# A roadmap connecting NLP research and language learning

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## Introduction

- Computers are widely used in foreign language teaching to help learners experience a foreign language & culture.
  - multimedia, web-based TV, chat with native speakers, ...
  - includes Computer-Assisted Language Learning tools, but such CALL tools do not analyze language
- Despite decades of research, NLP plays no role in real-life foreign language teaching and learning.
- This is striking given that NLP can address real-life needs for interactive feedback and adaptive materials in
  - schools with increasingly heterogeneous students
  - individualized and distance education
  - life-long learning

## The talk is based on:

Detmar Meurers (2012). Natural Language Processing and Language Learning. Encyclopedia of Applied Linguistics, edited by Carol A. Chapelle. Blackwell. 4193-4205.

- Detmar Meurers (2015). Learner Corpora and Natural Language Processing. The Cambridge Handbook of Learner Corpus Research, edited by Sylviane Granger, Gaëtanelle Gilguin and Fanny Meunier. Cambridge University Press.
- Luiz Amaral & Detmar Meurers (2011). On Using Intelligent Computer-Assisted Language Learning in Real-Life Foreign Language Teaching and Learning. ReCALL 23(1), 4-24.
- Markus Dickinson, Chris Brew & Detmar Meurers (2013). Language and Computers, Ch. 3: Language Tutoring Systems. Blackwell.

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- References
- Foreign Language Teaching and Learning (FLTL)

Second Language Acquisition (SLA) research

address teacher needs, while keeping them in charge

tasks, instructional interventions, relevance of input/output

and meaning, focus-on-form, developmental sequences

learner language to provide feedback, assess capabilities

native language to search, adapt and enhance materials

Cognitive Psychology

NLP supports the analysis of

- attention, memory, learning, motivation, lab studies
- Empirical Educational Science

Needed: a roadmap connecting NLP with

intervention studies, real-life evaluation, multi-level modeling



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#### The need for a roadmap and Language Learning Detmar Meurers



#### Intelligent Tutoring Systems

- online analysis of learner language aimed at supporting language acquisition
- provide immediate, individualized scaffolding feedback:
  - meta-linguistic feedback in form-focused activities
  - incidental Focus-on-Form in meaning-based activities
  - feedback on meaning (rare in ITS, but needed in TBL)
- determine progression through pedagogical material

#### Why analyze learner language?

#### Writer's aid tools

- feedback aimed at producing text
- identify and correct errors in orthography, grammar, usage
- could support tailoring of material to specific audiences (readability analysis, manual or automatic simplification)

#### Language Testing

- support assessment of learner competence
  - automate (some) grading
  - support more efficient grading by grouping learner answers
- draw valid inferences about a learner's state of knowledge
  - also central (but little discussed) for ITS

Why analyze learner language?

#### Learner Corpora

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- offline analysis of learner language
- gold-standard training & testing data for NLP development
- support effective search in annotated data
  - to provide insights into typical student needs in FLT
  - to provide empirical evidence for SLA research
    - developmental sequences
    - linguistic correlates of CEFR proficiency levels
    - L1 transfer (→ NLI shared task)
    - ▶ ...

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### Under the hood: Interpreting learner data

- How would you analyze the following sentences from the Hiroshima English Learners' Corpus (Miura 1998)?
  - (1) I didn't know
  - (2) I don't know his lives.
  - (3) I know where he lives.

They are taken from a translation task, for the Japanese of

- (4) I don't know where he lives.
- $\Rightarrow$  To reliably interpret learner language in ITS and learner corpus research, we should more seriously consider
  - the particular task, and
  - the learner characteristics.



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## Under the hood: Interpreting learner data Interpreting learner data requires task and learner modeling

- The best approach to grammatical error correction only reaches 39.7% precision, 30.1% recall (Ng et al. 2014)
- Inter-annotator agreement for error annotation of learner corpora is only starting to be reported (Rosen et al. 2014).
  - agreement good on some error tags (agreement, subcat.)
  - target hypotheses required for reliable error diagnosis
- Effective analysis of learner language requires modeling the expected well-formed and ill-formed variation.
- We need to constrain the space of variation using
  - well-designed tasks (Amaral & Meurers 2011; Quixal 2012)
  - learner modeling (Michaud et al. 2001; Amaral & Meurers 2008)

## Under the hood: Readability-ranking the web (Vajjala & Meurers 2013)

#### Are state-of-the-art readability models actually useful for classifying texts as found on the web?

- Can we re-rank search results based on reading levels?
- Implementation details:
  - feature set inspired by SLA measures
  - WEKA linear regression, since we want output on a scale
  - trained model on 5-level WeeBit corpus
- We applied the readability model to search results obtained through BING search API.
  - took 50 search queries from a public query log
  - computed reading levels for Top-100 results

# Why analyze native language for learners?

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### Searching for appropriate materials for learners

- learner or teacher specifies a keyword as in web search
  - to obtain authentic materials
  - on a current topic of individual interest
  - fully contextualized language use
- making search engines language-aware allows us to target
  - materials appropriate in readability
  - including language forms to be learned, in representative contexts

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### Enhanced presentation of materials, adapted to learner

- visual input enhancement supporting noticing
- generation of exercises

1 2 3 4 5

1.8 4.6 1.4 2.7

1.7

3.6 4.9

2.4 2.9 4.2

3.9 4.7

4.5 4.5

6.3

4.7

Vajjala & Meurers (2013)

Result Rank:

Query:

generation of annotations (e.g., vocabulary)

Results: Reading levels of top search results

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Avg.

Top100

4.6

4.0

4.3

3.6

3.2

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Results:

copyright copy law

europe union politics

euclidean geometry

halley comet

shakespeare

- avg. reading level of search results high (5 = GCSE)
- full range of reading levels among most relevant results returned by search engine

4.6 6.2 2.7

2.2

4.5 4.6 4.0 4.1

4.2 2.4 4.1 4.9

4.0

4.7 4.7

4.3

Readability-based re-ranking of results is potentially useful.



7 8

1.5 1.6

6

4.5

3.9 1.5 9 10

3.9 5.6

4.9 6.3

2.6 4.0

3.5 2.6

1.1

3.6 4.2 3.6

2.1



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## Conclusion/Outlook

- In language learning, there are real-life and research needs that NLP can help address.
- Development and evaluation requires collaboration with SLA, FLTL, Empirical Educational Science.
  - LEAD graduate school is a step in this direction
    - ReadingDemands: Is school book language taking the grade level and school type into account?
    - CogCorr: Linking linguistic and cognitive measures of text complexity
  - NLP benefits, but also feeds back into SLA & education
  - iCALL facilitates experimentally testing SLA hypotheses
    - WERTi/VIEW (Meurers et al. 2010)
    - eCALL (Presson et al. 2013)

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