

Expressive Neural Voice Cloning - Supplementary

1. User Study Templates

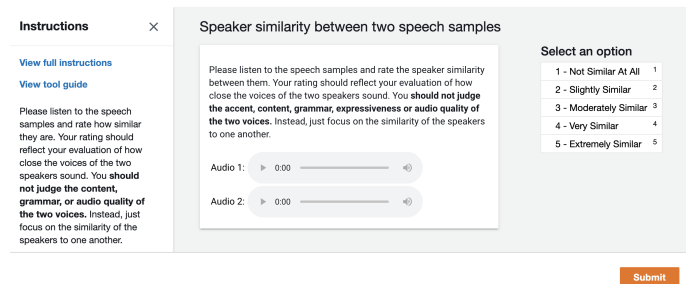


Figure 1: Interface for MOS evaluations on speaker similarity

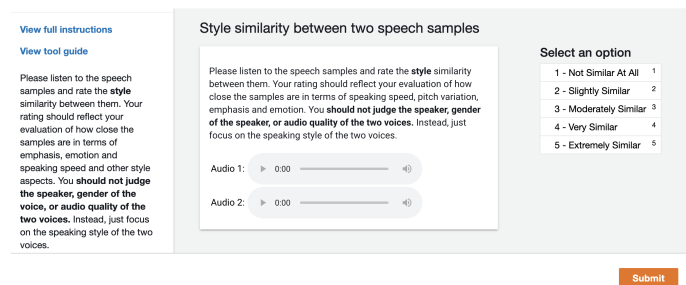


Figure 2: Interface for MOS evaluations on style similarity

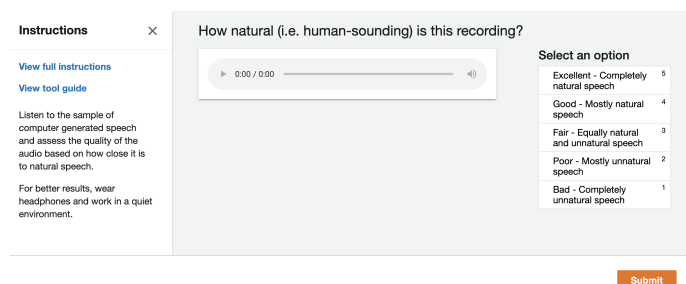


Figure 3: Interface for MOS evaluations on speech naturalness

2. Speaker Similarity MOS

We conduct a crowd-sourced user study on Amazon Mechanical Turk in which we ask users to rate the speaker similarity between two speech samples on a 5 point scale. This is a cross dataset evaluation (similar to all other evaluations conducted in our paper) where the mel synthesizer model is trained on LibriTTS dataset and evaluated on VCTK dataset. We use 10 target speaker samples for each cloning technique and synthesize 25 samples for the Text task for 20 randomly selected speakers in the VCTK dataset. Table 2 presents our speaker similarity evaluations for different approaches. While our methods outperform the different speaker baseline, they fall short in comparison to the same speaker MOS. This is likely due to the difference in accents of the synthesized speech and the ground truth audio. Our synthesizer is trained on the Libri-TTS dataset containing primarily North American English accented speakers but the VCTK dataset predominantly contains British speakers. The scores obtained for our techniques are similar to those reported in [Jia et al. \(2018\)](#) for their independently conducted study on cross dataset speaker similarity evaluation.

| Approach | Speaker similarity MOS |
|-------------------------------------|------------------------|
| Real data same speaker | 3.78 ± 0.11 |
| Real data different speaker | 1.61 ± 0.12 |
| Tacotron2 + GST - Zero Shot | 2.05 ± 0.10 |
| Proposed Model - Zero Shot | 2.21 ± 0.11 |
| Proposed Model - Adaptation Whole | 2.98 ± 0.11 |
| Proposed Model - Adaptation Decoder | 2.78 ± 0.11 |

Table 1: Speaker similarity MOS with 95% confidence interval for the Text task.

3. Model Adaptation hyper-parameter details

| Technique | Parameters | Samples | Iterations | Wall clock time |
|--------------------|------------|---------|------------|-----------------|
| Adaptation Whole | 32 million | 1 | 100 | ~ 3 mins |
| Adaptation Whole | 32 million | 5 | 100 | ~ 3 mins |
| Adaptation Whole | 32 million | 10 | 100 | ~ 3 mins |
| Adaptation Whole | 32 million | 20 | 200 | ~ 6 mins |
| Adaptation Decoder | 21 million | 1 | 100 | ~ 3 mins |
| Adaptation Decoder | 21 million | 5 | 200 | ~ 5 mins |
| Adaptation Decoder | 21 million | 10 | 200 | ~ 5 mins |
| Adaptation Decoder | 21 million | 20 | 200 | ~ 6 mins |

Table 2: Hyper-parameter details and wall clock time for model adaptation techniques. The column *Samples* refers to the number of *target speaker samples* used for model fine-tuning. We use Adam optimizer with learning rate of $1e-4$ across all model adaptation experiments.

References

Ye Jia, Yu Zhang, Ron Weiss, Quan Wang, Jonathan Shen, Fei Ren, zhifeng Chen, Patrick Nguyen, Ruoming Pang, Ignacio Lopez Moreno, and Yonghui Wu. Transfer learning from speaker verification to multispeaker text-to-speech synthesis. In *NeurIPS*. 2018.