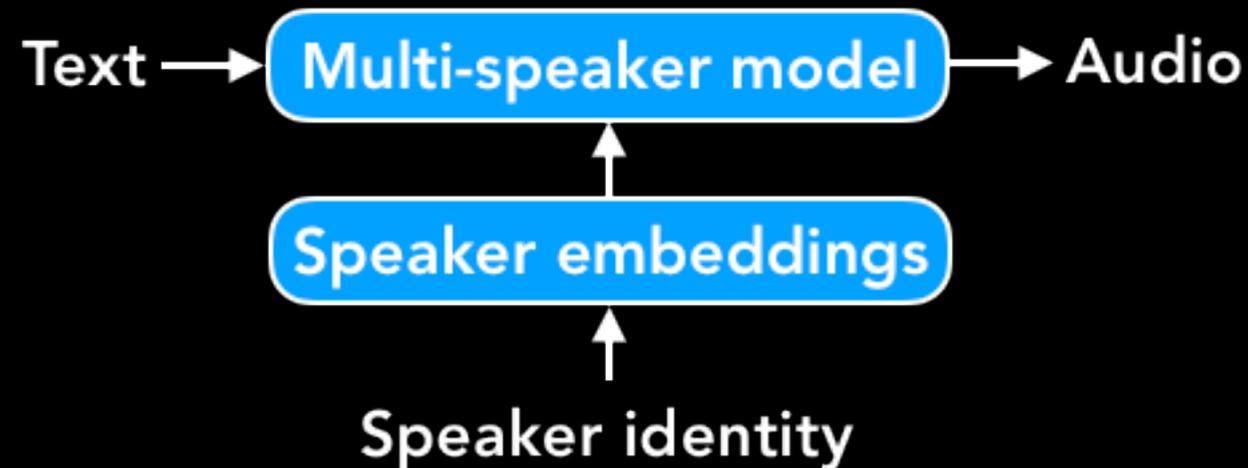


# Neural Voice Cloning with a Few Samples

Sercan O. Arik, Jitong Chen, Kainan Peng\*, Wei Ping, Yanqi Zhou

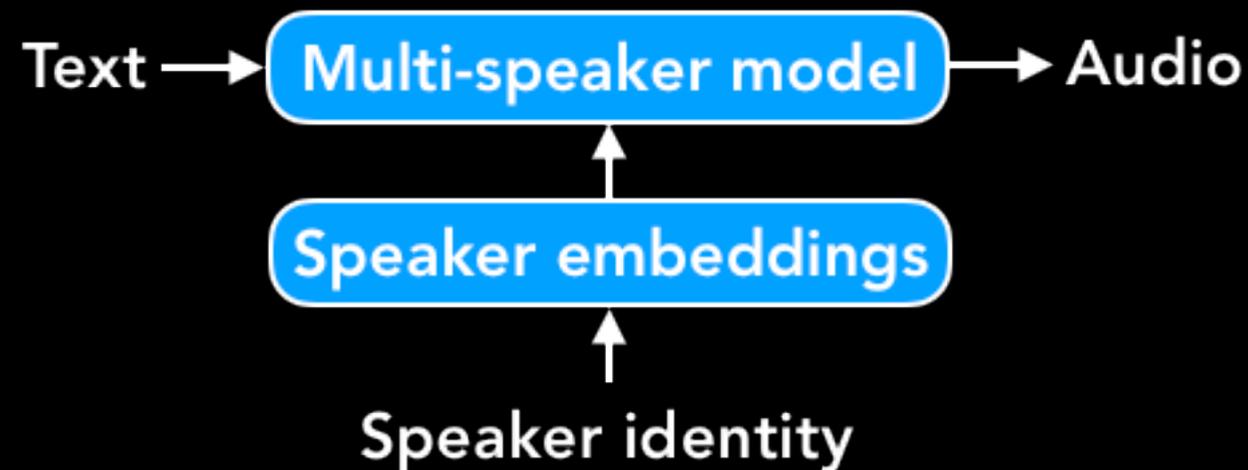
# Motivations

- Text-to-speech (TTS) models can be conditioned on text and speaker identity.
  - Text: linguistic information, content of the generated speech.
  - Speaker identity: speaker information (accent, pitch, speech rate...).



# Motivations

- Text-to-speech (TTS) models can be conditioned on text and speaker identity.
  - Text: linguistic information, content of the generated speech.
  - Speaker identity: speaker information (accent, pitch, speech rate...).



- Limitations:
  - Can only generate speech for observed speakers during training.
  - Require lots of speech samples per speaker (e.g., Deep Voice 2).

# Voice Cloning

- Voice cloning: synthesize the voices of new speakers from a few speech samples (few-shot generative model).
- Applications: personalized speech interfaces, content creation, assistive technology...



# Voice Cloning

- Voice cloning: synthesize the voices of new speakers from a few speech samples (few-shot generative model).
- Applications: personalized speech interfaces, content creation, assistive technology...
- Challenges:
  - Generalization: learn the voice of a new speaker.
  - Efficiency: extract the speaker characteristics from a few speech samples.
  - Computational cost: cloning with low latency and small footprint.
- Two approaches:
  - Speaker adaptation.
  - Speaker encoding.

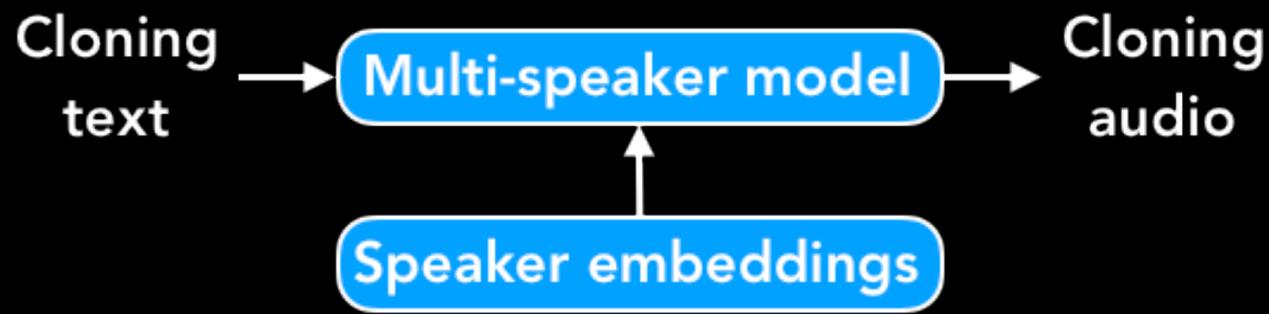


# Speaker Adaptation

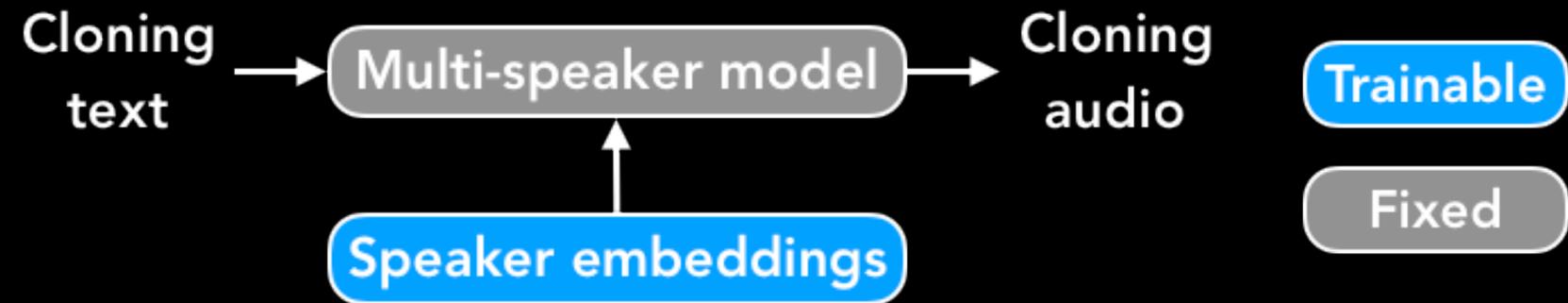
- Fine-tune a pre-trained multi-speaker model for a new speaker.
- Training data: a few text and audio pairs.

# Speaker Adaptation

- Fine-tune a pre-trained multi-speaker model for a new speaker.
- Training data: a few text and audio pairs.
- Two options for speaker adaptation:



Fine-tune the whole model



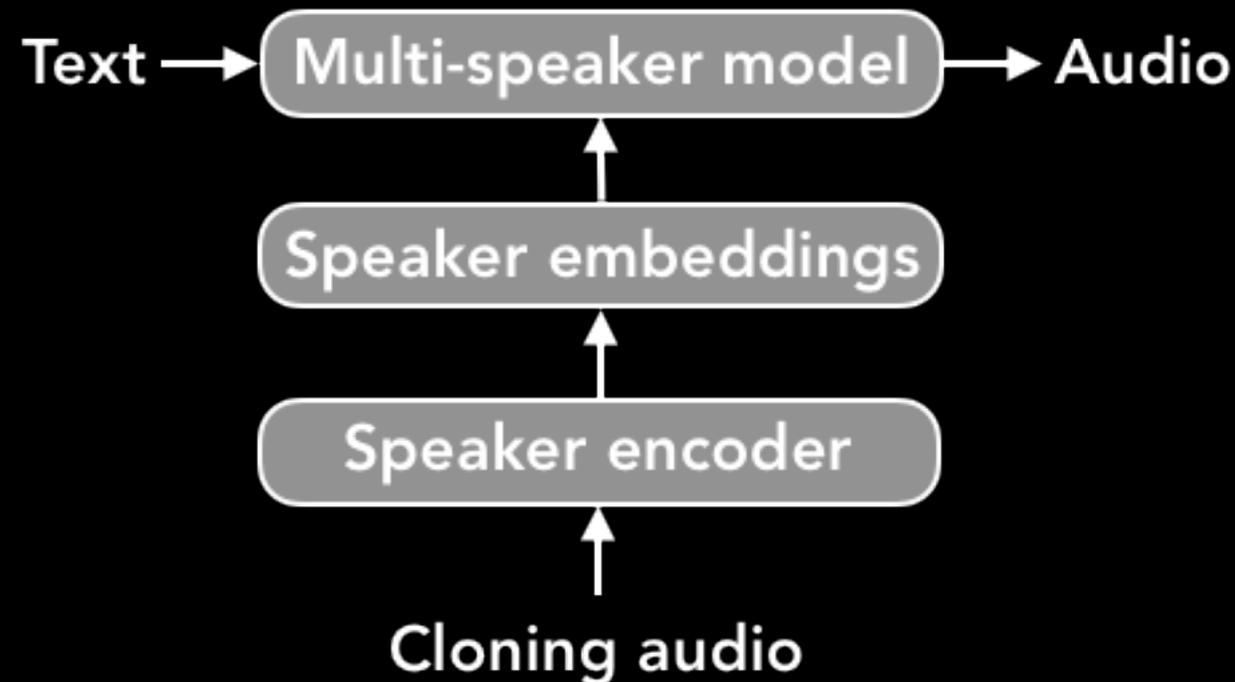
Fine-tune the speaker embedding only

# Speaker Adaptation Analysis

Approaches	Speaker Adaptation	
	Embedding-only	Whole-model
Cloning time	8 h	5 min
# of parameters per speaker	128	25 million

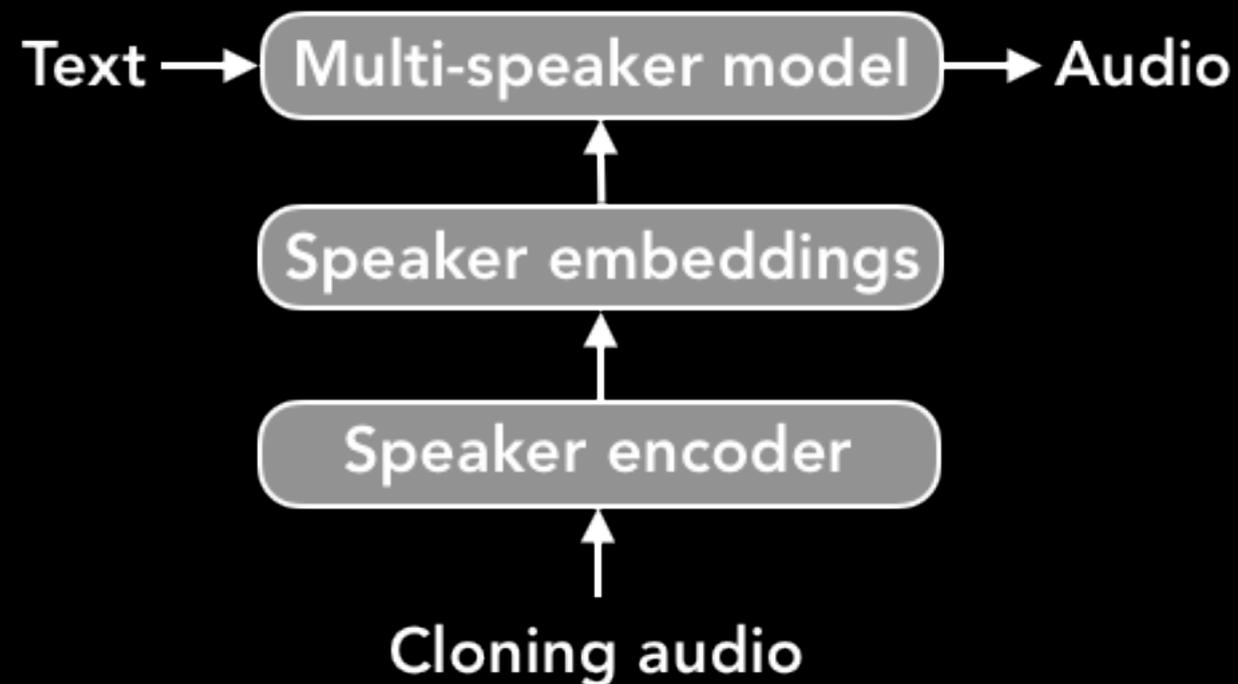
# Speaker Encoding

- Directly predict a new speaker embedding for a multi-speaker model.
- Train a speaker encoder with audio and speaker embedding pairs.



# Speaker Encoding

- Directly predict a new speaker embedding for a multi-speaker model.
- Train a speaker encoder with audio and speaker embedding pairs.
- Cloning time: a few seconds, more favorable for low-resource deployment.



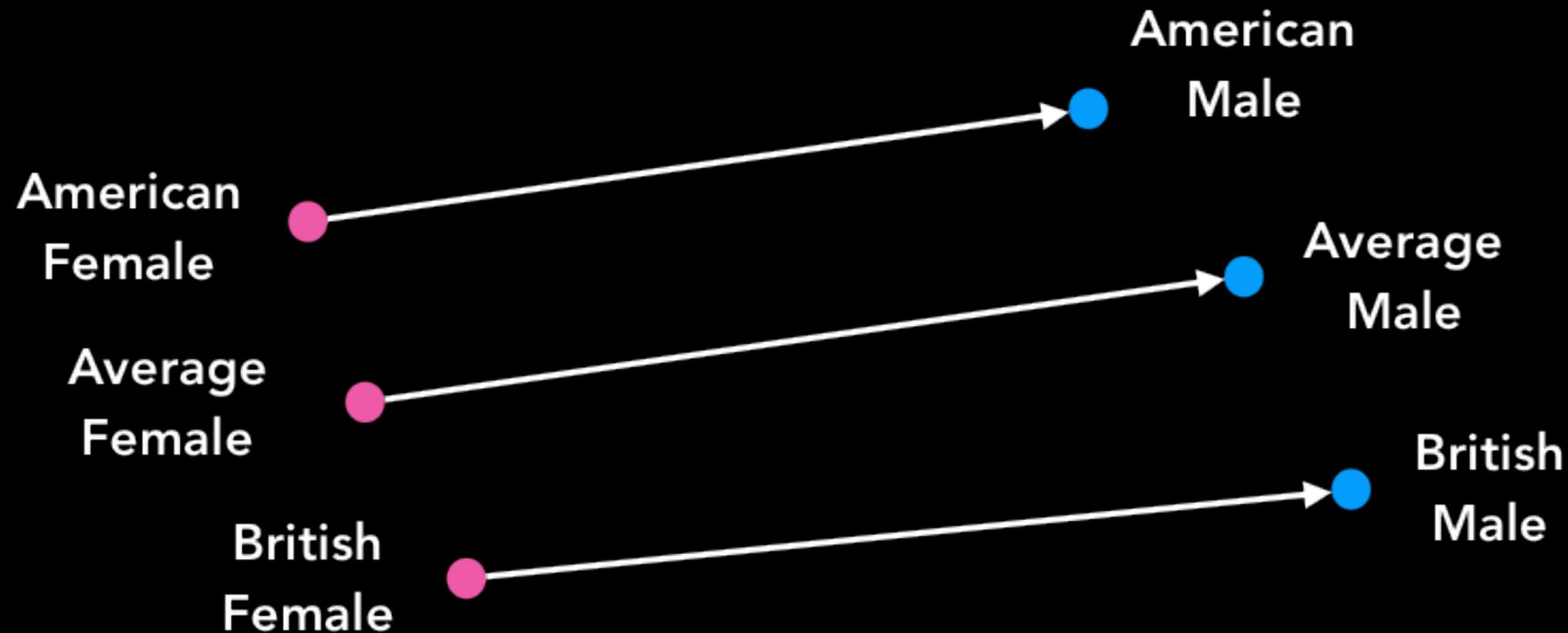
# Results

- Vocoder: classical Griffin-Lim algorithm.
- Demo website: <http://audiodemos.github.io>

Approaches		Speaker Adaptation		Speaker Encoding
		Embedding-only	Whole-model	
Mean Opinion Score (MOS)	Naturalness (5-scale)	2.67	3.16	2.99
	Similarity (4-scale)	2.95	3.16	2.85

# Voice Morphing via Embedding Manipulation

- $\text{BritishMale} + \text{AveragedFemale} - \text{AveragedMale} = \text{BritishFemale}$
- $\text{BritishMale} + \text{AveragedAmerican} - \text{AveragedBritish} = \text{AmericanMale}$



Thank you!

Welcome to our poster,  
and listen to samples!

Today, Session B, #91