



EU-BRIDGE Bridges Across the Language Divide

Sebastian Stüker, KIT

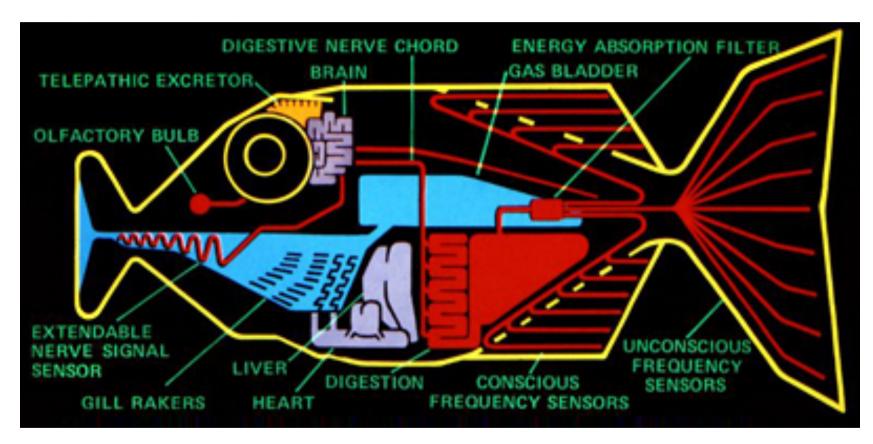


The work leading to these results has received funding from the European Union under grant agreement n° 287658



Not this project, + different architecture



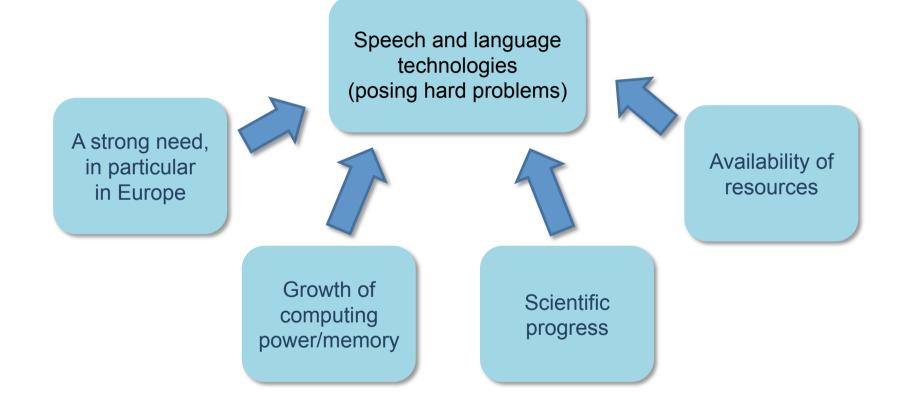


BABEL FISH

from "The hitch-hiker's Guide to the Galaxy" (TV series)
1981 © BBC
orig. animation artwork by Rod Lord

It's Happening Right Now





The Vision



- Bridges Between Languages in Europe:
 - Technology to bridge the European language
 - Building/maintaining a sustainable European language infrastructure
 - Language divide is expressed in language and speech
 - Easily accessible services
- Bridges Between Science and Society
 - Language technology requires continued scientific attention
 - Exploitation and insertion requires suitable adaptation
 - Metrics:
 - WER, BLEU, TER, F-Scores
 - User Friendliness, Productivity, Sales, Distribution Channels, Customer Support
 - Identify use cases and applications
 - Effective transition and insertion

EU-BRIDGE in a Nutshell



- Here: Not science fiction, but results coming soon (2014/2015)
- FP7 IP EU-BRIDGE: Bridges Across the Language Divide
- Development of a speech translation service infrastructure
- Targeted to make progress in particular regarding market insertion
- Project footprint:
 - Feb 2012 Jan 2015
 - Budget € 10.5m, EC funding € 7.8m
 - 10 partners
 - 1 service infrastructure, 4 use cases

Goals and Project Plan



Four use cases

- 1. Captioning and translation of subtitles for TV programs
- 2. Simultaneous translation of academic lectures
- 3. Speech translation services for the European Parliament
- 4. Translation of webinars

Four major objectives

- 1. Performance
- Language portability
- Reduction of dependency on data
- 4. Rapid technology transition and market insertion























































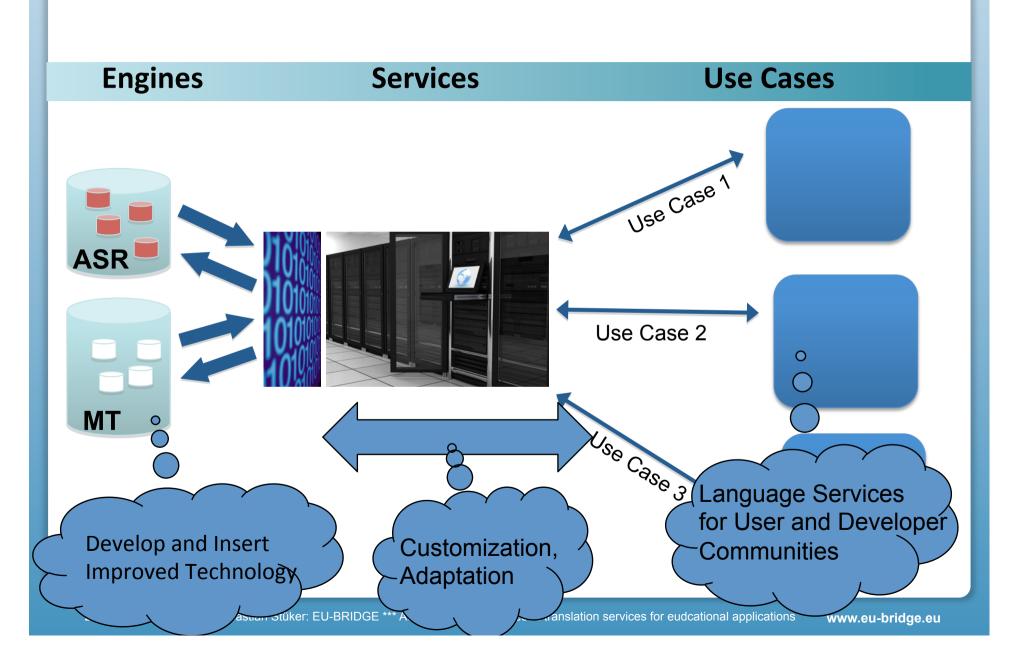




ACCIPIO PROJECTS

Language Service





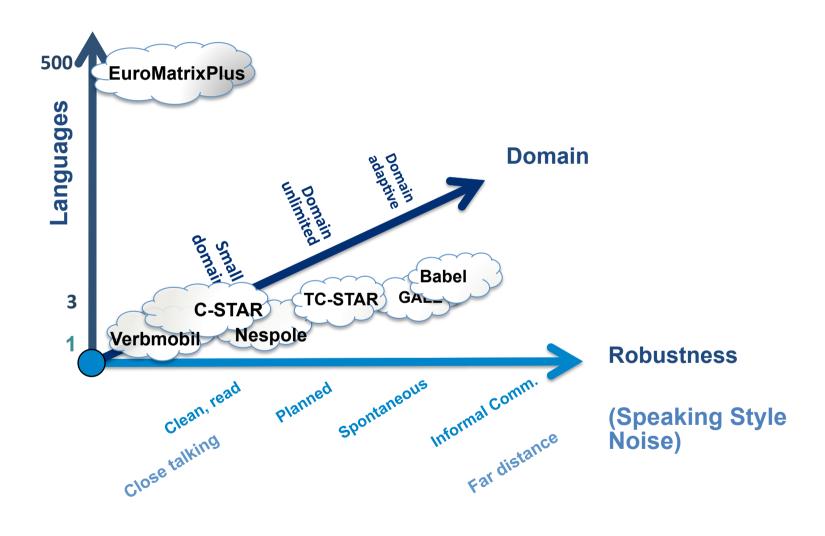
Language Service



Services **Engines Use Cases** Cross-Lingual
Captioning
Use Case 1 **ASR** Lecture **Translation** Use Case 2 Webinar Translation MT **SERENTY**

Technical Challenges



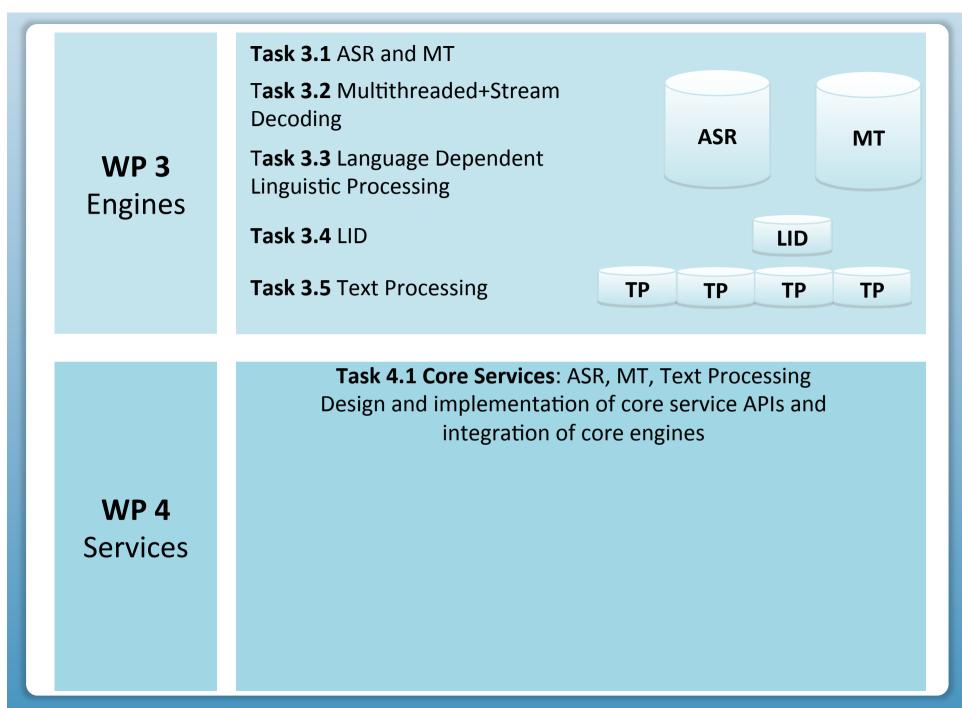


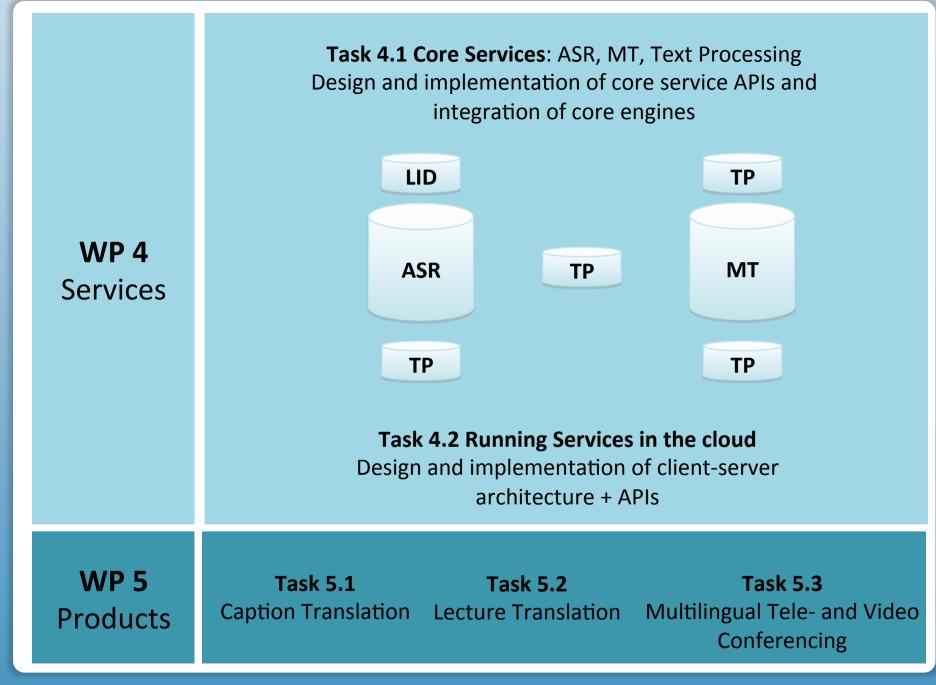
Project Organization



Laboratory

	Research	WP 1	W	P2	Standarized, inhouse tests:		
	build & improve	Transcription Translation	Data D	ependency Data Costs	WER, BLEU		
	Engines WP 3				External campaigns		
	run and schedule	Deployable En		WP6: Ev			
	Services		Evaluation				
	utilize	Transcription a	nscription and Translation Services			tion	
	Products	Caption Transl. RBM	WP 5 Lecture Transl. EUP/RBM/KIT	Webinar ADX	Field testing: dev costs cost reduction, usability		
Market Insertion							





Evaluations



"You Improve what you Evaluate"

- Evaluations organized around Use Cases
- Align Metrics with Product/Service Goals

Co-opetition

- Partners engage in friendly competition
- Goal is progress/complementarity, not site-to-site "horserace"
- Partners participate with standard or optimized/tuned systems

Evaluations



- Standard open benchmarks
 - Calibrate technology internationally in open competition
 - IWSL, WMT: cover lecture task and multilingual MT
 - EU-Bridge partners are (co-) organizers of these evals and can thus influence the process to suit EU-BRIDGE's needs
 - Don't need to create/market a new campaign
- Internal EU-Bridge Tasks
 - Need to evaluate & optimize technology for EU-BRIDGE use cases
 - Different goals/sub-goals pursued (e.g. captioning, NE, shortening, ..)
 - Evals/Assessment need to change frequently during the project
 - Comparisons with outside teams not needed and not helpful
- Field Tests and Measures:
 - Evaluate on Living online 'Organism'
 - Optimize for extrinsic measures, not intrinsic ones

IWSLT 2014



- Evaluation of Speech Translation Technology on Talks and Lectures
- Working on TED data, because of efficiency of creating training and evaluation data



Use Cases



Captioning and Translation of Multimedia Content

- BBC Weather Data
- Euronews
- Skynews

Multilingual Lectures, Meetings

- TED Lectures
- University Lectures
- European Parliament Voting Sessions

European Parliament Interpreter Support

- Terminology Support: Tool Field Tests
- Named Entities: Eval, Tool Integration & Tests

Webinar Translation

Integration into the Andrexen platform

Use Cases



Captioning and Translation of Multimedia Content

- BBC Weather Data
- Euronews
- Skynews

Multilingual Lectures, Meetings

- TED Lectures
- University Lectures
- European Parliament Voting Sessions

European Parliament Interpreter Support

- Terminology Support: Tool Field Tests
- Named Entities: Eval, Tool Integration & Tests

Webinar Translation

Integration into the Andrexen platform

Lecture Translation



- Continuous Monologue
- Speaking-Style
 - Planned, not read, not memorized
 - Fast, spontaneous, fragmentary, and no punctuation!
 - Noises, coughing, non-verbal events (e.g., singing)
- Vocabulary
 - Very large
 - Specialized terms
 - Foreign Words
- Speed, Rea-Itime
- Service-Infrastructure
 - Many parallel lectures;
 - Automatic, robust assignment of compute power

Lectures

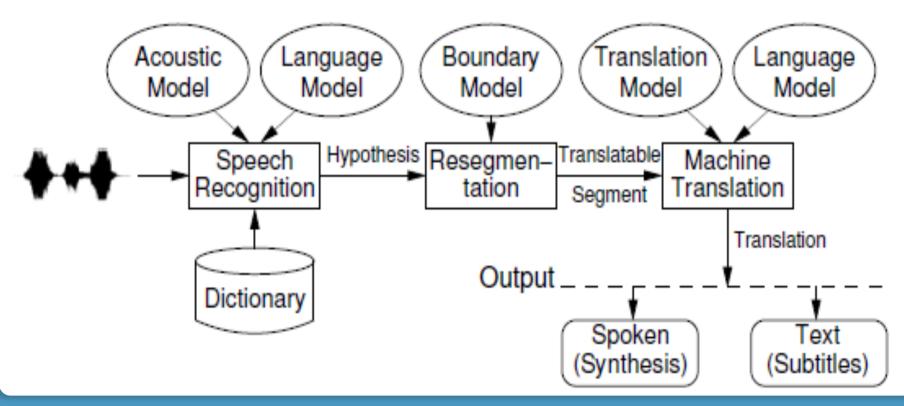




Speech-Translation Systems

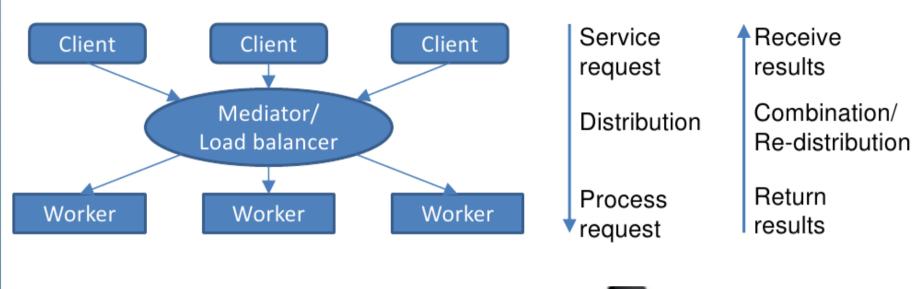


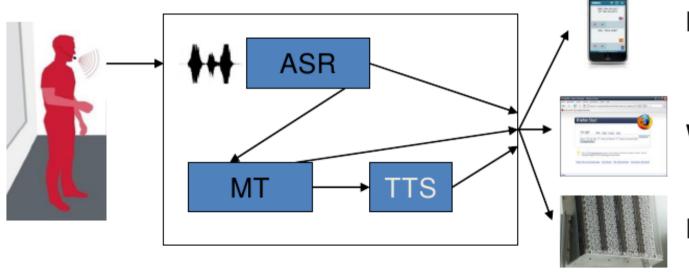
- Human Interpreters are too expensive
- Automatic Speech Translation an affordable solution:
 - Still lots of errors, room for improvement
 - But, better than nothing



Service Infrastructure







Mobile Devices

Web Browsers

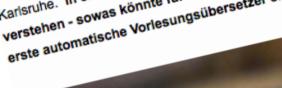
Loudspeakers

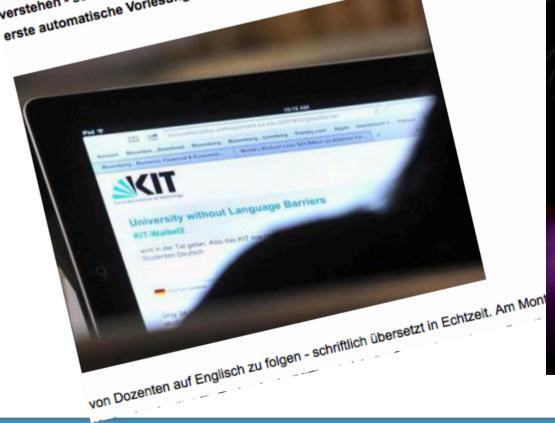
Launch, June 11 2012



Mehr als nur Bahnhof verstehen - Weltweit Karisruhe. In einer deutschen Vorlesung sitzen und wegen der Sprachbarriere nur Bal verstehen - sowas könnte für ausländische Studenten bald Vergangenheit sein. Der W erster Vorlesungsübersetzer

erste automatische Vorlesungsübersetzer ermöglicht Studierenden künftig, dem Vorlesungsübersetzer ermöglicht Studierenden künftig.









Deployment





Data Resources

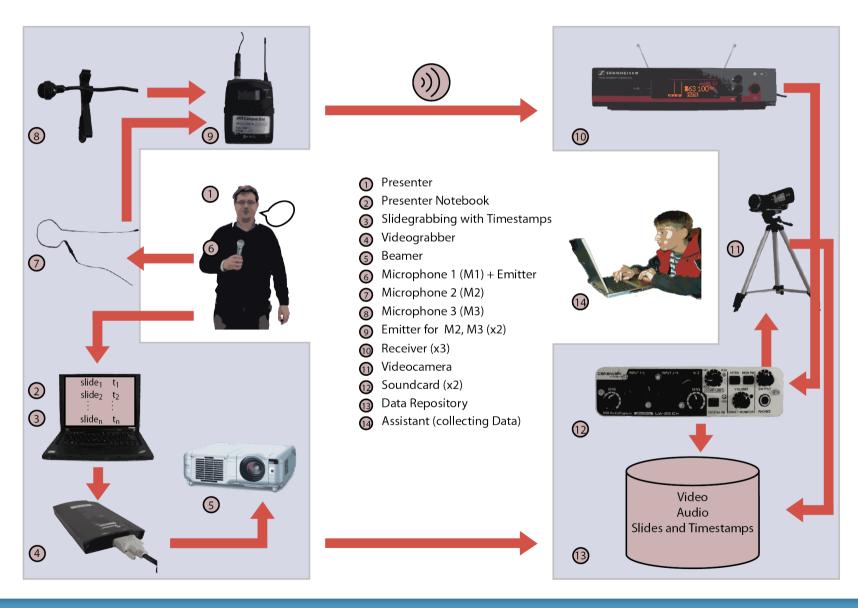


KIT lecture corpus:

- Collected to fit the needs for training the systems:
 - ASR AM: Large amounts of in-domain audio data with careful transcriptions
 - MT Translation Model: Parallel sentences of in-domain data for the translation
 - ASR+MT LM: Large amounts of monolingual sentences in the required domain
 - Any kind of meta data that might help (e.g., lecturers' slides)
- Data collection took place at KIT's lecture halls:
 - Started with computer science lectures
 - Gradually spread to lectures from all departments at the university

Data Resources





Automatic Speech Recognition



- Janus based ASR system: HMM/GMM based quinphone system with 4,000 models, MVDR front-end, 4gram LM
- Acoustic Model:
 - Trained on all lecture data in order to get a speaker independent model
 - Created 5 speaker adapted AMs: speaker independent model + Viterbi training and bMMIE training on the speaker dependent data
- Language Model:
 - Interpolation from 28 text corpora
 - Interpol. weights tuned on random selection from AM training data
 - 300k vocab selected by ML count estimation method
- German has a lot of compounds:
 - Sub-word vocabulary for compounds
- Vocabulary adaptation by deriving queries from slides (OOV 2.25%→0.75% at 300k vocab size)

ASR Performance

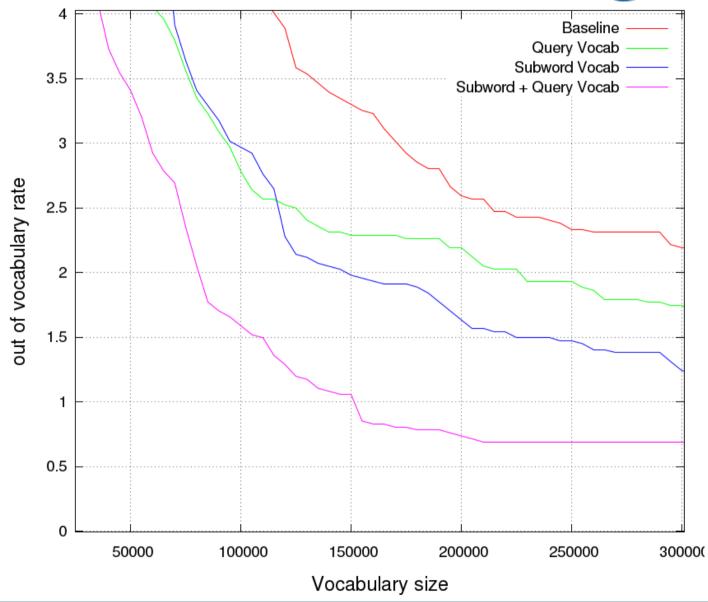


Tested the ASR system on the dev set of six lecturers

Lecturer	1	2	3	4	5	6
Speaker Independent AM	34.8%	21.1%	28.4%	22.9%	22.7%	19.0%
Speaker Dependent AM	_	18.9%	27.6%	22.6%	21.5%	17.8%
Adapted LM+Vocabulary	23.9%	17.3%	_	18.1%	_	15.4%

OOV Reduction





ASR Post-Processing



- Numbers are converted to digit sequences
- Common symbols are substituted, e.g., "Prozent"→'%'
- Punctuation, i.e. '.' and ',' is inserted:
 - Prediction via a 4gram model
 - Pause information used to adjust the priors
- Simple, rule based conversion of equations:
 - "F of x" → 'f(x)'
- Punctuation is used to structure text into sentences and chunks for translation

Machine Translation



Statistical Phrase Based MT System

- Trained on different out-of-domain data and the lecture corpus
- Applied the same compound word splitting as for the ASR system for consistency; names etc. were excluded from splitting by applying a named entity tagger
- Used a discriminatively trained word alignment approach
- Specific models for short and long range reorderings (also on the training data)
- Online system:
 - Phrase table filtered with ASR vocabulary
 - Simplified POS tagging

Machine Translation



Adaptation to the lecture domain

- Domain independent translation model trained on all data
- In-Domain TM only on the lecture data (re-use alignments)
- Combined via log-linear combination
- For LM log-linear combination of large LM, in-domain LM and TED LM
- Translations or special terms learned from Wikipedia and Wiktionary

MT Results



Lecturer	BLEU			
Lecturer 1	13.80			
Lecturer 2	22.58			
Lecturer 3	14.24			
Lecturer 4	20.83			
Lecturer 5	24.50			
Lecturer 6	24.13			

Thoughts on the Output Modality



Text instead of synthesized speech

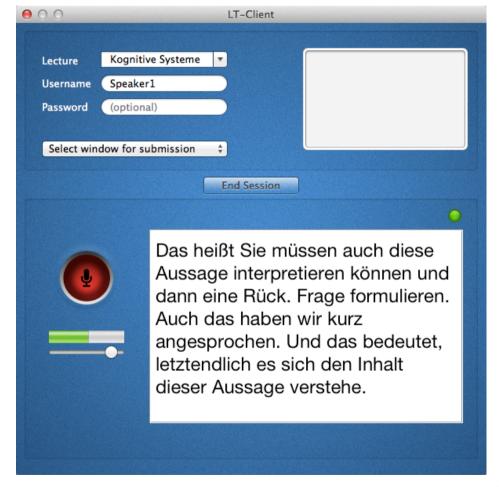
- Text can be easily distributed over the WWW
 - Laptops, smartphones, tablets
 - Nowadays ubiquitous
 - No proprietary software, just a browser
- Listening to synthesized speech can be very tiresome
 - Artificial voices not perfect
 - Original speech present in addition
- Translation system commits errors
 - Translated text contains errors; synthesis quality suffers from that
- Temporal Navigation
 - Once translation has been heard, it is gone
 - Text enables users to move back and forth in the translation; supports understanding the translation

Interface



Client for the lecturer:

- Must be as simple as possible!
- Client needs to know:
 - Who is speaking?
 - What lecture is it?
- Lecturer needs to know:
 - Is it turned on?
 - Is it doing something?

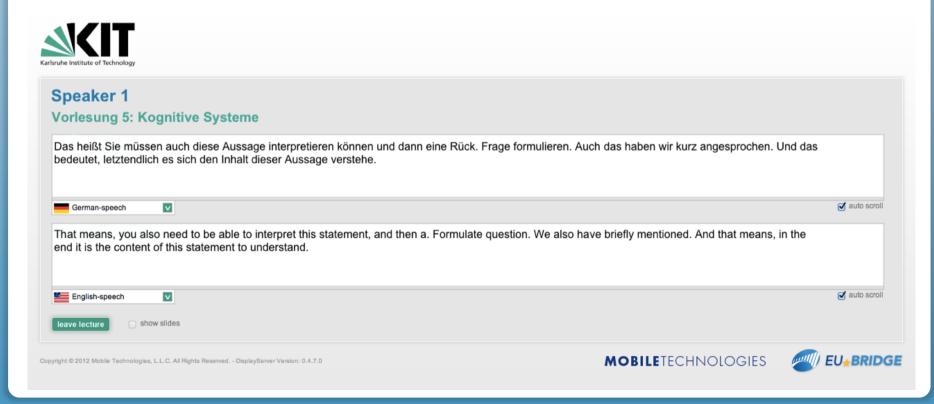


Interface



Displaying the results:

- Translation result + ASR result (?!)
- Scrolling back and forth in time
- Make text readable !!!



Where do we go?



- Interfaces need to get simpler!
- Systems need to become an omnivore:
 - Get all meta information as early as possible (before the lecture!)
 - Slides, papers, web sites, text books, etc.
 - Find a way for obtaining comparable corpora
- Make system self maintaining and autonomous
 - Automatic unsupervised acoustic model adaptation after every lecture
 - Automatically detect speaker, lecture and language: Access the university information system
- Offer additional services
 - Archive of the lectures for the students (for search)
 - Translation of the slides
 - Summary of the lecture
- Get the students in the loop
 - Automatic corrections by the students: during the lecture and afterwards
 - Make it a game

Andrexen - Webinar



- Role definition (Speaker / Listener)
 - Functionality was defined and implemented, and relevant MCloud services integrated
- Data flow (Architecture / Slideshow / Voice / Text)
 - Features implemented to support the parallel transmission of slides, speaker audio, translation text and intelligent user location to provide integrated UC experience

Andrexen - Webinar



Adapting to industry requirements:

- Real time streaming
- Support legacy systems:
 - Training narrow bandwidth (8KHz) acoustic model
- Specific vocabulary:
 - Vocabulary and language model adaptation

Andrexen - Webinar



- Translator integrated as virtual participant
- Is "chatting" the translation results

