



Imperceptible, Robust and Targeted Adversarial Examples for Automatic Speech Recognition

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Our Goals



- **Targeted**

Given an input audio x , a targeted transcription y , an automatic speech recognition system $f(\cdot)$, our target is to find a perturbation δ , that $f(x + \delta) = y$ and $f(x) \neq y$.

- **Imperceptible**


Humans cannot differentiate x and $x + \delta$ when listening to these examples.

- **Robust**

Played by a speaker and recorded by a microphone (over-the-air).

(We don't achieve this goal completely, but succeed at simulated rooms.)

Our Settings

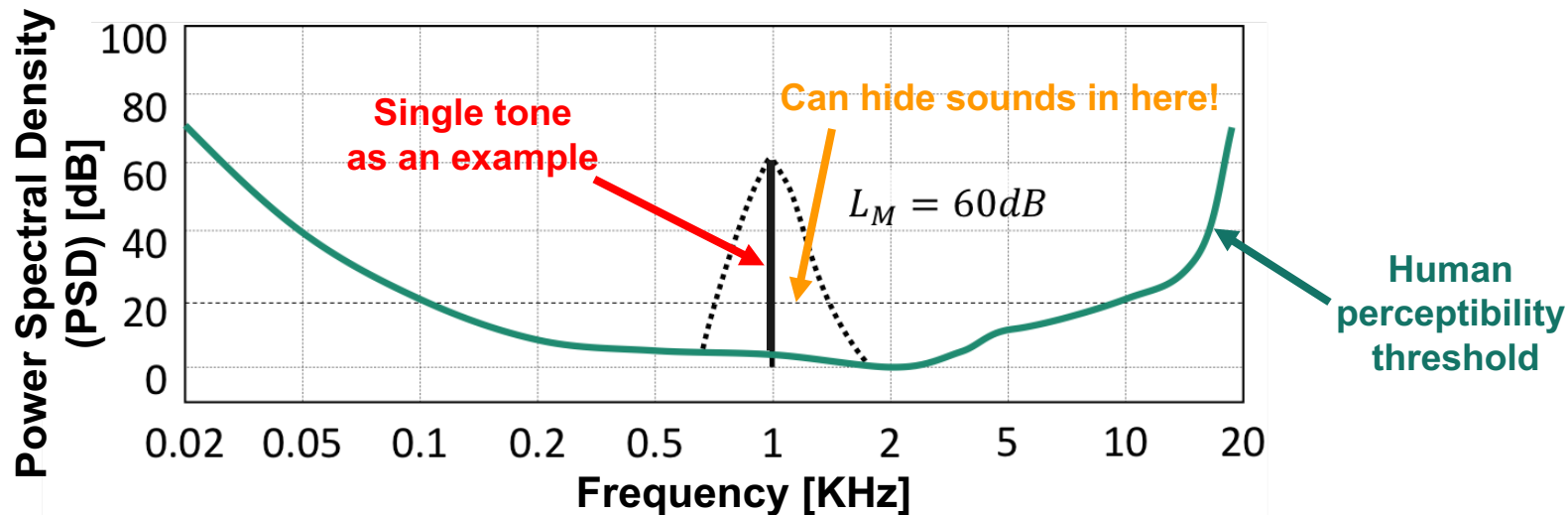
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- **Threat Model**
White-box Attack
 - **ASR Model**
Lingvo ASR system (state-of-the-art) [1]

[1] Shen, Jonathan, et al. "Lingvo: a Modular and Scalable Framework for Sequence-to-Sequence Modeling." arXiv preprint arXiv:1902.08295 (2019).


Imperceptibility

- **Frequency Masking**

A louder signal (the “masker”) can make other signals at nearby frequencies (the “maskees”) imperceptible.



Imperceptibility

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- **Loss function** $\ell(x, \delta, y) = \ell_{net}(f(x + \delta), y) + \alpha \cdot \ell_{\theta}(x, \delta)$
 - $\ell_{net}(f(x + \delta), y)$ is the cross-entropy loss function;
 - $\ell_{\theta}(x, \delta) = \max\{\bar{p}_{\delta}(k) - \theta_x(k), 0\}$ is the imperceptibility loss

Where δ is the perturbation, $\bar{p}_{\delta}(k)$ is the psd of δ and $\theta_x(k)$ is the masking threshold

Robustness



- **Room Simulator**

- Simulate room impulse r based on room configurations
- Convolve speech with reverberation $t(x) = x * r, t \sim T$

- **Robustness Loss Function**


- Minimize $\ell(x, \delta, y) = E_{t \sim T} [\ell_{net}(f(t(x + \delta)), y)]$ such that $|\delta| < \epsilon$

Imperceptible and Robust Attacks

- **Combination Loss Function (imperceptibility & robustness)**

- Minimize $\ell(x, \delta, y) = \underbrace{\mathbb{E}_{t \sim \mathcal{T}} [\ell_{net}(f(t(x + \delta)), y)]}_{\text{Robustness loss}} + \alpha \cdot \underbrace{\ell_{\theta}(x, \delta)}_{\text{Imperceptibility loss}}$

Conclusions

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- Construct *effectively imperceptible* adversarial examples using frequency masking.
 - Develop robust adversarial examples that remain effective after playing over-the-air in the simulated rooms.
 - Generate adversarial examples for non- ℓ_p -based metrics.



Thanks!

Come to our poster #65 !

Project Webpage:

<http://cseweb.ucsd.edu/~yaq007/imperceptible-robust-adv.html>

Code:

https://github.com/tensorflow/cleverhans/tree/master/examples/adversarial_asr