

Collapsed speech segment detection and suppression for WaveNet vocoder Yi-Chiao Wu, Kazuhiro Kobayashi, Tomoki Hayashi, Patrick Lumban Tobing, Tomoki Toda

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 WaveNet : a deep autoregressive network capable of directly modeling speech waveform *1 • WaveNet vocoder : using acoustic features as auxiliary features to guide WaveNet generating speech samples ***2**

*1. [A. van den Oord et al., 2016] *2. [Tamamori et al., 2017]

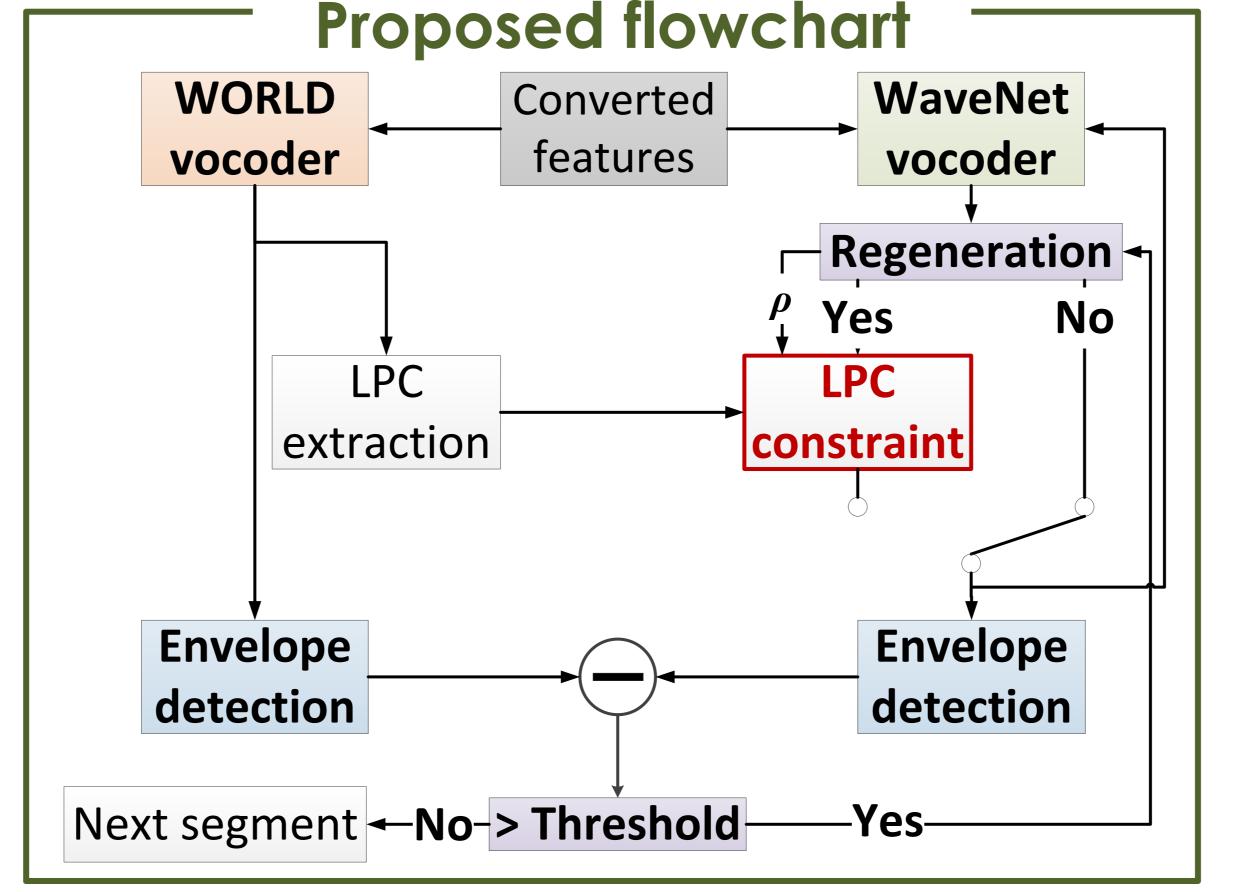
• For voice conversion (VC), speech generation using WaveNet conditioned on the converted features achieves better speech quality than WORLD vocoder • Conditioned on converted features, WaveNet sometimes generates collapsed segments - Type-I: a collapsed segment has extremely large power at all frequencies like white-noise - Type-II: a collapsed segment has irregular short impulse - These collapsed segment are caused by the mismatch between training data (natural speech) and testing data (converted speech) - Training WaveNet vocoder using converted speech is not straightforward because of the limited parallel corpus for VC and the time alignments errors of converted-target speeches.

Collapsed speech suppression based on the LPC-constrained WaveNet vocoder • Motivation: constrained the pmf of current sample using the relationship between current and past samples to prevent WaveNet from generating extremely non-speech like samples • LPC coefficient ϕ : each sample is described by a linear combination of previous samples • Constraint: (WaveNet predicted pmf)*(Gaussian(μ, σ^2))^{ρ}

- μ : the previous samples* ϕ ; σ^2 : the variance of LPC prediction error; ρ : hyper parameter • Problem: applying LPC-constraint causes over-smoothing side effects • Using small ρ first to ease the over-smoothing issue, and then increasing ρ when WaveNet

Collapsed speech detection based on the difference of waveform envelopes • Motivation: only applying LPC-constraint on collapsed speech segments • Observation I: collapsed speech segments are easily detected by the waveform shapes • Observation II: WORLD generated samples can be the reference waveform • Criterion: envelope(WaveNet) - envelope(WORLD) > threshold \rightarrow collapsed speech detected

Envelope detection algorithm *3



Experimental Evaluations

- Corpus for VC
 - SPOKE task of
 - 4 source spea
 - 81 training utte
- 35 testing utte
- Corpus for Wave - Multi-speaker data from CM training data
 - Speaker-depe target speake output layers
- Collapsed spee - Statistical hyp
- Collapsed spee (between Wave
- maxPOW: the
- maxMCD: the
- ENV: the differ

Verif Collapsed type Type-I Type-II

 Collapsed spee - Speech qualit - Speaker simila

> Speech qua Speaker simi The same Maybe the so Maybe diffe

Different

Conclusions



f Voice Conversion Challenge 2018 akers and 4 target speakers terances of each speaker erances of each source speaker veNet vocoder r WaveNet: using "bdl" and "slt" AU-ARCTIC (1132 utts *2), and all from VCC2018 (81 utts *12). endent WaveNet: using each er's training data to update the of the multi-speaker WaveNet ech detection evaluations oothesis test (verification) ech detection criterions eNet and World samples) e differences of maximum powers e maximum MCDs rences of envelopes (proposed)		
fication Equal Error Rate		
maxPOW maxMCD	ENV	
10% 40% 40% 50%	2% 20%	
ech detection& suppression tests ty preference test arity test (4 measurements)		
	Pe-7 313	
same 35% 38% 0.6 erent 34% 33% 0.7	310 567 752 963	

• The proposed collapsed speech detection method achieves 20% equal error rate • The proposed method gets 77 % preference in the quality test while keeping the same speaker similarity as the original WaveNet vocoder



