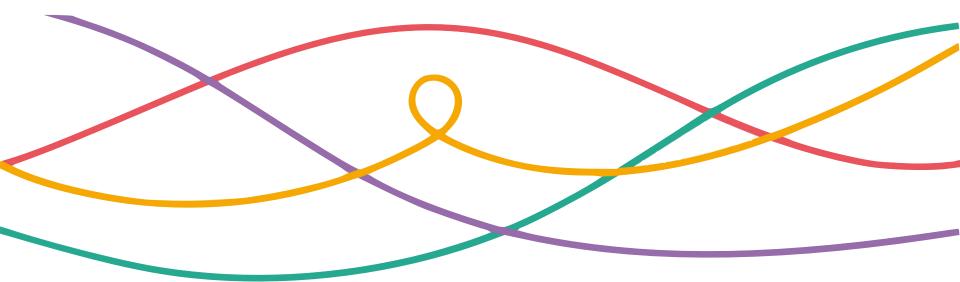
Voctro Labs – TROMPA Choir Singing Synthesis and Rehearsal Tools



Context



www.trompamusic.eu

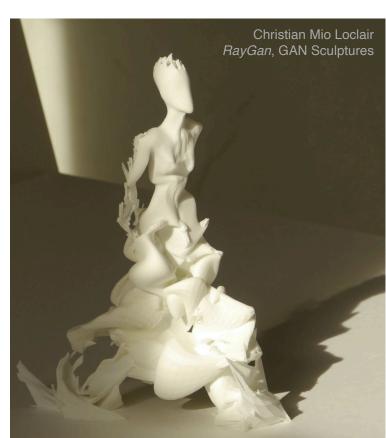
Facilitate access to public-domain digital resources with state-of-the-art technologies

Voctro's contribution

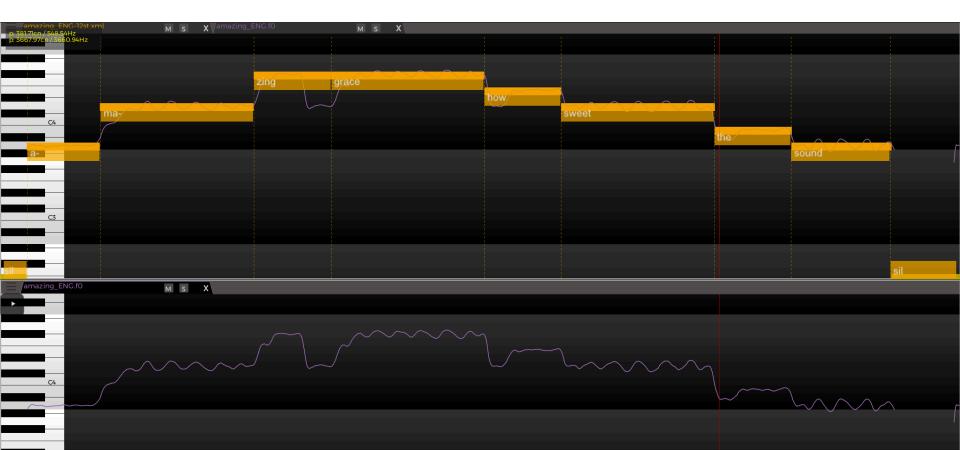
- Build a rehearsal tool for choir singers
- Synthesize a large multi-lingual public-domain repertoire

Singing synthesis

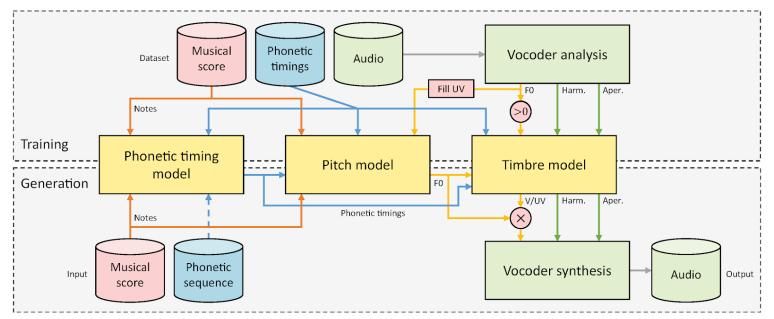
Pitch / intonation Timing / rhythm Timbre, identity, quality Phonetics / intelligibility Dynamics / loudness Style, vibrato, legato, ...



Modelling singing voice

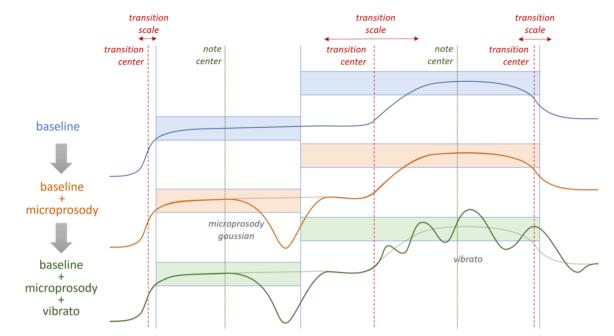


Singing synthesis



Baseline system: Blaauw, M. Bonada, J. (2017). **A Neural Parametric Singing Synthesizer**. In Interspeech 2017, Stockholm

Singing synthesis



Novel pitch model:

Bonada, J., Blaauw, M. (2020). HYBRID NEURAL-PARAMETRIC F0 MODEL FOR SINGING SYNTHESIS. In ICASSP 2020, Barcelona

Choir extensions

Specific datasets

16 professional singers (SATB)

Cor Francesc Valls, Barcelona

4 languages

Multiple voices

Voice cloning (DNN-speaker adaptation)

M. Blaauw, J. Bonada and R. Daido, "Data Efficient Voice Cloning for Neural Singing Synthesis," *ICASSP 2019*

https://mtg.github.io/singing-synthesis-demos/voice-cloning/

Large-scale repertoire

Choral Public Domain Library Since 1998 34,531 scores of free choral music



Prepared subset 4,107 scores in MusicXML format In the 4 supported languages

Demos

Sicut cervus







Rehearsal tool

Live demo https://trompa.netlify.app/

Rehearsal tool

Collaboration with choirs

9 choirs, mainly from Catalonia

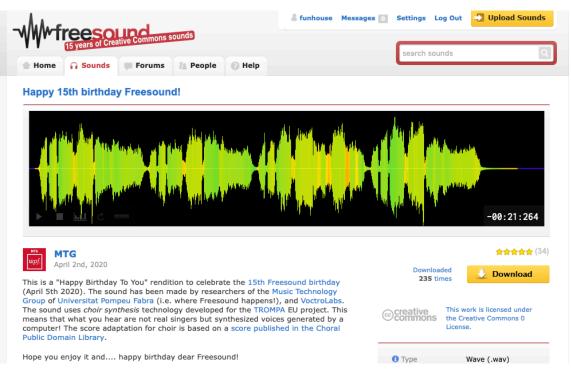
https://trompamusic.eu/node/105

Collaboration with Cantoría

Workshops for singers: access and practice Cantoría's repertoire Perform a concert that will bring together all the participants in spring 2021

https://trompamusic.eu/node/123

Semi-supervised singing synthesis



https://freesound.org/people/MTG/sounds/511618/

Semi-supervised singing synthesis

- New system under development submitted to ICASSP 2021
- · Can learn new voices from audio data only, without annotations
- Encoder-decoder model:
 - two encoders linguistic and acoustic
 - one (acoustic) decoder
- Training and inference:
 - 1. The entire system is trained in a supervised manner, using a labelled dataset.
 - 2. The system is adapted to a new target voice in an unsupervised manner, using the pretrained acoustic encoder
 - 3. At inference, the pretrained linguistic encoder is used together with the adapted decoder

https://mtg.github.io/singing-synthesis-demos/semisupervised/

Bonada, J., & Blaauw, M. (2020). Semi-supervised Learning for Singing Synthesis Timbre. arXiv preprint arXiv:2011.02809.

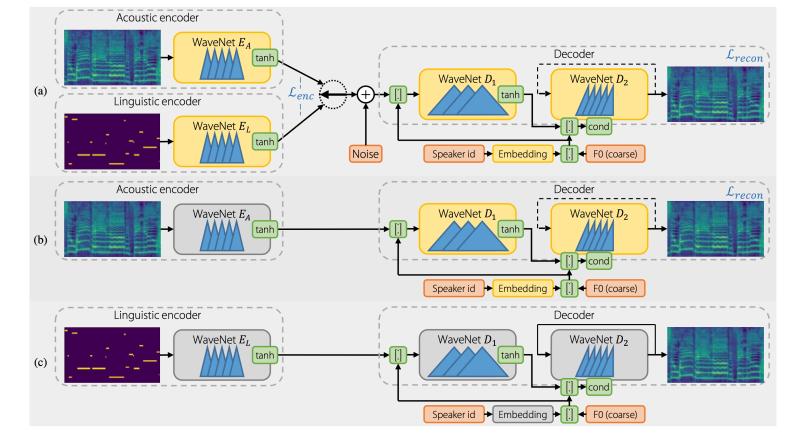


Fig. 1. A diagram of the model architecture in three different phases: (a) Training the encoder-decoder from annotated audio (supervised). (b) Training the decoder from audio (unsupervised). (c) Inference from linguistic features. Gray colored modules indicate their weights are kept fixed. The shape of the triangles in the WaveNet blocks represents the size of the receptive field and whether it is causal. A dashed autoregressive connection in the D_2 WaveNet block indicates teacher forced training with additive noise to avoid overfitting, while a solid connection indicates true autoregressive inference.

Thank you!

www.voiceful.io