



The Need for Research on Automatic Speech Recognition in Air Traffic Management

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An Introductory Experiment

Mister Chairman,

Again thank you very much for the invitation and for the opportunity to present my view – my personal view – of the application of automatic speech recognition in air traffic management domain.

Let's see, if at the end of the day, I am still happy.

46 words





The Output of my Smartphone

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Mr Shannon again thank you very much for the invitation and for the opportunity to prevent my view my personal view of the application of automatic speech recognition in a traffic management to me let's see if the end of the day I am still happy



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[E]





The Output of my Smartphone









Why not using, Alexa, Google ... in ATM?

- Special data privacy requirements (Cloud is not an option).
- Realtime aspect are important
- ATC has a special phraseology, currently not modelled
- Special Context Information (e.g. radar or weather data) is available
- Cyber Security issues
- Safety issues, i.e. software certification for ATM application arguments necessary
- ATM System manufacturers can't integrate ASR from Google or Amazon directly into their systems.







Why not using, Alexa, Google ... in ATM?

lufthansa one alfa one four descend flight level eight zero reduce two twenty knots contact tower one one nine decimal seven bye

22 words





Output of my Smartphone

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The Output of My Smartphone



9 substitutions
3 addition
1 deletion

Word Error Rate WER = 13 / 22 = 59%

Understanding is another challenge



What is Needed for ASR?









Machine Learning of Speech MALORCA Recognition Models for Controller Assistance



Instead of (highly skilled and paid) experts, machine learning is used.





Invention of MALORCA



Data selection: Select "good" or "bad" data



Learning Curve









Speech Recognition and Machine Learning Roadmap





ASR Applications of HAAWAII

Objective ATCo Workload Estimation





Highly Automated Air Traffic Controller Workstation with Artificial Intelligence Integration



ASR Applications of HAAWAII Integration of ASR and CPDLC



Highly Automated Air Traffic Controller Workstation with Artificial Intelligence Integration



CPDLC: Controller – Pilot Data Link Communication



ASR Applications of HAAWAII HAAWAII HAAWAII



ATCo Callsign Highlighting is "easy". The challenge and benefits is for pilot's voice.



























Understanding of ATCos and Pilots not Easy









ASR Applications of HAAWAII Readback Error Detection

- Detection Rate > 50%
- False Alarm Rate < 10%
- 2% of Commands contain Readback errors (seldom events)

R _{both} / E _{both}	0.1%	0.2%	0.3%	0.4%	0.5%	0.6%
98%	4.8%	9.1%	13.0%	16.7%	20.0%	23.1%
95%	4.9%	9.4%	13.4%	17.1%	20.5%	23.6%
90%	5.2%	9.8%	14.0%	17.9%	21.4%	24.6%
85%	5.5%	10.3%	14.7%	18.7%	22.4%	25.7%
80%	5.8%	10.9%	15.5%	19.7%	23.4%	26.9%
75%	6.1%	11.6%	16.4%	20.7%	24.6%	28.2%
70%	6.5%	12.3%	17.4%	21.9%	25.9%	29.6%
60%	7.6%	14.0%	19.7%	24.6%	29.0%	32.9%
50%	8.9%	16.4%	22.7%	28.2%	32.9%	37.0%
40%	10.9%	19.7%	26.9%	32.9%	38.0%	42.4%
20%	19.7%	32.9%	42.4%	49.5%	55.1%	59.5%
10%	32.9%	49.5%	59.5%	66.2%	71.0%	74.6%

- → Recognition Rates >50% on Command Level
- → Recognition Error Rates < 0.2% on Command Level





Conclusions

We need Research on Automatic Speech Recognition in Air Traffic Management

- COTS engines (smartphone, google etc.) are good, but not for ATM
- Speech Recognition does not include Speech Understanding
- Europe has an ontology
- Readback-Error detection is a challenge for research AND for subject matter experts
- Iterative approach is necessary AND possible







Thank you very much for staying in the webinar

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