



# Attack on Practical Speaker Verification System Using Universal Adversarial Perturbations

Weiyi Zhang<sup>1</sup>, Shuning Zhao<sup>1</sup>, Le Liu<sup>3</sup>, Jianmin Li<sup>1</sup> Xingliang Cheng<sup>2</sup>, Thomas Fang Zheng<sup>2</sup>, Xiaolin Hu<sup>\*1</sup>

<sup>1</sup>Department of Computer Science and Technology, Tsinghua University <sup>2</sup>Center for Speech and Language Technologies, BNRist, Tsinghua University <sup>3</sup>Beijing d-Ear Technologies Co., Ltd.

Contact: wy-zhang19@mails.Tsinghua.edu.cn

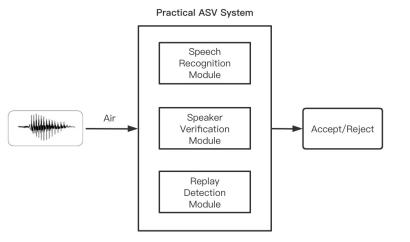




- Many authentication scenarios such as device access control, banking activities and forensics use automatic speaker verification (ASV) system for verification.
- Using **dynamic text** and **speaker verification** to ensure security.
- Performing attack on the **practical ASV system**.



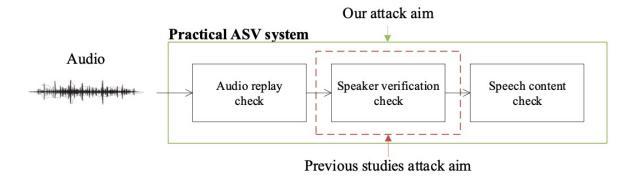
From: http://www.d-ear.com/article.jsp?s=1







- Threat model : speech recognition module and replay detection module are black box, speaker verification module is white box.
- Goal of attack :
  - *a.*  $R(x + \delta') = R(x)$ , where  $\delta' = Crop([\delta, \delta, ..., \delta], l)$
  - b.  $s(V(x + \delta'), V(y)) > \theta$
  - c.  $D(x + \delta') = D(x) = True$
  - *d.*  $\delta$  is independent of the text of *x*
  - *e.*  $\delta$  is robust to any transformation  $T(\cdot)$



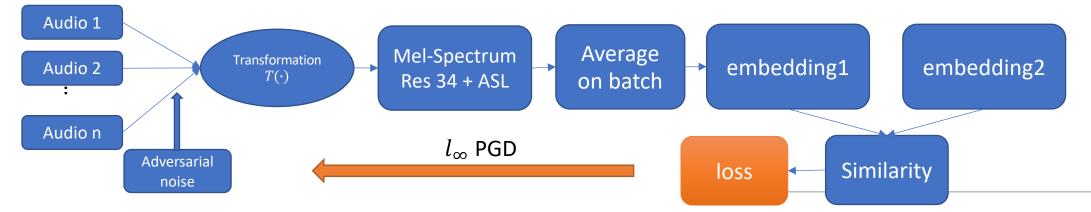




- Training set  $X = \{x_1, x_2, ..., x_N\}$  where each  $x_i$  contains different text contents. X covers great diversity about the adversary such as start offset, tune, emotion and etc.
- Loss function

$$L_1(X,\delta) = \sum_{n=1}^{N} \max(\theta - s(V(T(x_n) + T(\delta')), V(y)), -\kappa)$$
$$L_2(X,\delta) = \max(|STFT(\delta)|)$$
$$L(X,\delta) = L_1(X,\delta) + \gamma L_2(X,\delta)$$

• Two-step algorithm

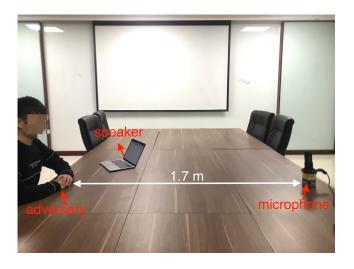


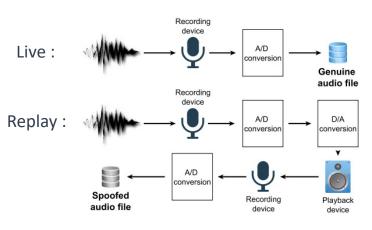




#### Experiments Introduction Attack Method Conclusion

- **Evaluation of Digital Attacks**
- **Evaluation of Physical Attacks**
- We play the adversarial perturbation as a separate source when the adversary is speaking.
- Our adversarial examples have a high success rate to pass the replay detection.





Attack type	Steps	ASR(%)	WER(%)	SNR(dB)
Clean data	N/A	0	12.95	N/A
intra-gender/baseline	236	98.43	32.33	16.90
intra-gender/ours	846	98.65	19.43	23.66
inter-gender/baseline	617	96.63	37.57	16.55
inter-gender/ours	1872	96.40	21.53	22.26

Attack ty	vpe ASR	(%)	WER(%)	CER(%)
Clean	C	)	11.42	5.78
Gaussia	in C	)	17.77	10.06
Baselin	e 80.	00	21.82	14.48
Ours	100	.00	14.97	7.53
Method	Number	Rate	e(%)	
Previous	45	37	.7	
Ours	120	67	.7	4
				4





- We proposed a two-step algorithm to generate universal adversarial perturbations for attacking the practical speaker verification system.
- We study the vulnerability of PSV system in physical world and help researchers to improve the security of such applications.





## Thanks

Presenter: Weiyi Zhang

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