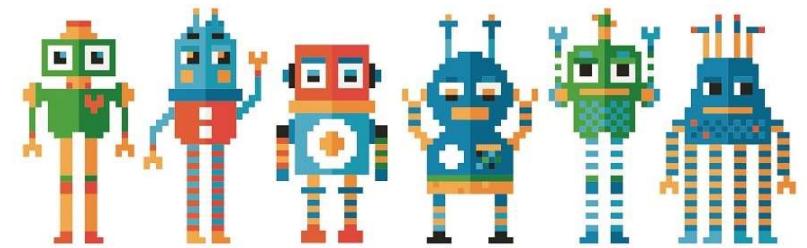


Future Teachers Are Wired Differently: Social Robots in Education

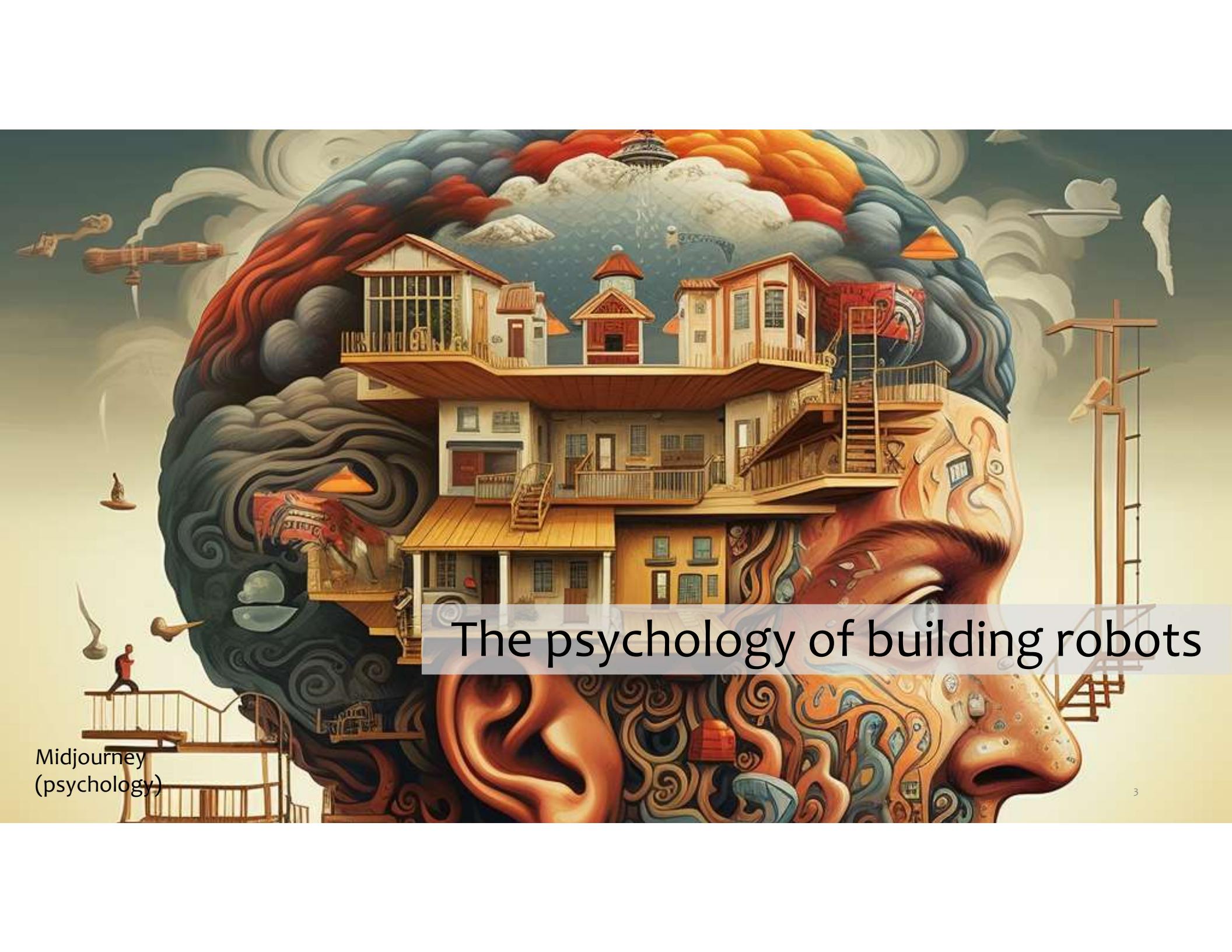
Tony Belpaeme

IDLab – imec, Ghent University, Belgium



Today's menu

- The psychology of social robots
- Social robots in education
- Autonomous social robots

A surreal illustration of a brain-shaped cityscape. The brain is filled with various architectural structures, including houses with balconies, a church with a steeple, and a lighthouse. The brain itself has a face with a wide-open mouth and a tongue. A smoking pipe is positioned near the mouth, with smoke rising. The background features a cloudy sky with a small airplane and a person riding a pipe.

The psychology of building robots

Midjourney
(psychology)

Robots tap into our social brain

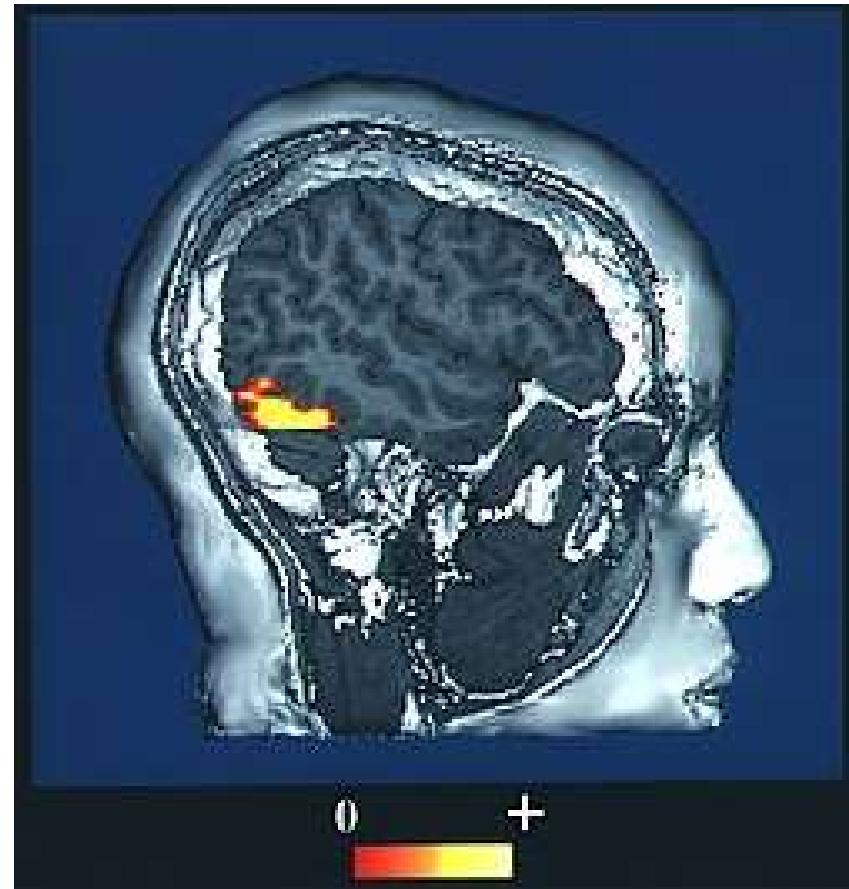
We attribute human-like characteristics to artefacts.

This effect is enhanced when the artefact is animated and responsive.



Pareidolia

- Perceiving human-like features in non-human stimuli.
- Evolutionary psychology explains pareidolia as a hyper response to face-like and human-like features.
- Better to respond to false positives than to not respond.



Fusiform Face Area responds to seeing faces and to pareidolia experiences







(Reddit)

9



Social robots

Social robots are designed to evoke anthropomorphisation.

But also dynamics, responsiveness, audible responses, linguistic and non-linguistic responses...

This can induce **attention, compliance, conformity** ...



University of Washington



Disney Imagineering

Social conformity

- How much do people conform to social pressure?
- Two kinds of social pressure
 - Informational: uncertainty, so you follow what the others are doing
 - Normative: influence by others
- Solomon Asch's social conformity experiments (1951)

S. E. Asch, Effects of group pressure upon the modification and distortion of judgments, in *Groups, Leadership, and Men*, H. Guetzkow, Ed. (Carnegie Press, 1951), pp. 222–236 .



<https://www.simplypsychology.org/asch-conformity.html>

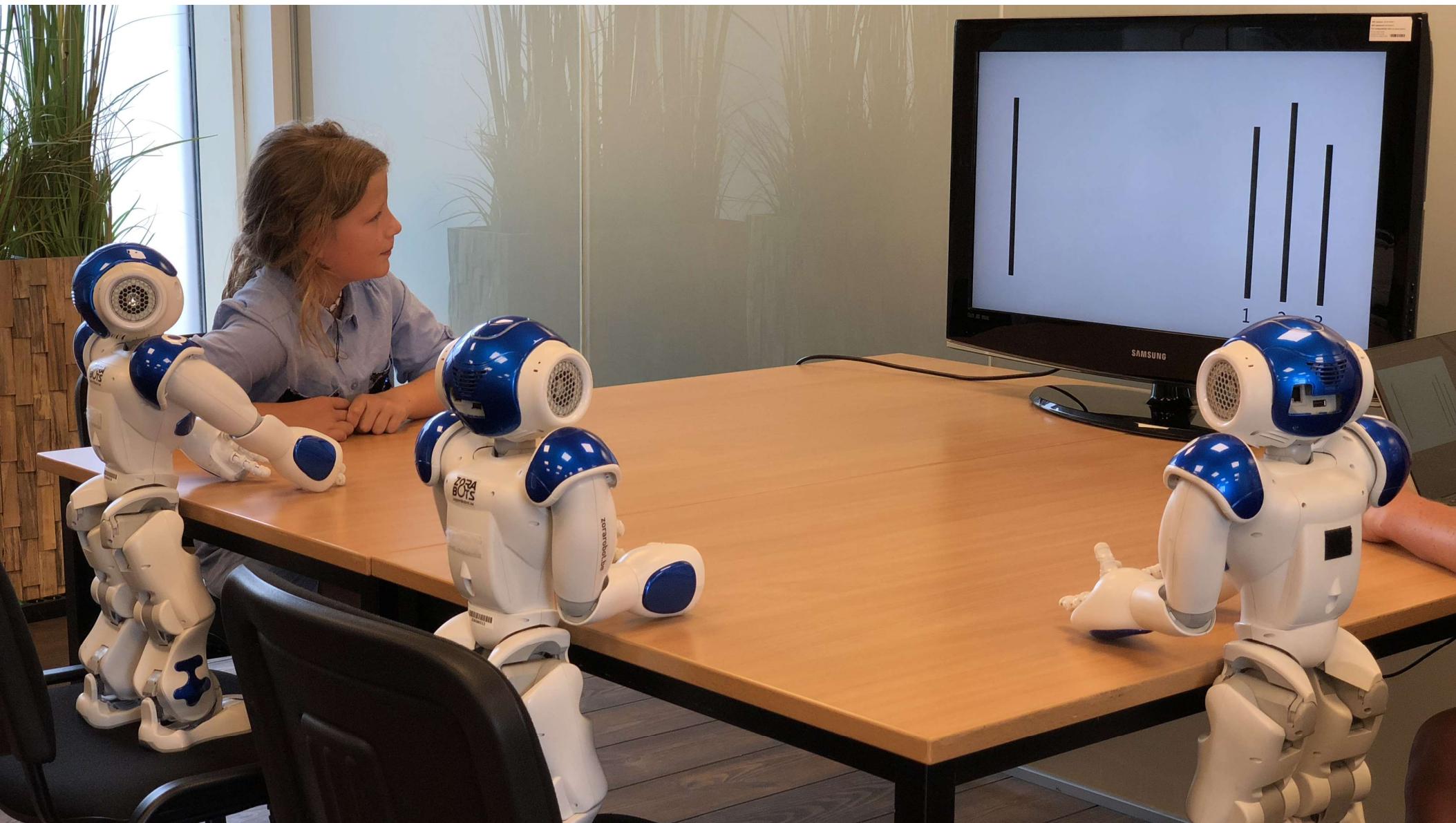


Social conformity

Asch found that people succumb to peer pressure

In 32% of the trials people copied the wrong response.

Would robots exert social peer pressure as well?





Method

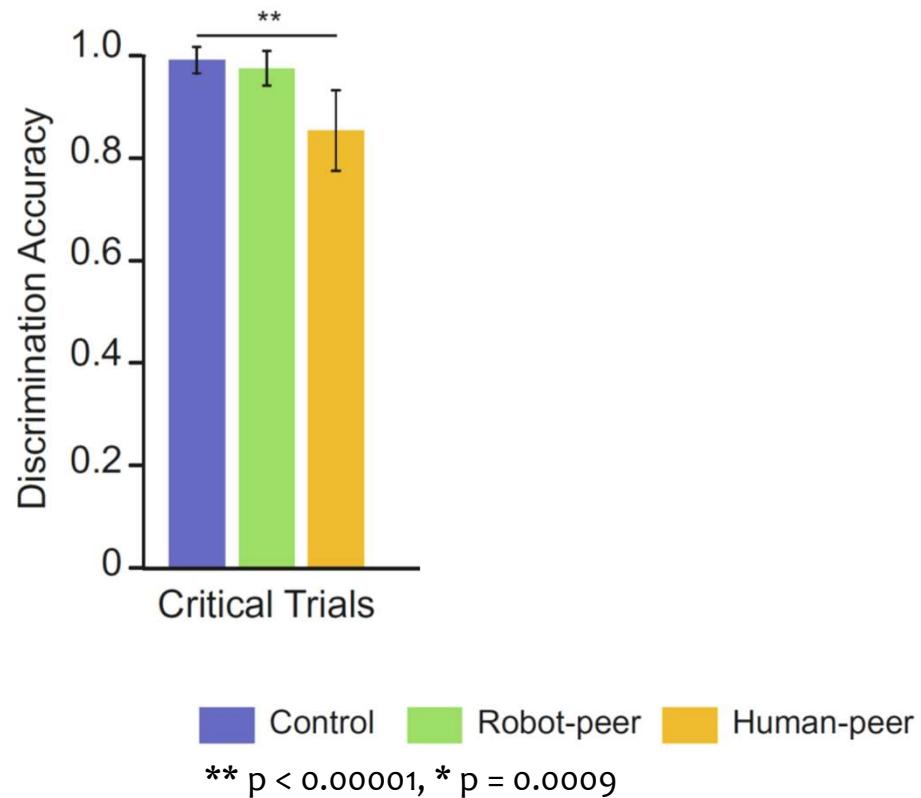
Five experimental conditions

1. Adult alone
2. Adult with robots
3. Adult with other adults
4. Child alone
5. Child with robots
6. Child with other children

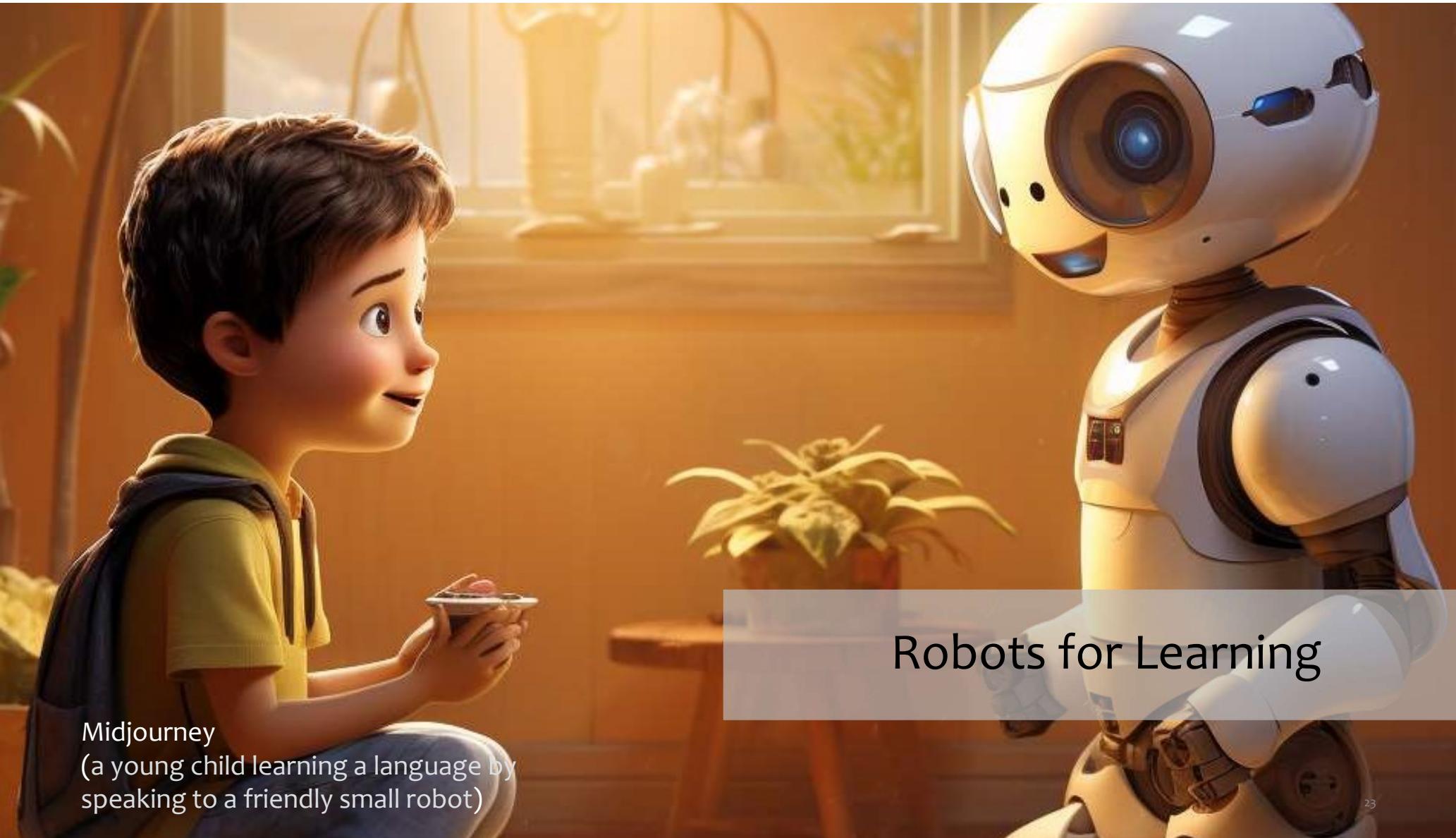
Participants

- Adults: $M_{age} = 31.0$; $SD = 14.2$
- Children: $M_{age} = 8.5$; $SD = 0.54$

adults



Anna-Lisa Vollmer, Robin Read, Dries Trippas, and Tony Belpaeme (2018) Children conform, adults resist: A robot group induced peer pressure on normative social conformity, *Science Robotics*, 3(21), eaat7111.



Robots for Learning

Midjourney
(a young child learning a language by
speaking to a friendly small robot)

Why robots?

Why indeed?

- Robots are expensive
- Robots are difficult to use at scale
- We already have enough technology (PC, tablets, smartphones, ...)

Good reasons

- Learning is **socially gated** (see Patricia Kuhl).
- Their physical and social presence makes robot tutors effective.
- Social robots can offer one-to-one tuition.
- They can achieve both cognitive and affective outcomes.
- Potential benefits for neurodiverse learners.

Using a Personal Robot to Teach Young Children

THOMAS W. DRAPER
WANDA W. CLAYTON
*Early Childhood Education Laboratory
Brigham Young University*

ABSTRACT. Seventy-five preschool children were instructed about birds by a human teacher, a moving personal robot, a stationary personal robot, and a tape recorder. How much the children learned and how much attention the children paid were compared for each type of instruction. The children learned when they were taught by the human teacher and when they were taught by the animated and the stationary robots. The children paid more attention to the live teacher and to the moving robot than they did to the stationary robot or to the tape recorder. The difference between the amount of attention the children paid to the animated robot and the amount of attention they paid to the human teacher was not statistically significant.

CHILDREN YOUNGER THAN 5 YEARS OF AGE learn differently from children older than 7 years of age. Young children's learning is more dependent on concrete, three-dimensional illustrations and social factors, such as personal liking of the teacher and animate teaching style, than is the learning of older children (Bredenkamp, 1987; Taylor, 1985). Because of the special needs of young learners, many experienced teachers and early childhood education experts have expressed doubts about the value of computer-controlled presentations for teaching young children. Much of the high technology used with young children has been labeled developmentally inappropriate because it consists of messages requiring relatively sophisticated symbolic reasoning, and hypothetical problem solving (Barnes & Hill, 1983; Brady & Hill, 1984; Cuffaro, 1984; Haugland & Shade, 1988; Tan, 1985; Zajone, 1984).

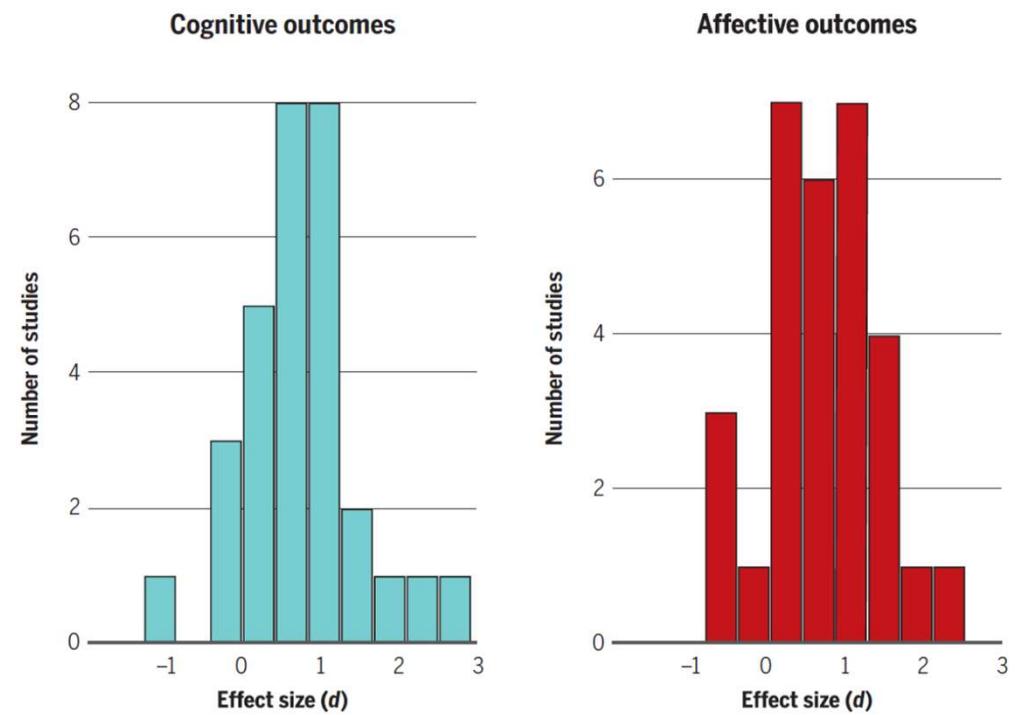
In the present study, we attempted to use technology to teach young children in a manner that is more consistent with the tenets of developmental

Address correspondence to Thomas W. Draper, Early Childhood Education Laboratory, Brigham Young University, Provo, UT 84602.



Effect sizes of outcomes

- 37 results compared a robot to alternative tech tutoring.
- **Cognitive $d = 0.70$**
- **Affective $d = 0.59$**
- Human tutor achieve cognitive outcomes of $d = 0.79$
- Positive affective outcomes do not mean positive cognitive outcomes, or vice versa.



Belpaeme, T., Kennedy, J., Ramachandran, A., Scassellati, B., & Tanaka, F. (2018). Social robots for education: A review. *Science Robotics*, 3(21), eaat5954.



Second language learning

- Current classroom setup is ill-suited for language tuition
 - “Broadcast mode” of education doesn’t fit how people acquire first and second languages.
 - Social interaction is important for language acquisition.
- Benefits of social robots
 - Interaction possible
 - One-to-one tuition
 - Pronunciation
 - Foreign language anxiety

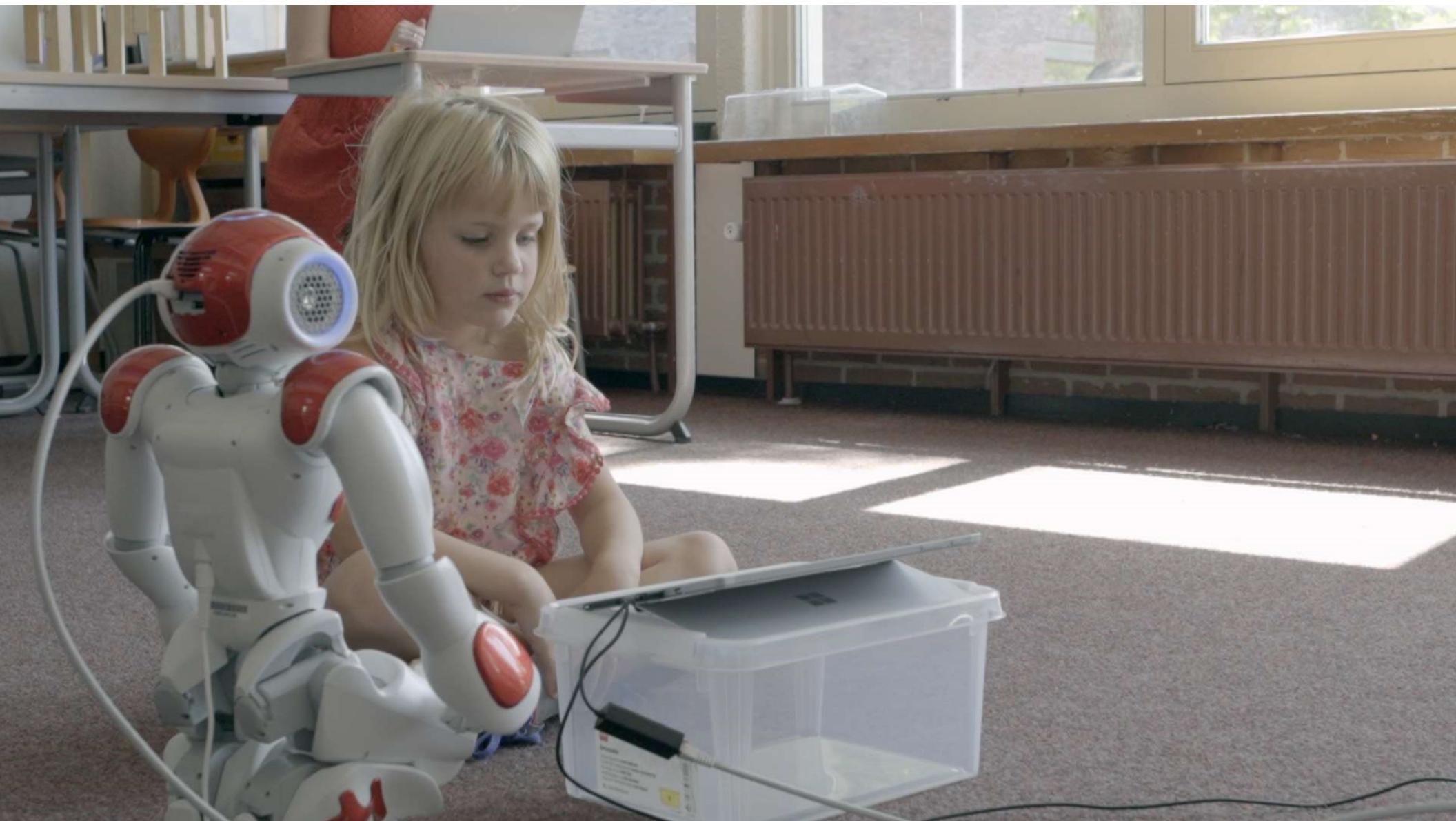
Kuhl, P. K. (2010). Brain mechanisms in early language acquisition. *Neuron*, 67(5), 713-727.

Snow, C. E., & Hoefnagel-Höhle, M. (1978). The critical period for language acquisition: Evidence from second language learning. *Child development*, 1114-1128.



Possibly the greatest challenge of all

- **Vocabulary learning works**, beyond that things become very complicated.
- **Conflicting age demands**. Critical Period Hypothesis: learn a language before puberty.
- **Technical challenges** prevent a dyadic conversation with the robot.



Video available at www.l2tor.eu

L2TOR is a European project that investigates how preschool children can learn a second language from a social robot.

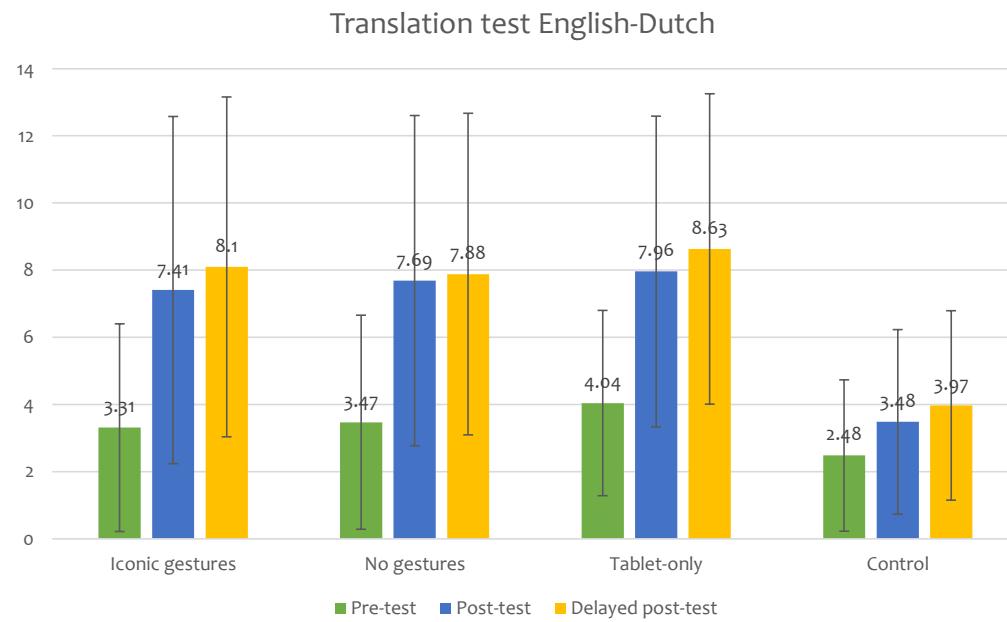


Large-scale study

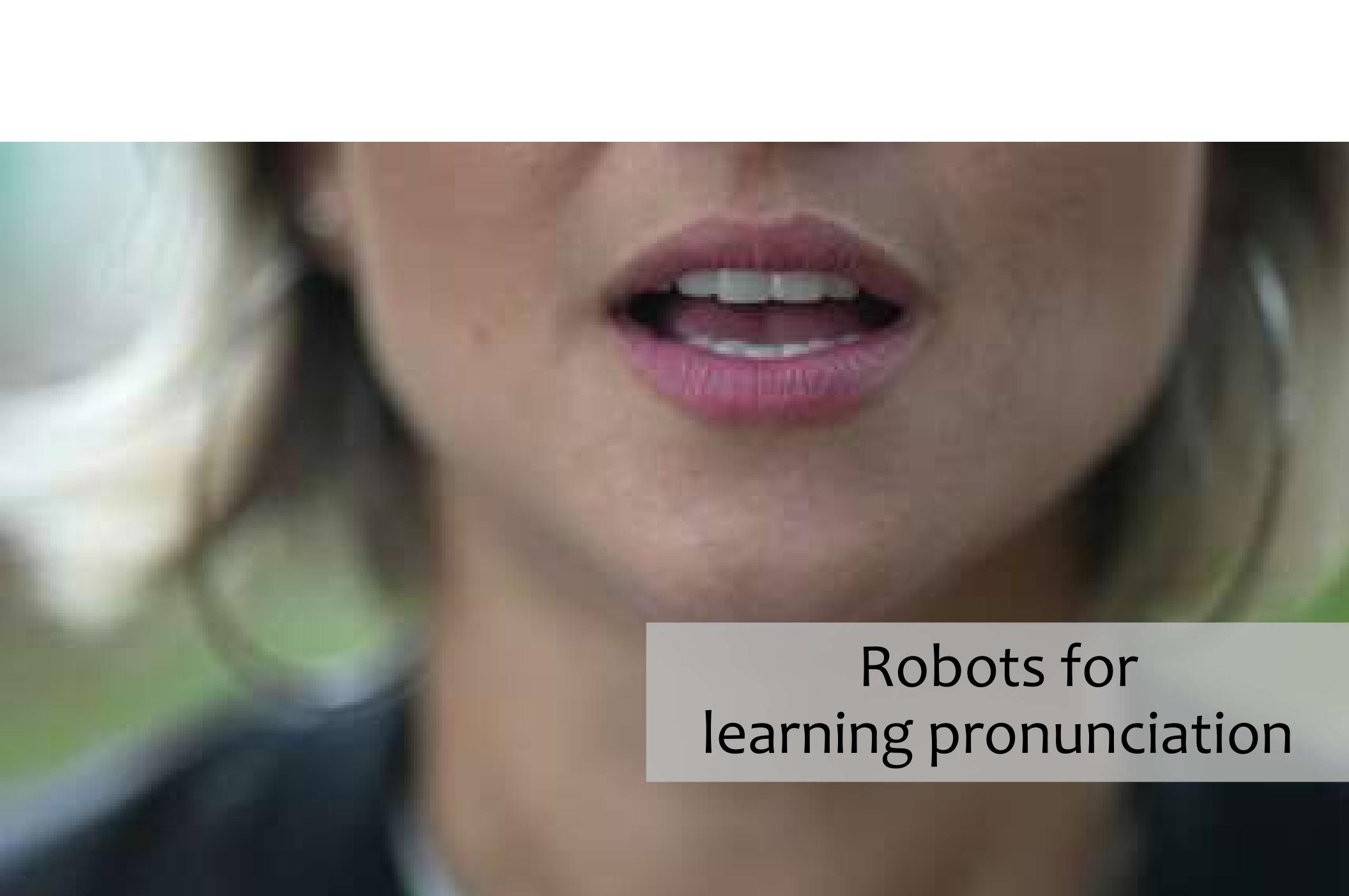
- **208 children** in 9 Dutch primary schools, 194 completed.
- Mean age = 5;8 (SD = 0;5), 34 English target words, embedded in grammatical context.
- 1 introduction, 7 learning sessions. **Two months per child.**
- Challenging words: space, number, actions
- Four conditions
 1. Robot supports teaching with iconic gestures.
 2. Robot does not use gestures.
 3. No robot, only tablet.
 4. Control conditions.

Results

- Learning is surprisingly slow.
- Children learn in all conditions, except in control condition: **no advantage for robot teaching**



Vogt, P., van den Berghe, R., de Haas, M., Hoffman, L., Kanero, J., Mamus, E., ... & Papadopoulos, F. (2019, March). Second Language Tutoring Using Social Robots: A Large-Scale Study. In 2019 14th ACM/IEEE Int. C. on Human-Robot Interaction (HRI) (pp. 497-505). IEEE.



Robots for
learning pronunciation

Audiovisual speech

- Audiovisual speech perception augments L1 language perception.
- Some evidence that L2 learners also benefit from seeing lips.
- **Hypothesis: pronunciation will improve when learning with a robot that uses audiovisual speech.**



Amioka, S., Janssens, R., Wolfert, P., Ren, Q., Pinto Bernal, M.J., Belpaeme, T. (2023) Limitations of Audiovisual Speech on Robots for Second Language Pronunciation Learning. *HRI2023*.

Methodology

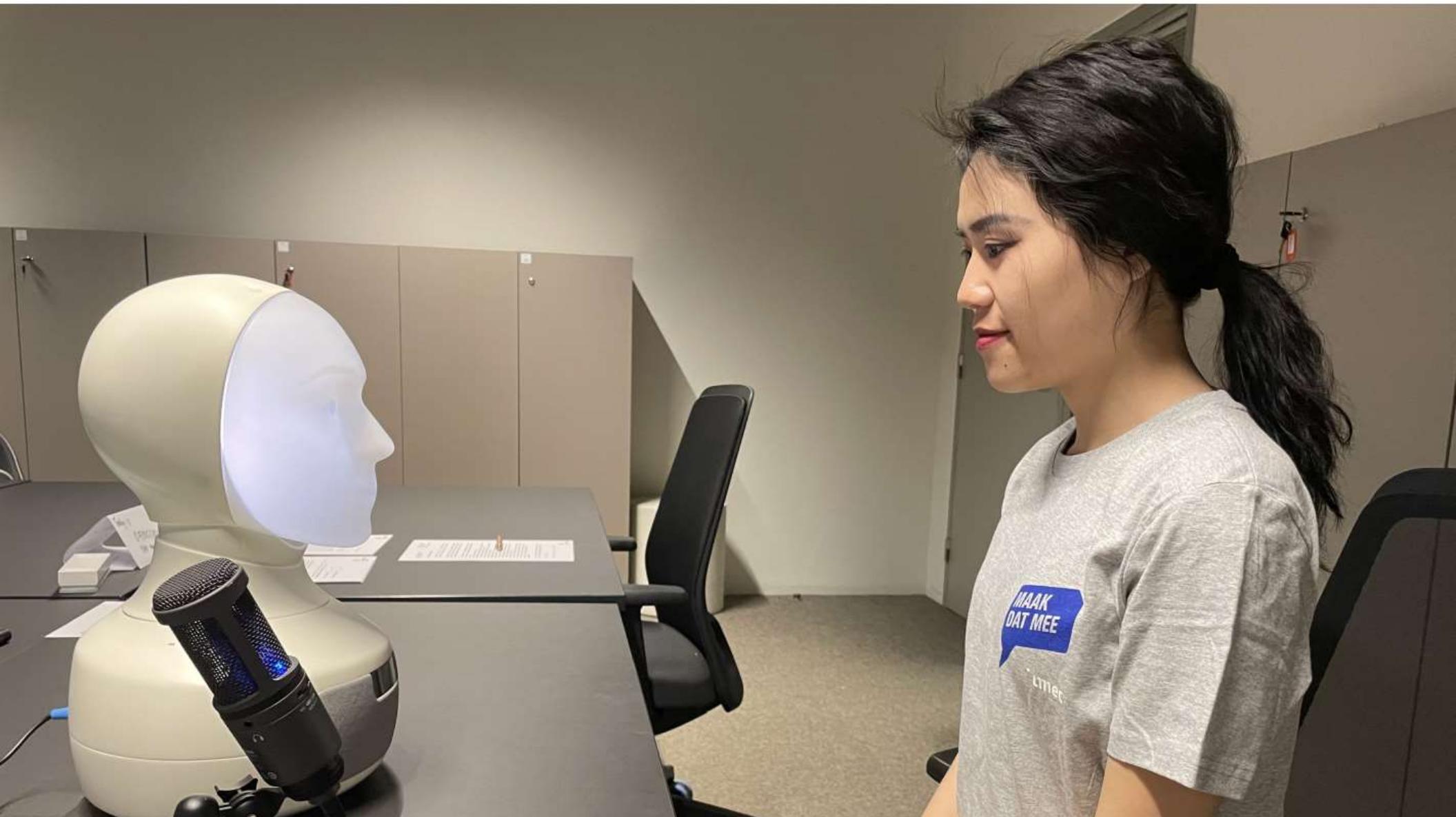
- Furhat robot, with motion captured audiovisual animation
- Robot teaches pronunciation of 30 Japanese words.
- 3 conditions
 - Matched lip synching
 - Mismatched lip synching
 - Computer generated

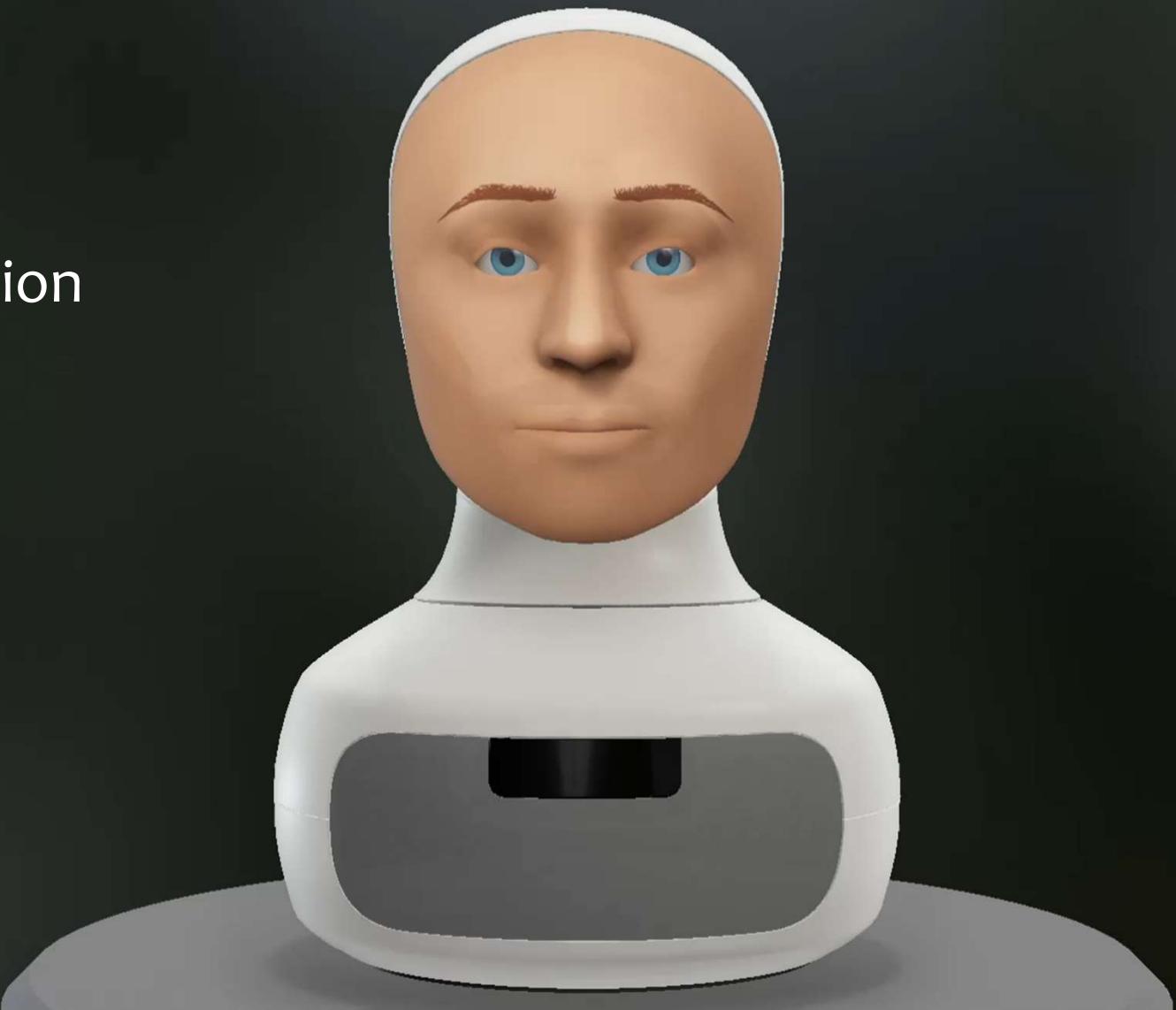


Methodology

- 27 participants recruited in Flanders (native Dutch speakers)
- Within subject design: all participants see all 3 conditions
- Their pronunciation is recorded and sent to Japan: scored by 20 native Japanese speakers from Tokyo region

Japanese	English
いち	one
ご	five
しち	seven
おはようございます	good morning
おやすみなさい	good night
ありがとうございます	thank you very much
どういたしまして	you're welcome
つたえられなかつた	I couldn't tell you
かぜぐすり	cold medicine
りょこう	travel

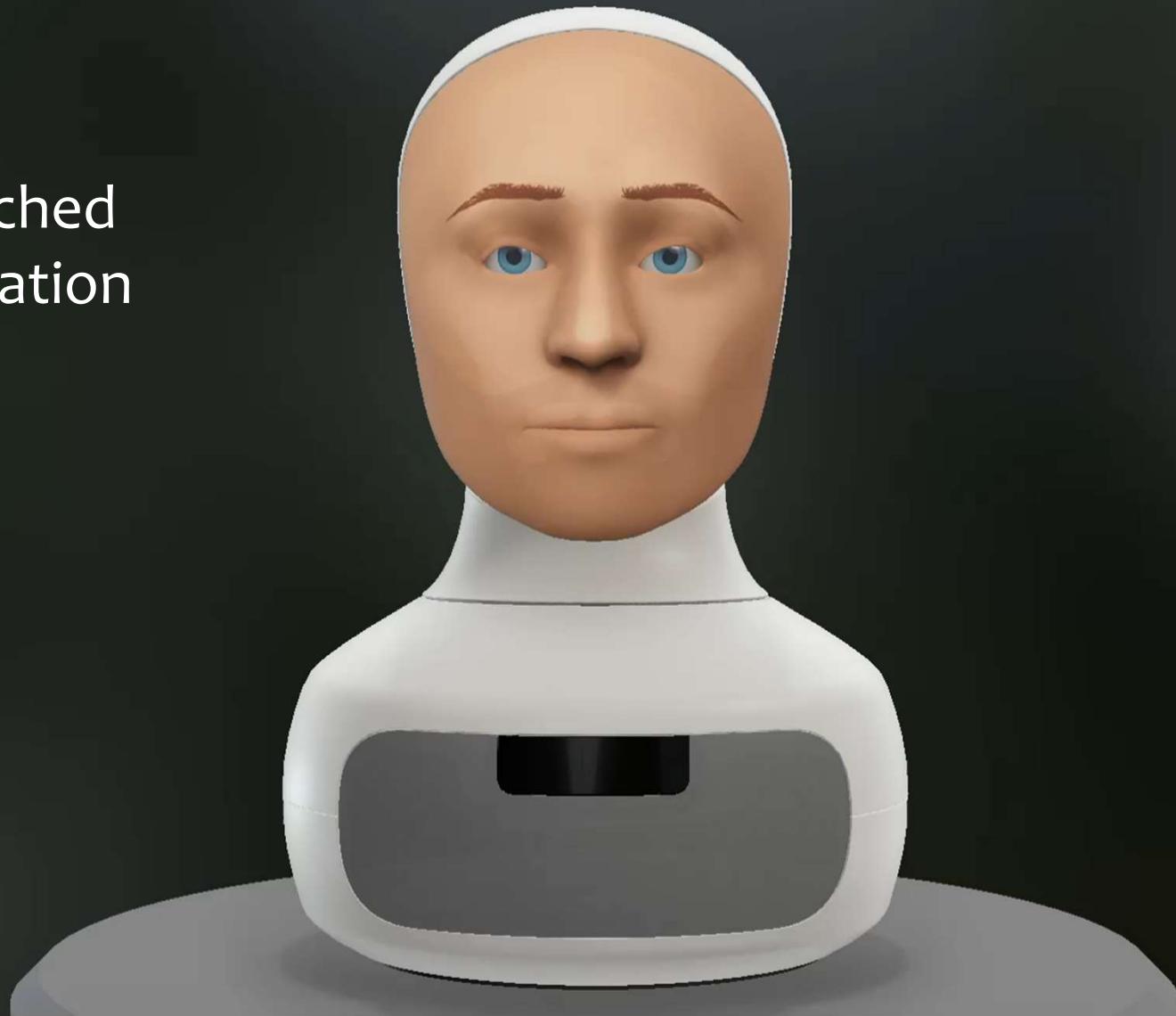




Matched
lip animation

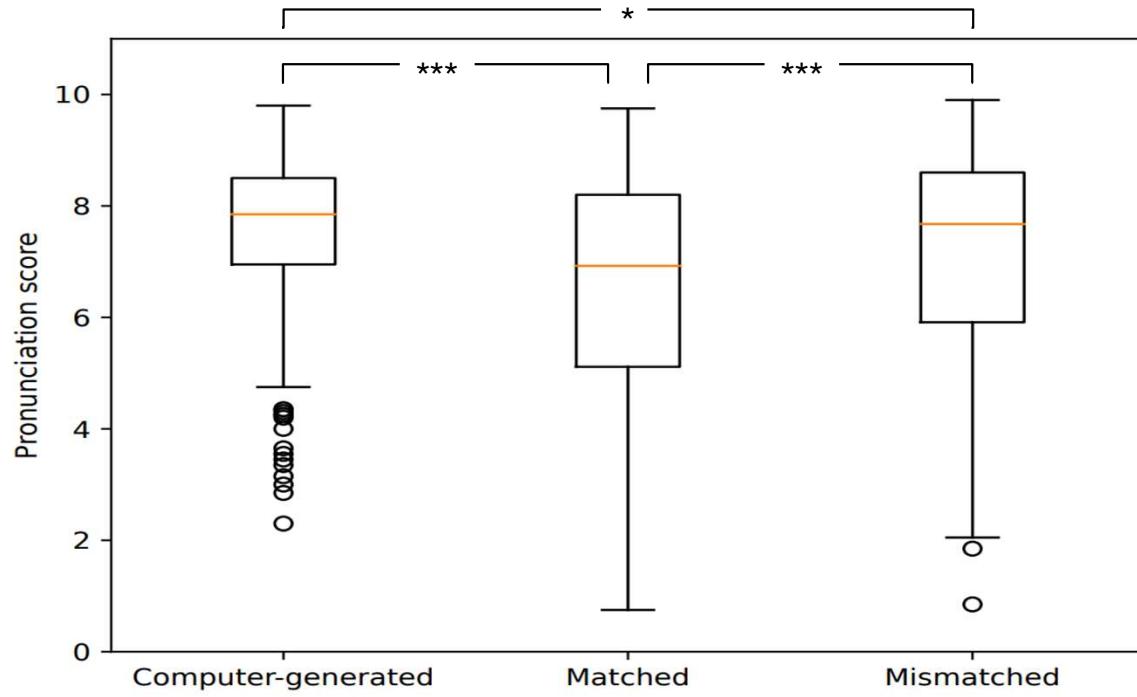


Mismatched
lip animation



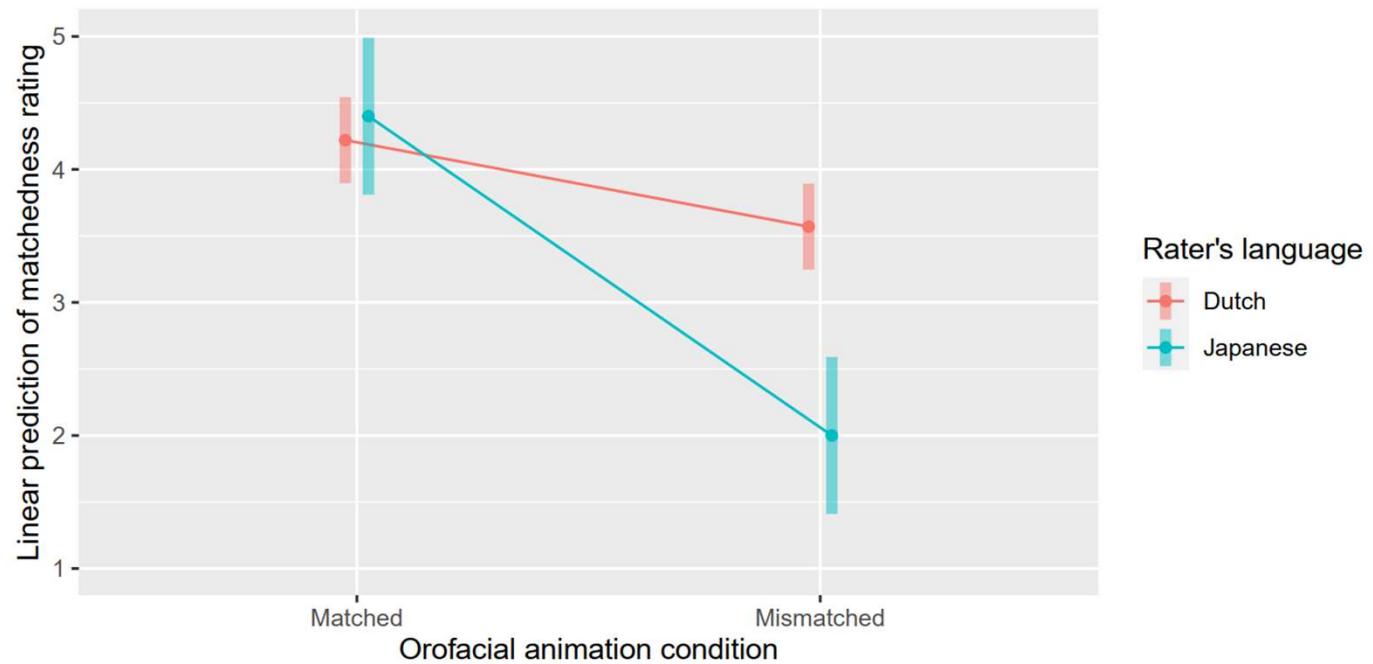
Results

Audiovisual speech does not aid pronunciation learning 😬



- * $p < 0,05$
- ** $p < 0,01$
- *** $p < 0,001$

What is going on?



Our participants did not notice the mismatched audiovisual speech.

Orofacial cues do not help learn a language that you do not master yet!



Building social robots

Midjourney
(a group of social robots)

VOL. LIX. No. 236.]

[October, 1950]

MIND
A QUARTERLY REVIEW
OF
PSYCHOLOGY AND PHILOSOPHY

I.—COMPUTING MACHINERY AND
INTELLIGENCE

By A. M. TURING

1. *The Imitation Game.*

I PROPOSE to consider the question, ‘Can machines think?’ This should begin with definitions of the meaning of the terms ‘machine’ and ‘think’. The definitions might be framed so as to reflect so far as possible the normal use of the words, but this attitude is dangerous. If the meaning of the words ‘machine’ and ‘think’ are to be found by examining how they are commonly used it is difficult to escape the conclusion that the meaning and the answer to the question, ‘Can machines think?’ is to be sought in a statistical survey such as a Gallup poll. But this is absurd. Instead of attempting such a definition I shall replace the question by another, which is closely related to it and is expressed in relatively unambiguous words.

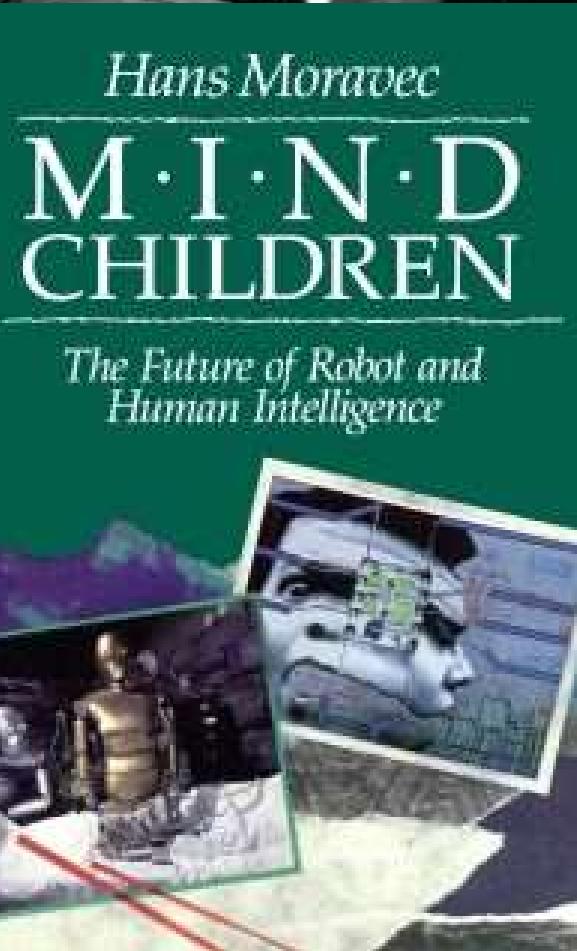
Out with the Wizard

Wizard of Oz approaches still amount for **50% of all research.**

Useful for a quick and cheap study or as a stub for underperforming technology.

But the goal is **autonomous human-robot interaction**





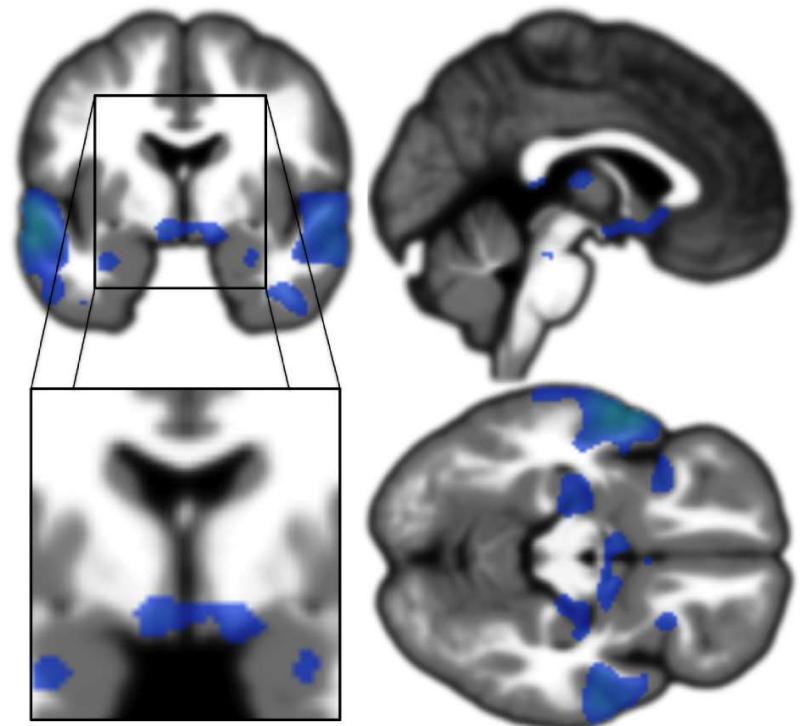
Moravec's Paradox

Building autonomous social robots is hard

Social interaction is **brain complete**.

All cognitive faculties are active and coordinated during social interaction.

Artificial social cognition needs artificial equivalents of these cognitive faculties and their interaction.

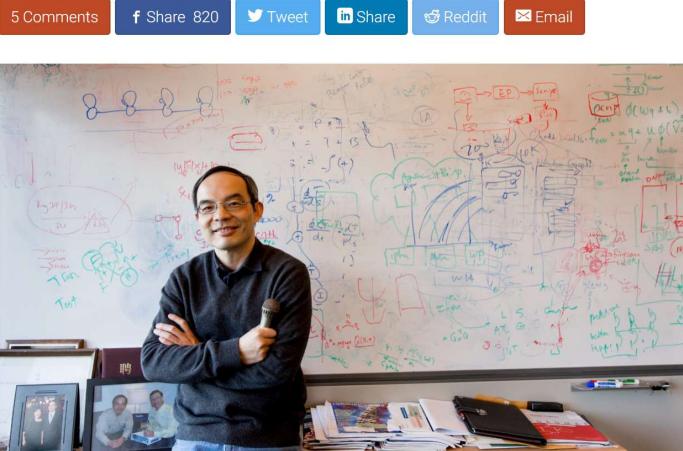


Rauchbauer, B., Nazarian, B., Bourhis, M., Ochs, M., Prévot, L., & Chaminade, T. (2019). Brain activity during reciprocal social interaction investigated using conversational robots as control condition. *Philosophical Transactions of the Royal Society B*, 374(1771), 20180033.

Speech recognition: super human?

Microsoft claims new speech recognition record, achieving a super-human 5.1% error rate

BY TODD BISHOP on August 20, 2017 at 7:44 pm



Xuedong Huang, a Microsoft Technical Fellow in AI and Research, leads Microsoft's Speech and Language Group. (Microsoft Photo)

5 Comments [f Share 820](#) [Tweet](#) [Share](#) [Reddit](#) [Email](#)

Cloud Tech Summit: Tickets on sale here!

GeekWire Newsletters

Subscribe to GeekWire's free newsletters to catch every headline

Enter your email address [Subscribe](#)

Send Us a Tip

Have a scoop that you'd like GeekWire to cover? Let us know.

[Send Us a Tip](#)

IBM speech recognition is on the verge of super-human accuracy

 **Chris Weller**
Mar. 9, 2017, 4:58 PM [1,494](#)

[FACEBOOK](#) [LINKEDIN](#) [TWITTER](#) [EMAIL](#) [PRINT](#)

In the world of speech recognition software, 5.1% is kind of a magic number.

Companies that can create software with error rates falling in that ballpark are essentially matching the capabilities of humans, who miss roughly 5% of the words in a given conversation.



ASR for atypical populations



Youtube: [Amazon Alexa Gone Wild!!!](#)

Still problematic for minorities, especially for atypical populations



Once upon a time in 1997

Methodology

- Children's speech in a school setting in England.
- 11 children, average age $M=4.9$, $SD=0.3$; 5F/6M
- Three kinds of utterances
 - Words ("one", "two", "three", ...)
 - Simple sentences ("The horse is in the stable", ...)
 - Spontaneous speech
- Three recording devices
 - NAO (V5.0, running NaoQi V2.1.4).
 - Studio grade microphone (Rode NT1-A)
 - Portable audio recorded (Zoom H1)

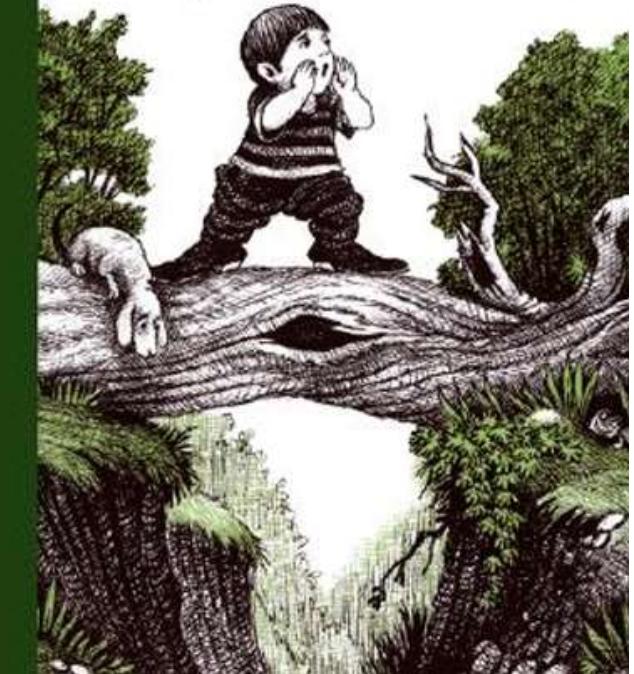


Counting



Spontaneous
speech

frog, where
are you? by mercer mayer



ASR performance for child speech

	Google M LD [95%CI]	Bing % rec.	Sphinx M LD [95%CI]	Nuance M LD [95%CI]
fixed (n=34)	0.34 [0.24,0.44]	11.8 [38]	0.64 [0.56,0.71] [0]	0.68 [0.64,0.73] [0]
spontaneous (n=222)	0.39 [0.36,0.43]	6.8 [17.6]	0.64 [0.61,0.67] [2.4]	0.80 [0.77,0.84] [0]
spontaneous clean only (n=83)	0.40 [0.35,0.45]	6.0 [16.9]	0.63 [0.58,0.68] [1.2]	0.78 [0.72,0.85] [0]

Only 6% to 12% of words recognised correctly

Kennedy, J., Lemaignan, S., Montassier, C., Lavalade, P., Irfan, B., Papadopoulos, F., ... & Belpaeme, T. (2017, March). Child speech recognition in human-robot interaction: evaluations and recommendations. In *Proceedings of the 2017 ACM/IEEE International Conference on Human-Robot Interaction* (pp. 82-90).

Fast forward to October 2023: OpenAI's Whisper 1550M model

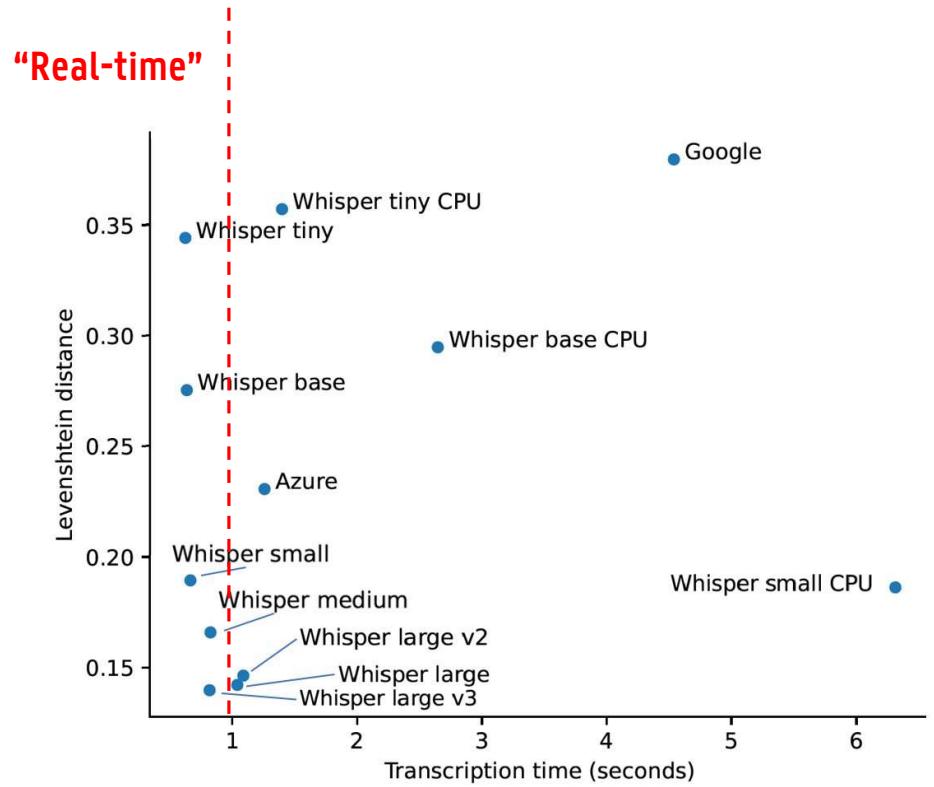
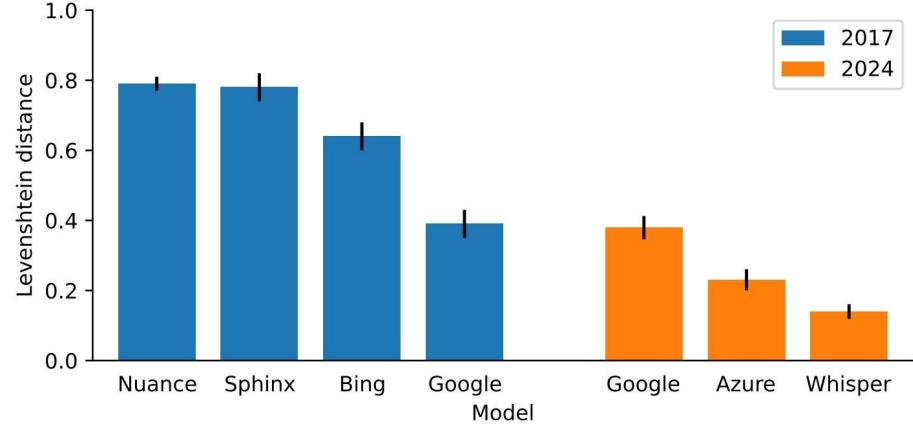


A boy looking at the frog. He's looking at it too. He's on the whole pack in the pond.
The frog is gone. The boy found a shoe. The dog fell out.



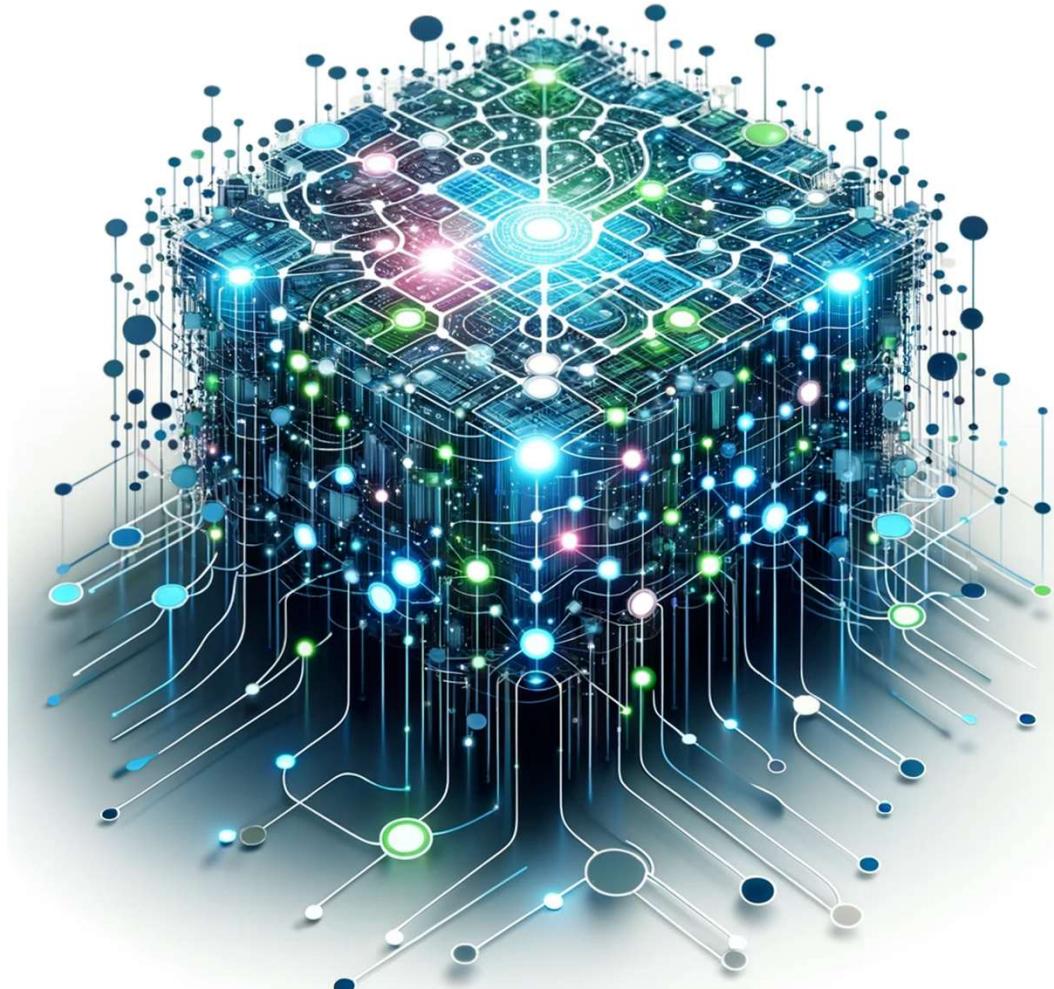
The dog is looking and the kid is watching. The dog is looking and the kid is watching.
The frog is getting out of the jar when the kid and the dog is asleep. The kid woke up,
so did the dog and he saw that the frog was gone. He was searching for him
everywhere but they couldn't find him.

Performance of Whisper



Large Language Models

- Unprecedented leap in quality of generative language performance
- Interactive and robust NLI.
- Adaptation through one shot/few shot prompting.
- Democratised access to LLMs.



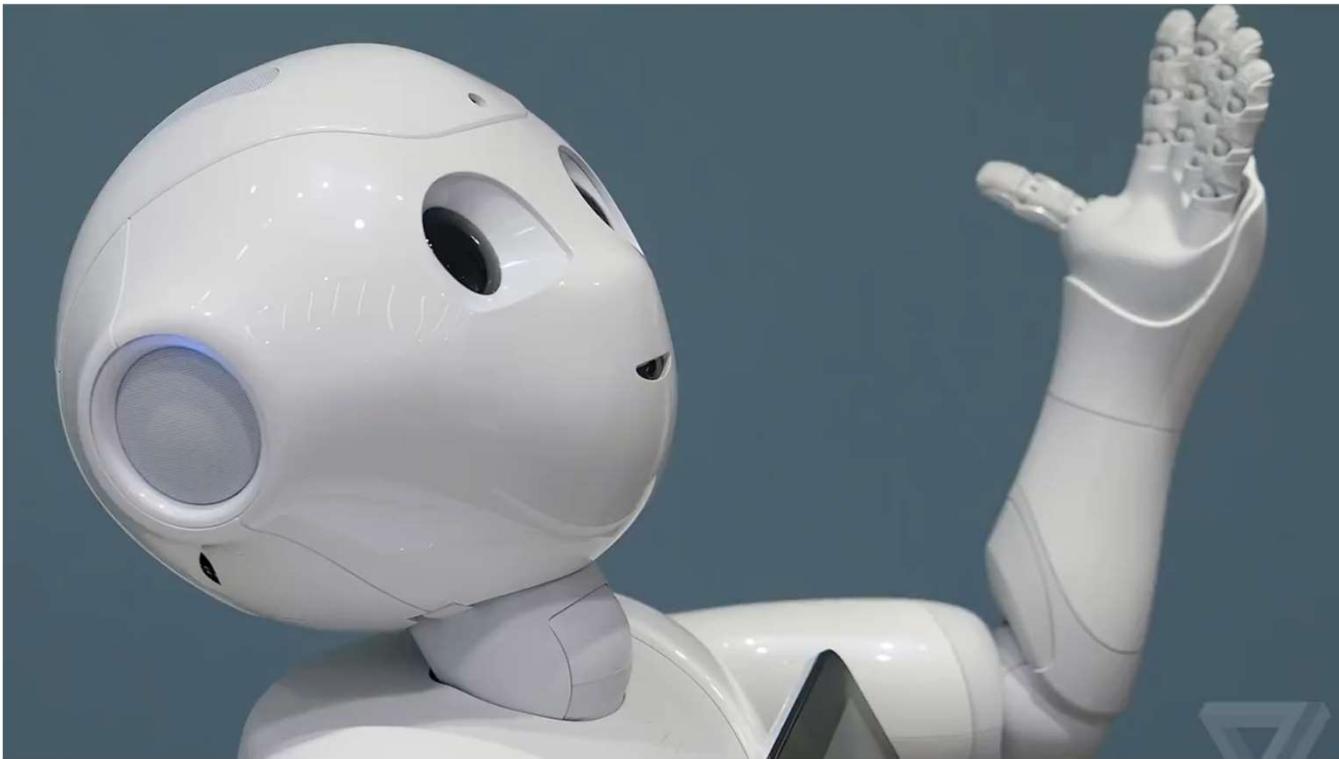
LLMs and robots

- Abilities needed for social robots
 - Open-domain dialogue capabilities
 - Storing state within a conversation
 - Can cover for poor performance by ASR
- Speed “good enough” for conversation
- Prompting to create robot persona
- Storing salient elements for next conversation

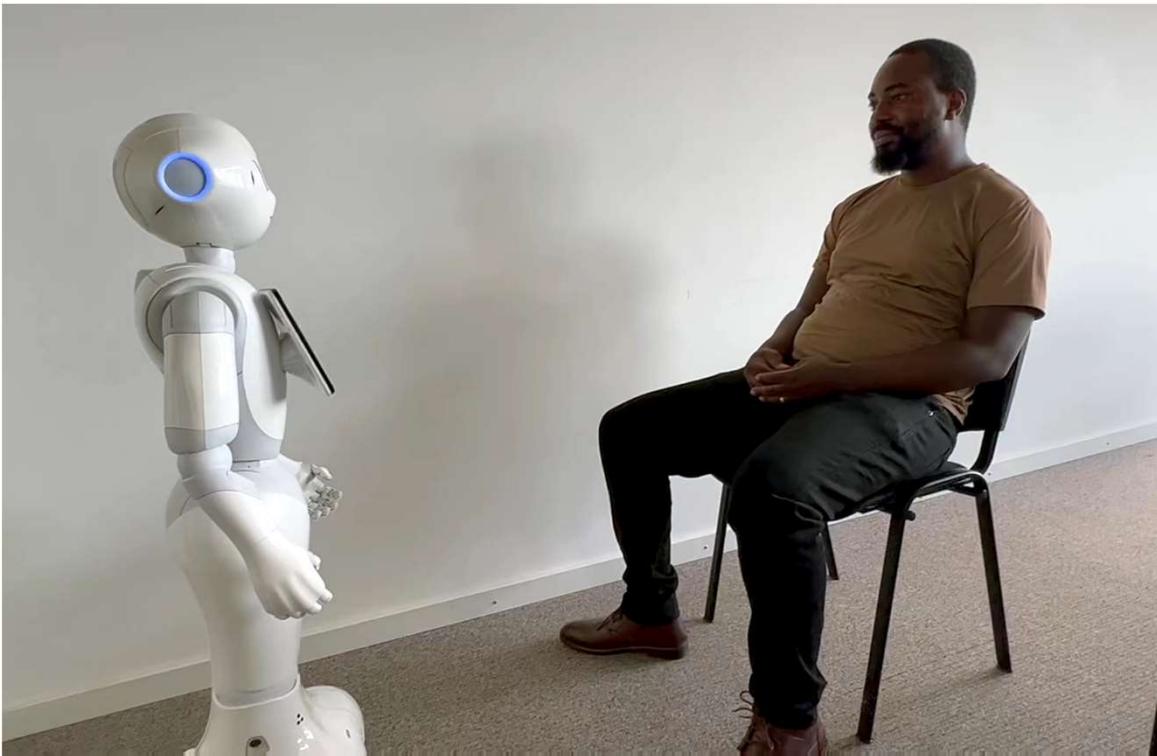
You are Pepper. Pepper is a helpful, creative, clever, empathic, friendly, and emotional robot. Now, you will have a natural and friendly conversation with your friend called {user_name}. The next information is about the last conversation and the information that you already know about your friend. Information: ###{last_conversation}###. The current date and time are {now.strftime('%m/%d/%Y %H:%M')} and the location where you are speaking is {location}. The conversation will be in {language}. Keep your responses in that language, brief, short and chatty.

Work by Maria Jose Pinto Bernal

Large Language Models for embodied interaction



Tanzanian Swahili: 0.0080% of ChatGPT training data



Source <https://commoncrawl.github.io/cc-crawl-statistics/plots/languages>

Multimodal interaction

LLMs incorporate **knowledge** about the physical and social world.

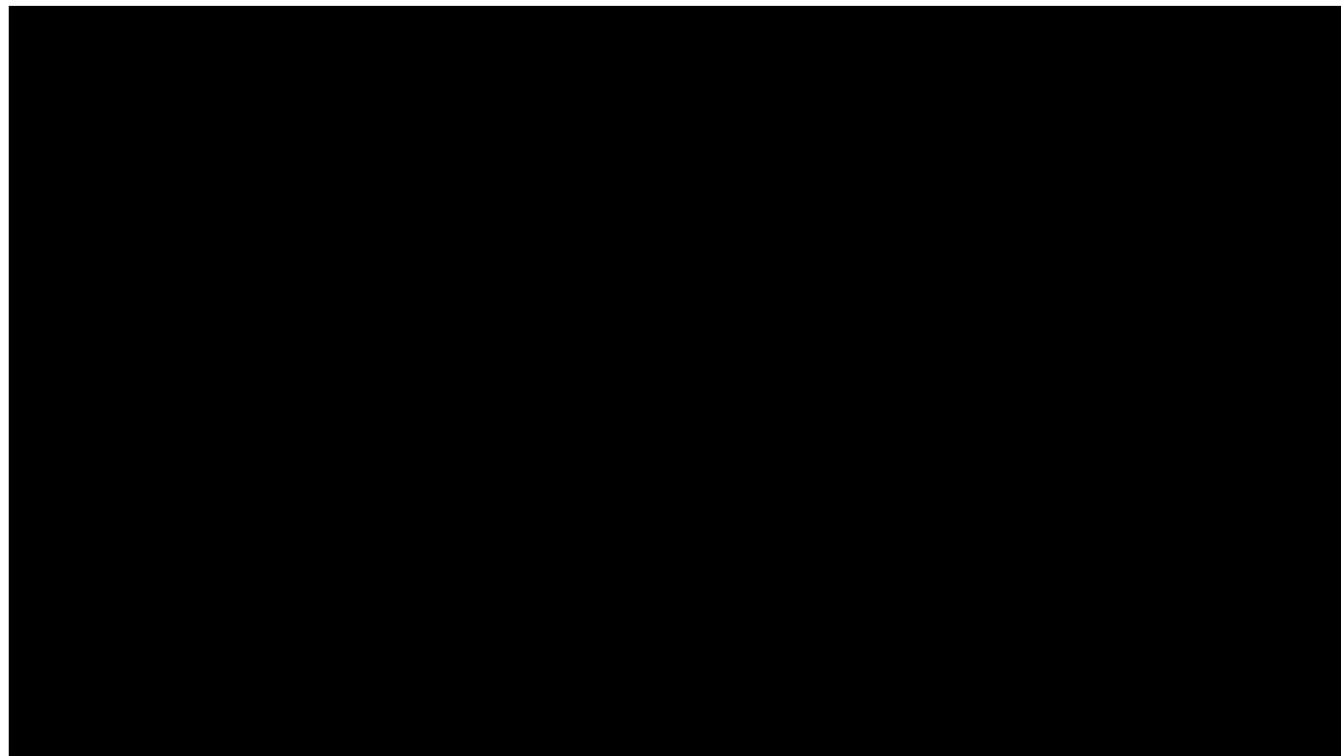
Even when LLMs are **disembodied**.

This is exactly the **world knowledge** we have been waiting for in AI.



Ahn, M., Brohan, A., Brown, N., Chebotar, Y., Cortes, O., David, B., ... & Yan, M. (2022). Do as I can, not as I say: Grounding language in robotic affordances. *arXiv preprint arXiv:2204.01691*.

Multimodal interaction with GPT4



Challenges

- Prompt engineering turns into an art form
- Integrating across modalities
 - The sequential nature of processing (TTS, LLM, ASR) is not the right way.
 - Dynamics and timing
- Integrating with services
- Computational effort needed
 - But the expense of running multimodal models is coming down – Llama 3
- Different embodiments and physical capabilities
 - “That is really great. I love swimming in the sea too.”
- ...

“Cognitive skills” of LLMs

- Understanding of physical and social world
- (Good enough) Understanding of causality
- Reasoning by analogy
- Creativity
- Theory of Mind
- Aligned to human values
- ...

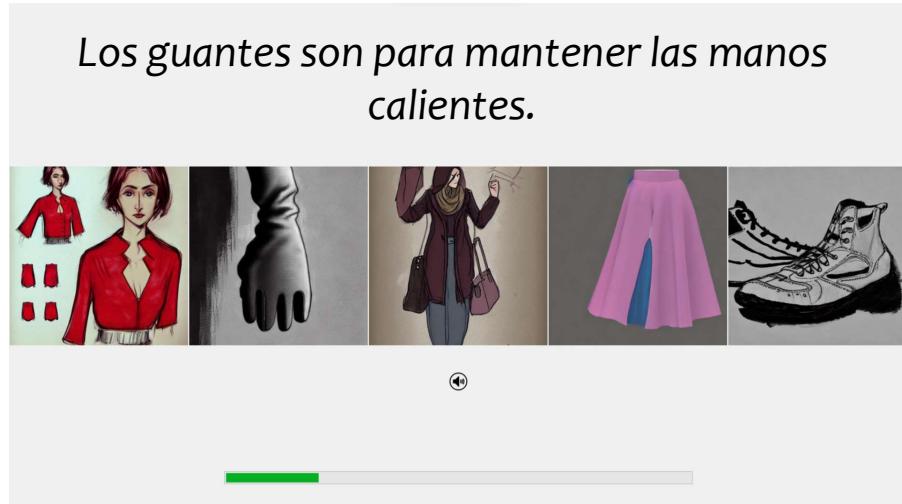
Note: all on a surface level, but LLMs pass the “duck test” for many cognitive skills.



Wikimedia

Generative AI for language education

LLM and Stable Diffusion to generate adapted learning material on the fly.



Verhelst, E., Janssens, R., Demeester, T., & Belpaeme, T. (2024, March). Adaptive Second Language Tutoring Using Generative AI and a Social Robot. In Companion of the 2024 ACM/IEEE International Conference on Human-Robot Interaction (pp. 1080-1084).



Are the social robots coming?

For the first time AI has broken through
Moravec's paradox.

Social interaction while being a hard technical problem is no longer an impossible technical problem



Wrap up

- Learning is socially gated and “social robots” are likely to play a role in our future learning.
- Artificial Intelligence is for the first time breaking through Marovc’s paradox.
- Social AI will lead to a learning technology revolution.



VALAWAI

fwo Research Foundation Flanders
Opening new horizons



Thank you!

Giulio Antonio Abbo

Madeleine Bartlett

Paul Baxter

Ruben Janssens

James Kennedy

Séverin Lemaignan

Qiaoqiao Ren

Edger Rutatola

Emmanuel Senft

Pieter Wolfert

Serge Thill

Eva Verhelst

A photograph of a traditional suspension bridge made of wood and ropes, spanning a river. The bridge has a central wooden tower and is supported by ropes from trees on both banks. The river water is clear and greenish-blue. In the background, there are rocky banks and some greenery.

1. narrow interactions

Teleoperation, aka **Wizard of Oz**, in HRI research makes up

50% of all studies

Useful for a quick and cheap study or as a stub for underperforming technology



Baxter, P et al. (2016). From characterising three years of HRI to methodology and reporting recommendations. In *2016 11th ACM/IEEE International Conference on Human-Robot Interaction (HRI)* (pp. 391-398). IEEE.

Programming social behaviour by hand is **hard and time consuming**

Limited interaction scope

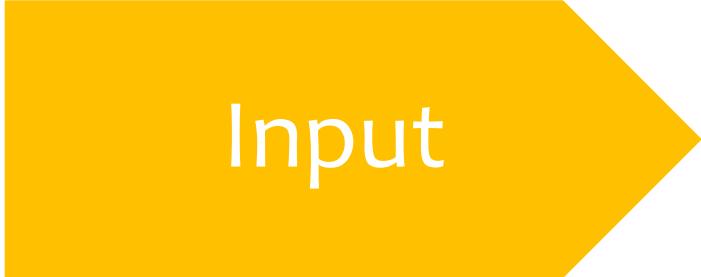
Not long-term: most interactions are well under 20 minutes

Commercial social robots **underwhelm**









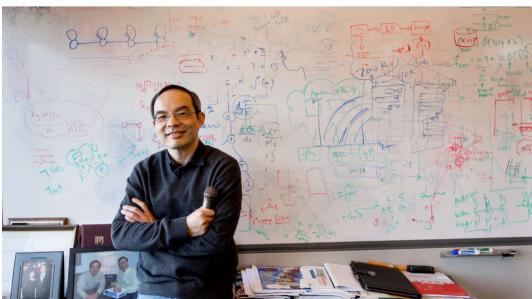
Input

Speech recognition: super human?

Microsoft claims new speech recognition record, achieving a super-human 5.1% error rate

BY TODD BISHOP on August 20, 2017 at 7:44 pm

5 Comments [f Share 820](#) [Tweet](#) [Share](#) [Reddit](#) [Email](#)



Xuedong Huang, a Microsoft Technical Fellow in AI and Research, leads Microsoft's Speech and Language Group. (Microsoft Photo)

Cloud Tech Summit: Tickets on sale here!

GeekWire Newsletters

Subscribe to GeekWire's free newsletters to catch every headline

Enter your email address

Send Us a Tip

Have a scoop that you'd like GeekWire to cover? Let us know.

IBM speech recognition is on the verge of super-human accuracy

 **Chris Weller**
Mar. 9, 2017, 4:58 PM  1,494

[FACEBOOK](#) [LINKEDIN](#) [TWITTER](#) [EMAIL](#) [PRINT](#)

In the world of speech recognition software, 5.1% is kind of a magic number.

Companies that can create software with error rates falling in that ballpark are essentially matching the capabilities of humans, who miss roughly 5% of the words in a given conversation.





ASR performance for child speech

	Google	Bing	Sphinx	Nuance
	M LD [95%CI]	% rec.	M LD [95%CI]	% rec.
fixed (n=24)	0.34 [0.24,0.44]	1.8 [38]	0.64 [0.56,0.71]	0 [0]
spontaneous (n=222)	0.39 [0.36,0.43]	6.8 []	0.64 [0.61,0.67]	0.5 []
spontaneous clean only (n=83)	0.40 [0.35,0.45]	6.0 []	0.63 [0.58,0.68]	1.2 []



6% to 38% words correctly recognised

Range of recognition success of Google ASR engine with no grammatical constraints. **Google is the best performing engine (compared to Sphinx, Bing, and Nuance).**

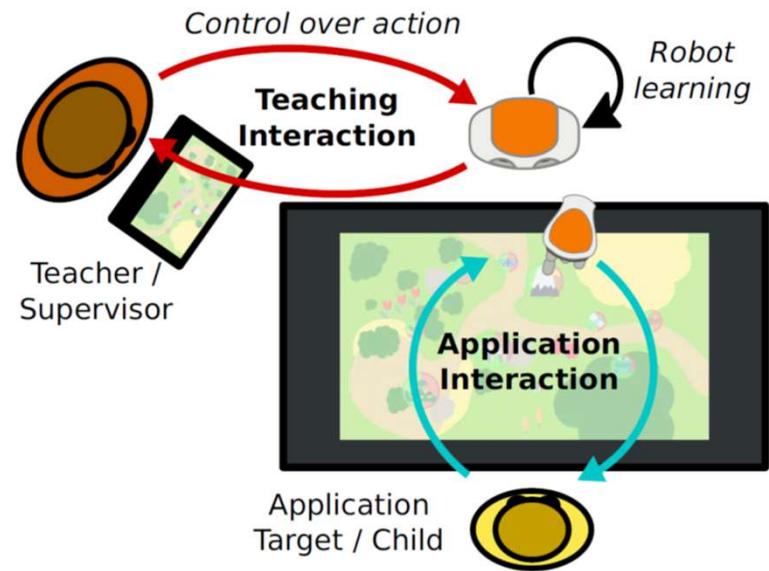
Kennedy, J., Lemaignan, S., Montassier, C., Lavalade, P., Irfan, B., Papadopoulos, F., ... & Belpaeme, T. (2017, March). Child speech recognition in human-robot interaction: evaluations and recommendations. In *Proceedings of the 2017 ACM/IEEE International Conference on Human-Robot Interaction* (pp. 82-90).



Action
selection

Data-driven action selection

- Learning for robots
 - Reinforcement Learning requires exploration and mistakes to be cheap.
 - Learning from Demonstration better suited for kinesthetic teaching.
- Needed
 - Online learning
 - Active involvement of the expert during learning.
 - Understanding of what the AI learns.
 - Retaining control over the AI.
- Solution is **SPARC: supervised progressively autonomous robot competencies.**

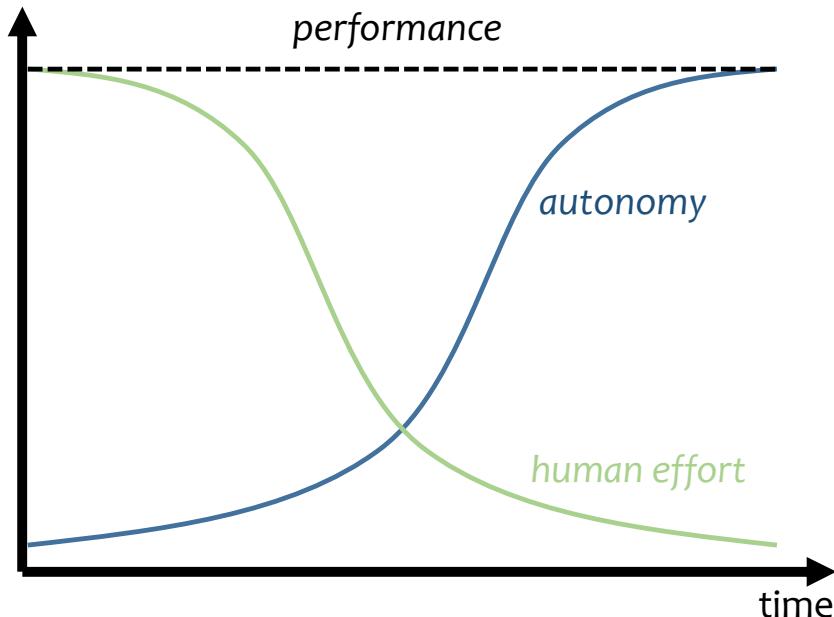


Senft, E., Lemaignan, S., Baxter, P. E., Bartlett, M., & Belpaeme, T. (2019). Teaching robots social autonomy₁₁₃ from in situ human guidance. *Science Robotics*, 4(35).



SPARC: Core concepts

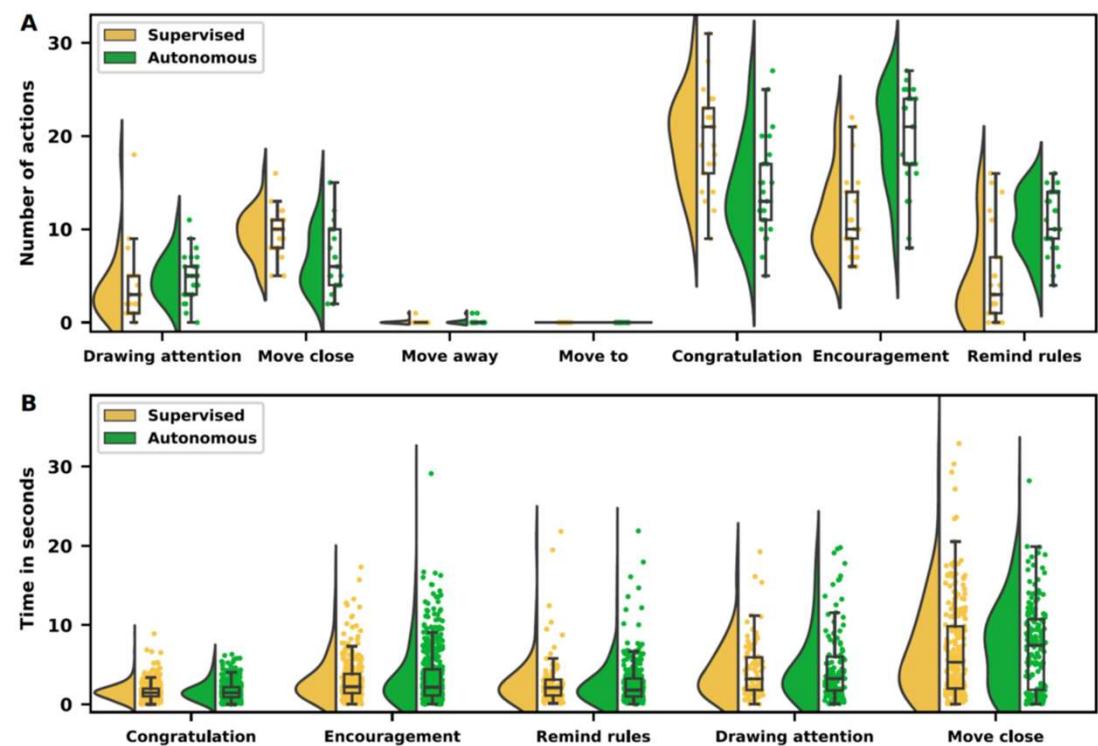
Robot progressively learns appropriate autonomous behavior from in situ human demonstrations and guidance.



Senft, E., Lemaignan, S., Baxter, P. E., Bartlett, M., & Belpaeme, T. (2019). Teaching robots social autonomy₁₁₅ from in situ human guidance. *Science Robotics*, 4(35).

Results

- Three conditions
 - SPARC robot
 - Autonomous robot
 - Passive robot (control condition)
- 75 children ($n = 75$; age: $M = 9.4$, $SD = 0.71$; 37 female), each of the three conditions was allocated 25 children.
- Challenging learning domain
 - 210 input dimensions and output action space of 655 actions.



Senft, E., Lemaignan, S., Baxter, P. E., Bartlett, M., & Belpaeme, T. (2019). Teaching robots social autonomy from in situ human guidance. *Science Robotics*, 4(35). ¹¹⁶



Output

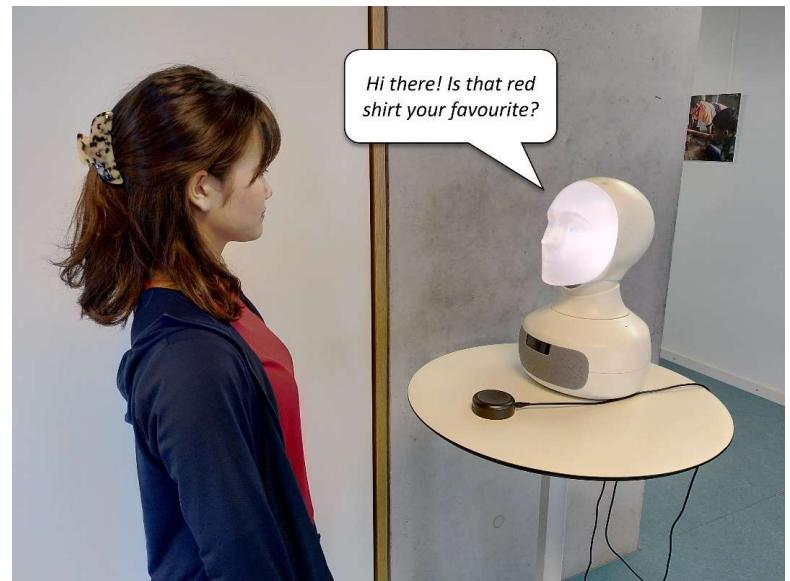
Multi-modal interaction

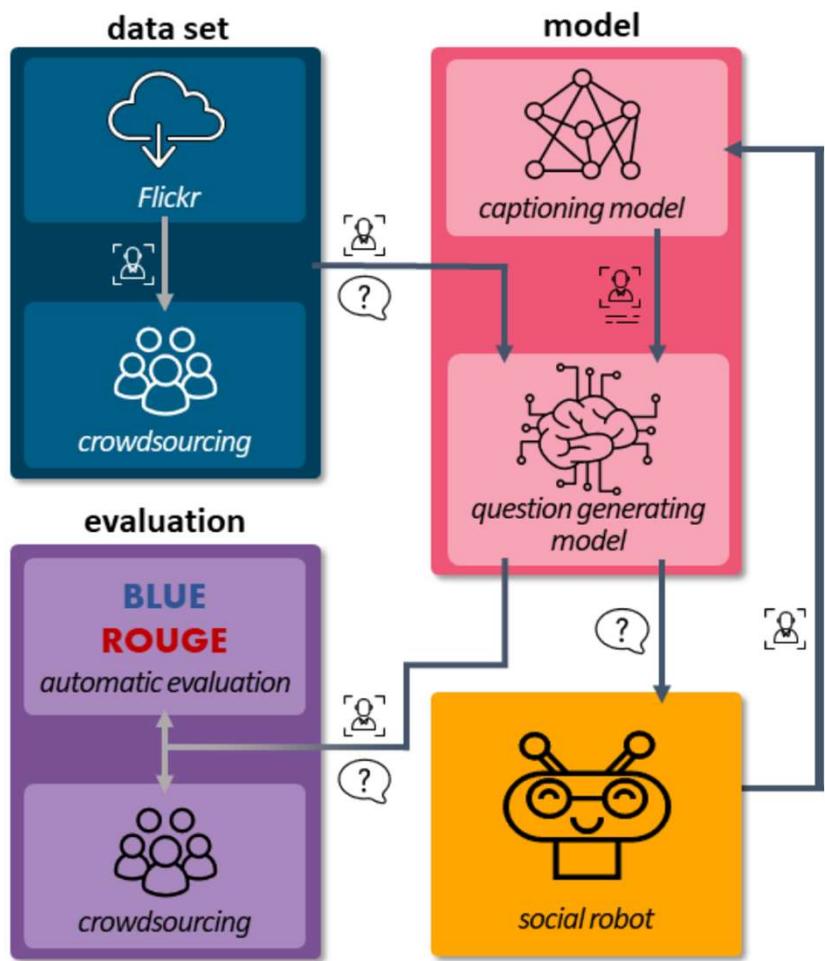
Current **chatbots** are uni-modal

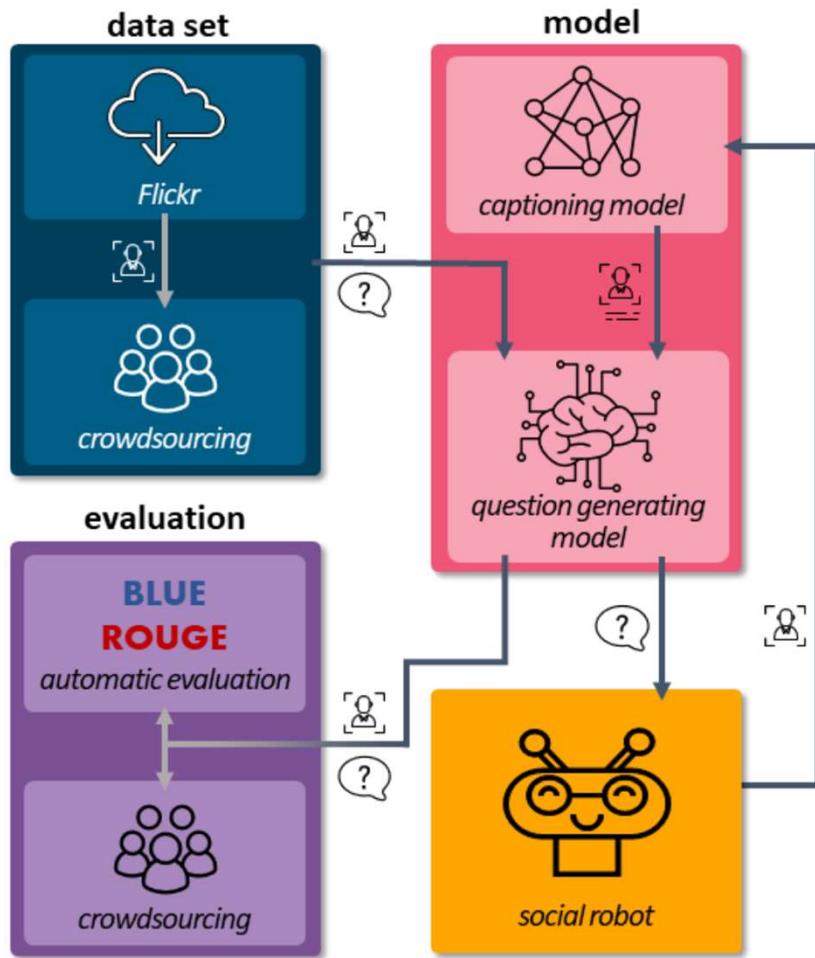
Phatic communication or small talk

All good conversation starts with a
conversation starter.

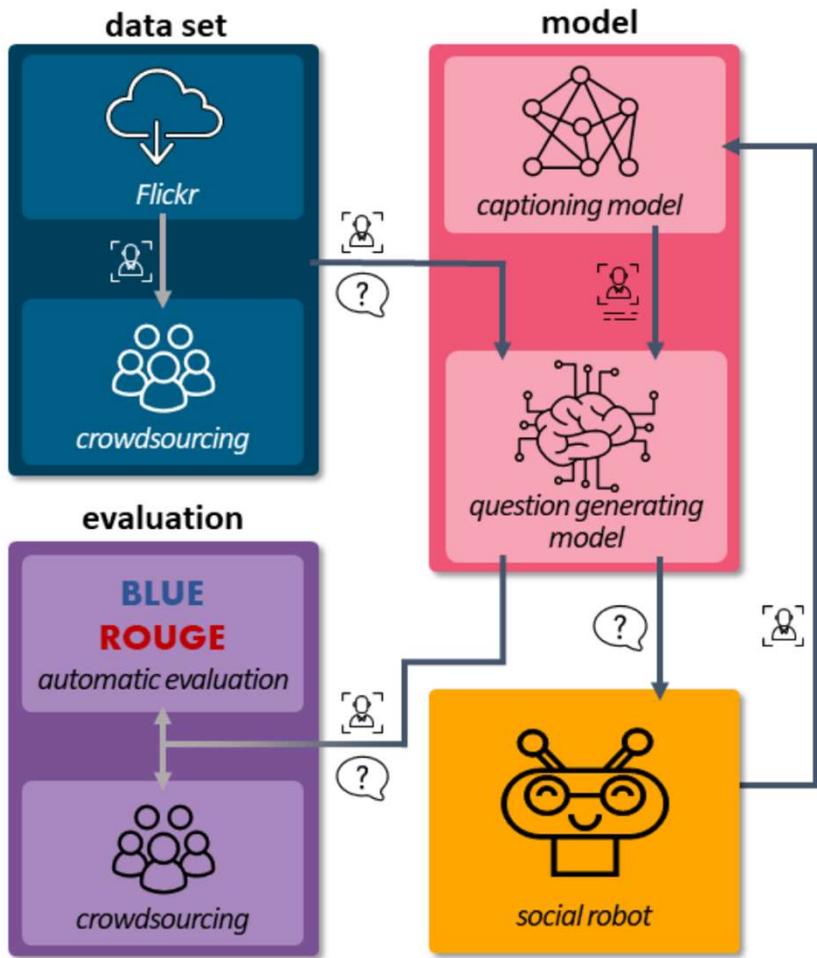
Can these be generated?







Have you always had freckles? Were you born with red hair? Is white your favourite colour? Is that cool hat typical of your country? Do your parents also have blue eyes? Where did you get that scarf? Do you go to bars often? What kind of drinks do you like? Where did you get that necklace? Do you put the braids in your hair?

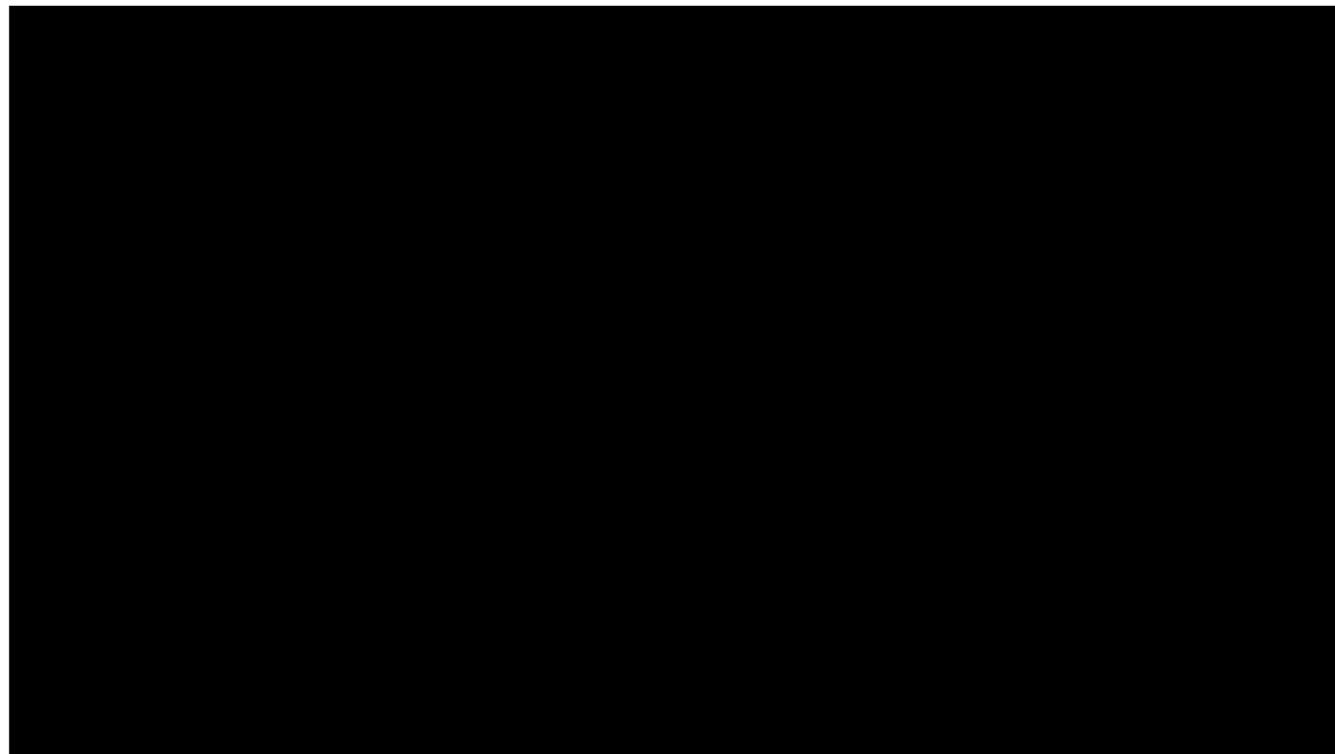


man with brown hair. the eyes are brown. the man has a beard. the man has short hair. the man is smiling. the man is wearing a black shirt. the man is wearing a necklace. the shirt is white. white letters on shirt.

Models

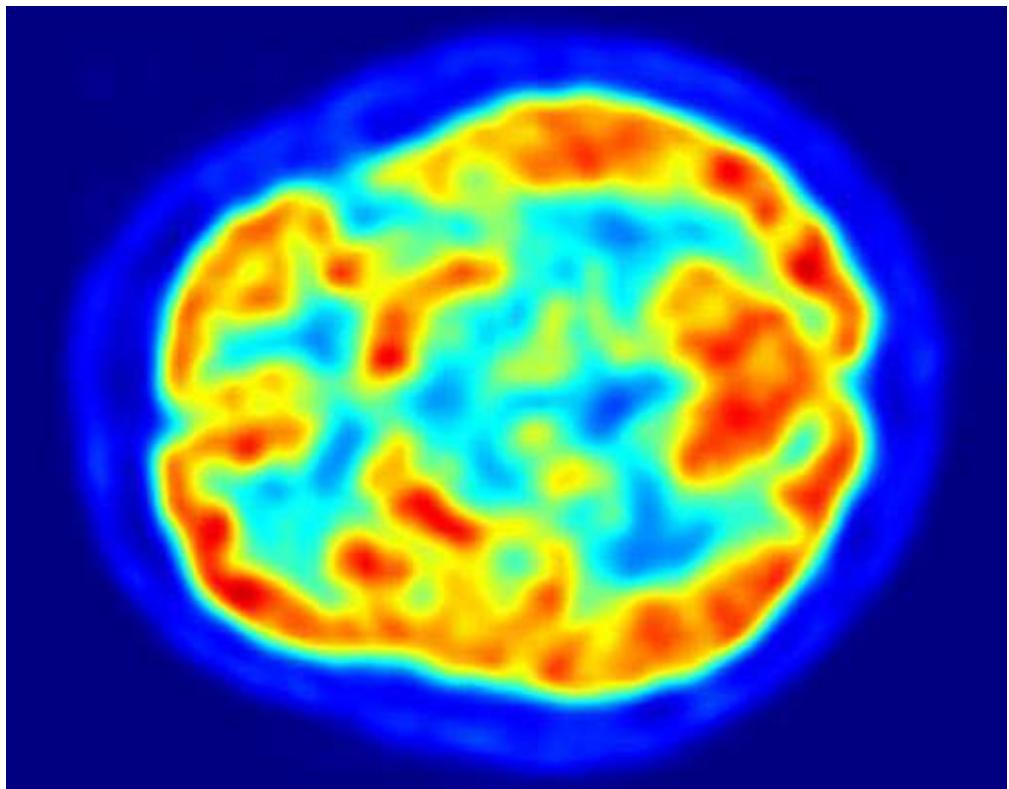
- Dense captioning network
- Neural generative model: BART
- Standard retrieval model: TF-IDF
- Neural retrieval model: Transformer-based Bi-Encoder

Demonstration



A call to arms for AI

- Interaction recruits all our cognitive faculties
- An effort by the entire AI department
- People and robots meet halfway
- Even simple systems can make a large impact



Sci-news.com