Automatic diagnosis and feedback for lexical stress errors in non-native speech: Towards a CAPT system for French learners of German

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Lexical stress



Some syllable(s) in a word more accentuated/prominent¹

um·FAHR·en vs. UM·fahr·en to run over to drive around

- German: variable stress placement, contrastive stress¹
- ► French: no word-level stress, final syllable lengthening²

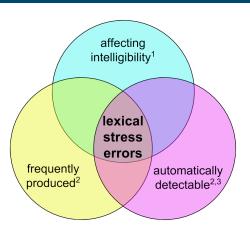
Goal: Computer-Assisted Pronunciation Training (CAPT) for lexical stress errors for French learners of German

¹A. Cutler. "Lexical Stress". In: *The Handbook of Speech Perception*. Ed. by D. B. Pisoni and R. E. Remez. 2005, pp. 264–289.

²M.-C. Michaux and J. Caspers. "The production of Dutch word stress by Francophone learners". In: *Proc. of the Prosody-Discourse Interface Conference (IDP)*. 2013, pp. 89–94.

Lexical stress errors in CAPT





¹U. Hirschfeld. *Untersuchungen zur phonetischen Verständlichkeit Deutschlernender.* Vol. 57. Forum Phoneticum. 1994

²A. Bonneau and V. Colotte. "Automatic Feedback for L2 Prosody Learning". In: *Speech and Language Technologies*. Ed. by I. lpsic. InTech, 2011

³Y.-J. Kim and M. C. Beutnagel. "Automatic assessment of American English lexical stress using machine learning algorithms". In: *SLaTE*. 2011, pp. 93–96

Outline



Lexical stress errors by French learners of German

Annotation of a learner speech corpus Inter-annotator agreement Frequency & distribution of errors

Diagnosis methods

Word prosody analysis
Diagnosis by comparison
Diagnosis by classification

Feedback methods

de-stress: A prototype CAPT tool

Conclusion

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Lexical stress errors in learner speech



- How reliably can human annotators identify errors in learner utterances?
- How frequently are errors actually produced by French learners of German?



Data: IFCASL corpus of French-German speech¹

- German utterances by French and German speakers
 - Adults (>18) and children (15-16)
 - Levels A2, B1, B2, C1 (children all A2/B1)
- Word- and phone-level segmentations (syllable level added automatically)
- Selected 12 word types (bisyllabic, initial stress)

Dataset for annotation:

668 German word utterances by ∼55 French speakers

¹C. Fauth et al. "Designing a Bilingual Speech Corpus for French and German Language Learners: a Two-Step Process". In: *9th Language Resources and Evaluation Conference (LREC)*. Reykjavik, Iceland, 2014, pp. 1477–1482.



15 Annotators, varying by:

- ► Native language (L1):
 - 12 German
 - 2 English (US)
 - 1 Hebrew
- Phonetics/phonology expertise:
 - 2 Experts
 - 10 Intermediates
 - 3 Novices



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Task: label utterances of 3 word types



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Praat annotation tool:



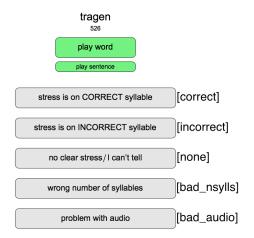


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Praat annotation tool:



Inter-annotator agreement



How reliably can human annotators identify errors in learner utterances?

- Agreement calculated for each pair of annotators who labeled the same utterances
- Quantified by:
 - Percentage agreement: N agreed/N both annotated
 - Cohen's Kappa¹ (κ): accounts for chance agreement

¹J. Cohen. "A Coefficient of Agreement for Nominal Scales". In: *Educational and Psychological Measurement* 20.1 (Apr. 1960), pp. 37–46.

Inter-annotator agreement



Overall pairwise agreement between annotators

	% Agreement	Cohen's κ
Mean	54.92%	0.23
Maximum	83.93%	0.61
Median	55.36%	0.26
Minimum	23.21%	-0.01

¹J. R. Landis and G. G. Koch. "The measurement of observer agreement for categorical data." In: *Biometrics* 33.1 (1977), pp. 159–174.

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- Rather low agreement ("fair" mean κ)
- Large variability among annotators, not explained by L1/expertise
- Single gold-standard label selected for each utterance

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Error distribution

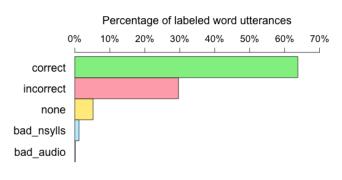


How frequently are errors actually produced by French learners of German?

Error distribution



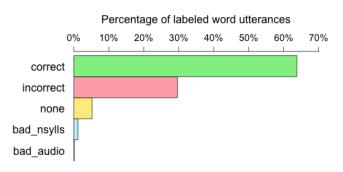
How frequently are errors actually produced by French learners of German?



Error distribution



How frequently are errors actually produced by French learners of German?



- Large variability across word types
- Beginners made more errors (vs. advanced)
- Children made more errors (vs. adult beginners)

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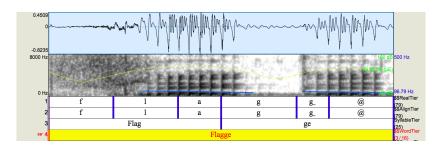
Conclusion

Word prosody analysis



Requires word, syllable, and phone segmentations

- Automatically produced via forced alignment¹
- This work uses existing IFCASL segmentations
- Syllable segmentations derived from words & phones



¹L. Mesbahi et al. "Reliability of non-native speech automatic segmentation for prosodic feedback." In: *SLaTE*. 2011.

Word prosody analysis: Duration



Duration (DUR)

- Perceptual correlate: length/timing
- Best indicator of German stress¹
- Simple to extract from segmentations
- Features: Relative syllable & nucleus (vowel) lengths

¹G. Dogil and B. Williams. "The phonetic manifestation of word stress". In: *Word Prosodic Systems in the Languages of Europe*. Ed. by H. van der Hulst. Berlin: Walter de Gruyter, 1999. Chap. 5, pp. 273–334.

Word prosody analysis: F0



Fundamental frequency (F0)

- Perceptual correlate: pitch
- 2nd best indicator of stress after duration¹
- ▶ Pitch contours computed using JSnoori^{2,3}
- Features: relative syllable & nucleus:
 - Mean F0 (in voiced segments)
 - Maximum F0
 - Minimum F0
 - F0 range (max-min)

¹G. Dogil and B. Williams. "The phonetic manifestation of word stress". In: *Word Prosodic Systems in the Languages of Europe*. Ed. by H. van der Hulst. Berlin: Walter de Gruyter, 1999. Chap. 5, pp. 273–334.

²isnoori.loria.fr

³J. Di Martino and Y. Laprie. "An efficient F0 determination algorithm based on the implicit calculation of the autocorrelation of the temporal excitation signal". In: *EUROSPEECH*. Budapest, Hungary, 1999, p. 4.

Word prosody analysis: Intensity



Intensity (INT)

- Perceptual correlate: loudness
- Worse predictor than DUR or F0, but still may have effect on stress perception¹
- Energy contours computed using Jsnoori
- Features: relative syllable & nucleus:
 - Mean energy (over 60dB "silence threshold")
 - Maximum energy

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Diagnosis by comparison



Comparison to a single reference utterance



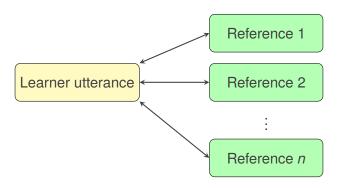
- Simplest approach, common in CAPT
- JSnoori (and predecessors) use this method¹
 - Assigns 3 scores (DUR, F0, INT)
 - ► Same syllable stressed?
 - Difference between stressed/unstressed syllables similar enough?
 - Overall score = weighted average of 3 scores
- Problem: extremely utterance-dependent!

¹A. Bonneau and V. Colotte. "Automatic Feedback for L2 Prosody Learning". In: *Speech and Language Technologies*. Ed. by I. Ipsic. InTech, 2011.

Diagnosis by comparison



Comparison to multiple reference utterances



- ▶ Less common in CAPT systems
- ► Less utterance-dependent than single comparison
- Overall score = average of one-on-one scores

Diagnosis by comparison



Options for selecting reference speaker(s)

- Manually
 - · Learner's choice
 - Teacher/researcher's choice
- Automatically
 - May be more effective to choose reference speaker most closely resembling the learner¹
 - Selected by comparing speakers' F0 mean and range (using all available recordings)

¹K. Probst et al. "Enhancing foreign language tutors - In search of the golden speaker". In: *Speech Communication* 37.3-4 (July 2002), pp. 161–173.



- More abstract representation of L1 pronunciation
- Not yet explored for German CAPT

Research questions:

- How well can lexical stress errors be classified?
- How does that compare with human agreement?
- Which features are most useful for classification?



Experiments:

- ▶ Trained CART classifiers using WEKA toolkit¹
- Used error-annotated dataset for training/test data (gold-standard labels)
- Used L1 utterances of the same words as training data (all automatically labeled [correct])

Evaluated in terms of:

- % accuracy (% agreement with gold-standard labels)
- κ with respect to gold standard

¹www.cs.waikato.ac.nz/ml/weka

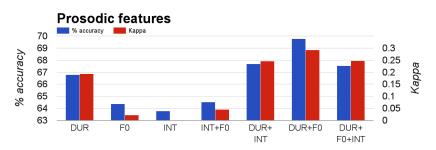


Which features are most useful for classification?

Feature set	Description
DUR F0 INT	Duration features Fundamental frequency features Intensity features
WD LV AG	Uttered word (e.g. <i>Tatort</i>) Speaker's skill level (A2 B1 B2 C1) Speaker's age/gender (Girl Boy Woman Man)

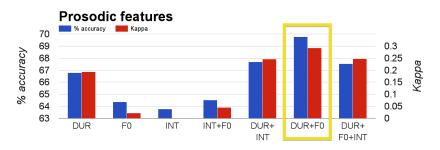


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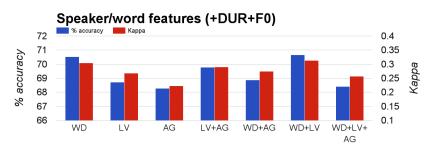


Best performance using only prosodic features: DUR+F0

- % Accuracy: 69.77%
- κ: 0.29

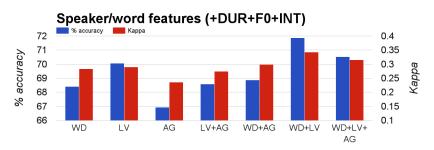


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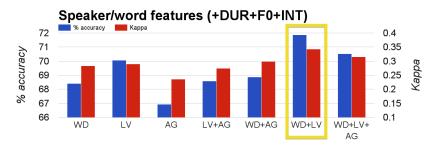


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Best performance overall: WD+LV+DUR+F0+INT

- % Accuracy: 71.87%
- ► *κ*: 0.34



How does classification accuracy compare with human agreement?

	% agreement	κ
Best classifier vs. gold standard Mean human vs. human	71.87% 54.92%	0.34 0.23

- Results are encouraging in this context
- Still want better performance for real-world use

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Implicit feedback



Allows learner to notice features of their utterance/reference utterance, without explicitly evaluating their pronunciation

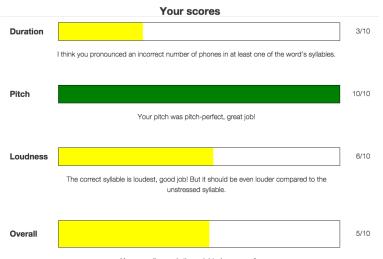
Im Frühling fliegen Pollen durch die Luft.



Explicit feedback



Directly calls learner's attention to error(s) and/or offers corrective instruction



Self-assessment as feedback



May be linked to progress and motivation¹

Self-assessment		
Listen to your utterance and the reference utterance(s).		
Then answer these questions:		
Which syllable did you stress?		
0	The first syllable (correct)	
0	The second syllable (incorrect)	
0	Neither syllable (incorrect)	
Is the stress as clear in your utterance as it is in the reference utterance?		
0	Just as clear as in reference	
0	Not as clear as in reference	
0	I don't know	
What co	What could you work on for next time?	
	Continue	

¹A. Neri et al. "The pedagogy-technology interface in computer assisted pronunciation training". In: *Computer Assisted Language Learning* (2002).

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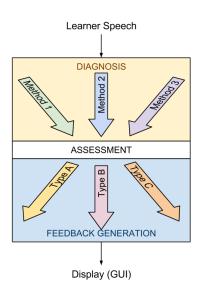
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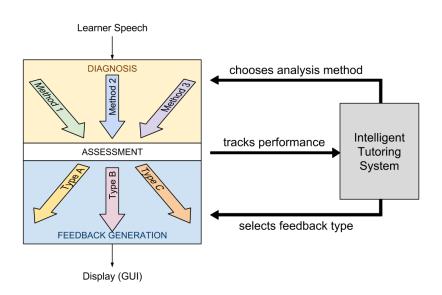
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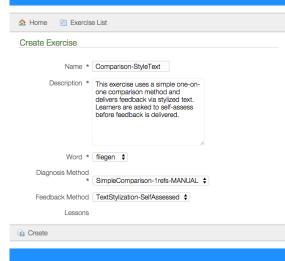


Teacher/Researcher interface



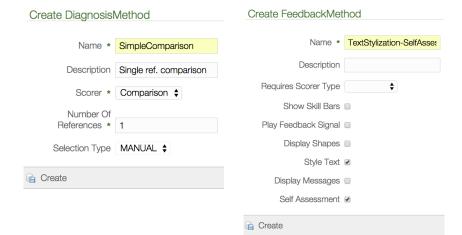
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Teacher/Researcher interface





Student interface





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- Annotation & analysis of lexical stress errors in small corpus of German spoken by French speakers
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Main contributions of the thesis:

- Annotation & analysis of lexical stress errors in small corpus of German spoken by French speakers
 - Rather low inter-annotator agreement
 - Roughly one-third of utterances contained errors
- Exploration of classification for error diagnosis
 - 71.87% accuracy, κ = 0.34 wrt. gold-standard labels
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- The de-stress CAPT tool
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Future work:

- In vivo studies using de-stress
- ► Improve classification performance (e.g. new algorithms)

Selected references



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Thanks for listening!



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- ▶ Bernd Möbius
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