# The Blizzard Machine Learning Challenge 2017

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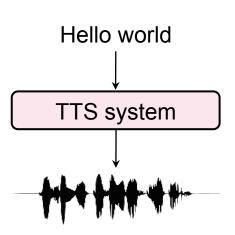
https://synsig.org/index.php/Blizzard\_Challenge\_2017

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#### Introduction

#### Text-to-speech (TTS) system

- Technique for generating for artificial speech given input text
- Evaluation of methods for TTS systems
  - Comparisons are difficult when the training corpus, task, and listening test are different



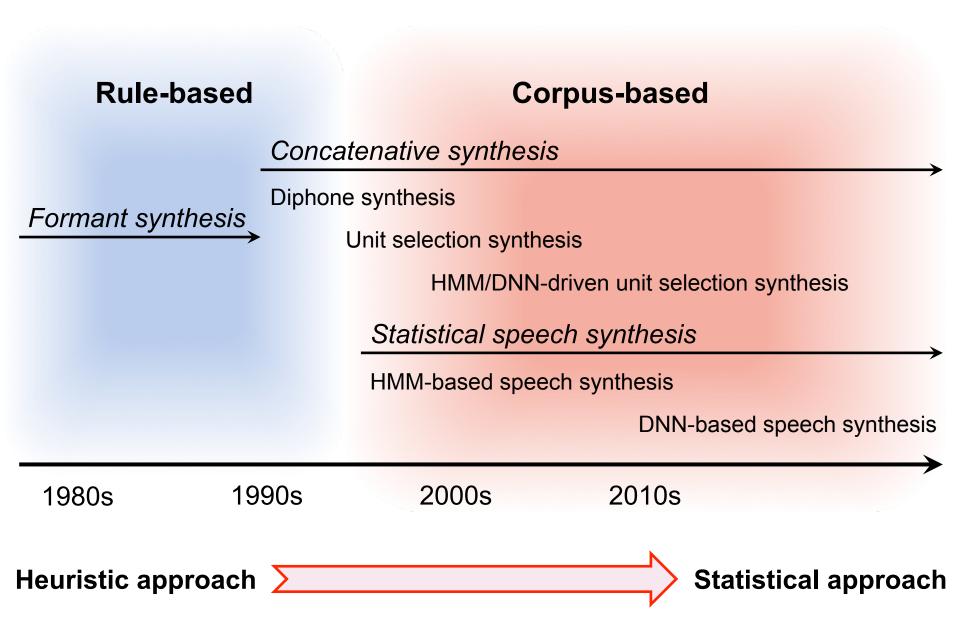
#### Blizzard Challenge [Black & Tokuda; '05]

- Better understand and compare research techniques in building corpus-based TTS systems with the same data
- A lot of time has to be spent on speech-specific tasks
  - ⇒ Not attractive to machine learning researchers

#### Blizzard Machine Learning Challenge

Focus on machine learning problems for speech synthesis

## **History of TTS system**



## Statistical speech synthesis

#### Statistical speech synthesis

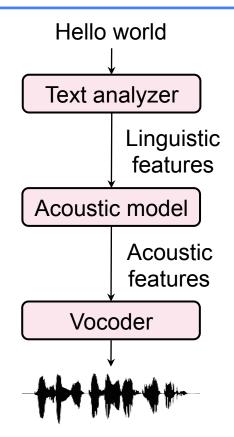
 Mapping to speech waveform from text on the basis of a statistical model

## HMM-based speech synthesis ('95~)

- Context-dependent subword HMMs
- Regression trees to cluster and tie HMM states

## DNN-based speech synthesis ('13~)

Replace regression trees with DNN



## More recent DNN-based speech synthesis ('16~)

- Integration of vocoder and acoustic modeling
  - WaveNet, SampleRNN, etc.
- Integration of text analyzer and acoustic modeling
  - Seq2seq model, Char2Wav, Tacotron, etc.

## **Blizzard Challenge**

### Evaluations of TTS systems

 Comparisons are difficult when the training corpus, task, and listening test are different

#### Blizzard Challenge [Black, Tokuda, King, et al.]

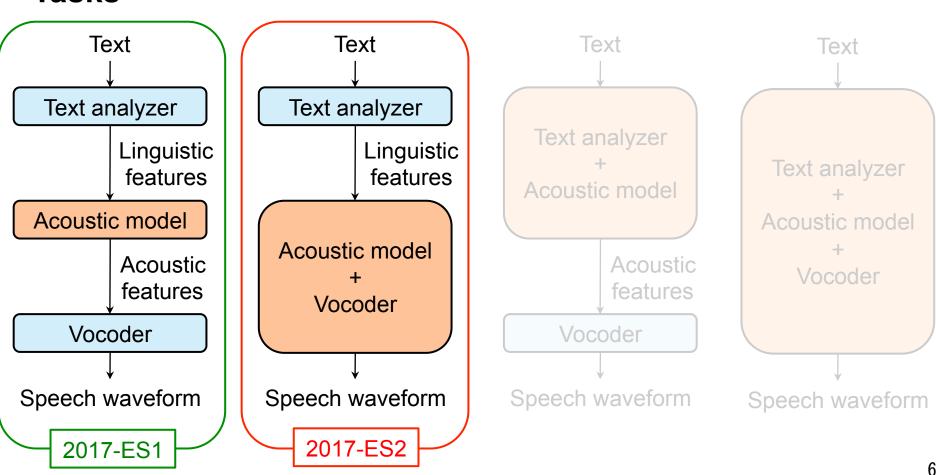
- Goal
  - Better understand and compare research techniques in building corpusbased TTS systems
  - Evaluation campaign rather than competition
     ⇒ Purpose of the challenge is to share knowledge
- Method
  - Participants build voices on a common dataset
  - Organizers evaluate them in a single listening test
- Annual Blizzard Challenge 2005-2017
  - Need of construct all components for a complete TTS system
  - A lot of time has to be spent on speech-specific tasks
    - ⇒ Not attractive to machine learning researchers

# **Blizzard Machine Learning Challenge 2017**

#### Blizzard Machine Learning Challenge

- Does not involve speech-specific tasks
- Allows participants to concentrate on machine learning problem

#### Tasks

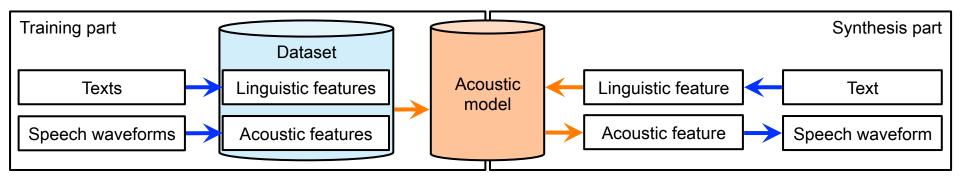


#### **Tasks**

Processes of the organizers
Processes of the participants

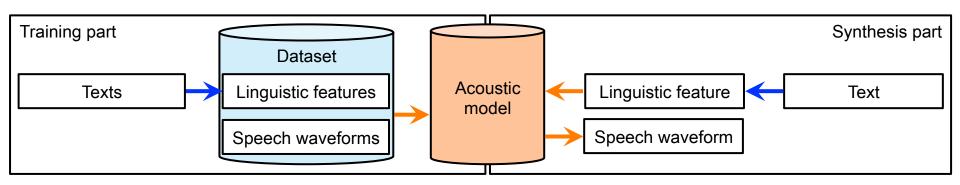
#### o 2017-ES1

Prediction of acoustic features from linguistic features



#### o 2017-ES2

Prediction of speech waveforms from linguistic features



## Datasets (1/2)

#### Data

- Commercial-quality children's audiobooks from Usborne Publishing Ltd.
- Same as the Blizzard Challenge 2016
- 5 hours of speech data



"I'm king of the jungle," roared Lion.
"I'm going to eat you all up."
"No!" cried the jungle animals.

Character1
Character2
Descriptive part

#### Data pruning

- Mismatches between speech waveform and text
- Excessively expressive speech data (e.g. scream, singing voice)
  - ⇒ Negative effect on acoustic model training
- Speech data including phoneme alignment errors were pruned
- 4 hours of speech data (4651 files when divided into sentences)

## Datasets (2/2)

#### Speech waveforms (2017-ES2)

44.1kHz 16 bits monaural Waveform Audio File Format (WAVE)

#### Acoustic features (2017-ES1)

- 77-dimensional acoustic features
  - Log F<sub>0</sub> (linearly interpolated values in unvoiced parts)
  - Voiced and unvoiced information
  - 50-dimensional mel-cepstrum representing spectral envelope
  - 25-dimensional mel-cepstrum representing aperiodicity measures

## Linguistic features (2017-ES1 and 2017-ES2)

- 687-dimentional linguistic features
  - Forced phoneme alignment ⇒ Frame-level linguistic features
  - Normalized to be within 0.0–1.0 based on minimum and maximum

# **Systems**

## 7 teams registered and 3 teams submitted

#### Pairs of team ID and name are confidential

ID	Category	Task	Model	Sampling frequency	Syn. speech
Α	Natural speech	-	_	44.1kHz	
X	Benchmark	2017-ES1	FFNN	44.1kHz	
Υ	Benchmark	2017-ES1	FFNN + Trajectory training	44.1kHz	
Н	Submitted	2017-ES1	LSTM	44.1kHz	
I	Submitted	2017-ES1	LSTM + GAN postfilter	44.1kHz	
Z	Benchmark	2017-ES2	WaveNet	16kHz	
G	Submitted	2017-ES2	LSTM + WaveNet	22.05kHz	

## Listening test

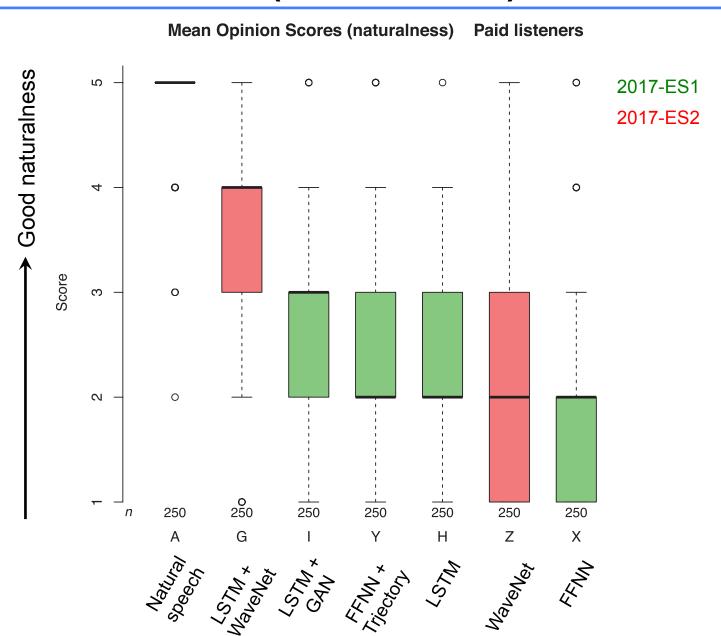
#### Design of listening test

- The evaluation combined the entries for 2017-ES1 and 2017-ES2 into a single listening test
- 50 paid native listeners

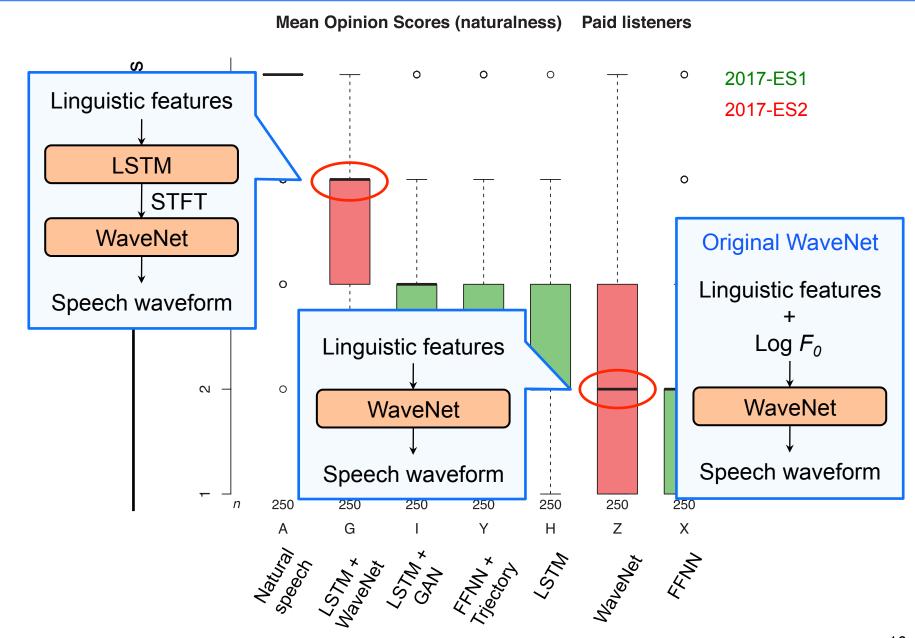
#### Evaluation criteria

- Naturalness
  - 5-point mean opinion score (MOS) test
  - 1: completely unnatural 5: completely natural
- Speaker similarity
  - 5-point MOS test
  - 1: sounds like a different person 5: sounds like the same person
- Intelligibility
  - Dictation test
  - Word error rate (WER)
  - Semantically unpredictable sentence (SUS)

# Result (naturalness)

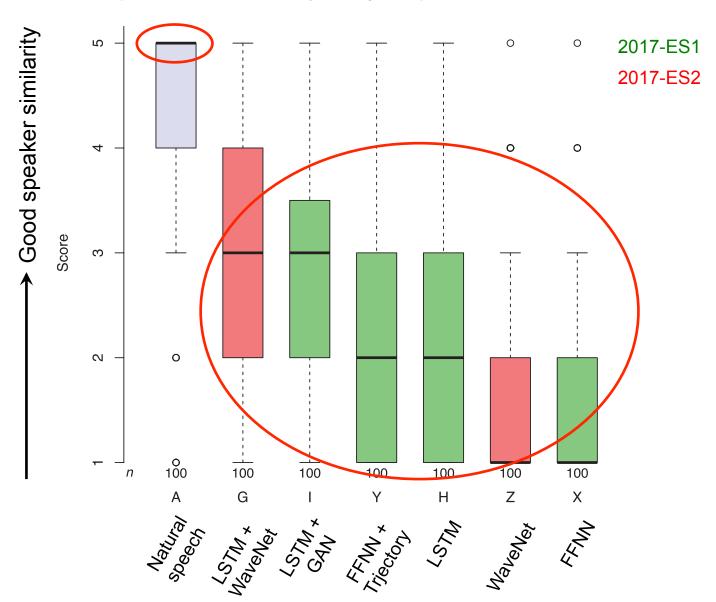


## Result (naturalness)



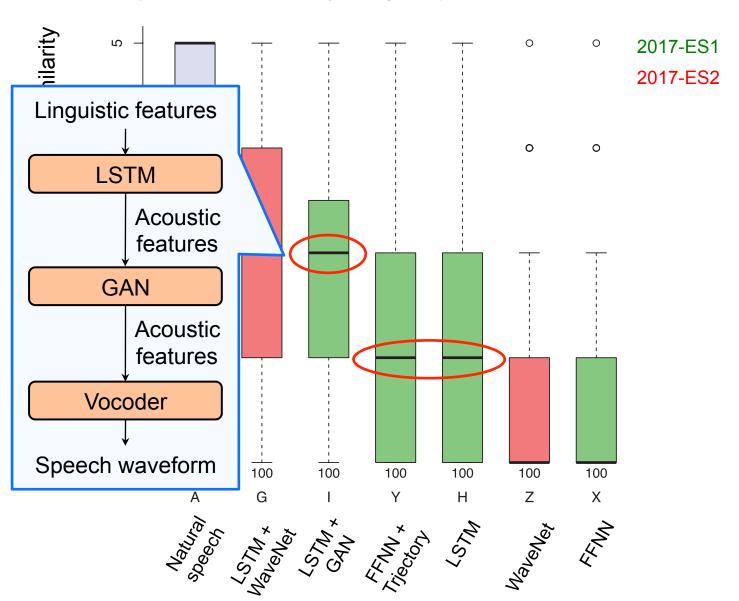
# Result (speaker similarity)

Mean Opinion Scores (similarity to original speaker) Paid listeners

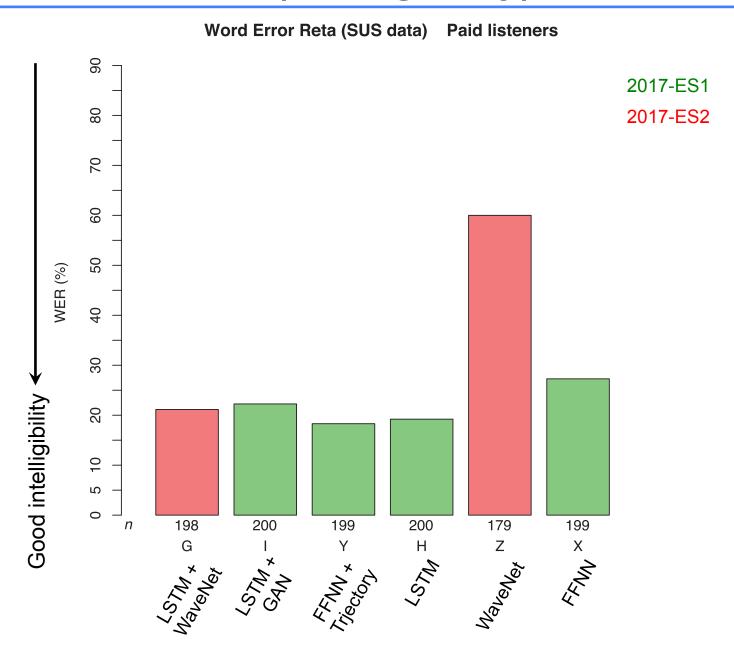


# Result (speaker similarity)

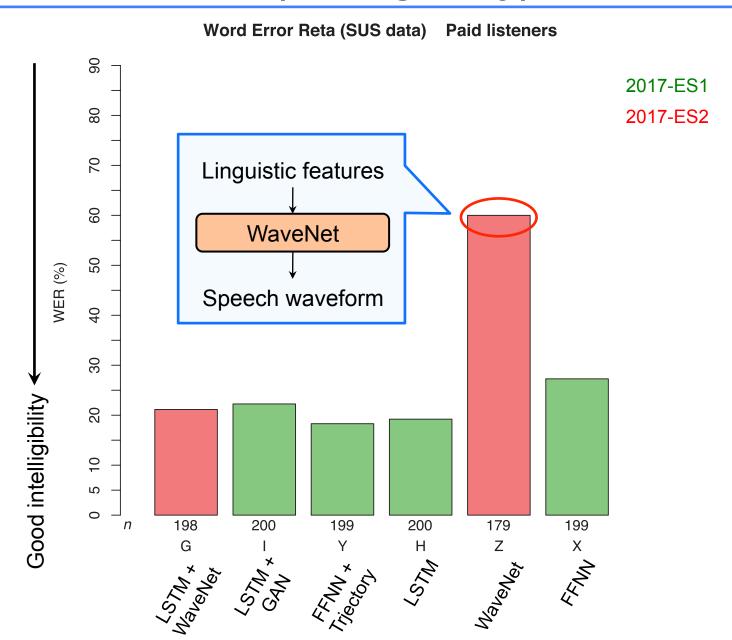
Mean Opinion Scores (similarity to original speaker) Paid listeners



# Result (intelligibility)



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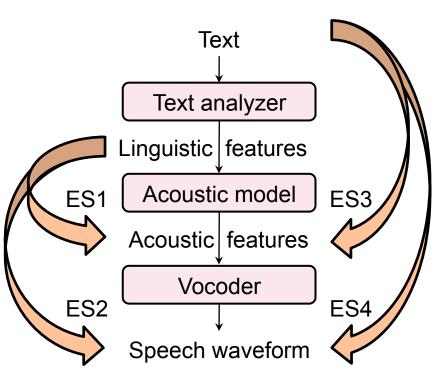
## Discussion and future plan

#### Recruit machine learning researchers

- Lack of advertisement
  - Difficult to control listening test if there are many participants
- Quality confirmation of synthesized speech
  - Release synthesized speech of benchmark system in advance
  - Release training script of benchmark system in advance
  - Release simple objective measure

## End-to-end speech synthesis

- Text → Acoustic feature
- Text → Speech waveform



#### **Conclusions**

#### Blizzard Machine Learning Challenge 2017

- 2017-ES1
  - Prediction of acoustic features from linguistic features
- 2017-ES2
  - Prediction of speech waveform from linguistic features
- Listening test
  - Naturalness, speaker similarity, and intelligibility evaluated
- Results
  - State-of-the-art machine learning approaches achieved higher scores