



UR Signals and Systems

Perceptual Long-Term Harmonic plus Noise Modeling for Speech Data Compression

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Long-Term Harmonic plus Noise Model (LT-HNM)

□ Perceptual LT-HNM for Data reduction



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Long-Term Harmonic plus Noise Model (LT-HNM)



Voiced Unvoiced Voiced ST-HNM LT-model LT-model LT-model

Harmonic plus Noise Models:

- harmonic band: multiples of the fundamental frequency (F0)
- noise band: peak picking frequencies

The two bands are separated by a voicing cutoff frequency (VCO) □Speech signal is segmented into voiced ($F_0 \neq 0$) and unvoiced ($F_0=0$) LT-sections

 \Box A Long-Term model is applied to the ST-HNM parameters (F₀, VCO frequencies and spectral amplitudes) for each LT-section.



Long-Term Harmonic plus Noise Model (LT-HNM)

LT-model: Discret Cosine Model (DCM)

$$\hat{X}(n) = \sum_{p=0}^{P} c_p \cos(\frac{p \pi n}{N})$$

□ Applying a DCM to the time trajectory of the ST-HNM parameters (F0, FV and amplitudes) in a LT time section.

□ Exploits the correlation between successive ST-parameters

□ Optimization of the model order P



For amplitudes, we apply a DCM twice:

□ first along the frequency axis to model the spectral amplitudes in a ST-frame (1D-DCM)

□ second along the time axis to model the time trajectory of the 1D-DCM coefficients in a LT-section (2D-DCM).



Long-Term Harmonic plus Noise Model (LT-HNM)

□ Perceptual LT-HNM for Data reduction



Auditory Masking



□ A masking threshold is computed for each ST-frame

□ The spectrum amplitudes (harmonic + noise) are compared to the masking threshold
→ Only amplitudes above the mask are selected as audible

ST-HNM Data Reduction

p-ST-HNM

Only audible amplitudes are considered in the ST-HNM, inaudible ones are discarded from the model

 \rightarrow The data size of the model parameters is considerably reduced: up to 50% in a ST-frame

 \rightarrow Reduction of the data-rate with the equivalent perceptual quality

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p-LT-HNM

The LT-modeling is applied to the parameters of the p-ST-HNM

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Double data-compression: auditory masking + LT-modeling



Long-Term Harmonic plus Noise Model (LT-HNM)

□ Perceptual LT-HNM for Data reduction



Overall Data Reduction



□ 72% of database frequencies are noise frequencies, while 27.8% are harmonics

□ Total frequency components compression ≈ 20%: noise band compression ≈ 23.3%, harmonic band compression ≈ 11.3%

■ 84.1% of achieved compression is due to the noise band, while the harmonic frequencies contribute only by 15.% to the total compression

Data Reduction in V/UV ST-frames



□ Data compression rate is up to 50% in a ST-frame

Compression rate is higher for unvoiced ST-frame (entirely composed of noise frequencies)

□ Higher contribution of noise band to the total compression



Reduction of the LT-HNM coefficients rate



The 1D-DCM order is considerably reduced when applying the auditory masking

The data size to be LT-modeled is reduced

Reduction of the LT-HNM data size



Parameters-Rate Gain





Listening Quality (PESQ and MOS Scores)



 Mean PESQ scores: 40 speech samples from TIMIT (English) and 20 samples from HINT (French)
MOS Subjective listening test applied to HINT samples (12 French speakers participants)

→ No auditory distortion when applying the p-ST-HNM (PESQ and MOS)

→ No significant auditory distortion when applying the p-LT-HNM (PESQ and MOS)



Conclusion and Perspectives

Conclusion:

Two stages of compression:

□ Perceptual based compression: 18% (ST-HNM → p-ST-HNM)
□ Perceptual LT modeling: 89% (p-ST-HNM → p-LT-HNM)

→ Total compression: 90% (ST-HNM → p-LT-HNM)

Perceptual Quality

The perceptual HNM and generic HNM provide equivalent quality scores

Perspectives:

A two stage vector quantization is currently being applied to the perceptual LT-HNM parameters and will to design a low bit-rate speech codec.





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