captcha	88.8	98.4	0.93		
Smile/Blink	85.5	100	5.01		





## **Captcha Breaking/Solving Attacks**

**Hum**<sub>aud</sub>: Users in our user study.

Atc<sub>typ</sub>: Man-powered Captcha solving services [12].

Atcocr: OCR-based Captcha decoding services [13].

Atc<sub>best</sub>: State-of-the art Captcha breaking tool [14].

Captcha Sample	Captcha	Recognition Accuracy (%)				Response Time (seconds)			
	Scheme	Hum <sub>aud</sub>	Atc <sub>typ</sub>	Atcocr	Atc <sub>best</sub>	Hum <sub>aud</sub>	Atc <sub>typ</sub>	Atcocr	Atc <sub>best</sub>
149172	reCaptcha <sub>numeric</sub>	87.1	96.7	0	77.2	0.90	22.11	2.98	10.27
17 2659	<b>Ebay<sub>numeric</sub></b>	94.1	100	0	58.8	0.73	12.33	2.79	5.98
Яндекс	Yandex <sub>numeric</sub>	87.7	96.7	0	2.2	0.89	15.05	3.30	15.50
bad apple	reCaptcha <sub>phrase</sub>	88.0	91.5	0	N/A	1.02	20.88	3.03	N/A





# **Captcha Breaking/Solving Attacks**

 $Atc_{typ}$ : Man-powered Captcha solving services.

Reported Avg. Accuracy (%) and Response Time (sec.) of Man- Powered Captcha Solving Services									
Service	Acc.(%)	Time	Service	Acc.(%)	Time				
anti-captcha	99.0	7	2captcha	96.6	10				
captchaboss	99.9	8	imagetyperz	99.0	12				
deathbycaptcha	95.8	10	9kw.eu	N/A	30				





#### **Captcha Breaking/Solving Attacks**

Atc <sub>best</sub> : ML-Based Captcha Breaking Tools.										
Captcha Scheme	Gao et al. [14]		Burzstein et al. [15]		Captcha	Gao et al. [14]		Burzstein et al. [15]		
	Acc.(%)	Time(s)	Acc.(%)	Time(s)	Scheme	Acc.(%)	Time(s)	Acc.(%)	Time(s)	
reCaptcha (old)	7.8	8.06	21.74	7.16	Microsoft	16.2	12.59	N/A	N/A	
reCaptcha	77.2	10.27	19.22	4.59	Amazon	25.8	13.18	N/A	N/A	
Yahoo!	5	28.56	3.67	7.95	Taobao	23.4	4.64	N/A	N/A	
Baidu	44.2	2.81	54.38	1.9	Sina	9.4	4.83	N/A	N/A	
Wikipedia	23.8	3.74	28.29	N/A	Ebay	58.8	5.98	47.92	2.31	
QQ	56	4.95	N/A	N/A	Yandex	2.2	15.5	N/A	N/A	





# **Conclusions**

- Smile/blink etc. detection is weak against spoofing.
- of the image of the image. The image of the image. The image of the image.
- rtCaptcha: Very limited time to;
  - \* Break Captcha
  - \* Synthesize voice/face of the victim.
- Limitation: rtCaptcha needs audible response, which could NOT be usable in certain environments.





# **Future Work**

- ortCaptcha is only a part of a bigger umbrella project to make facial recognition based authentication **both usable and secure**.
- Onnect protocol to make our face-based authentication a single sign on service.
- Currently working on the **privacy** issue of biometrics-based authentication: you want to log in with your biometrics, but you don't want the server to know what you look/sound like.



- [1] Taigman, Yaniv, et al. "Deepface: Closing the gap to human-level performance in face verification." IEEE CVPR. 2014.
- [2] Schroff, Florian, et al. "Facenet: A unified embedding for face recognition and clustering." IEEE CVPR. 2015.
- [3] https://azure.microsoft.com/en-us/services/cognitive-services/
- [4] http://ws.amazon.com/rekognition
- [5] https://www.faceplusplus.com/
- [6] http://kairos.com/
- [7] Jackson, Aaron S., et al. "Large pose 3D face reconstruction from a single image via direct volumetric CNN regression." IEEE ICCV. 2017.
- [8] Sharif, Mahmood, et al. "Accessorize to a crime: Real and stealthy attacks on state-of-the-art face recognition." ACM CCS. 2016.
- [9] Zhang, Zhiwei, et al. "A face antispoofing database with diverse attacks." IEEE ICB. 2012.
- [10] Wu, Zhizheng, et al. "A study of speaker adaptation for DNN-based speech synthesis." INTERSPEECH. 2015.
- [11] Wu, Zhizheng, et al. "ASVspoof 2015: the first automatic speaker verification spoofing and countermeasures challenge." INTERSPEECH. 2015.
- [12] https://anti-captcha.com/
- [13] <a href="http://www.captchatronix.com/">http://www.captchatronix.com/</a>
- [14] Gao, Haichang, et al. "A Simple Generic Attack on Text Captchas." NDSS. 2016.
- [15] Bursztein, Elie, et al. "The End is Nigh: Generic Solving of Text-based CAPTCHAs." WOOT. 2014.





# Thanks!

Any questions ?