

# MUSIKA! FAST INFINITE WAVEFORM MUSIC GENERATION

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A comprehensive collection of generated samples can be found on [marcoppasini.github.io/musika](https://marcoppasini.github.io/musika)

**State-of-the-art** waveform music generation systems are **slow**. We introduce **Musika**, which allows for much **faster than real-time generation of music of arbitrary length** on a consumer CPU.

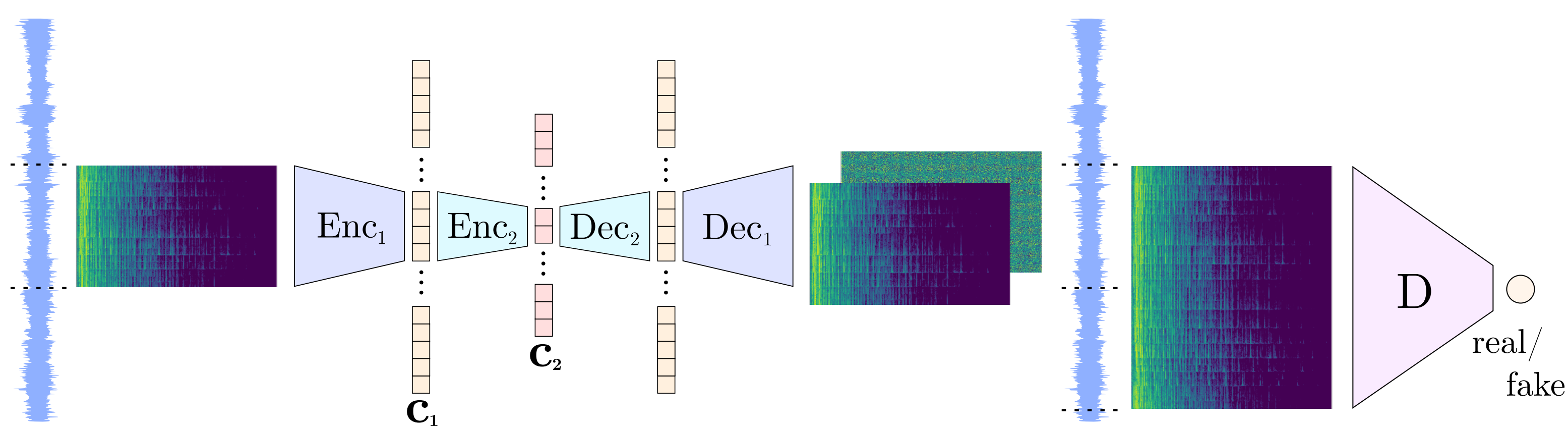


Fig. 1: **The proposed 2-level audio autoencoder**. A log-magnitude spectrogram is used as input, while the **decoder outputs magnitude and phase spectrograms** which are then inverted with iSTFT. A discriminator evaluates the magnitude spectrogram of two adjacent excerpts.

We train a **GAN** on compressed **invertible representations** from an efficient **Autoencoder**. **Coordinate** and **style vectors** allow parallel generation of arbitrarily **long** and **coherent** samples.

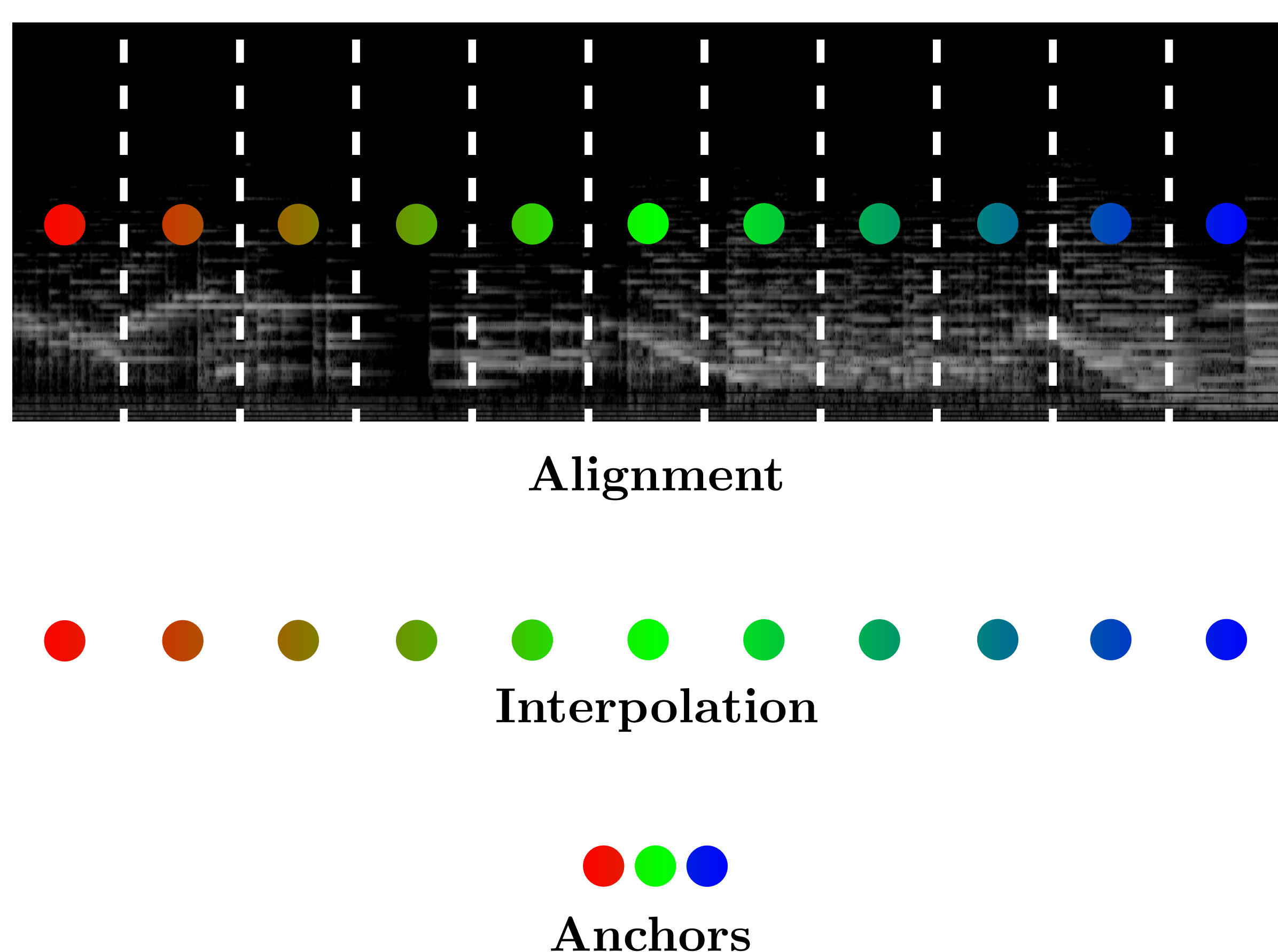


Fig. 2: **The alignment process**. We first **sample random anchor vectors** and linearly **interpolate between them**. The adversarial learning process **aligns the latent coordinate space with the generated latent vectors space**.

We first train the **Autoencoder** to produce **general representations**. We then train **GANs** to generate **Piano** and **Techno** music, with **note density** and **tempo** as conditioning.

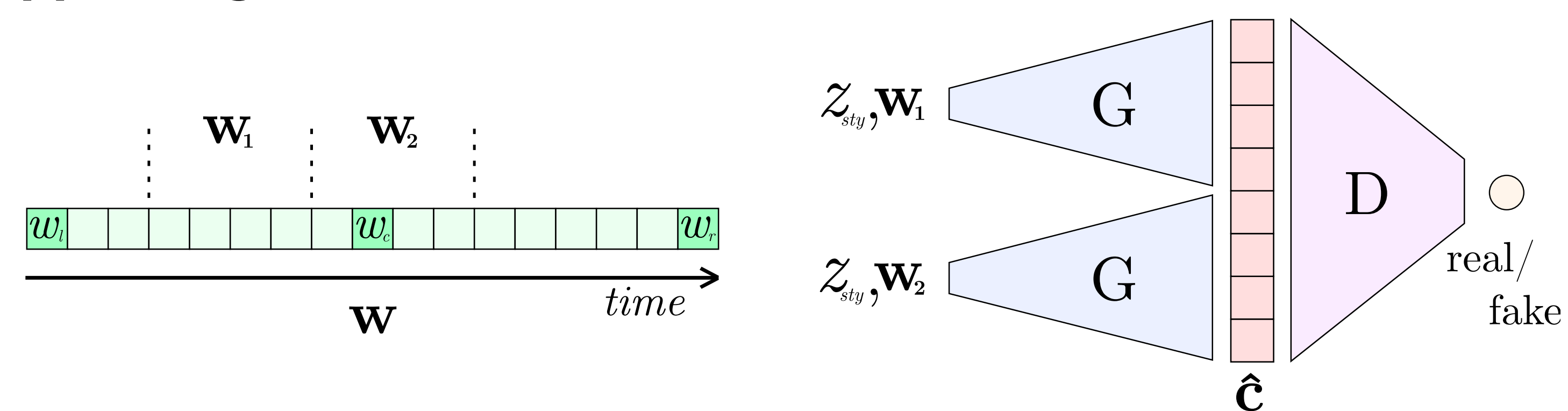


Fig. 3: **The proposed latent GAN training process**. Two adjacent latent coordinate sequences are randomly cropped from the linear interpolation between **three anchor vectors** and used as input to the generator, together with a **shared style vector**.

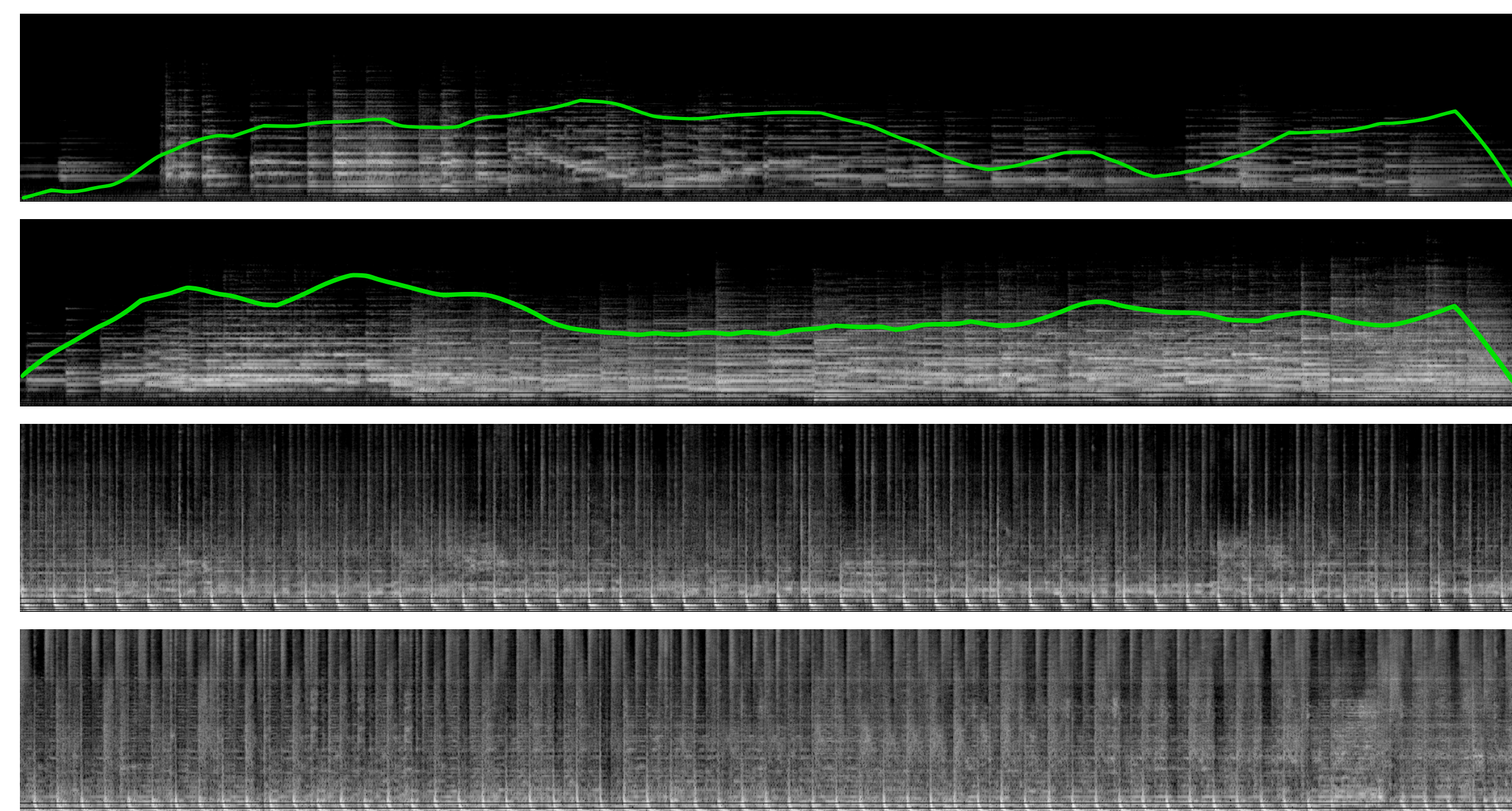


Fig. 4: Log-melspectrograms of **generated piano and techno samples from the conditional models**. For the piano samples (top row), we indicate the corresponding **note density conditioning** with a green line. The tempo used as conditioning for the techno samples (bottom row) is **120 bpm** and **160 bpm**, respectively.

| Model (Faster than real-time) | GPU         | CPU |
|-------------------------------|-------------|-----|
| Musika Uncond. Piano          | 972x        | 40x |
| Musika Cond. Piano            | 921x        | 40x |
| UNAGAN[1] Piano               | 28x         | 11x |
| Musika Uncond. Techno         | <b>994x</b> | 39x |
| Musika Cond. Techno           | 917x        | 39x |

Tab. 1: Comparison of **generation speed** between the different models.

| Model                    | FAD          |
|--------------------------|--------------|
| Musika Uncond. Piano     | <b>1.641</b> |
| Musika Cond. Piano Rand. | 2.150        |
| UNAGAN[1] Piano          | 11.183       |

Tab. 2: **FAD evaluation** for generated piano music. We evaluate a conditional Musika model using random values of note density as conditioning.

We see our system as solving an important technical challenge -- **real-time music generation of sufficient quality, conditioned on user input** -- and hope it can advance the field of **human-AI co-creation**.

[1] Jen-Yu Liu et al. "Unconditional Audio Generation with Generative Adversarial Networks and Cycle Regularization". In: INTERSPEECH. 2020.