

Phase reconstruction from amplitude spectrograms based on von-Mises-distribution deep neural network

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Hypothesis: phase reconstruction from amplitude spectrograms based on DNNs

Phase reconstruction

- Audio signal processing often processes amplitude spectrograms.
- Speech synthesis is shifting from vocoder params. to amplitudes.

Griffin-Lim phase reconstruction method^[1] provides **unnatural artifacts in speech**.

DNN-based phase reconstruction

- Can we train DNNs to predict the phase?
- Isotropic-Gaussian-distribution DNN (mean squared error training) is not suitable because phase is a periodic variable.**

Griffin-Lim method [1] —
A phase reconstruction method by iterating STFT and inverse STFT.

Our approach

- Propose a novel **DNN that has the von Mises distribution** which is a probability distribution for a periodic variable.
- Introduce **group delays** that has strong relationship to the amplitude of speech.

- DNN can predict group delay accurately more than phases, and
- our methods achieve better speech quality than the conventional Griffin-Lim method.

Proposed method: von-Mises-distribution DNN-based phase reconstruction

von Mises distribution and DNN-based phase reconstruction

von Mises (vM) distribution^[3]

- $P(y; \mu, \kappa) = e^{\kappa \cos(y-\mu)} / 2\pi I_0(\kappa)$
- μ : mean, κ : concentration
- $I_0(\cdot)$: Modified Bessel function

Negative log likelihood (μ : parameter)

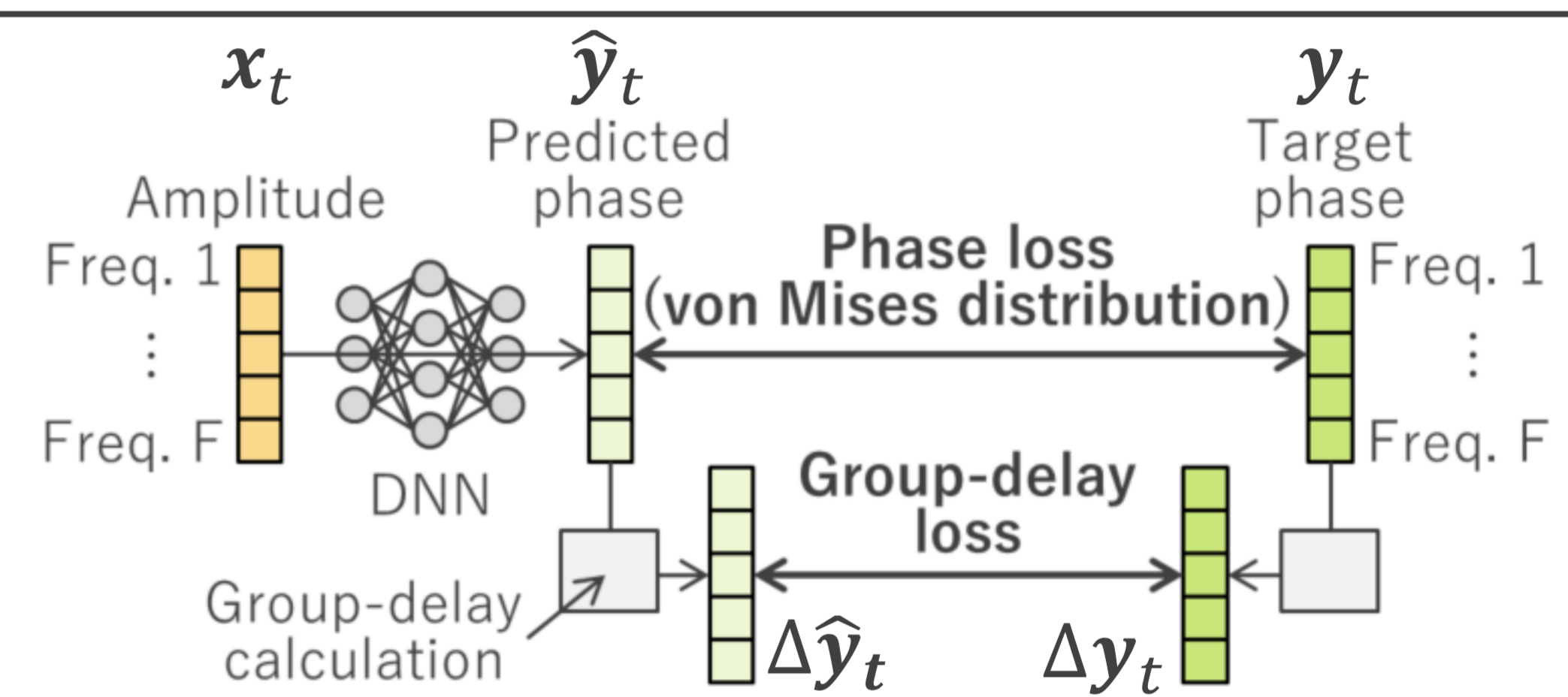
- $-\log P(y; \mu, \kappa) \propto -\cos(y - \mu)$

DNN-based phase reconstruction

- DNN that convert an amplitude x_t to phase y_t (t is the frame index.)

Loss for DNN training

- Phase loss** derived from vM dist.
- Group-delay loss**



1) Phase loss

- Maximum likelihood estimation of vM distribution.

$$L_{PH}(y_t, \hat{y}_t) = -\sum_f \cos(y_{t,f} - \hat{y}_{t,f})$$

$y_{t,f}$: phase at t -th frame and f -th freq. bin

2) Group-delay loss

- Approximate group delay with 1st-order difference.

$$L_{GD}(y_t, \hat{y}_t) = -\sum_f \cos(\Delta y_{t,f} - \Delta \hat{y}_{t,f})$$

$\Delta y_{t,f} = -(y_{t,f+1} - y_{t,f})$: group delay at t -th frame and f -th freq. bin

*Group delay and phase of AR models have strong relationship [3].

Evaluation: prediction accuracy, effects to speech quality, and improvements by group delay

Contents	Value/Settings
Training / test data	JSUT speech corpus ^[4] 5000 / 300 utts.
Sampling freq.	16 kHz
Frame shift, FFT taps	80 samples (5 ms), 512 samples
DNN input	Log amplitudes at current ± 2 frames
DNN output	Phase (3 types: 0-2kHz, 0-4kHz, 0-8kHz)
DNN architecture	Feed-Forward w/ gated activation units
Post-process	Phase refinement by the Griffin-Lim method

2) Speech quality

- Evaluation methods
- Preference AB tests on speech quality

Preference AB tests by 30 listeners on crowdsourcing

Method A	Scores	p -value	Method B
Griffin-Lim	0.497 vs. 0.503	0.871	PH (2 kHz)
Griffin-Lim	0.280 vs. 0.720	$< 10^{-9}$	PH (4 kHz)
Griffin-Lim	0.277 vs. 0.723	$< 10^{-9}$	PH (8 kHz)
Griffin-Lim	0.453 vs. 0.547	0.022	PH+GD (2 kHz)
Griffin-Lim	0.233 vs. 0.767	$< 10^{-9}$	PH+GD (4 kHz)
Griffin-Lim	0.247 vs. 0.753	$< 10^{-9}$	PH+GD (8 kHz)
Griffin-Lim	0.447 vs. 0.553	0.009	GD (2 kHz)
Griffin-Lim	0.463 vs. 0.537	0.073	GD (4 kHz)
Griffin-Lim	0.490 vs. 0.510	0.619	GD (8 kHz)

Results

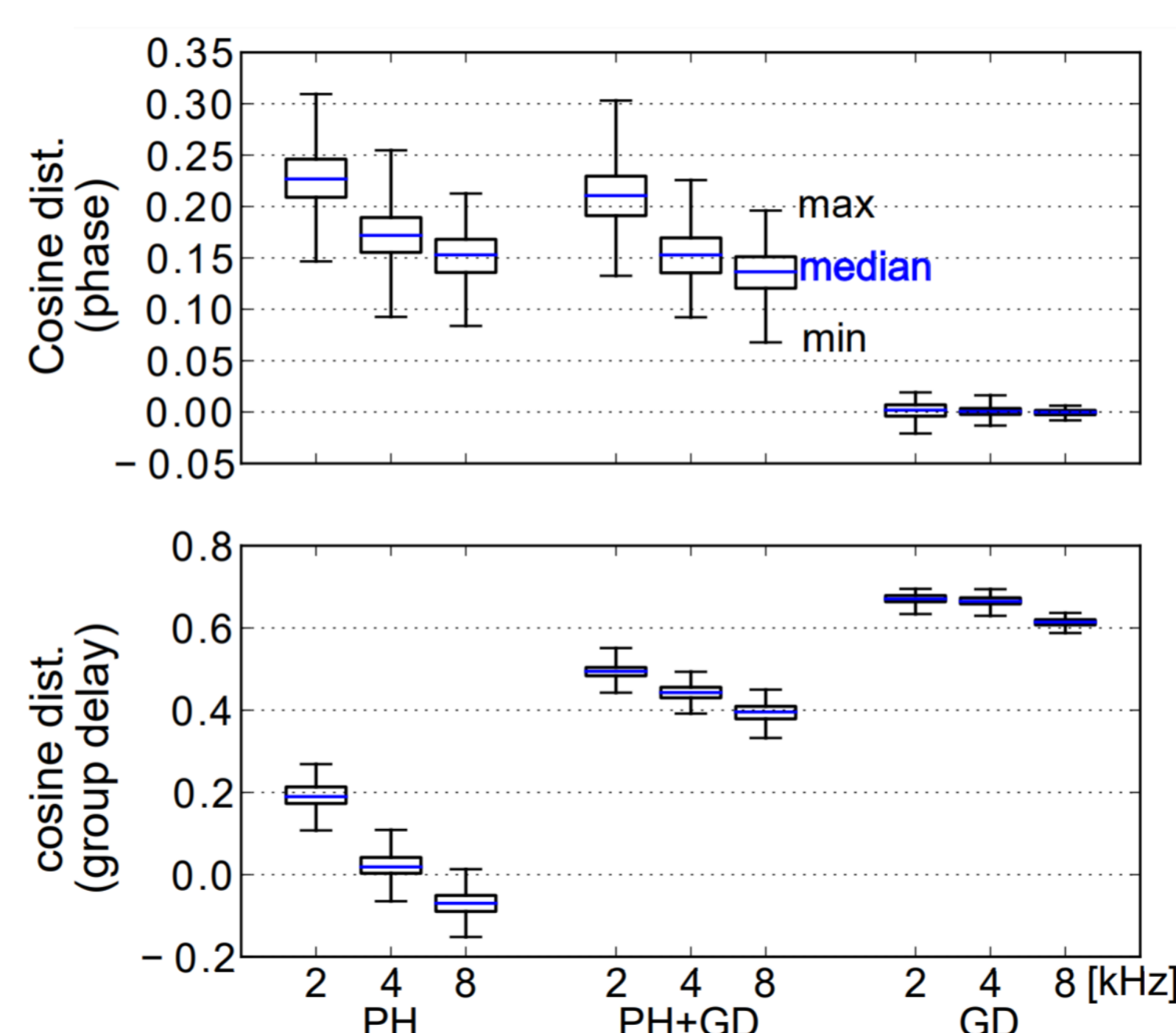
- **Better than Griffin-Lim**
- **Multi-task learning achieves better in all settings**

1) Prediction accuracy

- Compared systems
- PH: Phase loss only
- GD: Group-delay loss only
- PH+GD: Multi-task learning

- Evaluation method
- Cosine distance

- Results
- **Group delay is estimated accurately more than phase.**



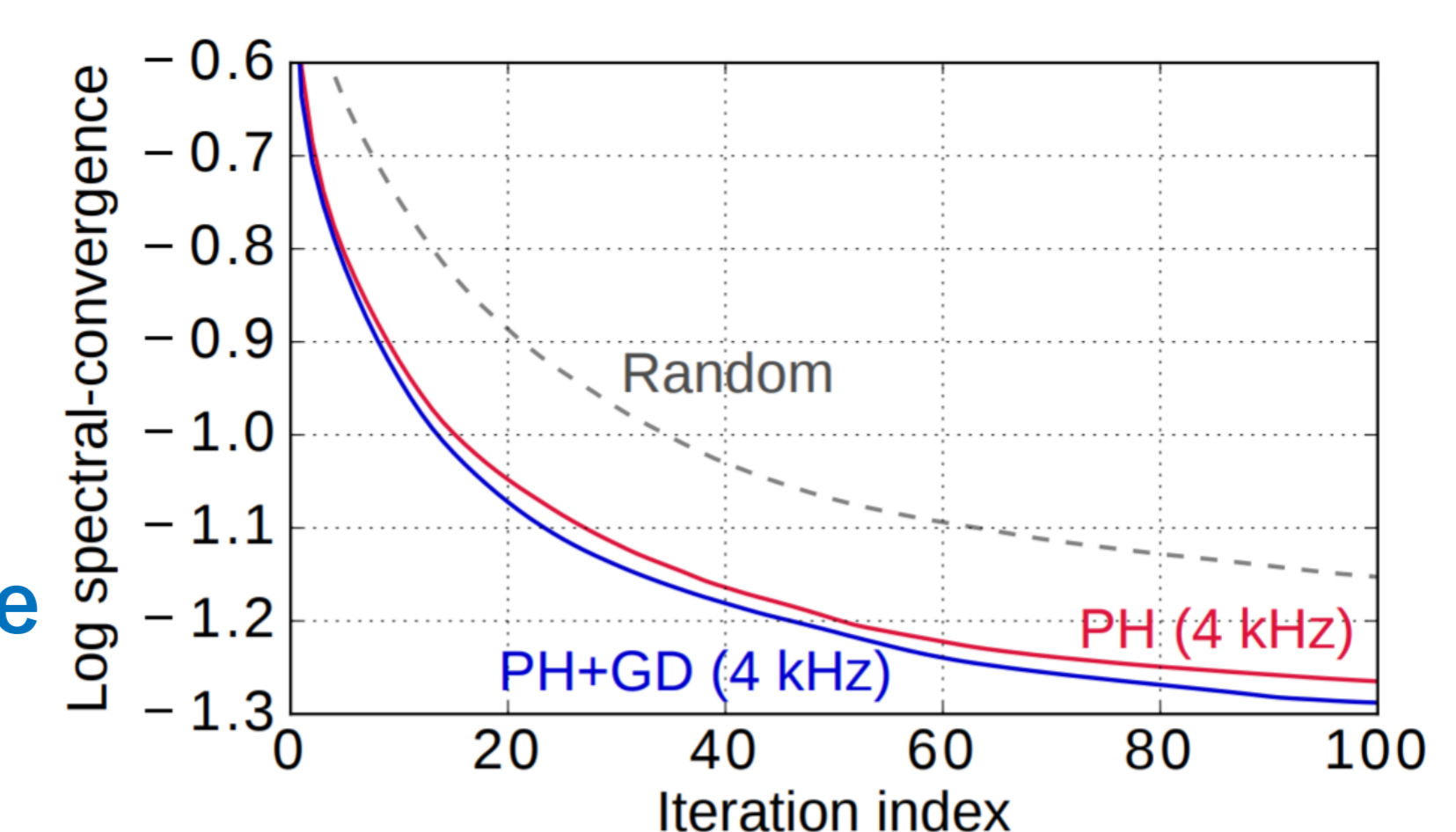
3) Improvements by group-delay loss

- Speech quality
- Better than phase loss

- Convergence in post-process
- Spectral convergence^[5]: reconstruction performance through STFT & inverse STFT

- **Group-delay loss provides phases that are closest to the perfect reconstruction**

Method A	Scores	p -value	Method B
PH (2 kHz)	0.487 vs. 0.513	0.514	PH+GD (2 kHz)
PH (4 kHz)	0.486 vs. 0.514	0.500	PH+GD (4 kHz)
PH (8 kHz)	0.545 vs. 0.455	0.031	PH+GD (8 kHz)



Reference

- [1] Griffin et al., IEEE Trans., 1984. [2] Bishop, Springer, 2006. [3] Itakura et al., Proc. ICASSP, 1987. [4] Sonobe et al., arXiv, 2007. [5] Sturm et al., Intl. Conf. on Digital Audio Effects, 2011.

