









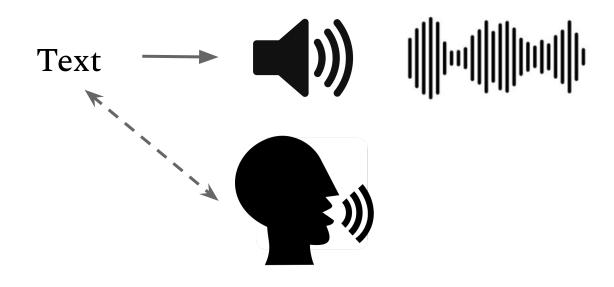




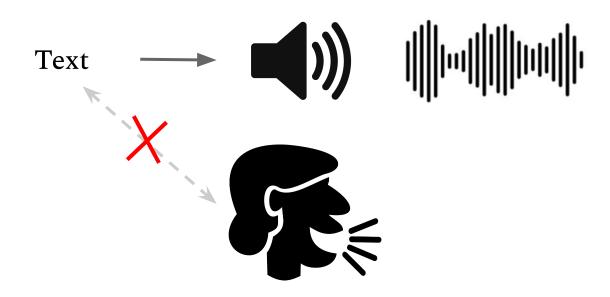
YourTTS: Towards Zero-Shot Multi-Speaker TTS and Zero-Shot Voice Conversion for everyone

E. Casanova^{1,2}, J. Weber², C. Shulby³, A. Candido Junior⁴, E. Gölge², Moacir A. Ponti^{1,5}

Zero-shot Text-to-Speech

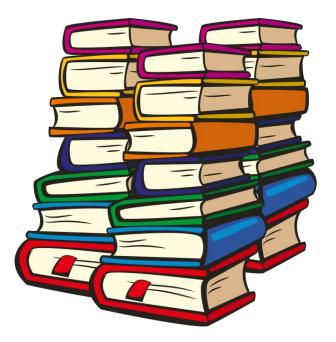


Zero-shot Text-to-Speech



Resources and Performance Gap

English



Mandarin & a few others



Remaining languages



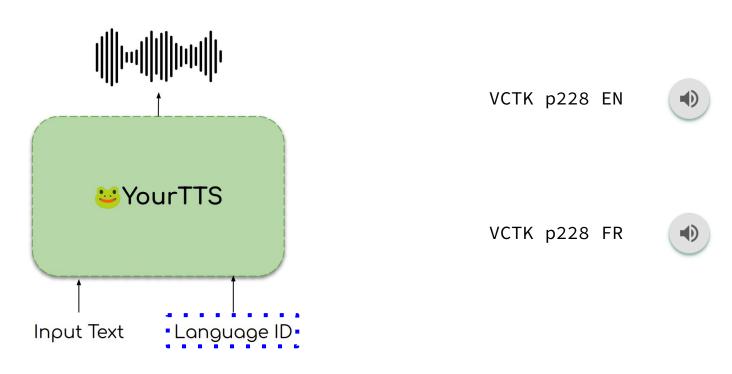
YourTTS, a multilingual model taking advantage of the high speaker count of English

Contribution

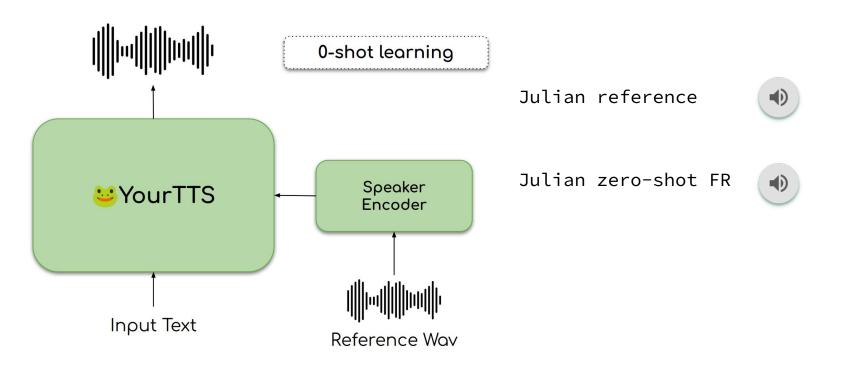
The first to explore a multilingual approach in ZS-TTS achieving state-of-the-art results +

- multilingual TTS
- zero shot TTS
- cross-lingual zero shot
- voice conversion

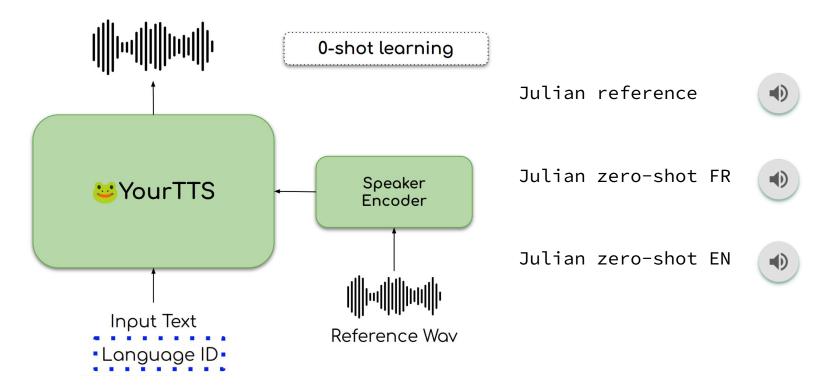
Multi-lingual TTS System



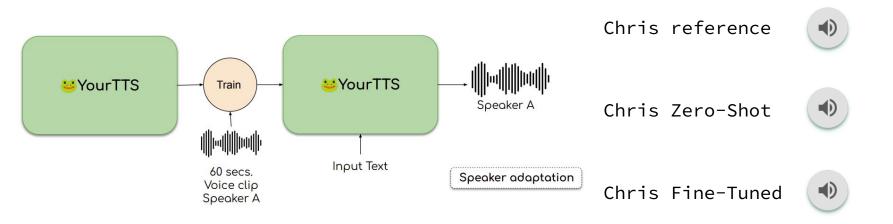
Zero Shot Learning



Cross-lingual Zero-Shot

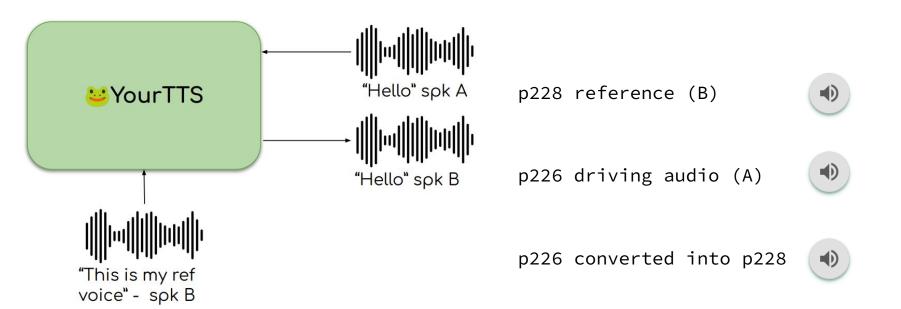


Finetuning a TTS model with a short sample





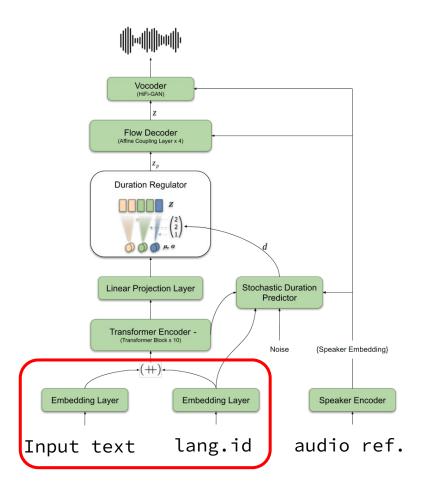
Voice conversion



YourTTS inference

Language id concatenated to every char embedding:

- allows for code switching



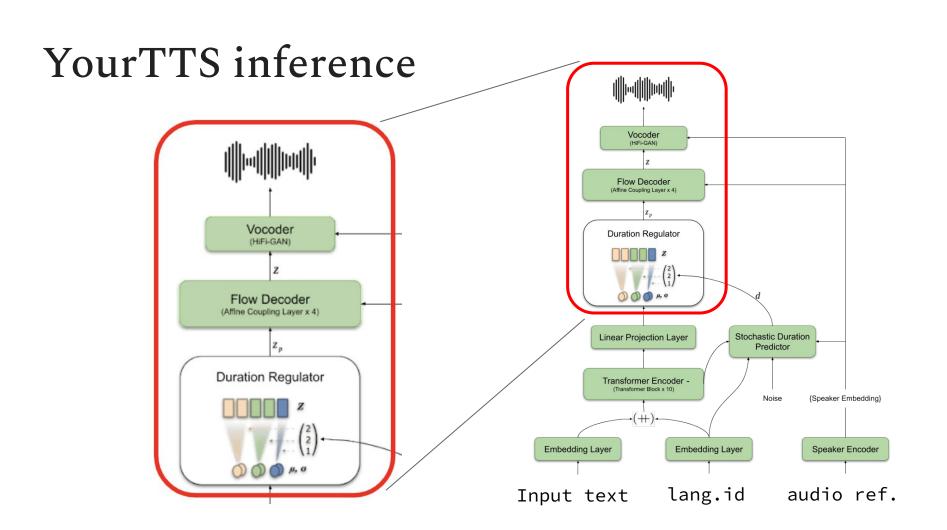
YourTTS inference

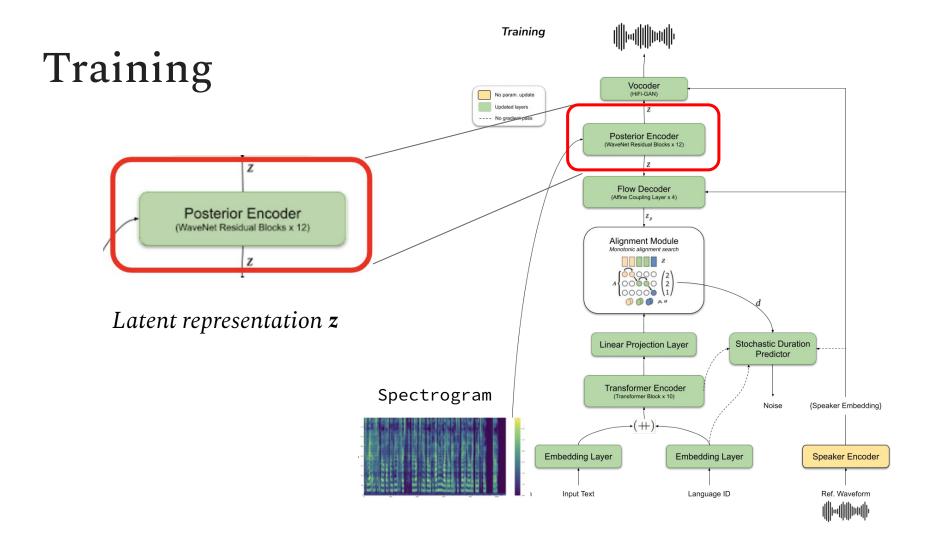
Vocoder Transformer encodes text+language input (HiFi-GAN) into sequence of pseudo-phonemes Flow Decoder (Affine Coupling Layer x 4) **Duration Regulator** Stochastic Duration Linear Projection Layer Predictor Linear Projection Layer Stochastic Duration Predictor Transformer Encoder -Transformer Encoder -{Spearer Embedding} (Transformer Block x 10) Noise Noise {Spealer **Embedding Layer Embedding Layer** Speaker Encoder

Input text

lang.id

audio ref.



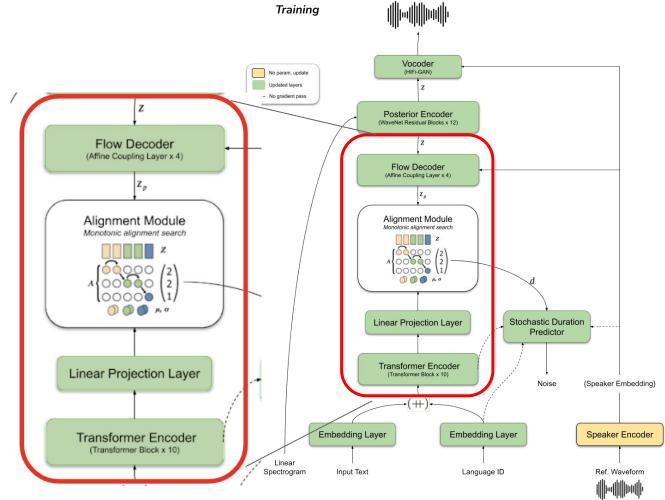


Training

 z_p

VS

pseudo-phonemes



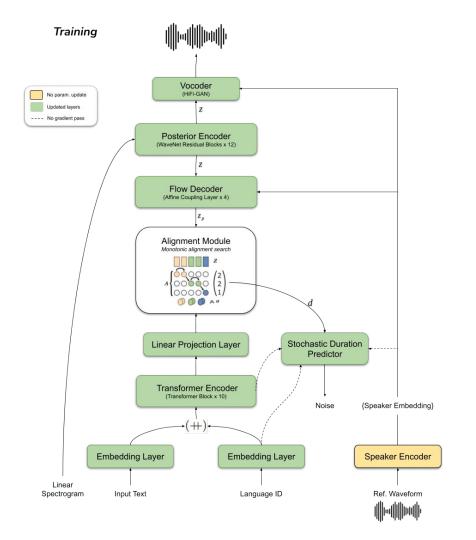
Loss Functions

Speaker Embedding pretraining:

Softmax + Prototype Angular with VoxCeleb dataset

TTS Model:

 Speaker Consistency Loss
 Cosine similarity between ground truth and generated audio embeddings



| • | Portuguese | O modelo me ouviu falando apenas em Português. Mas, com YourTTS, sei falar também em Inglês e Francês. |
|----------|------------|--|
| 1 | English | The model heard me speaking only Portuguese. But with YourTTS I can also speak English and French |
| • | French | Le modèle m'a entendu parler uniquement en |

portugais. Mais avec YourTTS, je peux aussi

parler en anglais et en français.

Results

Experiments

- 1. VCTK dataset (98 speakers);
- 2. VCTK and TTS-Portuguese Corpus (1 speaker);
- 3. VCTK, TTS-Portuguese and M-AILABS french dataset (5 speakers);
- 4. VCTK + TTS-Portuguese + M-AILABS + LibriTTS (1151 speakers).

Experiments Setup

• All experiments were implemented using **coqui** TTS: github.com/coqui-ai/TTS an open source TTS framework.

 Audio samples and checkpoints of all experiments are available on:

github.com/Edresson/YourTTS

ZS-TTS results

| | VCTK | | | LIBRITTS | | | MLS-PT | | |
|----------------|---------|-----------------|-------------------|----------|---------------|-----------------|--------|-----------------|----------------|
| EXP. | SECS | MOS | SIM-MOS | SECS | MOS | SIM-MOS | SECS | MOS | SIM-MOS |
| GROUND TRUTH | 0.824 | 4.26±0.04 | 4.19±0.06 | 0.931 | 4.22±0.05 | 4.22 ± 0.06 | 0.9018 | 4.61±0.05 | 4.41 ± 0.05 |
| ATTENTRON ZS | (0.731) | (3.86 ± 0.05) | (3.30 ± 0.06) | _ | _ | _ | _ | _ | _ |
| SC-GLOWTTS | (0.804) | (3.78 ± 0.07) | (3.99 ± 0.07) | _ | _ | _ | _ | _ | _ |
| Exp. 1 | 0.864 | 4.21±0.04 | 4.16±0.05 | 0.754 | 4.25±0.05 | 3.98 ± 0.07 | _ | _ | _ |
| Exp. $1 + SCL$ | 0.861 | 4.20 ± 0.05 | 4.13±0.06 | 0.765 | 4.21 ± 0.04 | 4.05 ± 0.07 | _ | _ | _ |
| Exp. 2 | 0.857 | 4.24±0.04 | 4.15±0.06 | 0.762 | 4.22±0.05 | 4.01 ± 0.07 | 0.740 | 3.96 ± 0.08 | 3.02 ± 0.1 |
| EXP. 2 + SCL | 0.864 | 4.19±0.05 | 4.17±0.06 | 0.773 | 4.23±0.05 | 4.01 ± 0.07 | 0.745 | 4.09 ± 0.07 | $2.98{\pm}0.1$ |
| EXP. 3 | 0.851 | 4.21±0.04 | 4.10±0.06 | 0.761 | 4.21±0.04 | 4.01±0.05 | 0.761 | 4.01±0.08 | $3.19{\pm}0.1$ |
| EXP. 3 + SCL | 0.855 | 4.22±0.05 | 4.06 ± 0.06 | 0.778 | 4.17 ± 0.05 | 3.98 ± 0.07 | 0.766 | 4.11±0.07 | 3.17 ± 0.1 |
| Exp. 4 + SCL | 0.843 | $4.23{\pm}0.05$ | 4.10 ± 0.06 | 0.856 | 4.18 ± 0.05 | $4.07{\pm}0.07$ | 0.798 | $3.97{\pm}0.08$ | 3.07 ± 0.1 |

- SOTA results in similarity and speech quality for unseen speakers
- Produce female voice in PT without seen female voice during training

Zero-shot voice conversion results

| REF/TAR | M-M | | M-F | | F-F | | F-M | | ALL | |
|---------|-----------------|-----------------|-----------------|-----------------|-----------------|-----------------|-----------------|-----------------|-----------------|-----------------|
| | MOS | SIM-MOS |
| EN-EN | 4.22±0.10 | 4.15±0.12 | 4.14±0.09 | 4.11±0.12 | 4.16±0.12 | 3.96 ± 0.15 | 4.26±0.09 | 4.05±0.11 | 4.20 ± 0.05 | 4.07±0.06 |
| PT-PT | 3.84 ± 0.18 | 3.80 ± 0.15 | 3.46 ± 0.10 | 3.12 ± 0.17 | 3.66 ± 0.2 | 3.35 ± 0.19 | 3.67 ± 0.16 | 3.54 ± 0.16 | 3.64 ± 0.09 | 3.43 ± 0.09 |
| EN-PT | 4.17±0.09 | 3.68 ± 0.10 | $4.24{\pm}0.08$ | 3.54 ± 0.11 | 4.14±0.09 | 3.58 ± 0.12 | 4.12±0.10 | 3.58 ± 0.11 | 4.17±0.04 | 3.59 ± 0.05 |
| PT-EN | 3.62 ± 0.16 | 3.8 ± 0.10 | 2.95 ± 0.2 | 3.67 ± 0.11 | 3.51 ± 0.18 | 3.63 ± 0.11 | 3.47 ± 0.18 | 3.57 ± 0.11 | 3.40 ± 0.09 | 3.67 ± 0.05 |

- Intra-lingual results comparable to SOTA in VCTK
- Cross-lingual results EN2PT similar as PT2PT

Speaker Adaptation results

| | SEX | DUR. (SAM.) | Mode | SECS | MOS | SIM-MOS |
|----|-----|-------------|------|-------|---------------|-----------------|
| EN | | | GT | 0.875 | 4.17 ± 0.09 | 4.08±0.13 |
| | M | 61s (15) | ZS | 0.851 | 4.11 ± 0.07 | 4.04 ± 0.09 |
| | | | FT | 0.880 | 4.17 ± 0.07 | 4.08±0.09 |
| | | | GT | 0.894 | 4.25 ± 0.11 | 4.17 ± 0.13 |
| | F | 44s (11) | ZS | 0.814 | 4.12 ± 0.08 | 4.11 ± 0.08 |
| | | | FT | 0.896 | 4.10 ± 0.08 | $4.17{\pm}0.08$ |
| PT | М | 31s (7) | GT | 0.880 | 4.76 ± 0.12 | 4.31 ± 0.14 |
| | | | ZS | 0.817 | 4.03 ± 0.11 | 3.35 ± 0.12 |
| | | | FT | 0.915 | 3.74 ± 0.12 | 4.19 ± 0.07 |
| | | 20s (5) | GT | 0.873 | 4.62 ± 0.19 | 4.65 ± 0.14 |
| | F | | ZS | 0.743 | 3.59 ± 0.13 | 2.77 ± 0.15 |
| | | | FT | 0.930 | 3.48 ± 0.13 | 4.43 ± 0.06 |

- good results with 1 min of speech, presenting naturalness over zero-shot
- 44 seconds or less reduces the quality when compared to the zero-shot.

Limitations and room for improvement

- Accent supressing
- Monotonic tones for long sentences
- Unnatural speeds for some speaker/language combinations
- Reduce audio artefacts

Possibilities

- Capture different accents and particularities
- Learn from few speakers of (very) low resource languages
 - o indigenous languages
 - dialects
- Learn speech of persons that may lose their voices

Thanks















