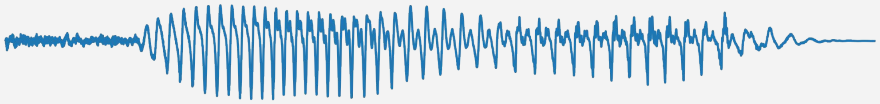
Vector Quantized Neural Networks for Acoustic Unit Discovery



Benjamin van Niekerk, Leanne Nortje, Herman Kamper

The Generative Factors of Speech

HH / Y / UW / M / ER

HUMOUR

Content:

- Discrete phonetic units.
- ≅44 phonemes in English.

Prosody:

- Rhythm
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- Stresses

Timbre:

- Quality of a particular voice.
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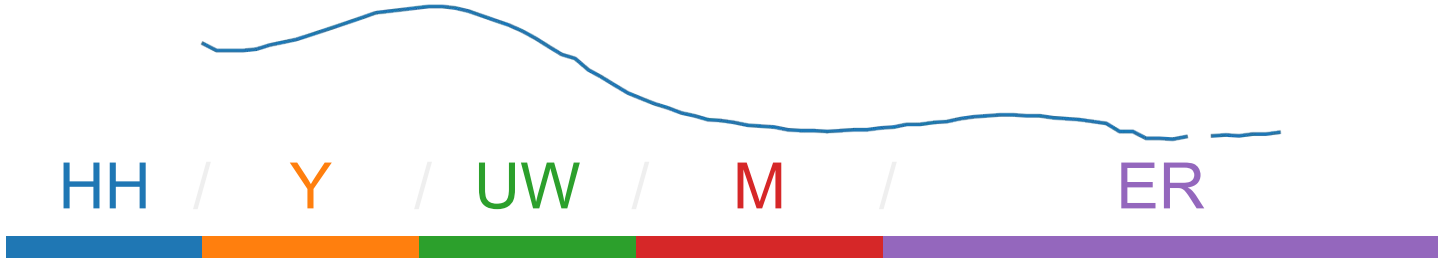

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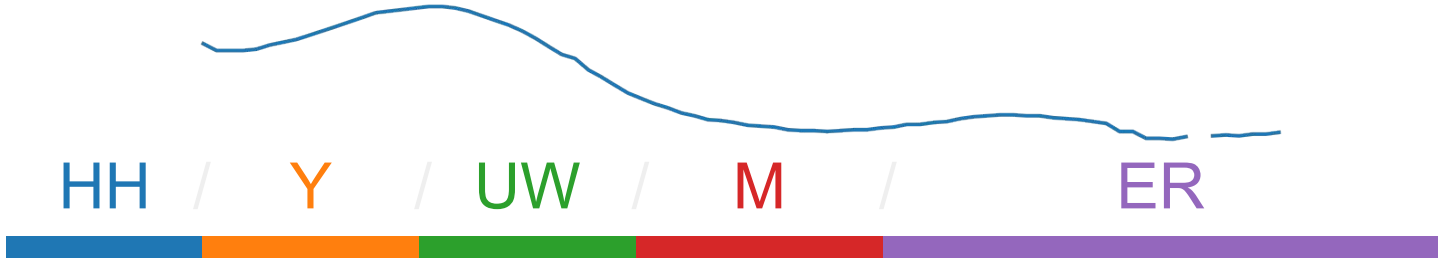

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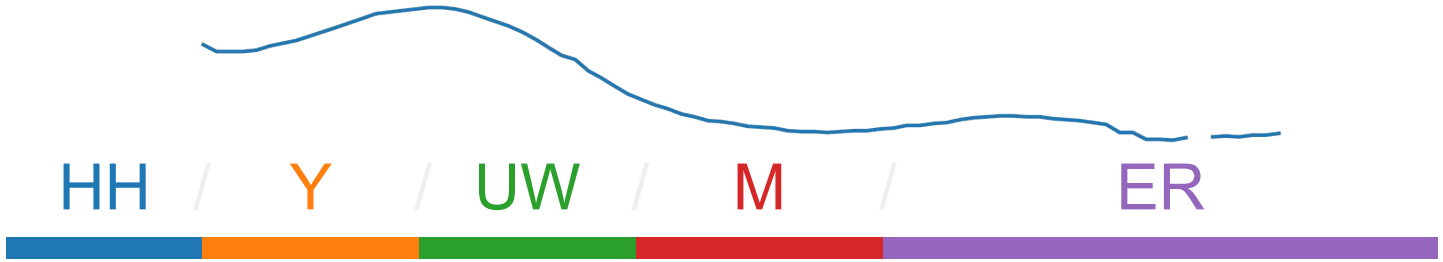

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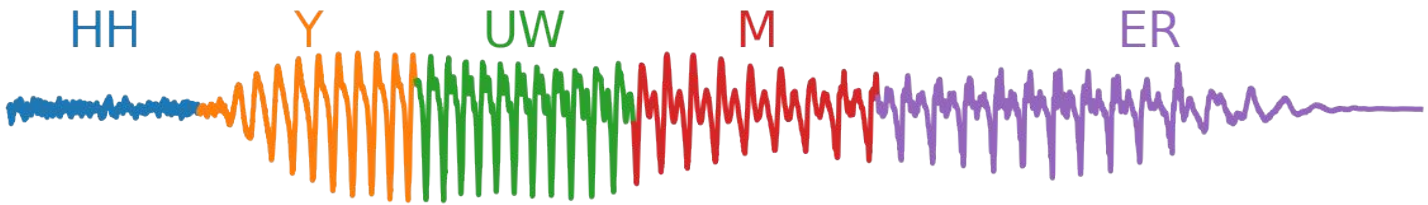

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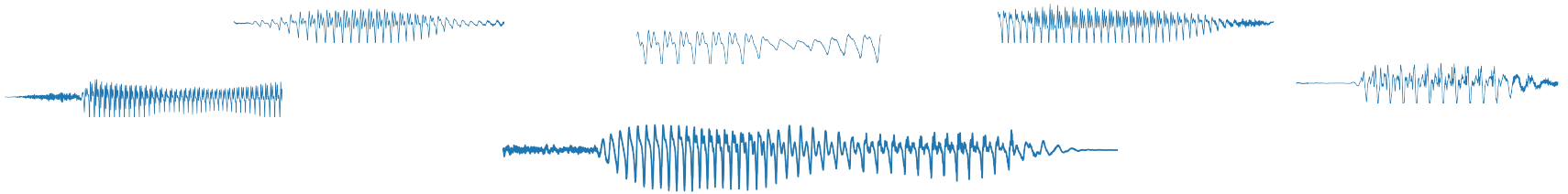

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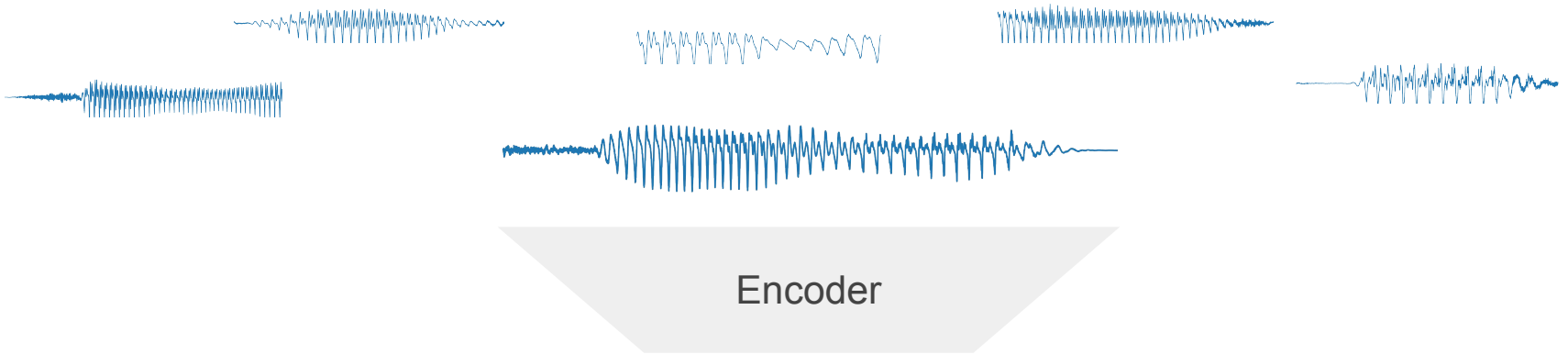

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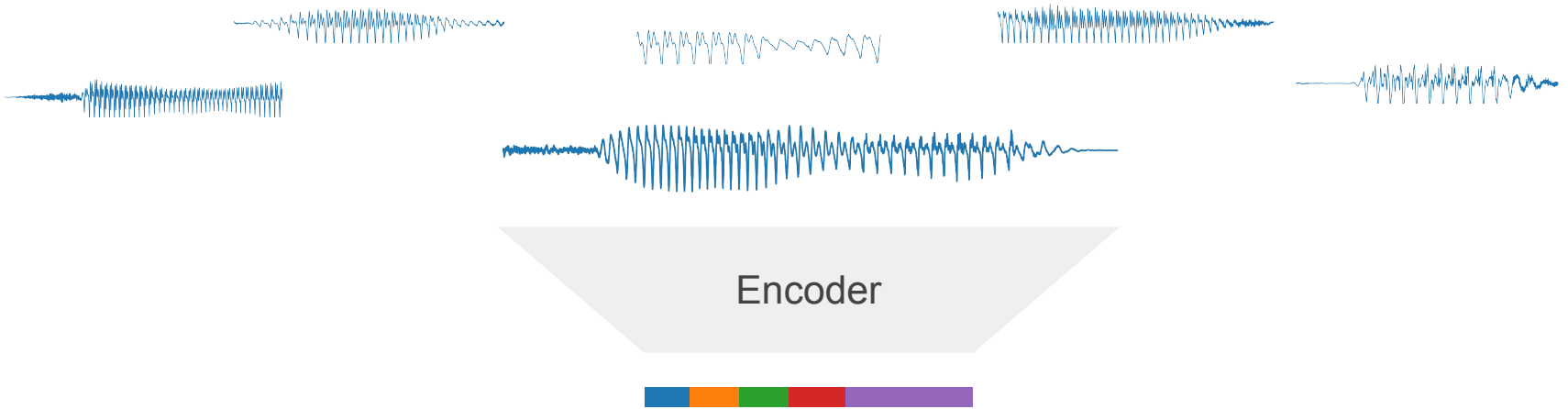

Applications

Bootstrap training of **low-resource** speech systems:

Automatic speech recognition

Text-to-speech

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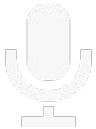

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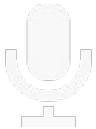

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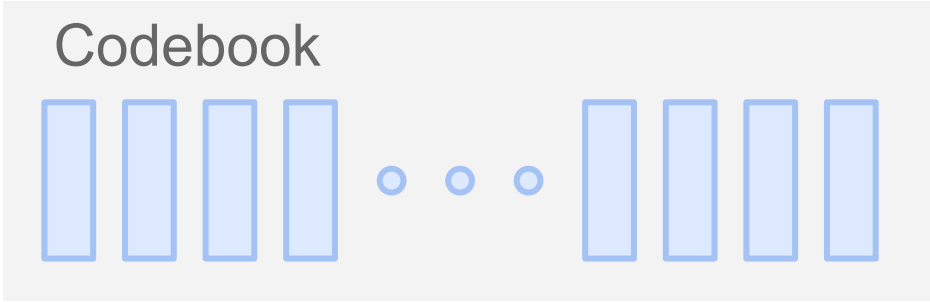

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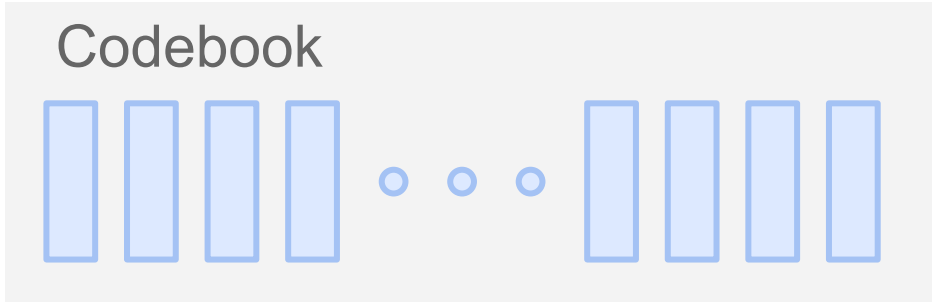

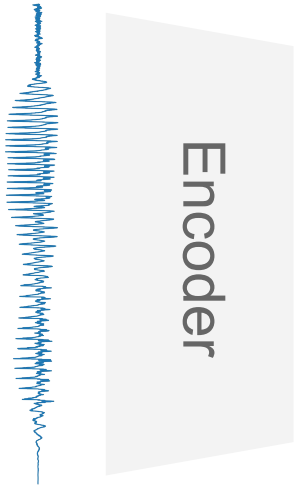

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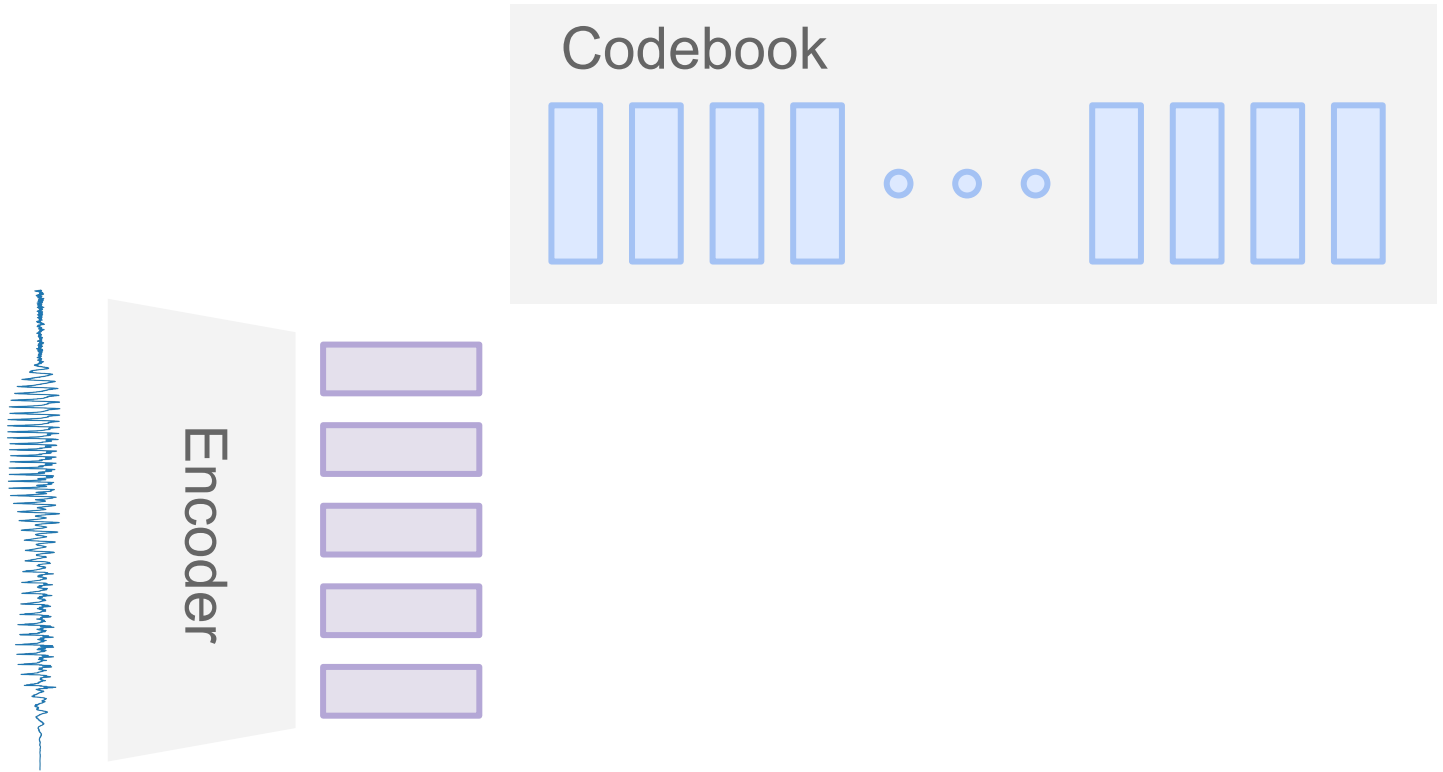

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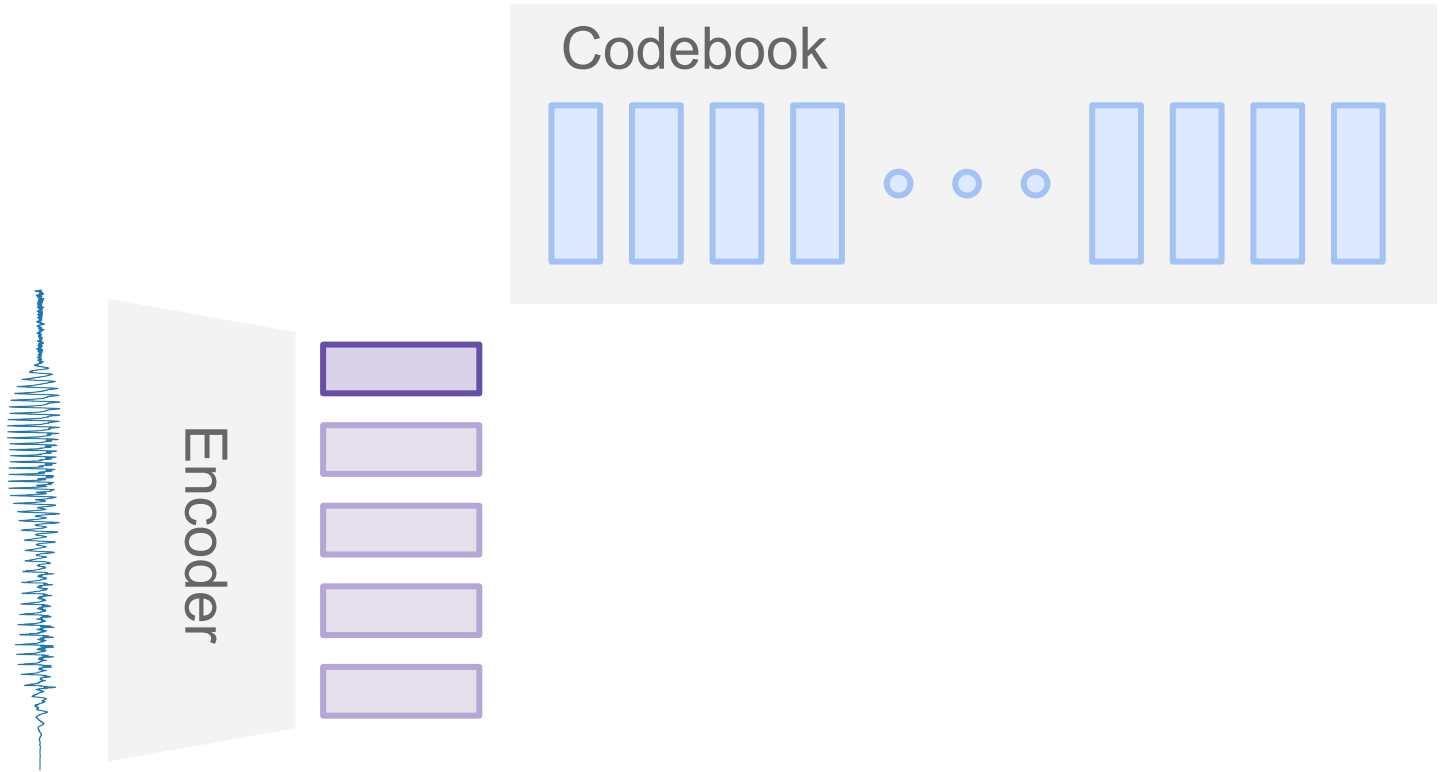

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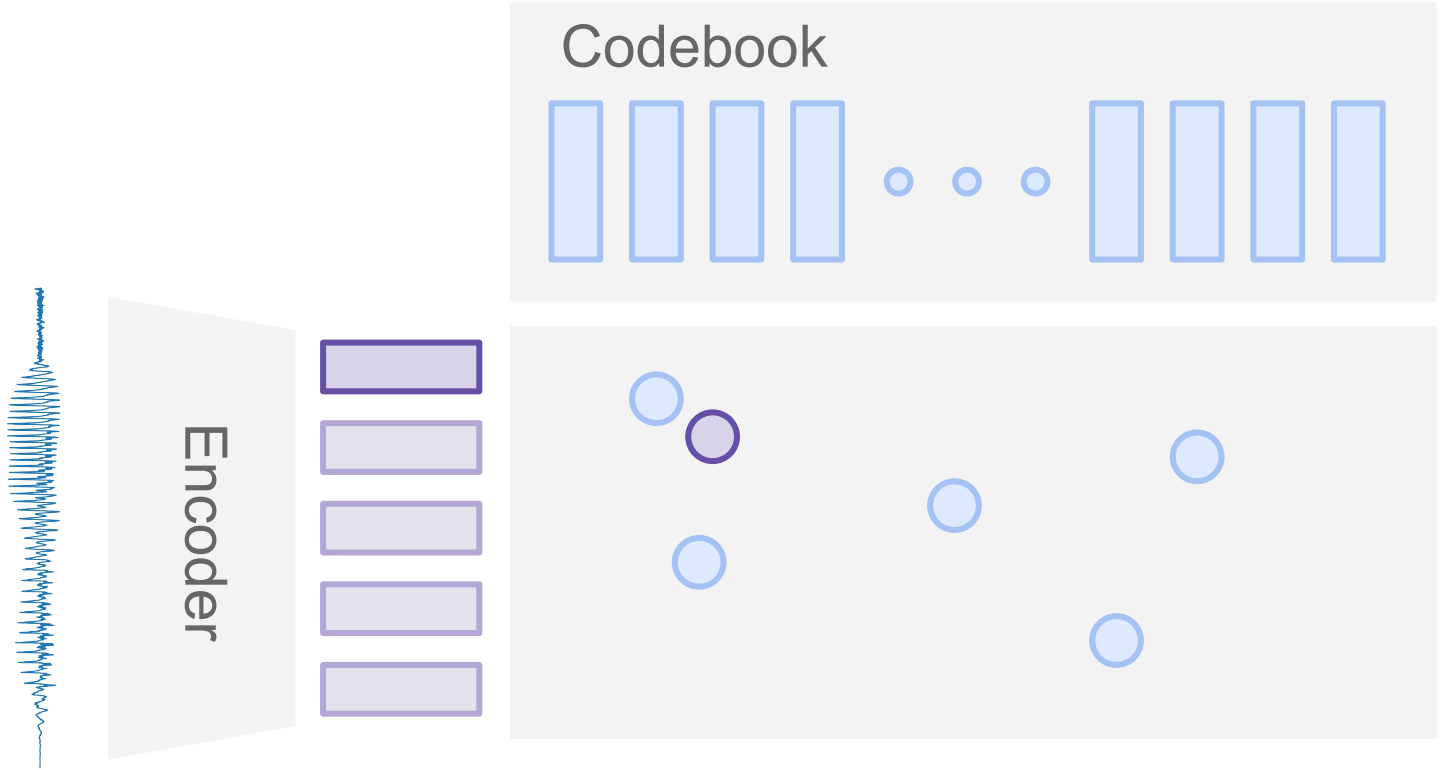

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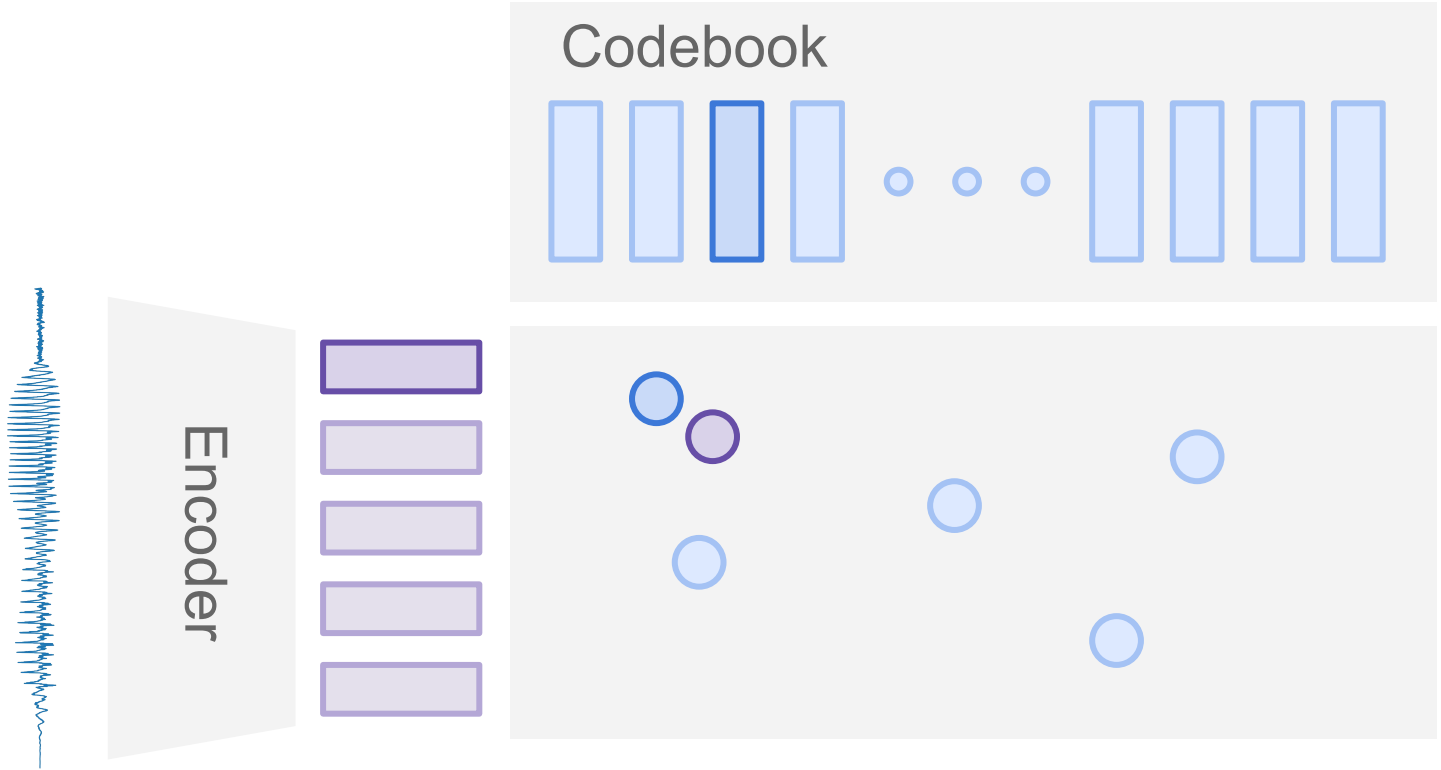

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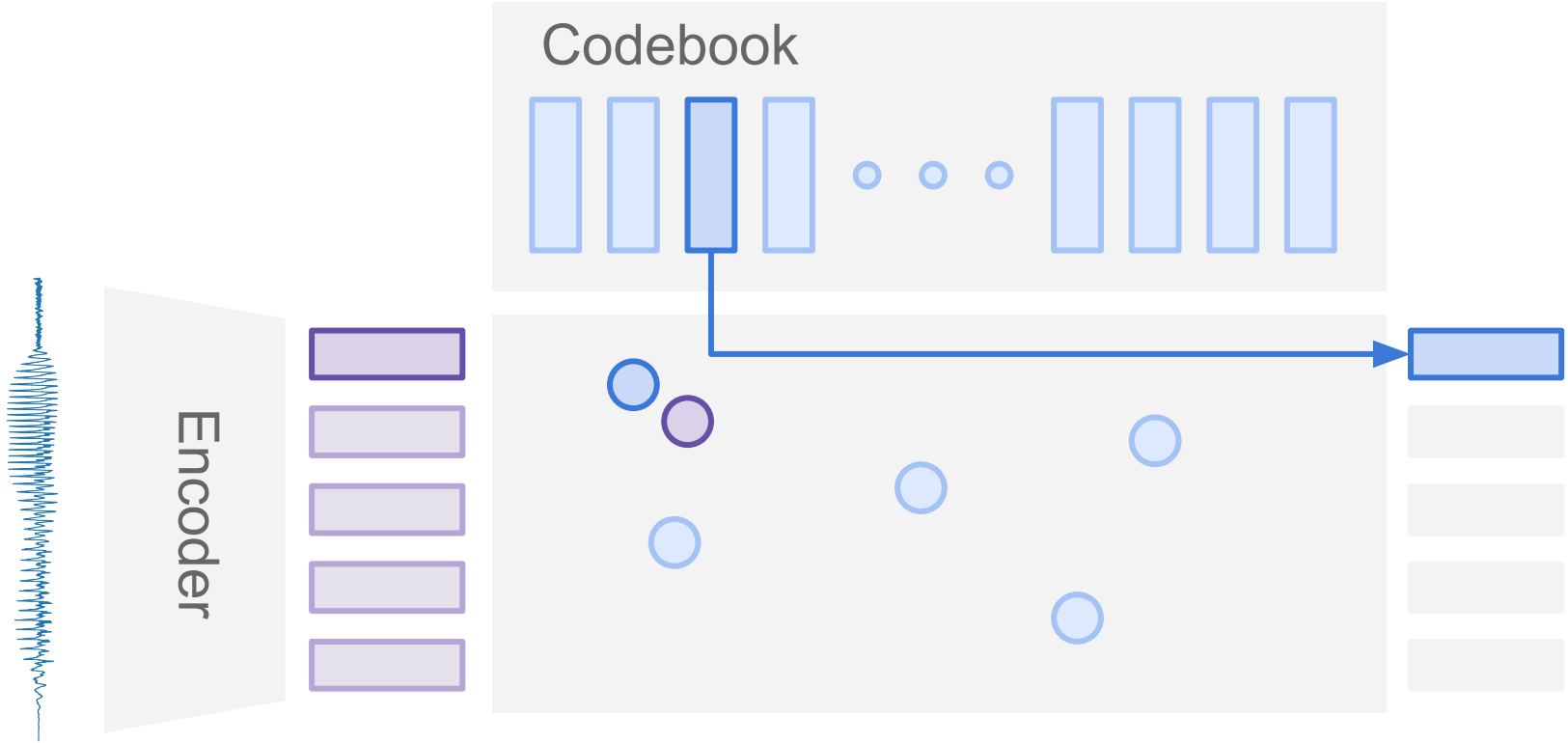

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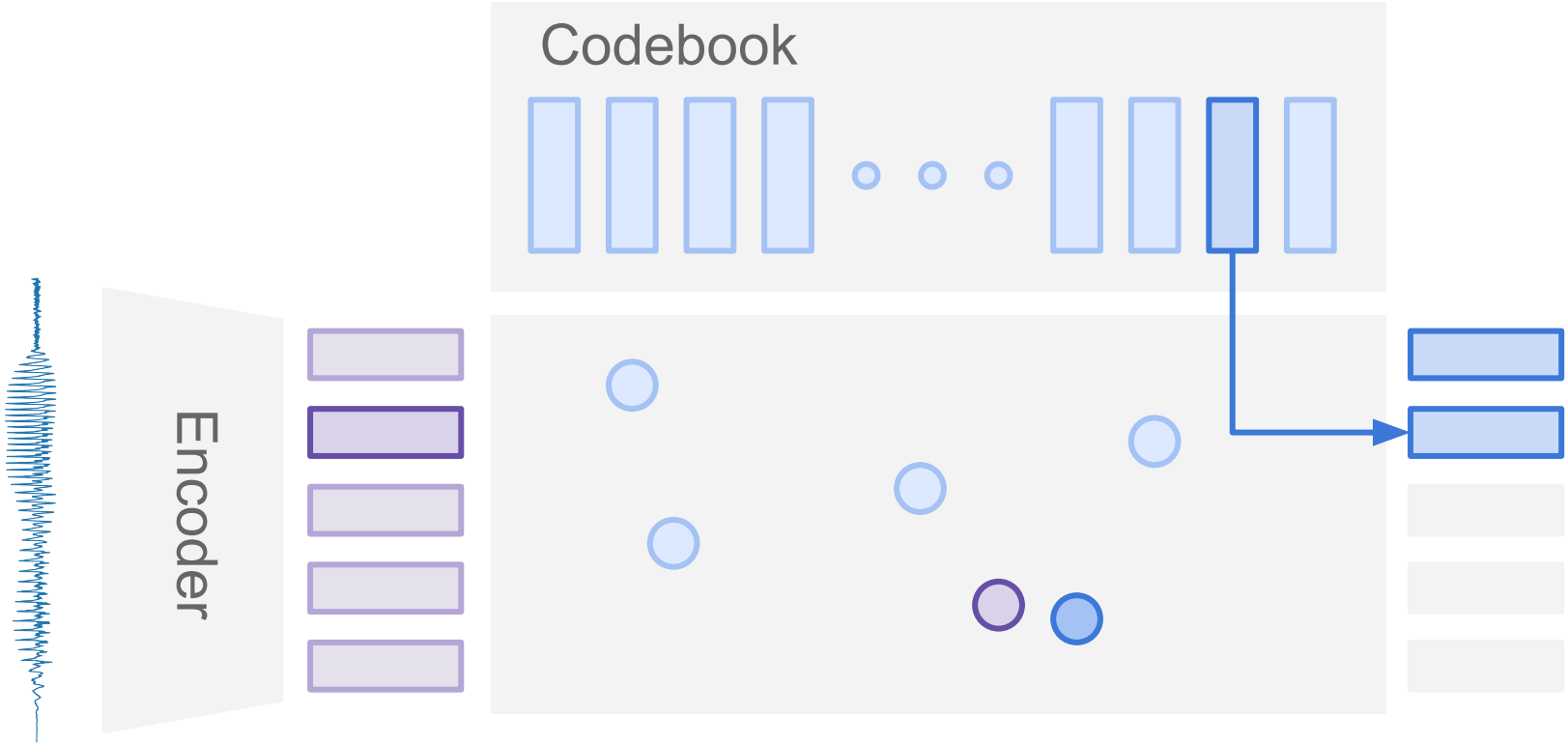

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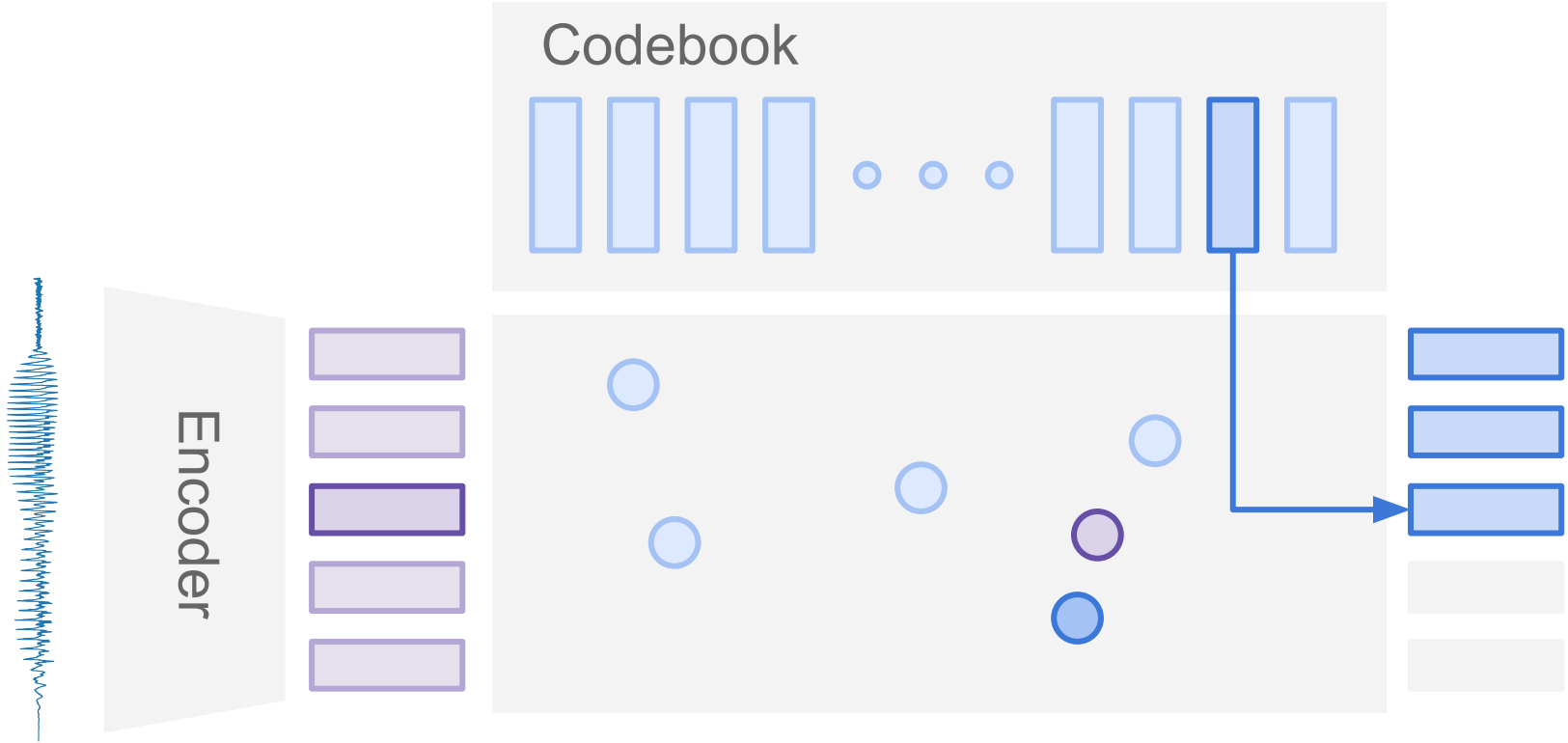

Our contribution: we propose and compare two models for acoustic unit discovery in the *ZeroSpeech 2020 Challenge*.

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Inspired by: J. Chorowski, et al. "Unsupervised speech representation learning using wavenet autoencoders." IEEE/ACM transactions on audio, speech, and language processing. 2019.

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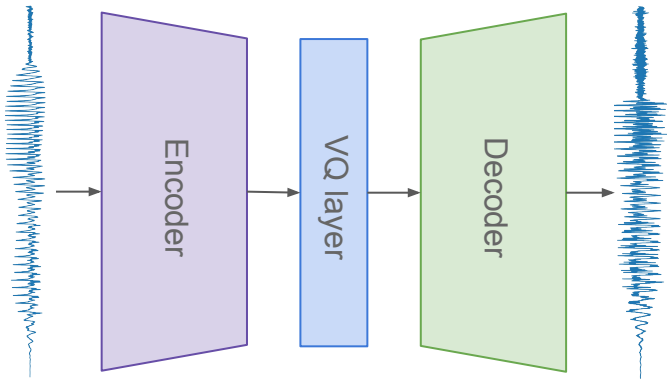

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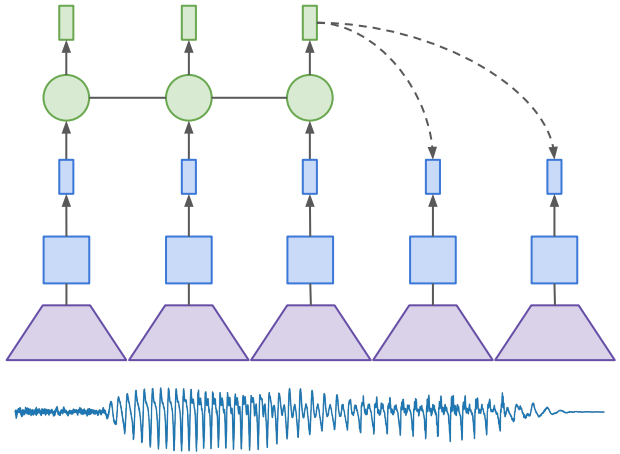

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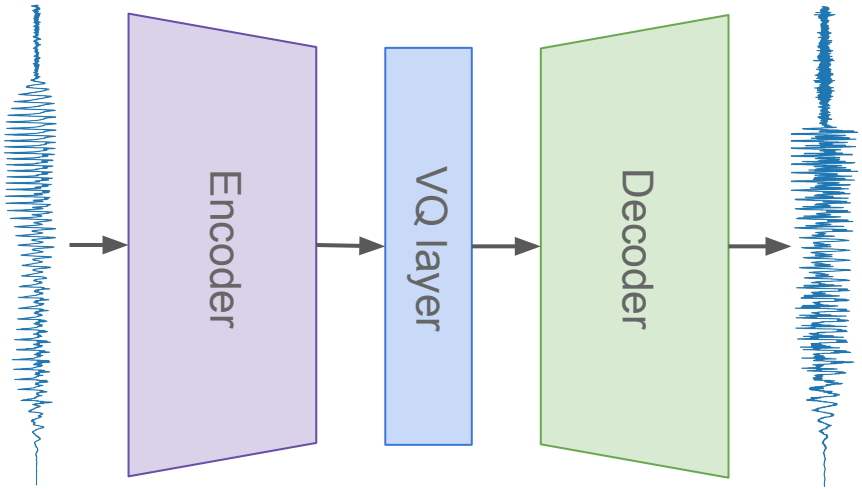

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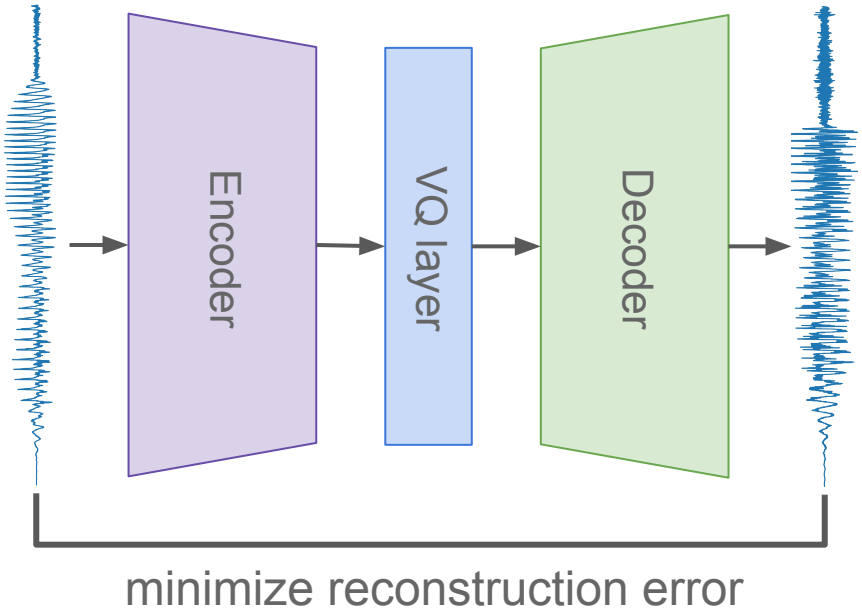

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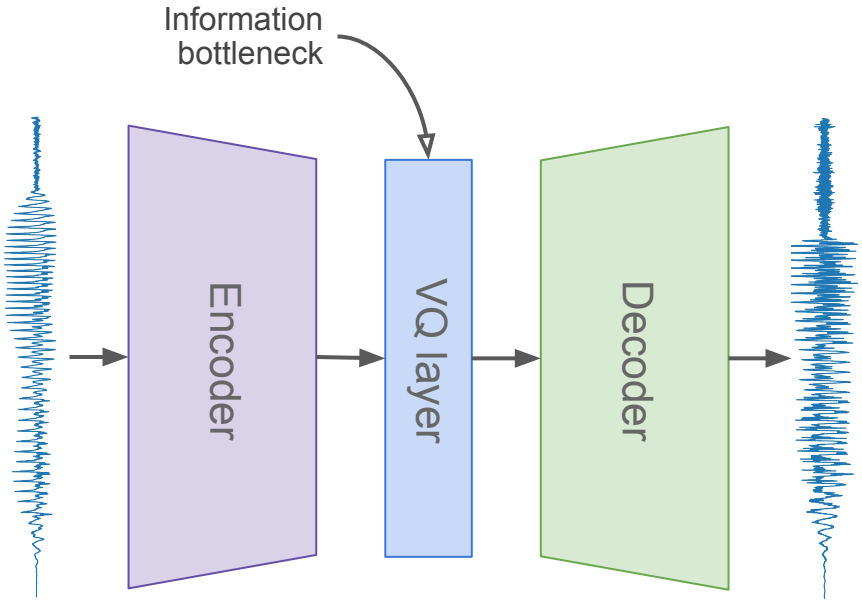

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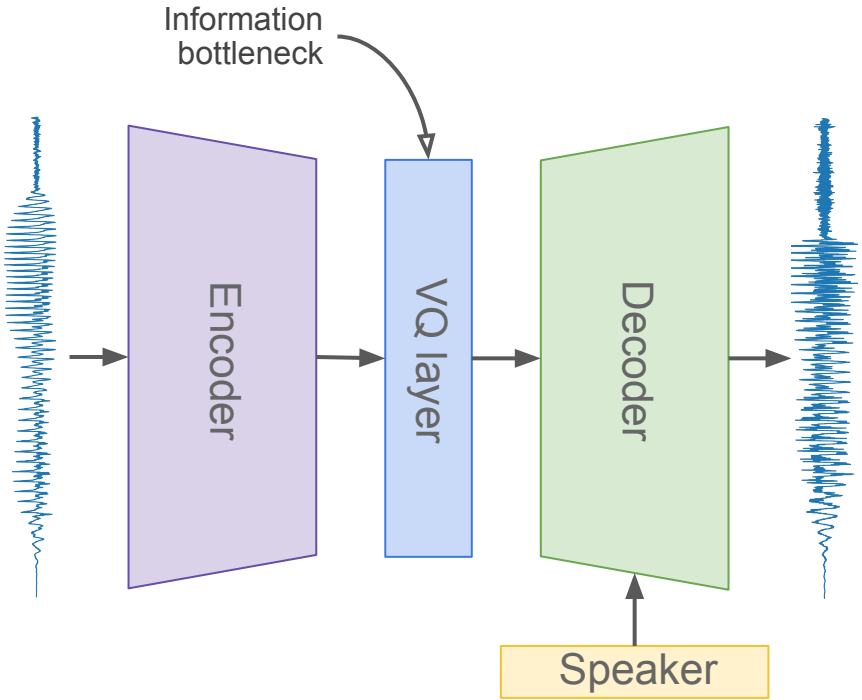

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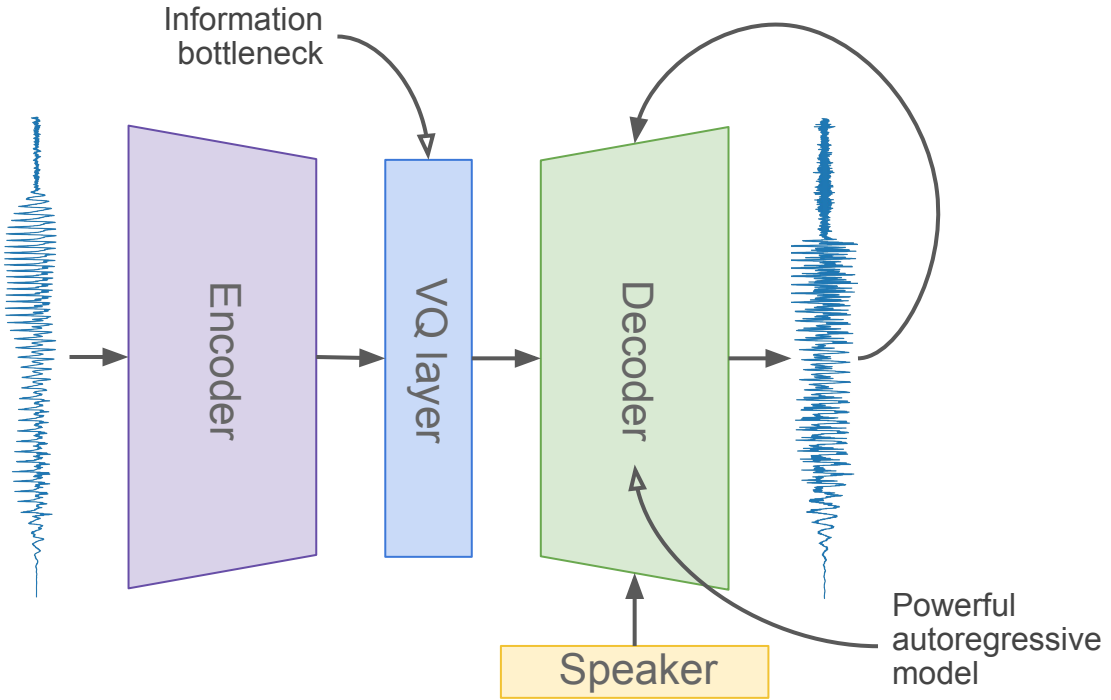

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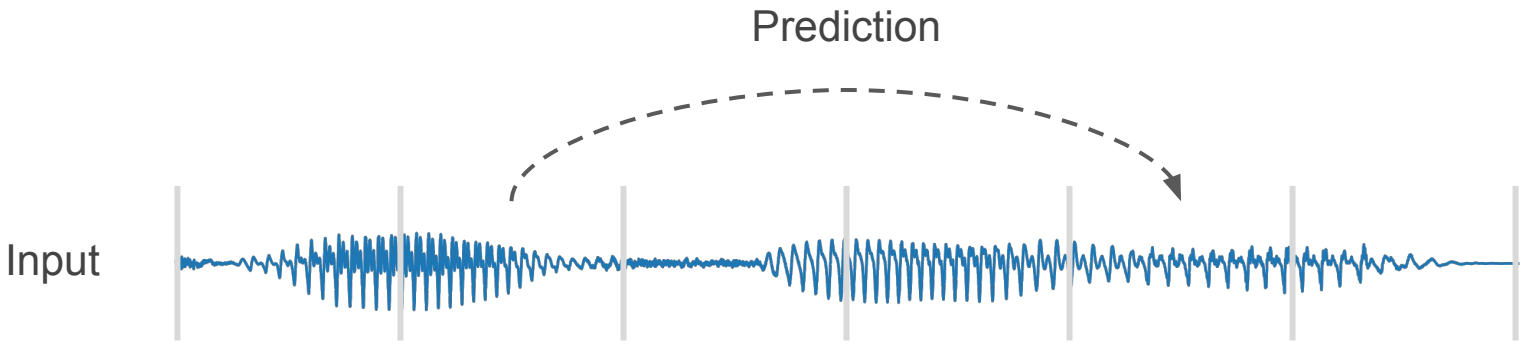

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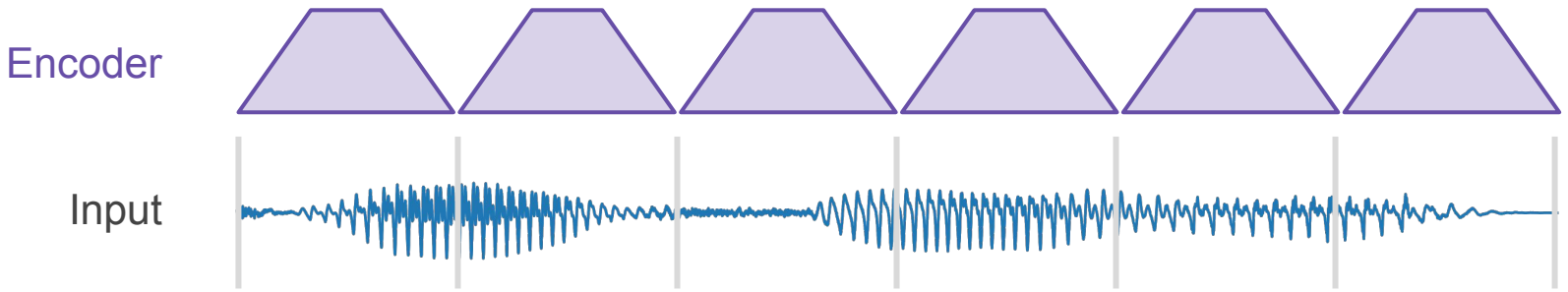

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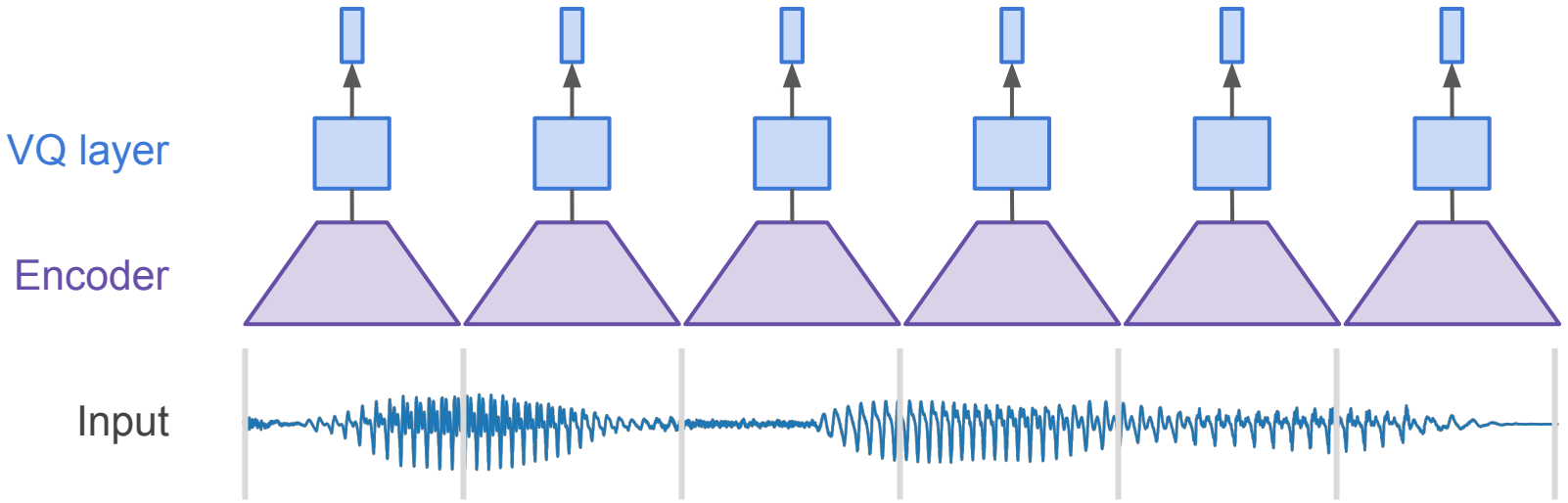

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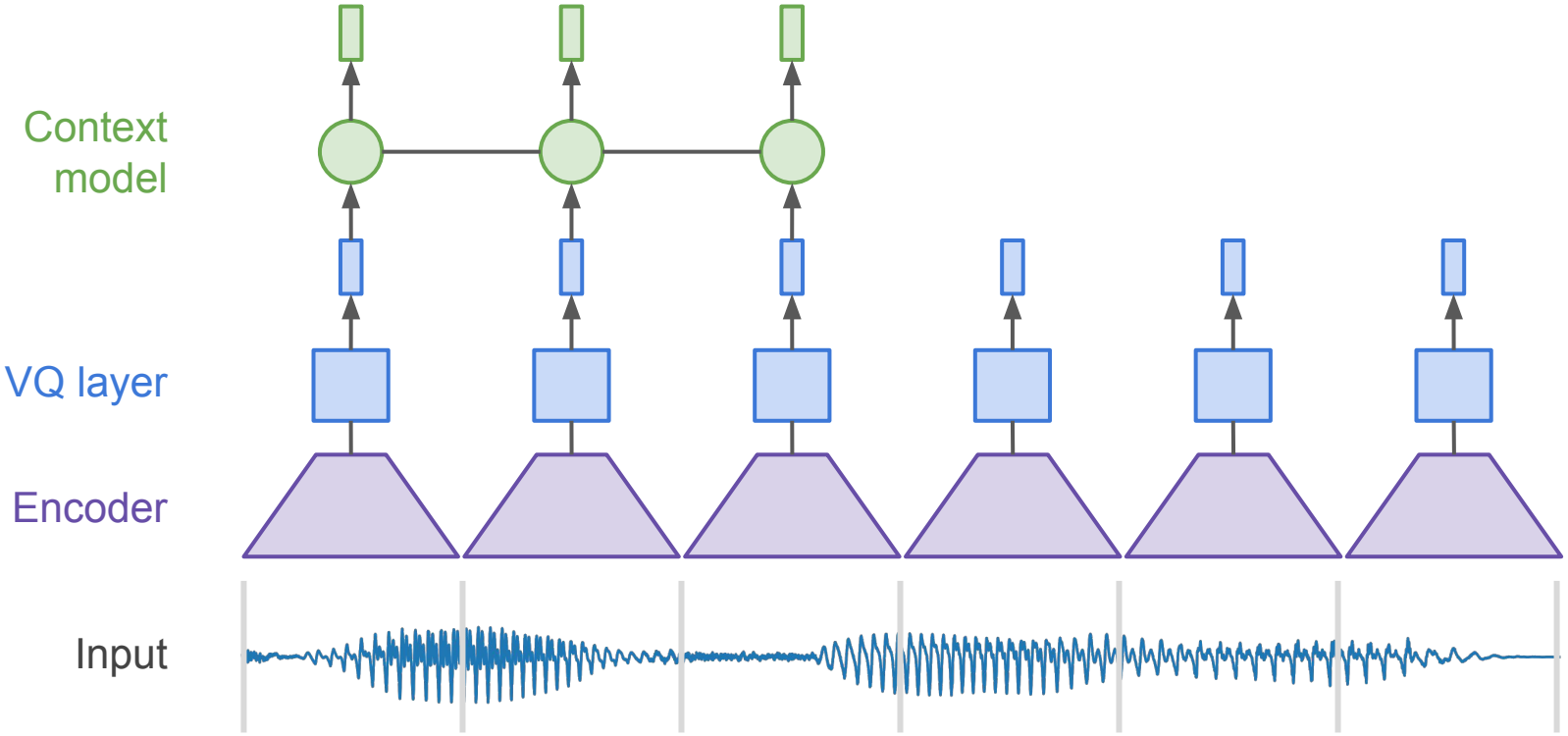

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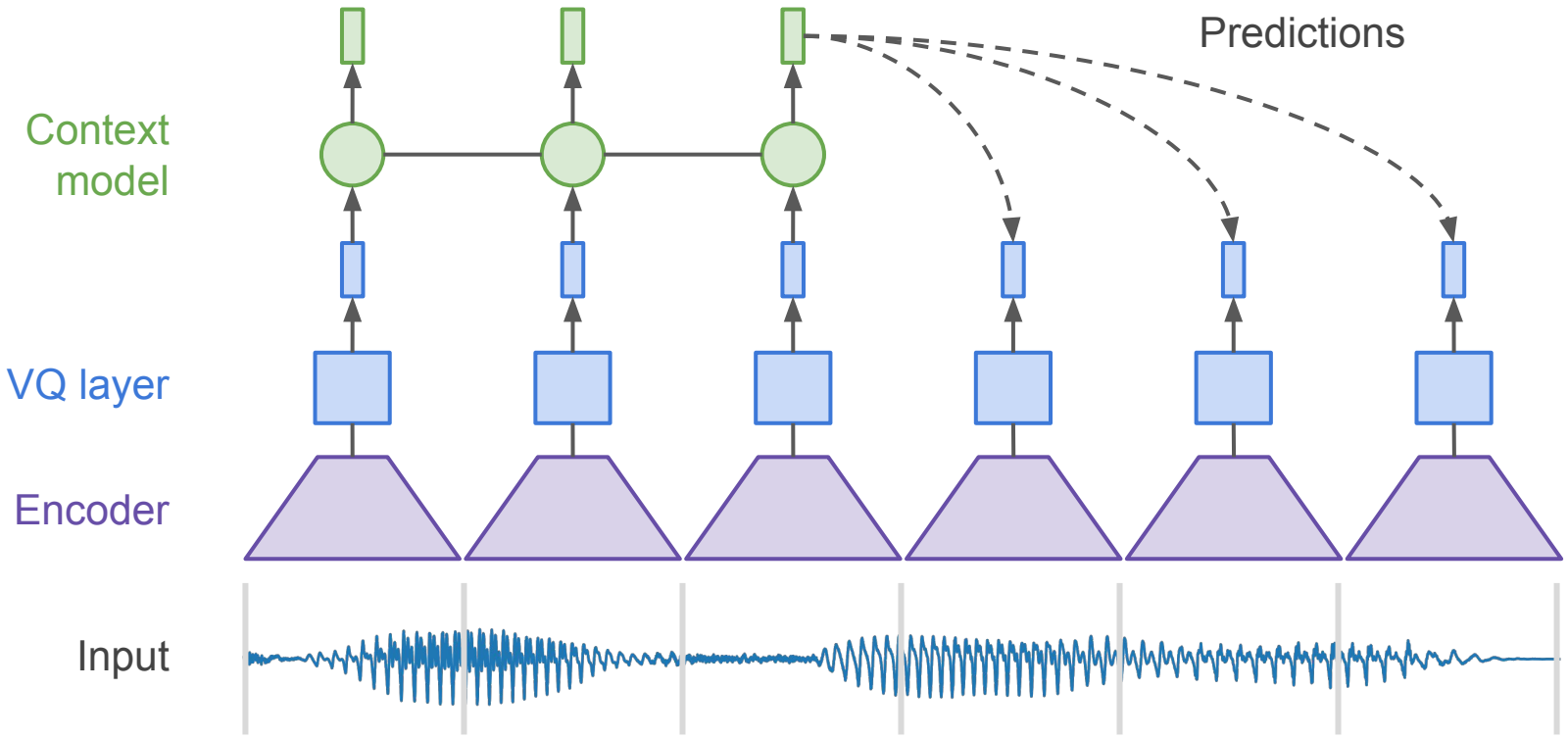

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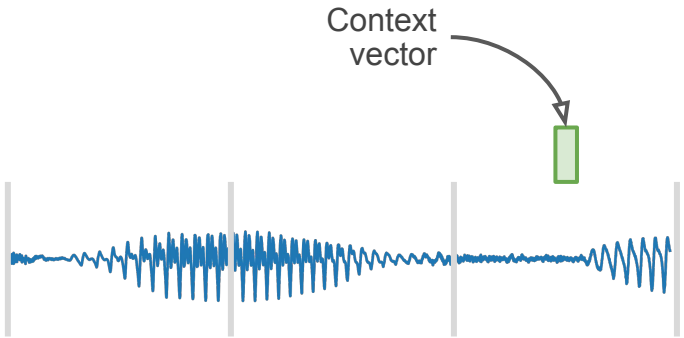

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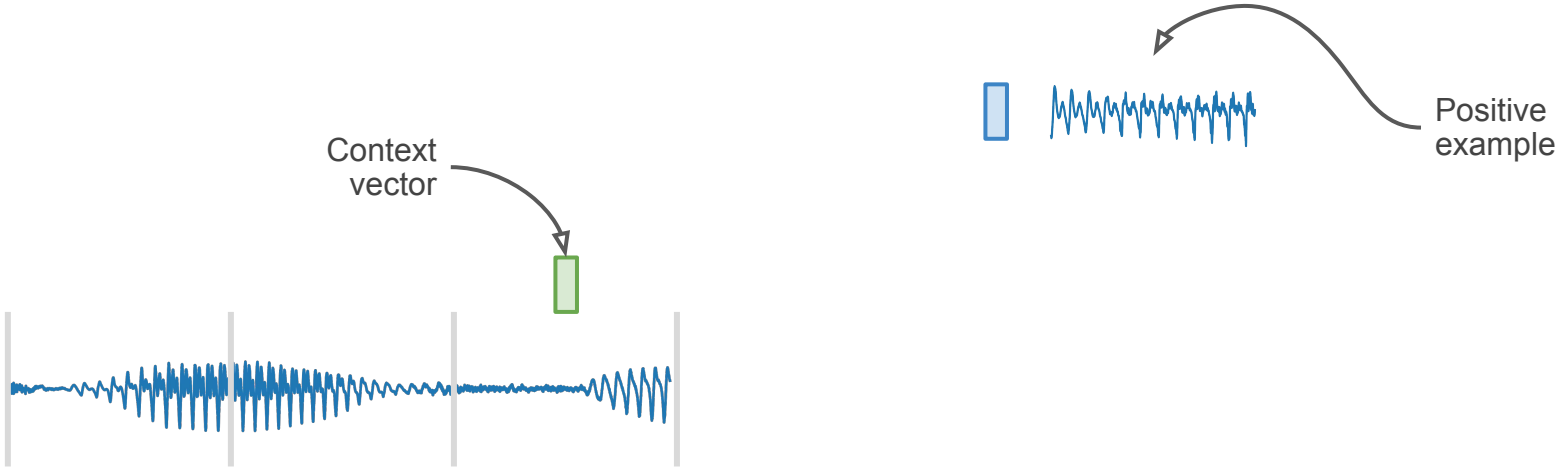

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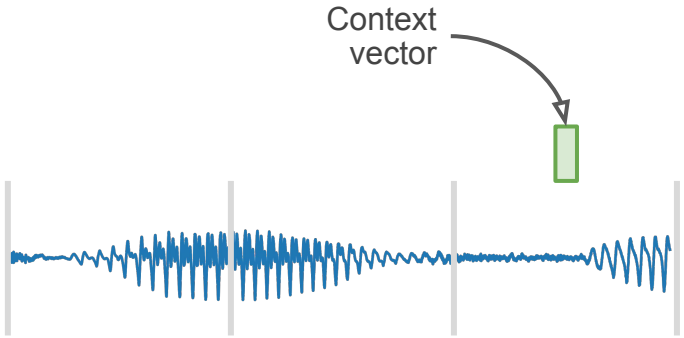

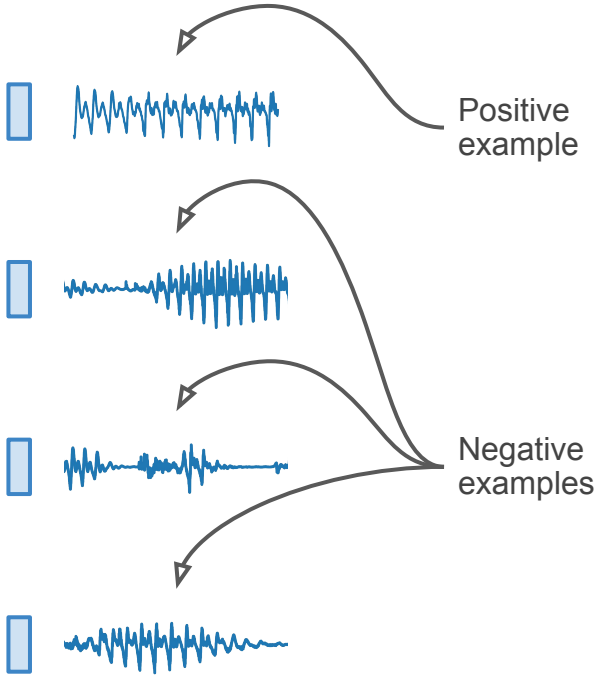

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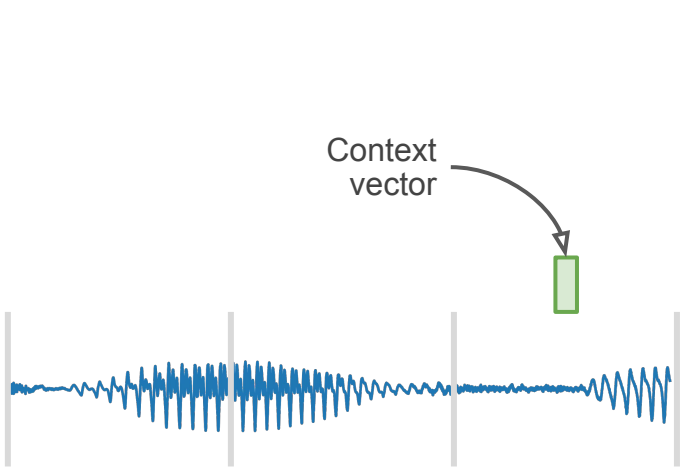

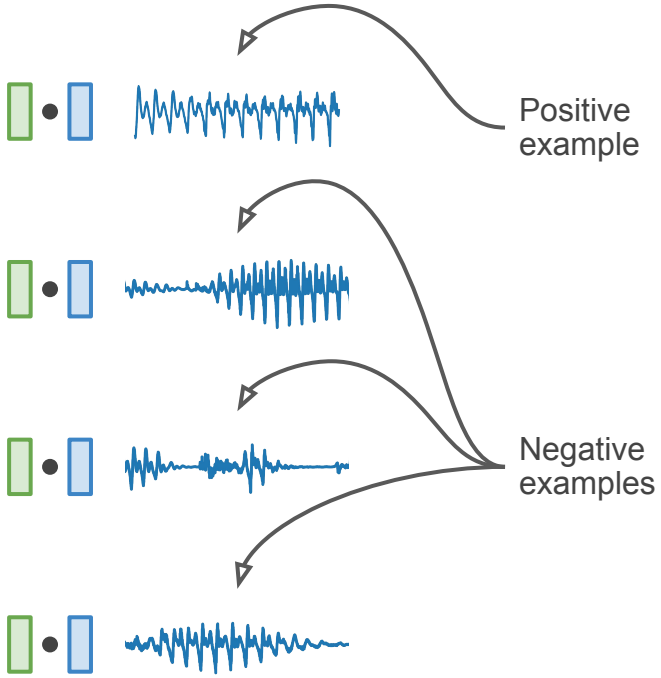

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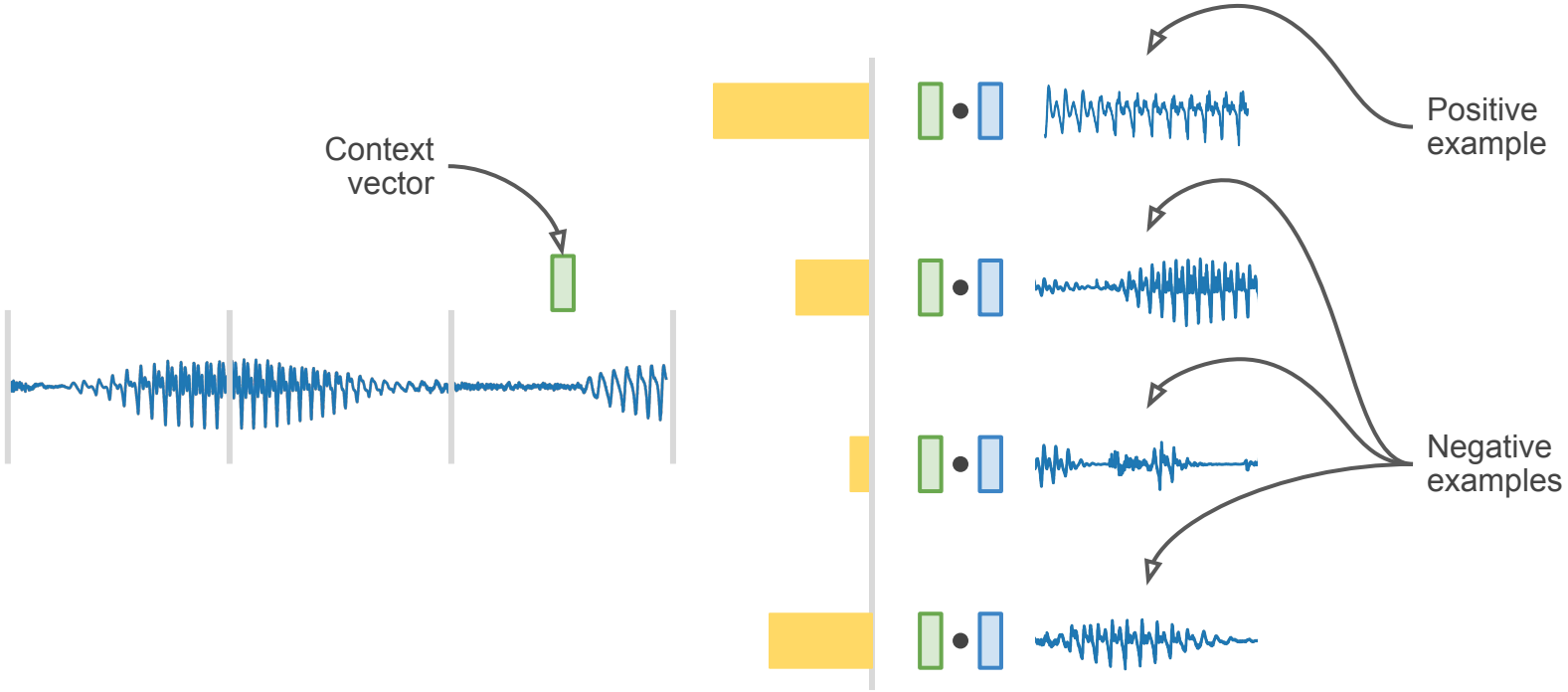
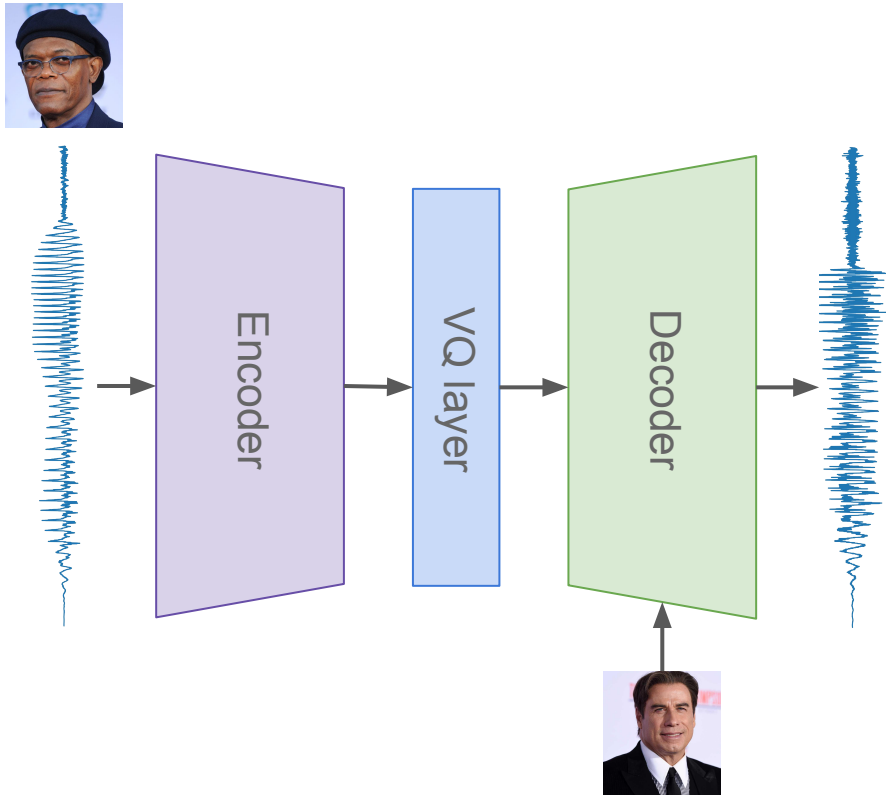


Evaluation - Voice Conversion



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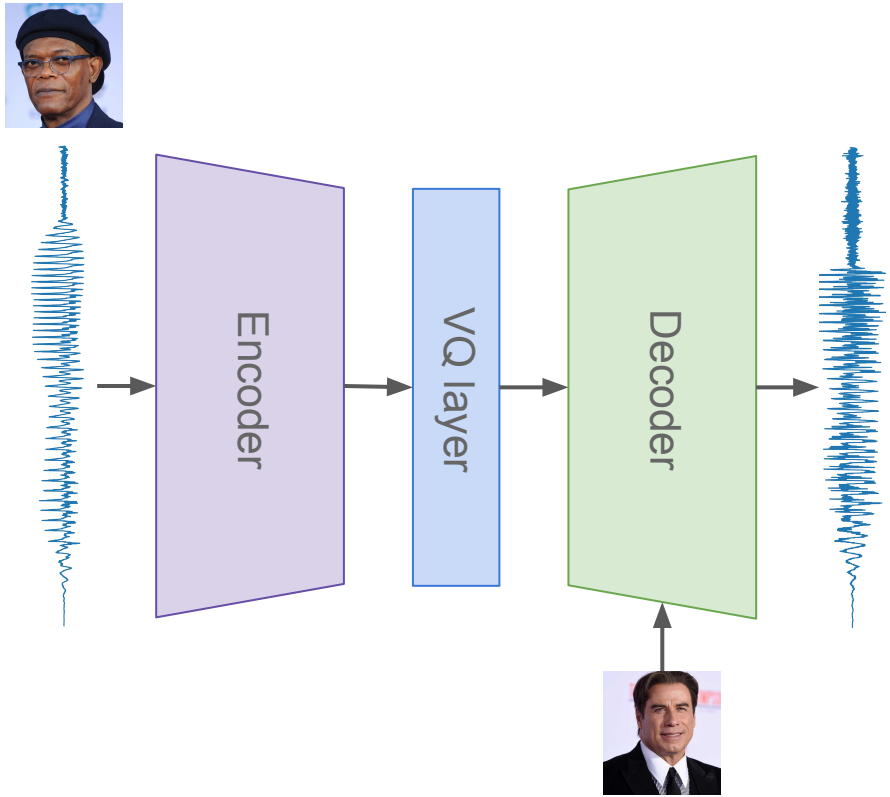

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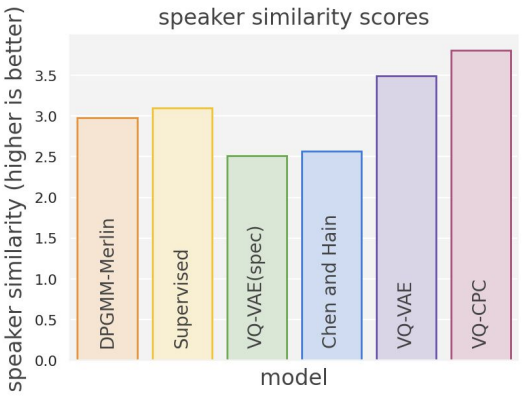
Evaluation - Voice Conversion

Source	Converted	Target	Other Conversion
			
			

Evaluation - Voice Conversion



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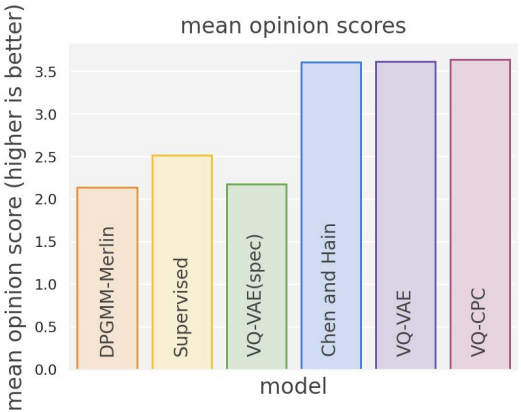

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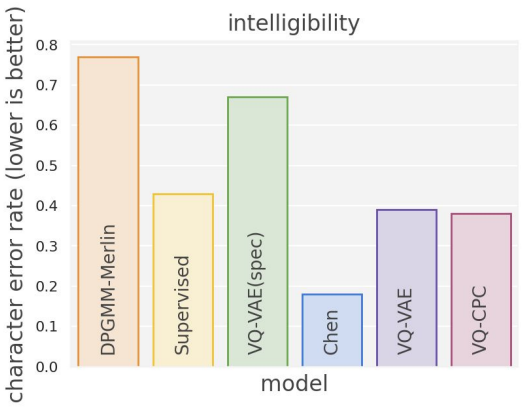

Evaluation - ABX Score

Triphone A:

beg

Encoder

Evaluation - ABX Score

Triphone A:

beg

Encoder

Triphone B:

bag

Encoder

Evaluation - ABX Score

Triphone A:

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Encoder

Triphone X:

beg

Encoder

Triphone B:

bag

Encoder

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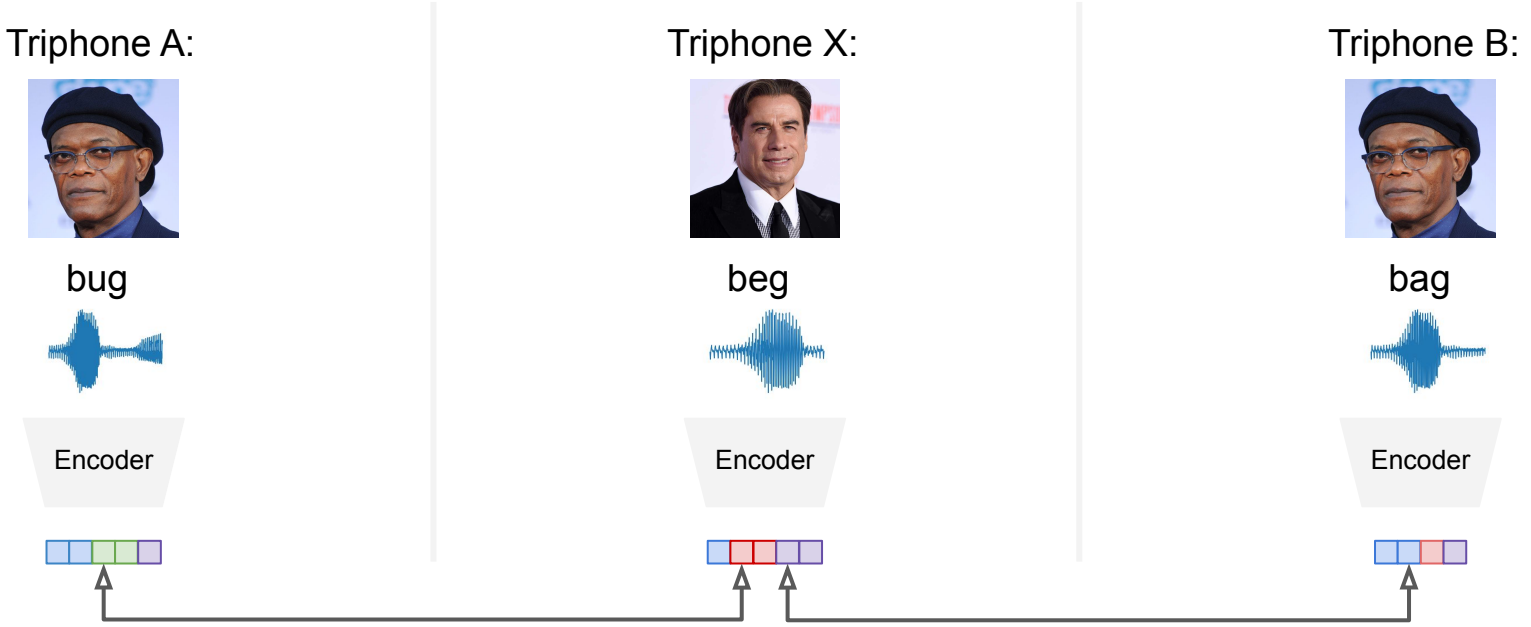

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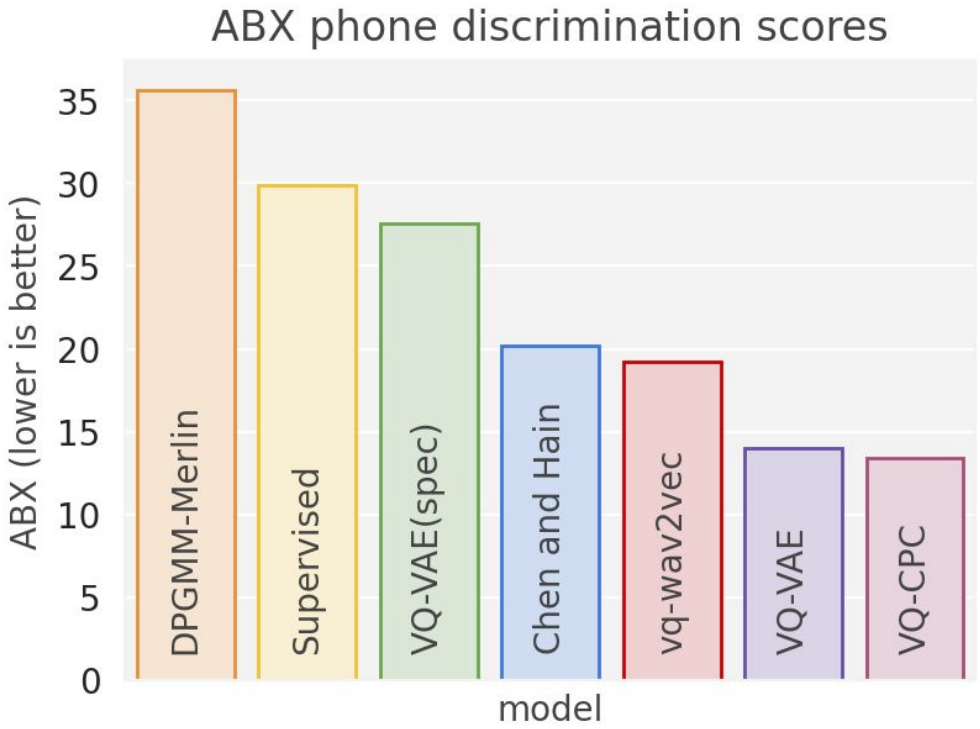
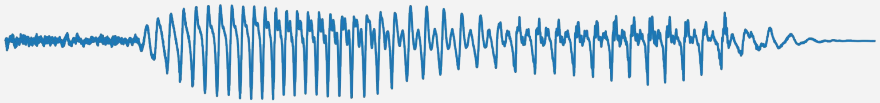


Questions?



Vector Quantized Variational Autoencoder