

A template-based approach for speech synthesis intonation generation using LSTMs

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Gustav



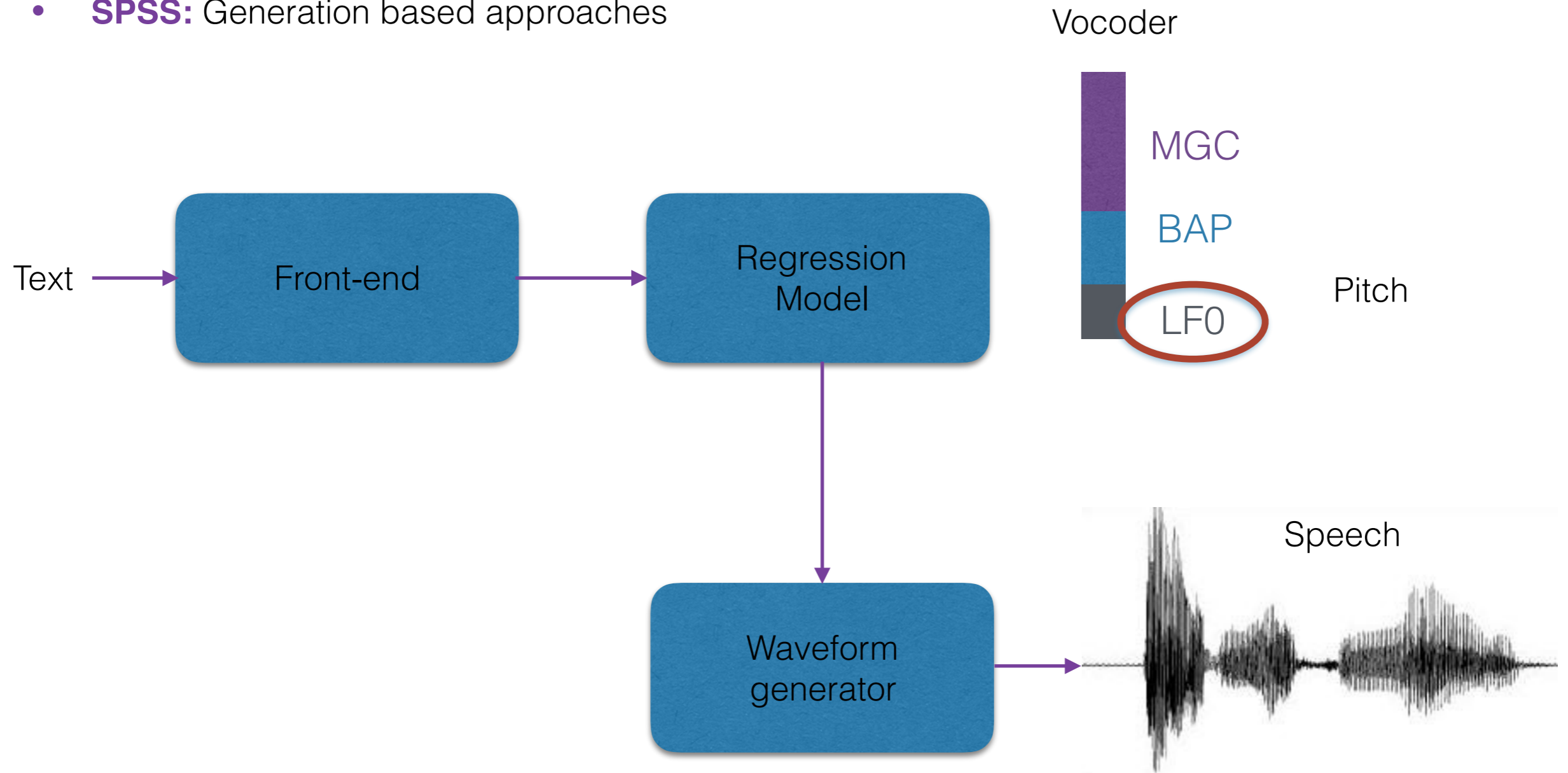
Zhizheng



Simon

Introduction: Statistical speech synthesis

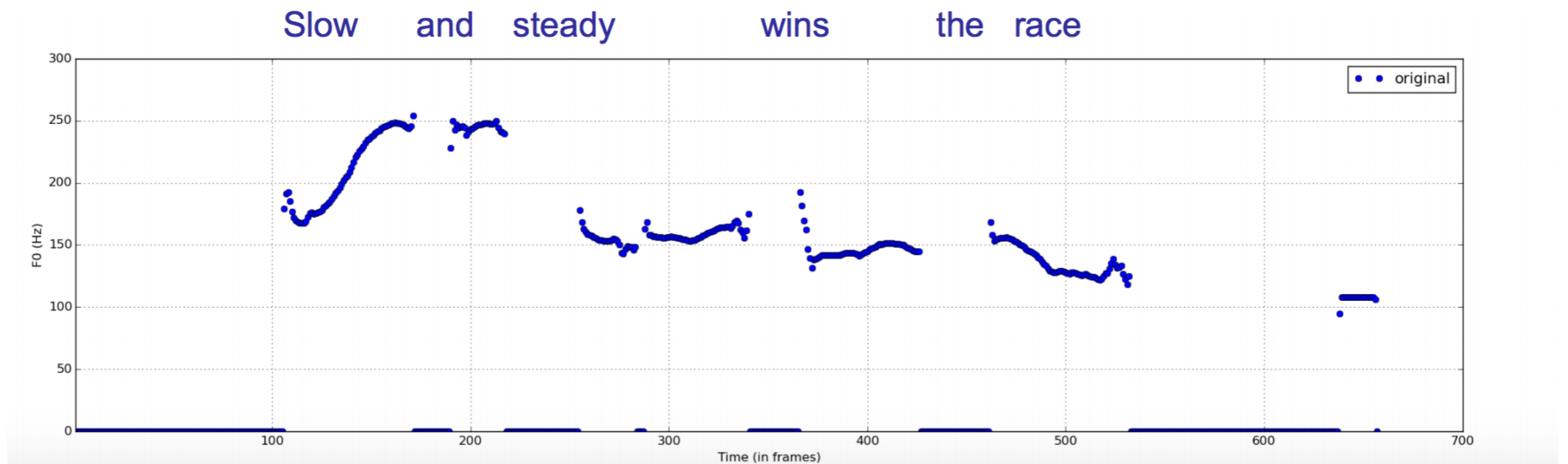
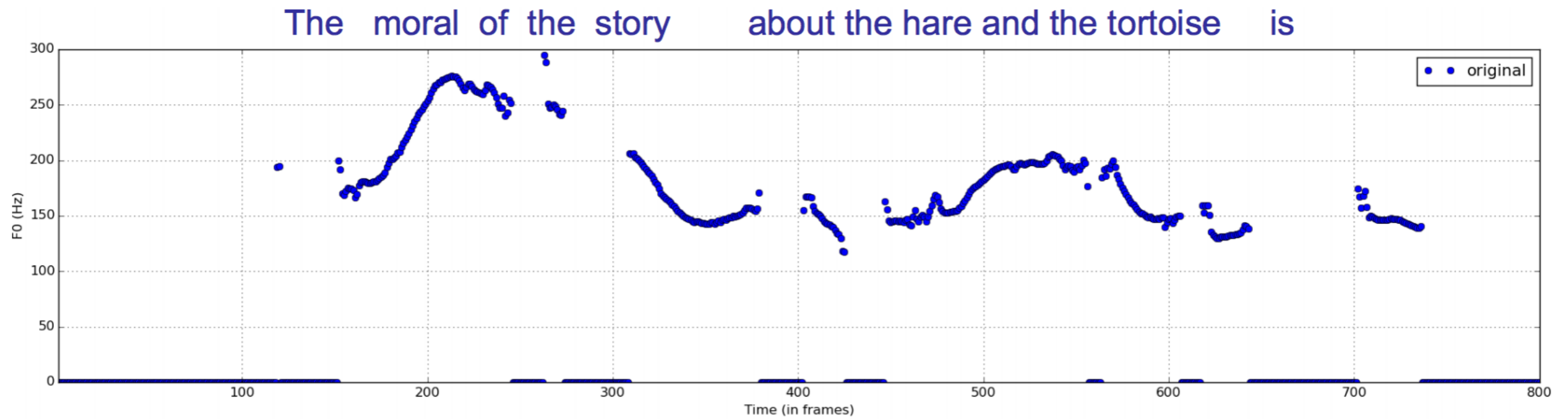
- **SPSS:** Generation based approaches



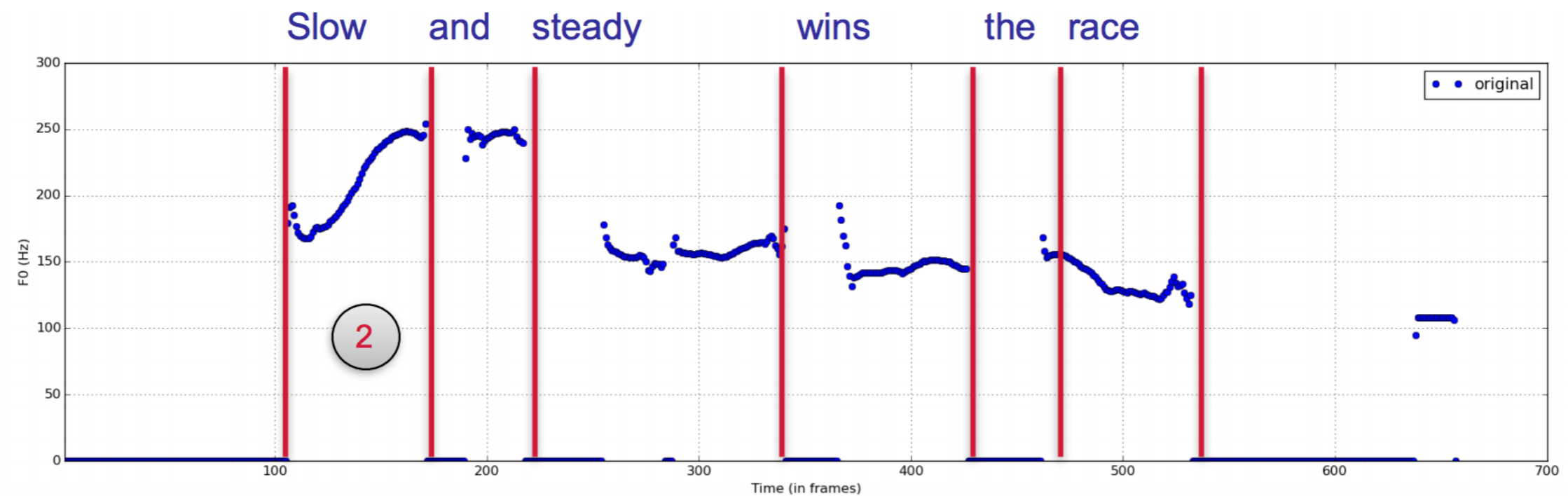
Why template-based approach?

- Lack of convincing intonation makes current parametric systems sound dull and lifeless.
- Typically, these systems predict F0 frame-by-frame using regression models.
- This approach leads to overly-smooth pitch contours and fail to construct an appropriate prosodic structure.
- Templates retain the dynamic range of F0 within the segment.
- We propose a classification-based approach to automatic F0 generation.

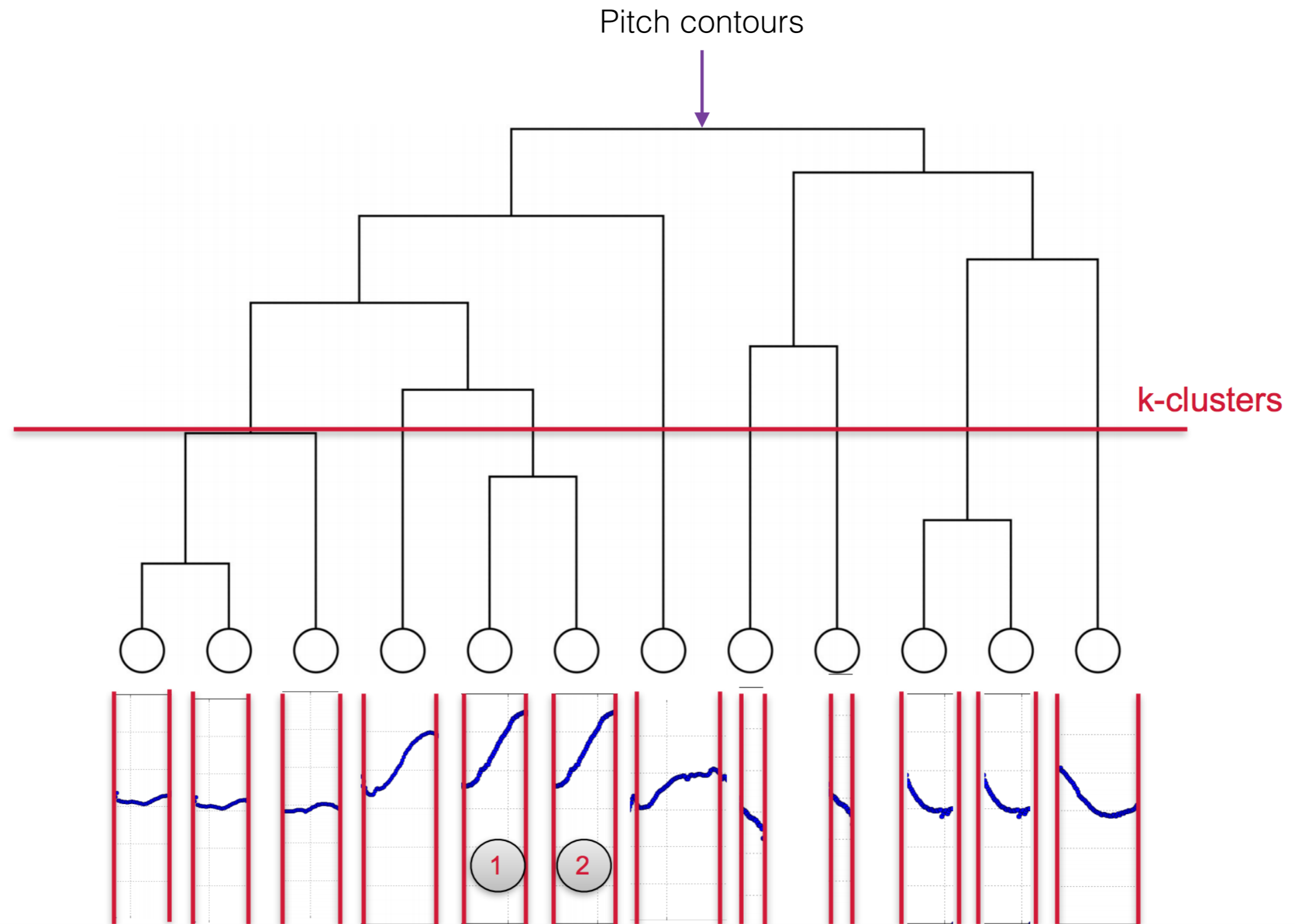
Pitch contour



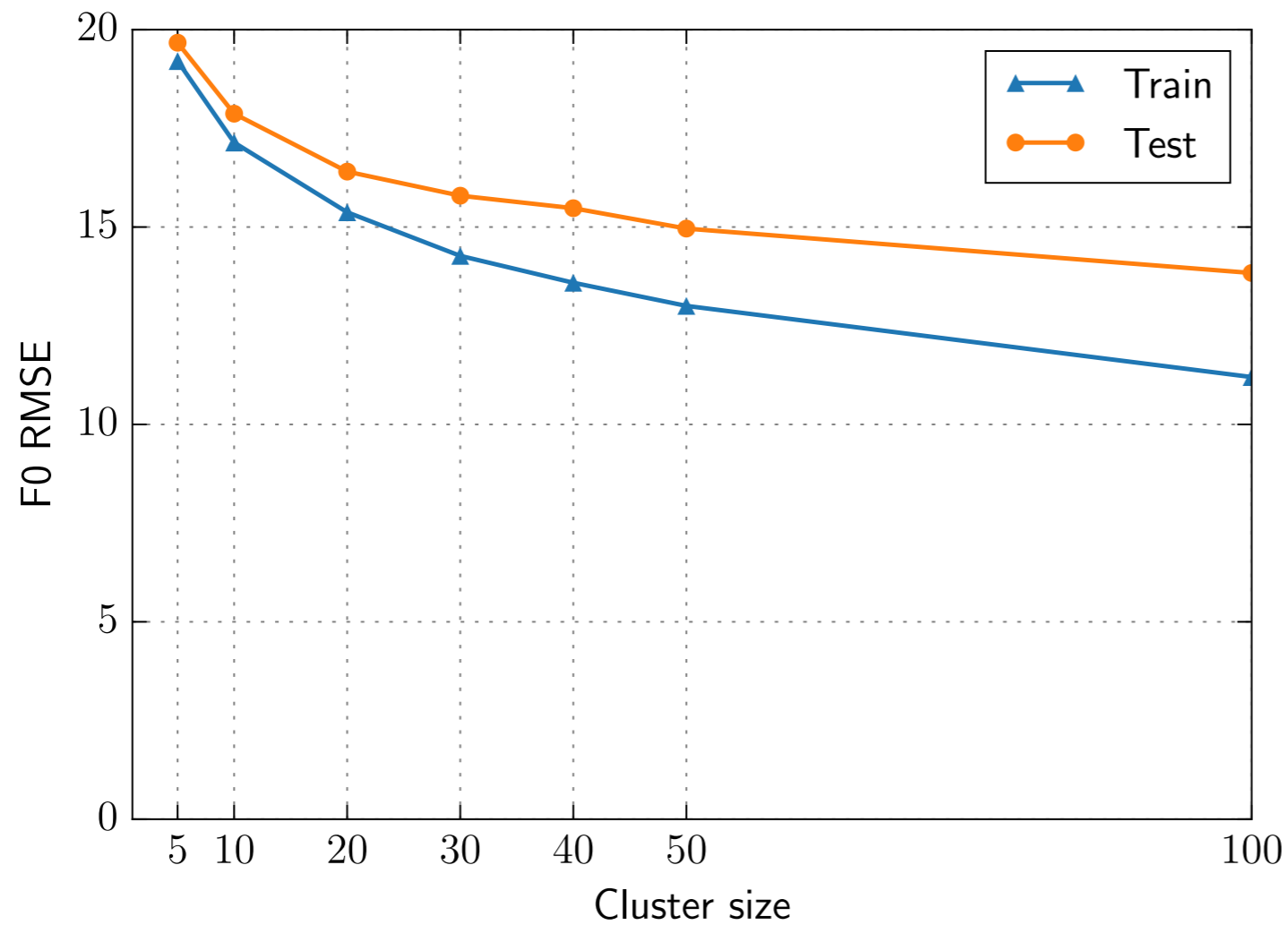
Pitch contour segmentation



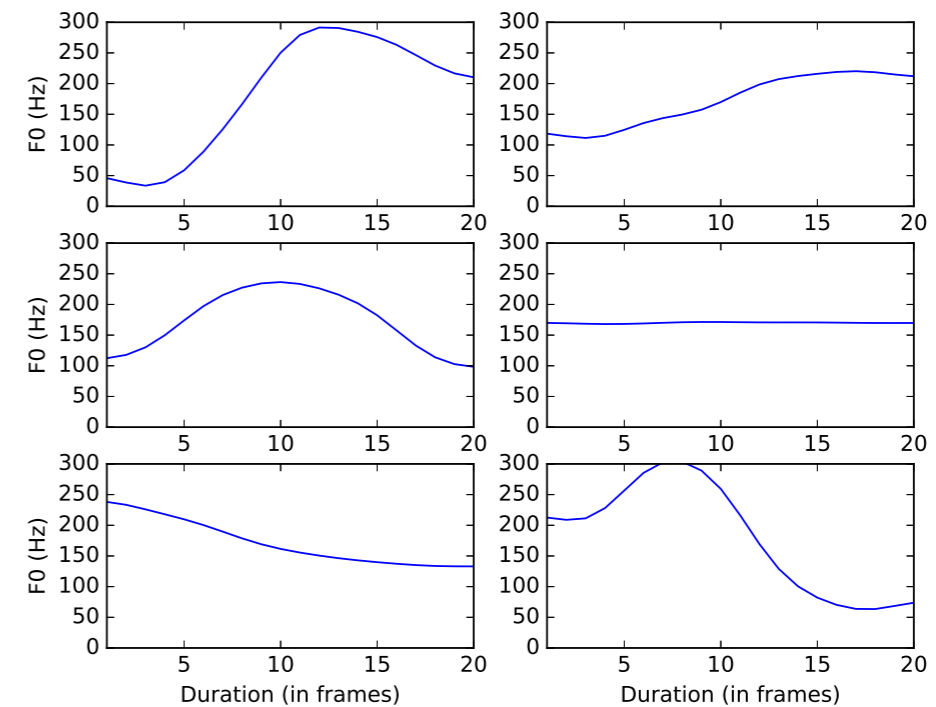
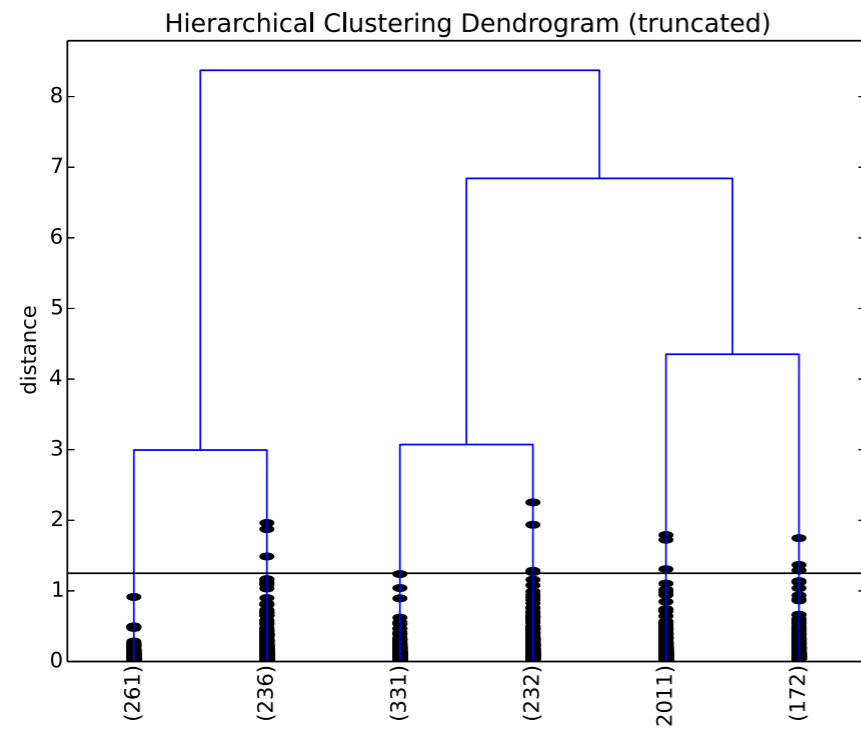
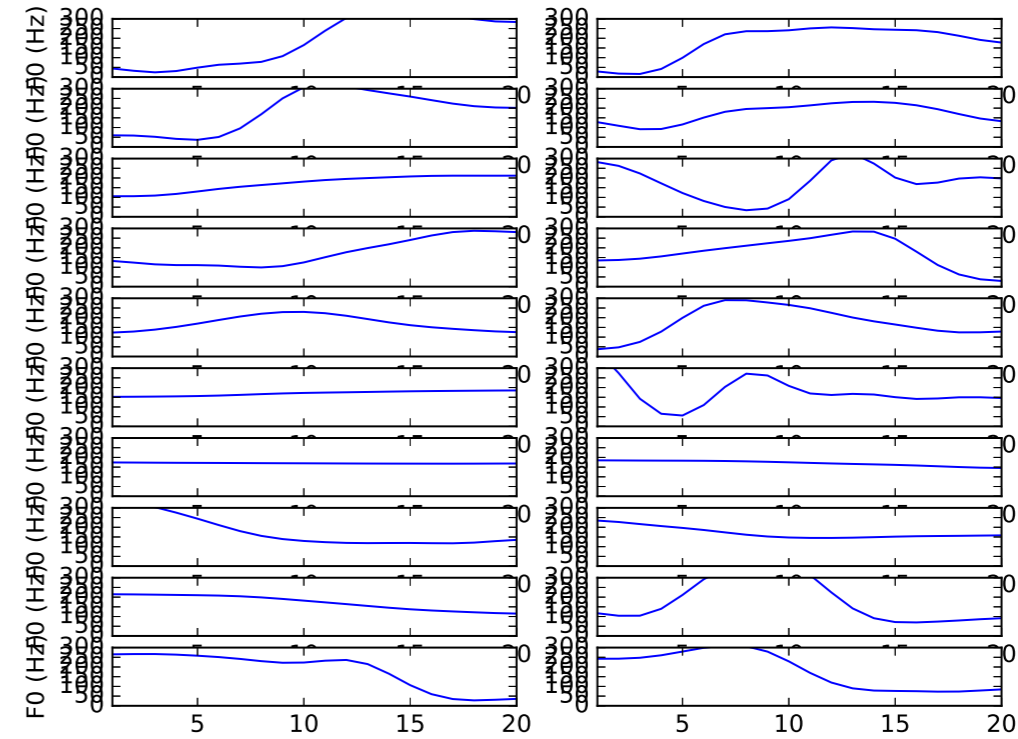
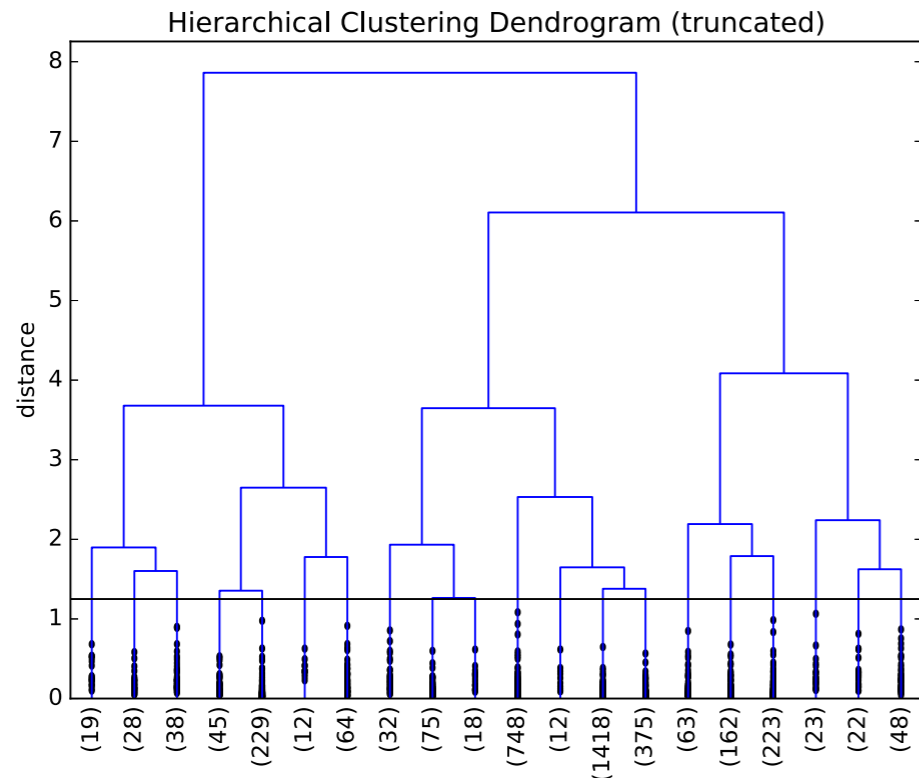
Hierarchical clustering



How to determine number of clusters?



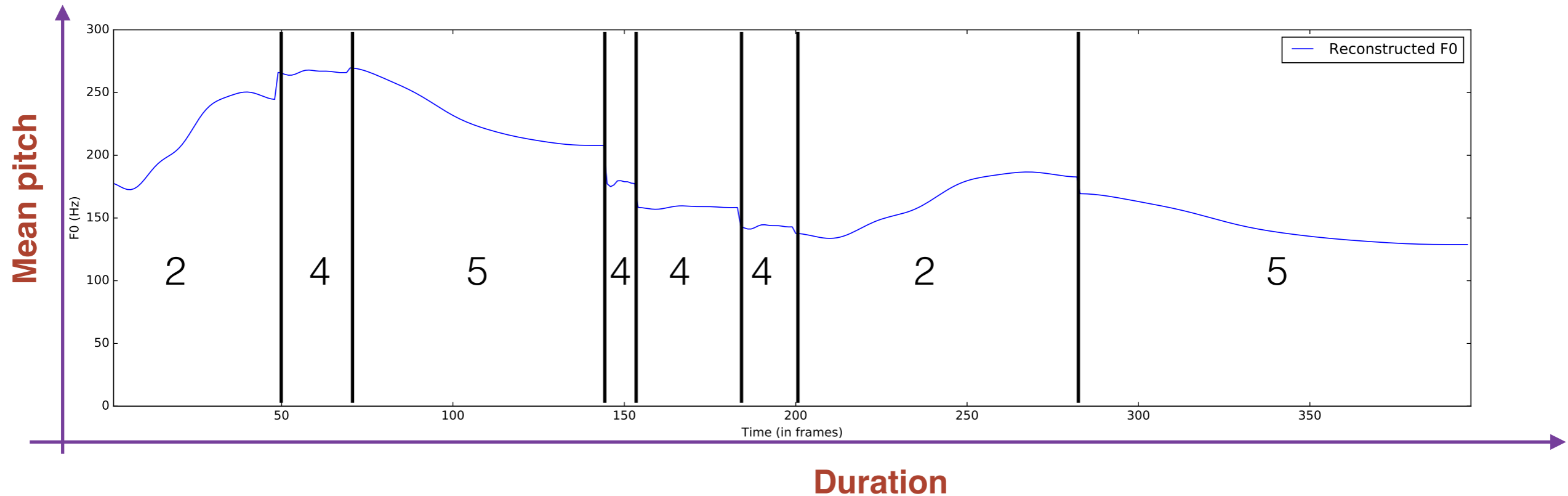
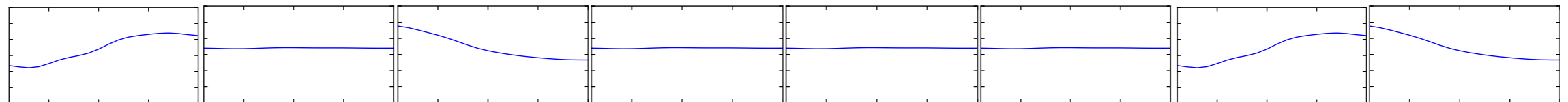
A set of templates (clusters)



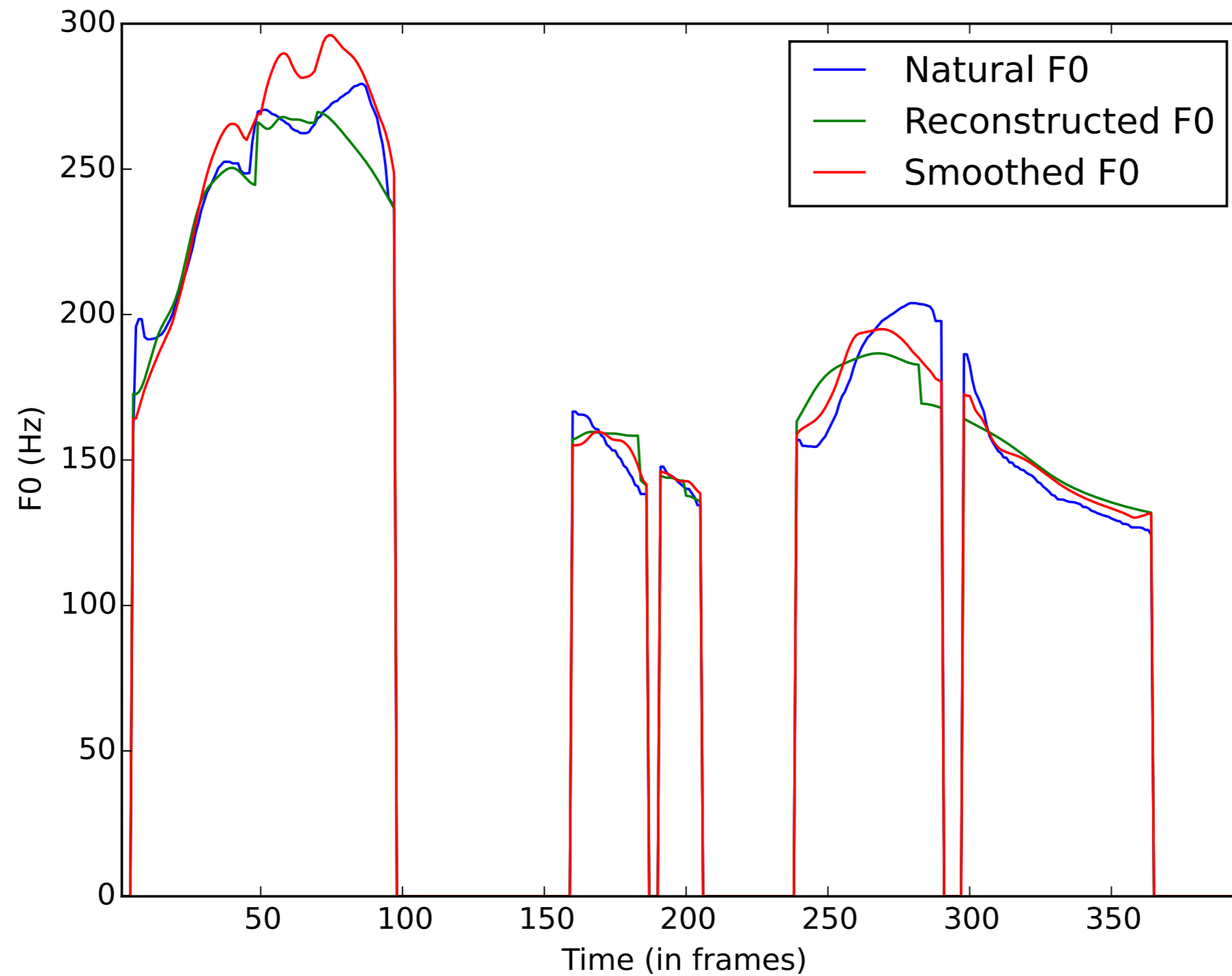
Intonation reconstruction from templates

Goldilocks and the three bears

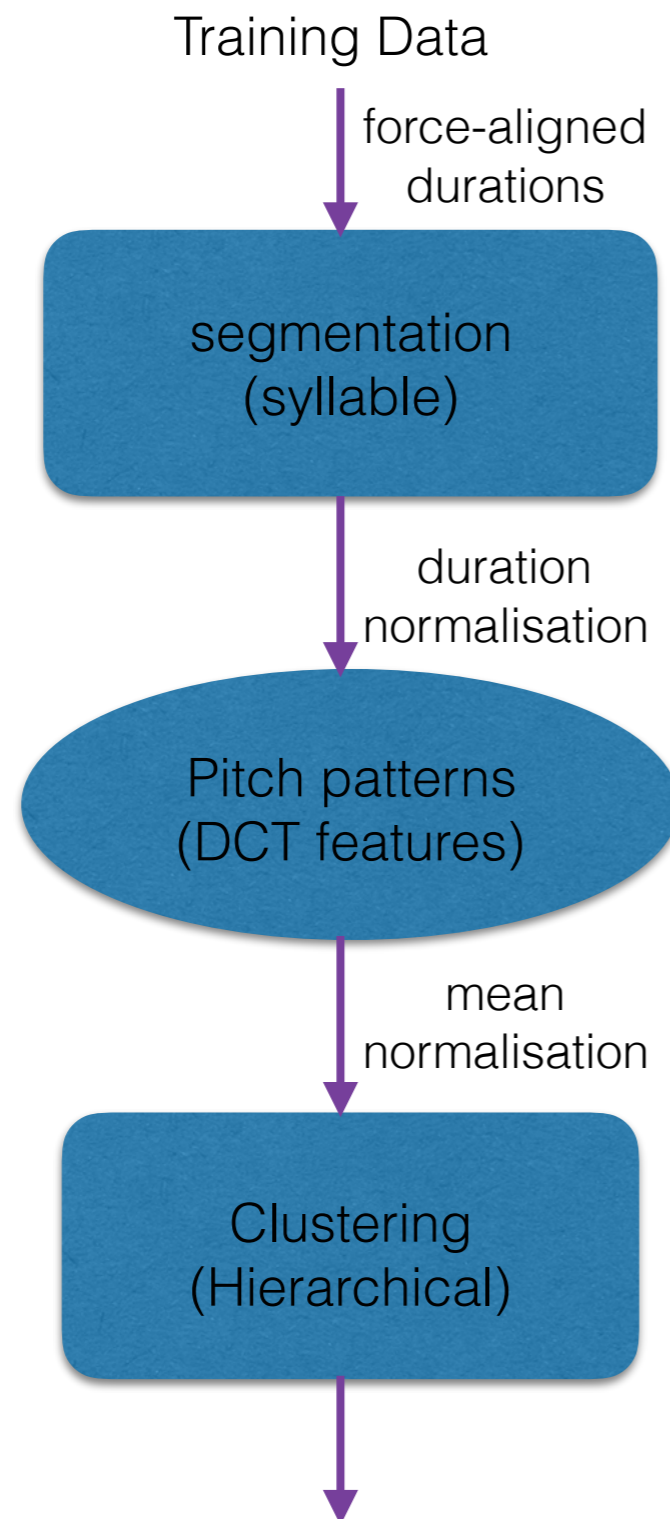
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2 4 5 4 4 4 2 5



Intonation reconstruction



Hierarchical clustering - Recap



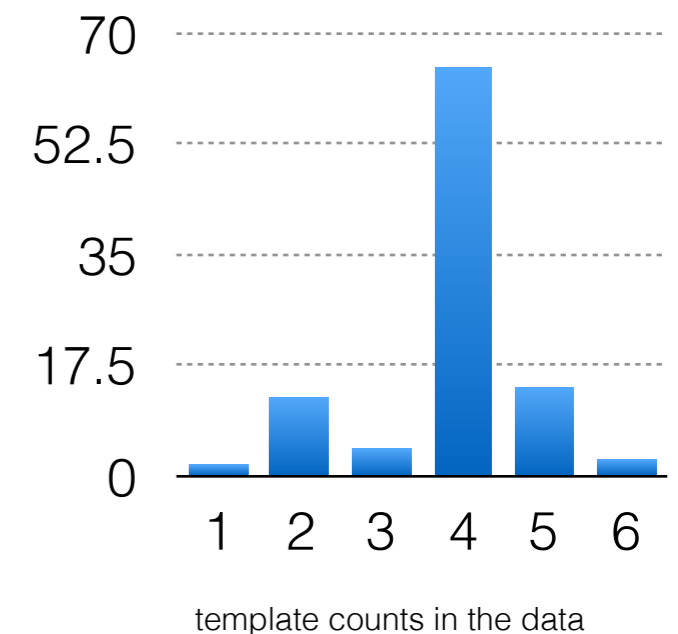
- Interpolate the F0 contour of each utterance and segment into syllables
- Apply DCT based decomposition: c_0 representing the mean over syllable, $\mathbf{c} = [c_1, \dots, c_{N-1}]$, representing the shape of the contour
- Perform top-to-bottom hierarchical clustering over the patterns (\mathbf{c}).

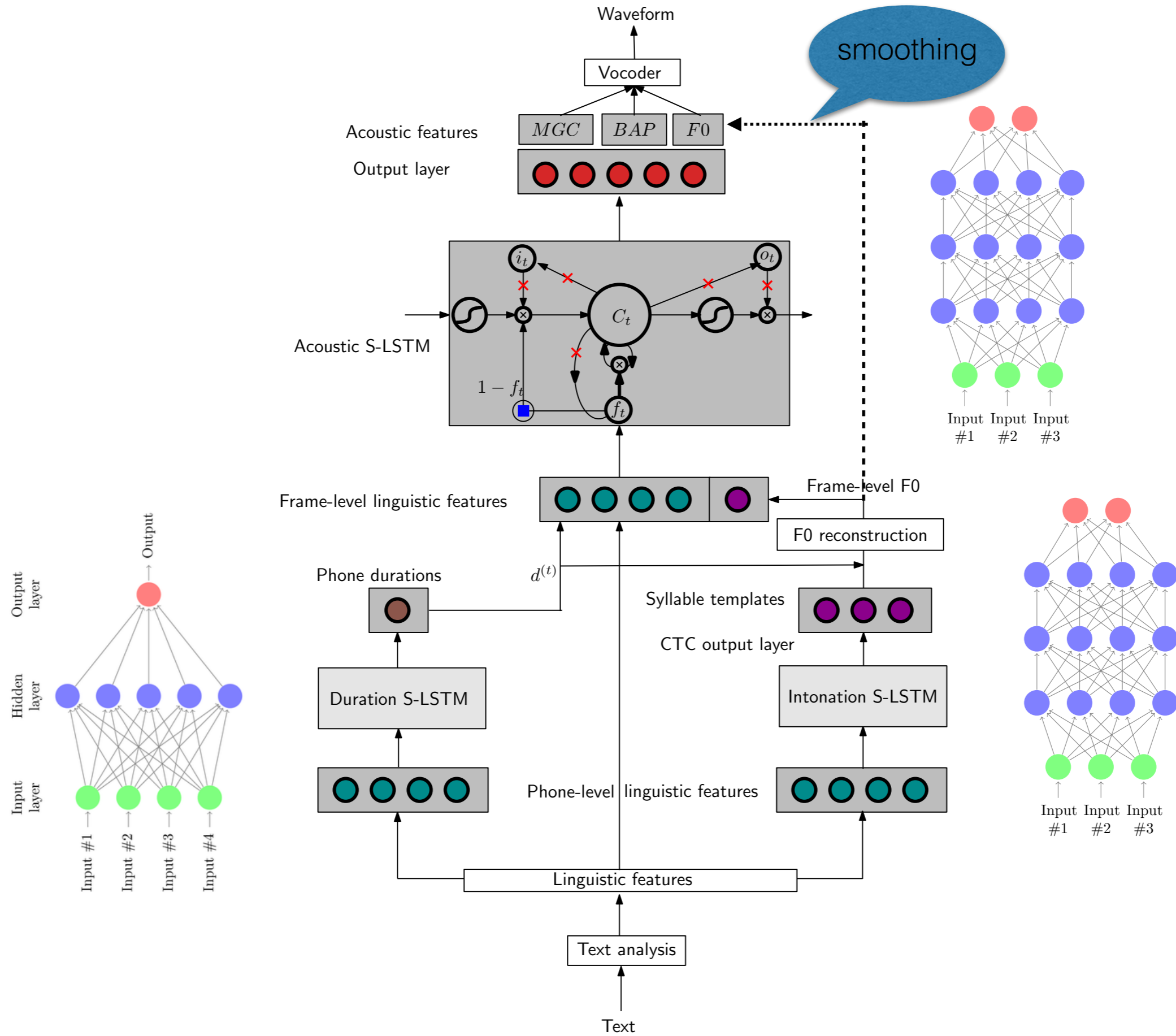
Proposed approaches



Neural Network classifiers:

- A hierarchical deep neural network classifier (HC).
 - ▶ The first DNN chooses between flat and non-flat template.
 - ▶ The second DNN chooses among rest of the non-flat templates.
- A simplified LSTM with a CTC output layer (CTC).
 - ▶ Connectionist temporal classification coupled with S-LSTM to predict the sequence of templates given sequence of phonemes.





Results: systems

- Baseline system
 - ▶ MSE - A frame-wise regression baseline predicting F0 using LSTMs.
- Proposed systems
 - ▶ HC - A hierarchical deep neural network classifier
 - ▶ CTC - A simplified LSTM coupled with CTC output layer
 - ▶ Oracle - A oracle system using templates derived from natural F0 contour but with *predicted F0 mean and duration*

Objective evaluation

- Classification measures
 - ▶ Accuracy - percentage of templates correctly classified
 - ▶ F1 score - is a measure of test's accuracy (precision and recall)

Model	Accuracy	F1 score
HC	61.1%	0.590
CTC	63.8%	0.593

Objective evaluation

- F0 prediction measures
 - ▶ RMSE - Root mean square error
 - ▶ CORR - Pearson correlation

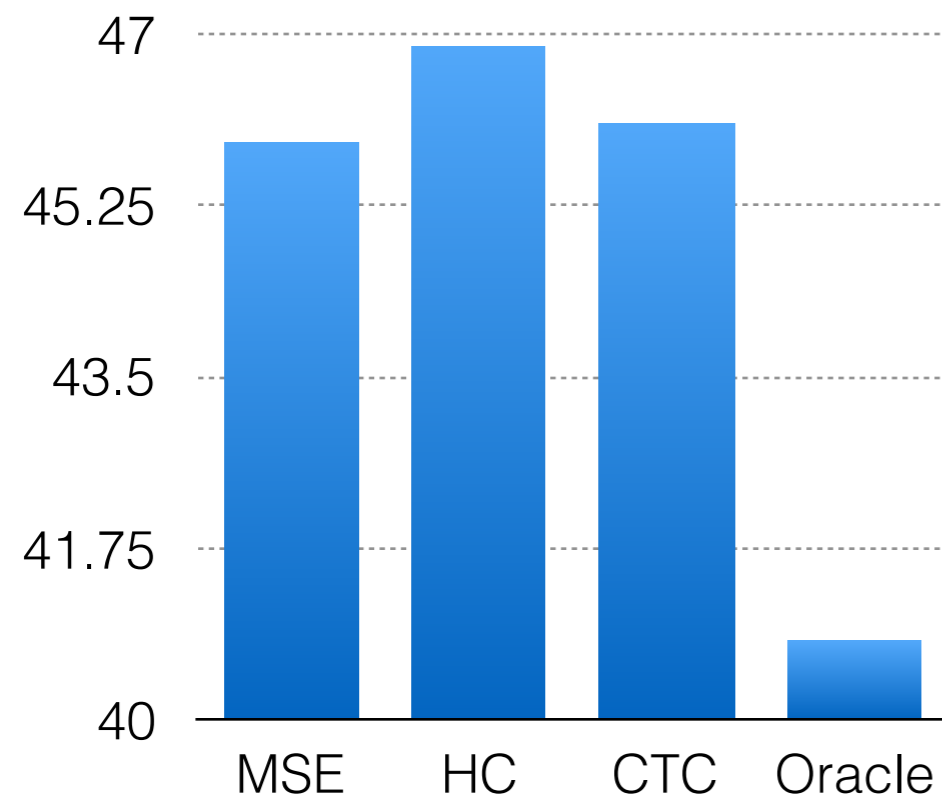


Fig: RMSE of predicted F0

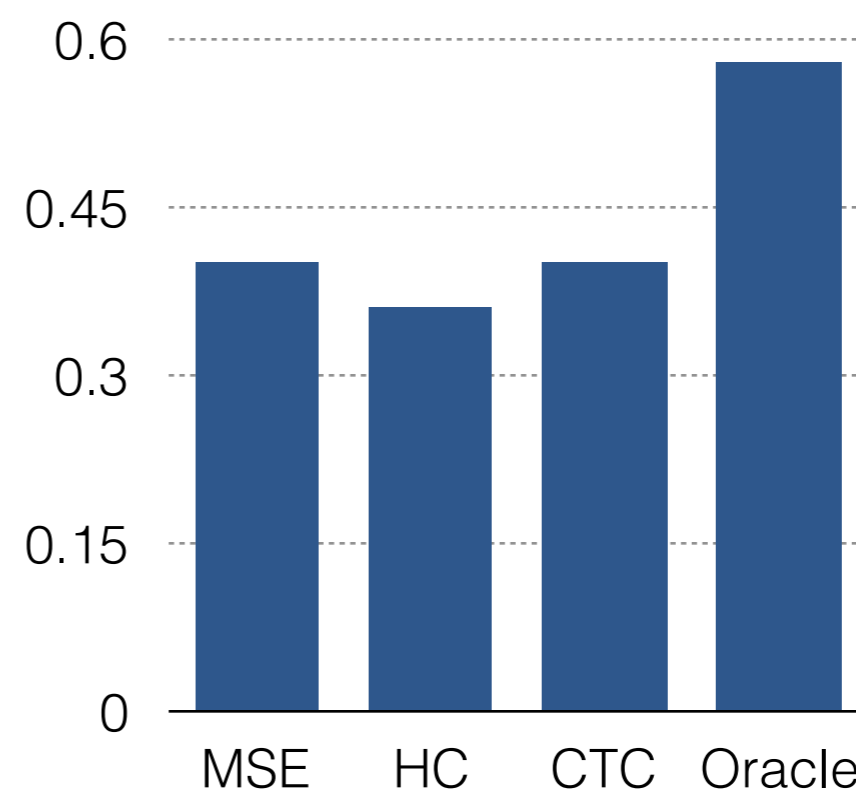


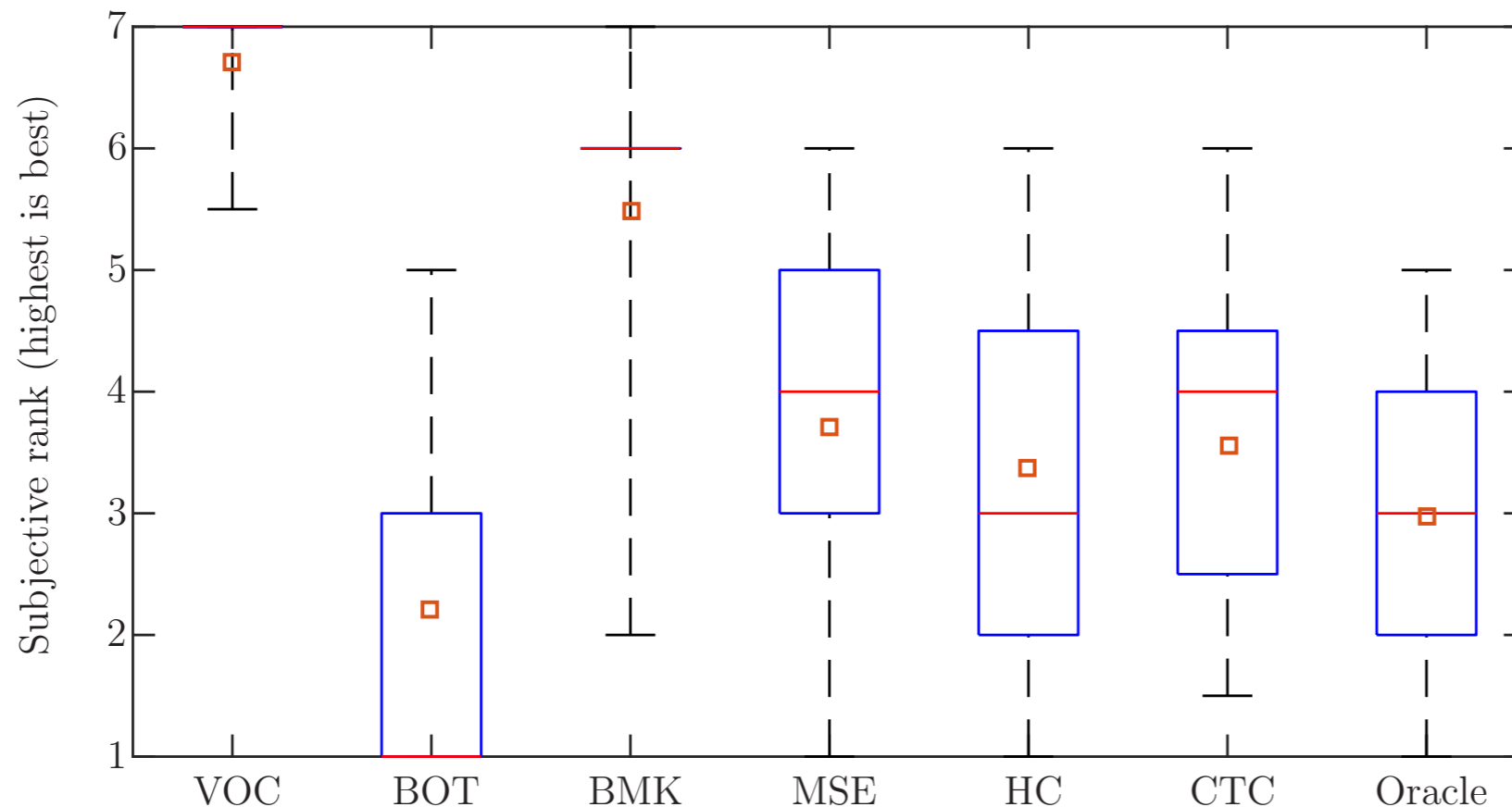
Fig: Correlation of predicted F0

- Oracle templates + Oracle F0 mean - 0.89 (corr.)

Subjective evaluation

- Reference systems
 - ▶ MSE - A frame-wise regression baseline predicting F0 using LSTMs.
 - ▶ BOT - A bottom line using piecewise-constant F0 per syllable (the mean natural F0)
 - ▶ BMK - A benchmark system using force-aligned durations and natural F0 contours
 - ▶ VOC - A top line of vocoded speech (STRAIGHT in this work)

Subjective evaluation: MUSHRA



- 20 listeners
- 20 out of 32 test stimuli

Fig: Box plot of aggregate ranks from listening test. Red lines are medians, orange squares means.

Summary and conclusions

- A classification approach to intonation prediction with syllable F0 templates
- Proposed approach matches the performance of conventional approach
- Has potential to exceed it once the issues with oracle template system are overcome
- Future work:
 - ▶ Better smoothing techniques and word-level templates
 - ▶ Use the prediction probabilities as features for frame-level regression approaches

Code

- Code for templates and clustering
 - ▶ https://github.com/ronanki/Hybrid_prosody_model
- Code for training neural networks
 - ▶ <https://github.com/CSTR-Edinburgh/merlin>

