

My background:

Human-Chatbot Interaction Design

How to design chatbots?

Supported by IKTPLUSS, Research Council of Norway



Chatbots og interaksjonsdesign

Naturlig språk blir en stadig vanligere måte å bruke digitale tjenester på. Eksempler er Apples Siri og Googles Assistant. Facebook Messenger gir i dag tilgang til mer enn 30.000 chatbots.

Til tross for en rivende teknologiske utvikling innen chatbots gjenstår mange forskningsutordringer. I prosjektet **Human-Chatbot Interaction Design** kombinerer vi interaksjonsdesign med kunstig intelligens og maskinlæring.

Målet er å bidra til kunnskap om brukersentrert design av chatbots, slik at disse kan bli nyttige og engasjerende innganger til informasjon og tjenester – uansett brukerens kjønn, bakgrunn og alder.

OM PROSJEKTET

Prosjektperiode 2017-2021

Program IKTPLUSS

PARTNERE





STØTTET AV



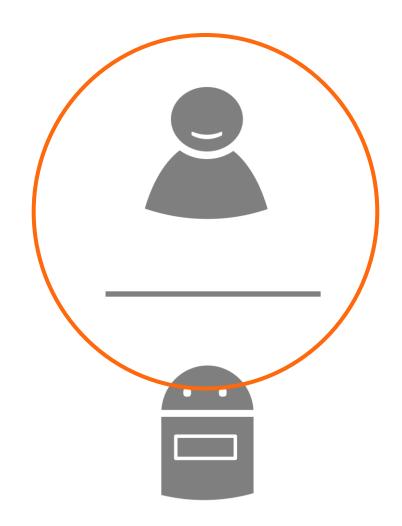
Kontakt: Prosjektleder Asbjørn Følstad | 93293013 | asf@sintef.no

Les mer: www.sintef.no/humanchatbotinteractiondesign

My background:

SINTEF, HCl group

Psychology (!?!)



Interaction with AI – module 2

Interaction design

Four sessions

Design of interaction with AI

Asbjørn Følstad

September 22

October 6

Understanding interaction with Al

Morten Goodwin

October 13

October 20



Literature



Amershi, S., Weld, D., Vorvoreanu, M., Fourney, A., Nushi, B., Collisson, P., ... & Teevan, J. (2019). Guidelines for human-Al interaction. In Proceedings of the 2019 CHI Conference on Human Factors in Computing Systems (paper no. 3). ACM.

Kocielnik, R., Amershi, S., & Bennett, P. N. (2019). Will You Accept an Imperfect AI?: Exploring Designs for Adjusting End-user Expectations of AI Systems. In Proceedings of the 2019 CHI Conference on Human Factors in Computing Systems (paper no. 411). ACM.

Liao, Q. V., Gruen, D., & Miller, S. (2020). Questioning the AI: Informing Design Practices for Explainable AI User Experiences. In Proceedings of the 2020 CHI Conference on Human Factors in Computing Systems (paper no. 463). ACM.

Yang, Q., Steinfeld, A., Rosé, C., & Zimmerman, J. (2020). Re-examining Whether, Why, and How Human-Al Interaction Is Uniquely Difficult to Design. In Proceedings of the 2020 CHI conference on human factors in computing systems (Paper no. 164).

Literature

Følstad, A., & Brandtzæg, P. B. (2017). Chatbots and the new world of HCI. interactions, 24(4), 38-42.

Luger, E., & Sellen, A. (2016). Like having a really bad PA: the gulf between user expectation and experience of conversational agents. In *Proceedings of the 2016 CHI Conference on Human Factors in Computing Systems* (pp. 5286-5297). ACM.

Frey, C. B., & Osborne, M. A. (2017). The future of employment: how susceptible are jobs to computerisation?. *Technological forecasting and social change*, 114, 254-280.

Carter, S., & Nielsen, M. (2017). Using artificial intelligence to augment human intelligence. *Distill*, 2(12), e9.

Wilson, H. J., & Daugherty, P. R. (2018). Collaborative intelligence: humans and AI are joining forces. *Harvard Business Review*, 96(4), 114-123.

Noessel, C. (2017). Designing Agentive technology: Al that works for people. Rosenfeld Media.

Hall, E. (2018). Conversational design. A Book Apart

McAfee, A., & Brynjulfsson, E. (2016). Machine, Platform, Crowd. Harnessing Our Digital Future. Norton & Company.

Individual assignment

Building on and extending the individual assignment in Module 1. Startup today – finish October 23)

Mid-term report (Oct. 23)

Group assignment

Building on and extending the group assignment in Module 1. Startup today – finish October 23)

- + Task on user and interaction design perspective (startup today finish October 6, include in mid-term report as Appendix 1)
- + Task on machine learning perspective (startup Oct. 13 finish Oct. 20, include in mid-term report as Appendix 2)

Midterm report - individual assignment

Three topics:

- Characteristics of Al-infused systems.
- Human-Al interaction design.
- Chatbots / conversational user interfaces.

Language: English or Norwegian.

Max. pages: 6

Min. articles referenced 4.



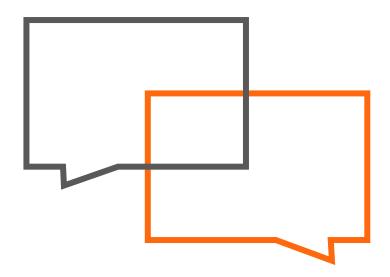
Midterm report – group assignment

Content – 5-7 pages

- A description of the group, who you are names.
- A description of what area of "interaction with AI" you are interested in working with.
- (new) Background section: Position your work relative to existing knowledge and practice
- Minimum 1 maximum 2 questions that you want to address.
 Please write some sentences about the questions. These questions can change and evolve later in the midterm report and in the final report as you go about investigating your questions.
- **(updated)** Method section overall approach, design process (optional, but encouraged), data collection methods
- (new) Sketches and/or prototypes (optional, but encouraged)
- (new) Findings (progress, initial outcomes)
- (updated) Minimum five references to literature.

Appendices – approx. 1 page each

- Appendix 1: Chatbot design task briefly describe the process and outcome. Detail reflections and lessons learnt.
- Appendix 2: Machine learning task briefly describe the process and outcome. Detail reflections and lessons learnt.



Brief status on the group task – each group say a few words

Group 1

- 1. Can a human establish an emotional bond with an Al chatbot?
- 2. To what extent could an AI chatbot replace a human psychologist?

Group 2

1. How can a chatbot be influenced by people's behaviour?

Group 3

- 1. Could AI-powered speech recognition contribute to older adults' communication with family and friends? And possibly contribute to their autonomy?
- 2. Additionally, speech recognition is not perfect. How are these incomprehensive handled by systems and users today? How should/could errors be handled?

Group 4

- 1. Interagerer folk forskjellig med chatbotten når den har et menneske som bilde, enn når den er en tydelig chatbot? Stoler man mer på informasjonen hvis man tror det er et ekte menneske man snakker med?
- 2. Hva skal til for at en person reagerer over det chatbotten sier og hvordan påvirker dette videre interaksjon? Hva blir atferden etter man har chattet?

Group 5

- 1. What are the expectations of how an AI should behave?
- 2. Is it possible to make an AI more human in the way it acts?

Group 6

- 1. How does the speech pattern differ from human to human interaction?
- 2. How can design make people understand the capability of their personal assistant?

Group assignment – suggestion (I):

Improve existing design of Al-infused system

Wonder document

Interation design

Evaluation and trust

Identify interesting, existing Alinfused system

Use (selected) guidelines for Human-Al Interaction to discuss interaction design. Identify good design and possible improvements

Suggest and prototype redesign in response to discussion

Test with users and refine

Group assignment – suggestion (II):

Develop early prototype of simple Al-infused system

Wonder document

Interation design

Evaluation and trust

Identify need for simple AI-infused system (e.g. a chatbot?)

Concept development and prototype

Use (selected) guidelines for Human-Al Interaction to motivate design choices.

Test with users and refine

Group assignment – suggestion (III):

Understand use of Al-infused systems and discuss guidelines

Wonder document

Interation design

Evaluation and trust

Identify one or more interesting, existing Al-infused systems

Set up study to assess or better understand how the system is used or experienced

Discuss theoretical and practical implications of findings

Reflect on correspondance between findings and guidelines for Human-Al Interaction. Critical discussion and possible suggestions

Agenda

Today

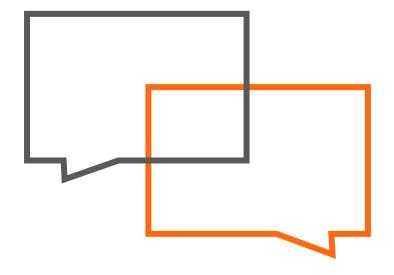
Next

- 1 Interacting with AI an overview
- Chatbots interacting with AI in natural language

- User-centred design of Al
- 4 User-centred design of chatbots

Interaction with AI - overview

A1?



AI?



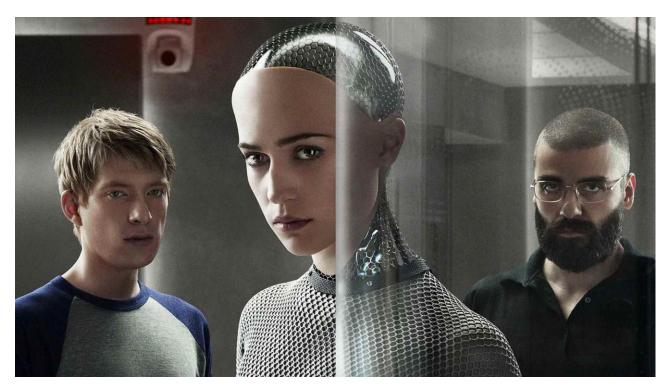


Devs (2020)

Artificial super intelligence

Artificial general intelligence

Artificial narrow intelligence

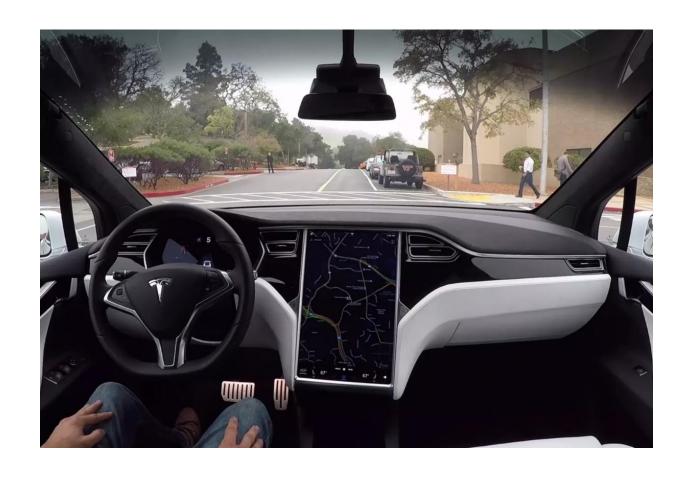


Ex Machina (2014)

Artificial super intelligence

Artificial general intelligence

Artificial narrow intelligence



Artificial super intelligence

Artificial general intelligence

Artificial narrow intelligence

Interaction with AI, and designing for interactions with AI, concerns artificial narrow intelligence (narrow AI)

Artificial super intelligence

Artificial general intelligence

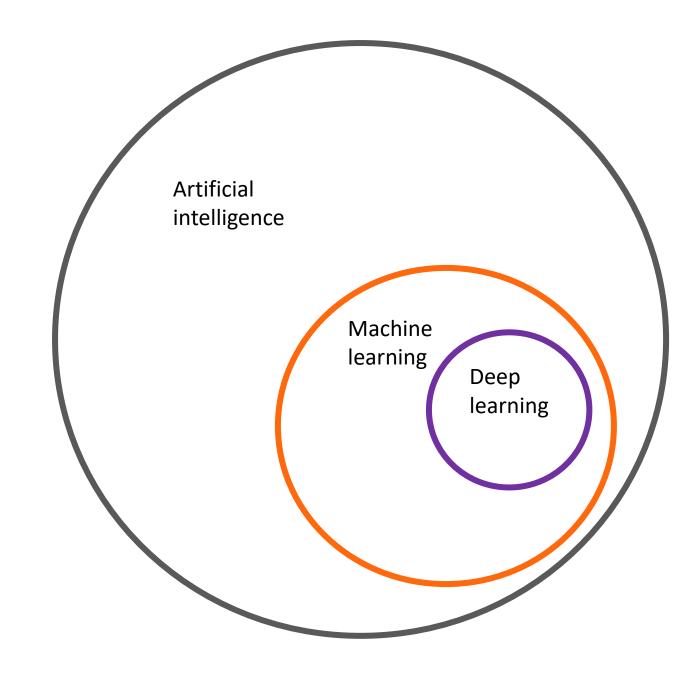
Artificial narrow intelligence

Interaction with AI, and designing for interactions with AI, concerns artificial narrow intelligence (narrow AI)

Need to know:

- Characteristics of narrow AI
- Opportunities and limitations of narrow Al

... from now on, when talking about Al we mean narrow Al



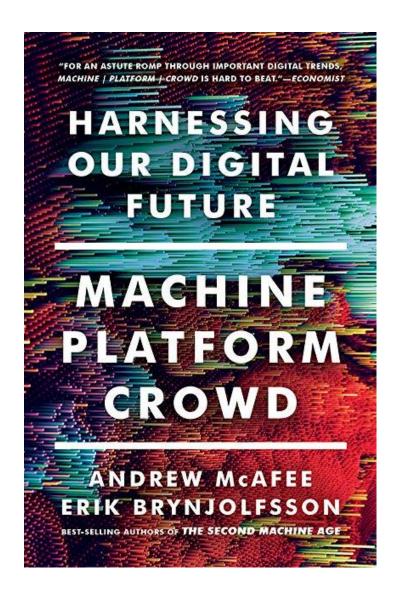
Two drivers:

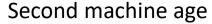
Large datasets

Computational power

An understanding of recent Al systems may be:

Computer systems
learning and improving
on the basis of large data
sources



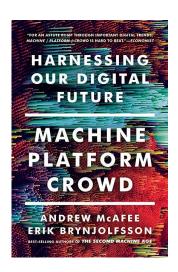


- Phase 1: Computers drives productivity
- Phase 2: Computers can do work we never thought of as programmed or routine

Michael Polanyi on tacit knowledge - "We know more than we can tell." (Polanyis paradox)

Computers now taking on tasks reflecting Polanyis paradox

The board game Go paramount example



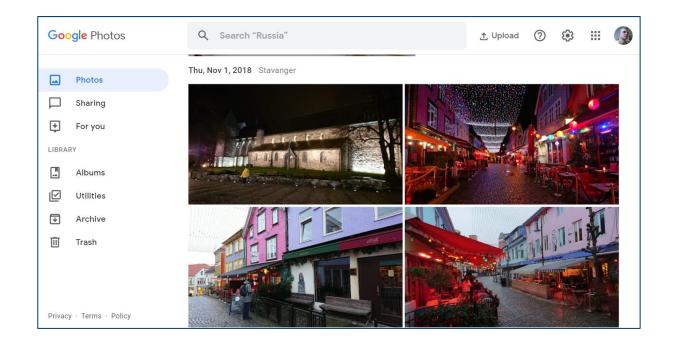


De fleste datasett som kan brukes til å trene dype nett til å bli gode på en eller annen funksjon, er kjempestore.

[...]

Et nytt triks som mange bedrifter nå bruker, er å først designe en tjeneste som mange brukere vil ha. [...] Gjennom bruken av tjenesten gir brukerne fra seg verdifulle data som igjen brukes til å trene nett for å gi brukerne nye funksjoner de setter pris på.

Bjørkeng, P. K. (2018). Kunstig intelligens – den usynlige revolusjonen. Vega.

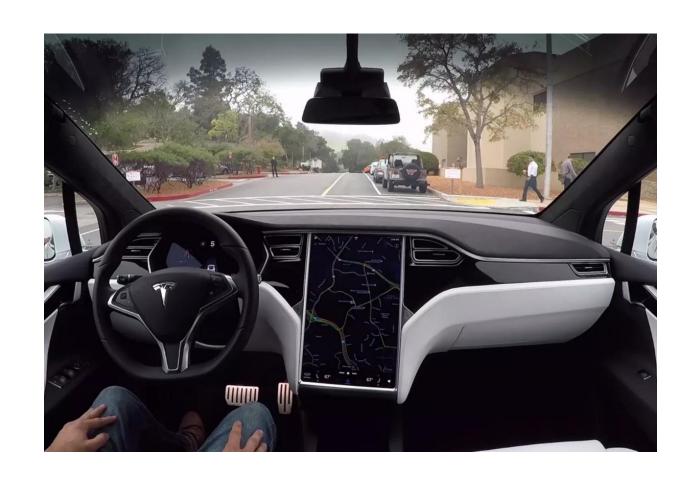


En datafelle er begrepet som brukes om Teslas særegne tilnærming til datainnsamling.

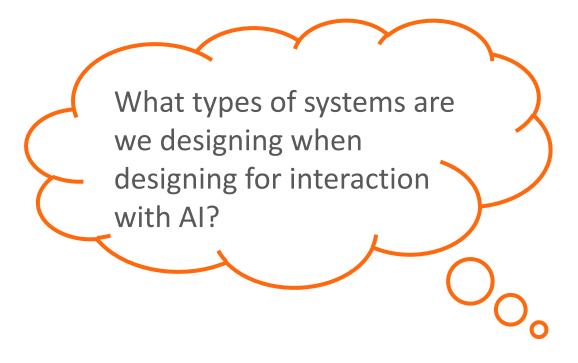
Google-eier Alphabet og deres selskap Waymo har nå brukt ti år bare på å samle inn data om mange nok ulike trafikksituasjoner. (menneskelige sjåfører i selvkjørende biler klare til å ta over)

Elon Musk og Tesla hadde slett ikke 10 år til overs. I stedet etablerte de en datafelle. [...] Her eneste nye Tesla-eier er nå med i dette gigantiske datainnsamlingsprosjektet (shadowmode)

Bjørkeng, P. K. (2018). Kunstig intelligens – den usynlige revolusjonen. Vega.



Al-infused systems



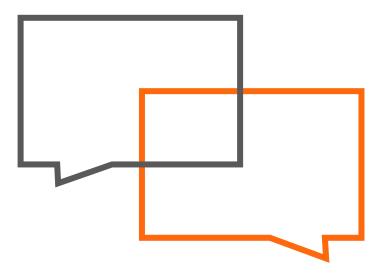
Al-infused systems

"Systems that have features harnessing AI capabilities that are directly exposed to the end user."

Amershi, S., Weld, D., Vorvoreanu, M., Fourney, A., Nushi, B., Collisson, P., ... & Teevan, J. (2019). Guidelines for human-Al interaction. In Proceedings of the 2019 CHI Conference on Human Factors in Computing Systems (paper no. 3). ACM.

Al-infused systems

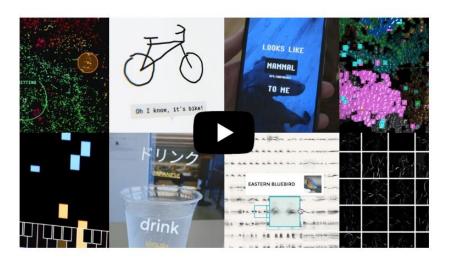
favouriteexamples?



COLLECTION

Al Experiments

AI Experiments is a showcase for simple experiments that make it easier for anyone to start exploring machine learning, through pictures, drawings, language, music, and more.



AI + Learning

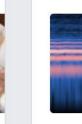


New Project

- ⚠ Open an existing project from Drive.
- Open an existing project from a file.







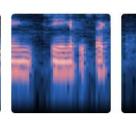








Image Project

Teach based on images, from files or your webcam.

Audio Project

Teach based on one-second-long sounds, from files or your microphone.

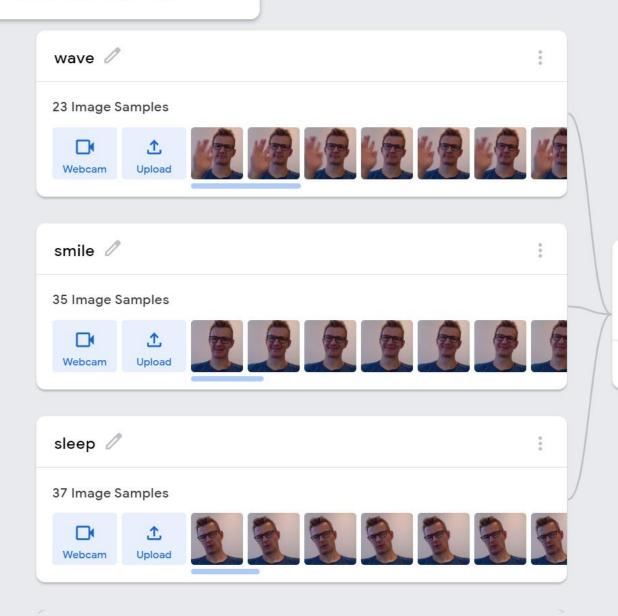
Pose Project

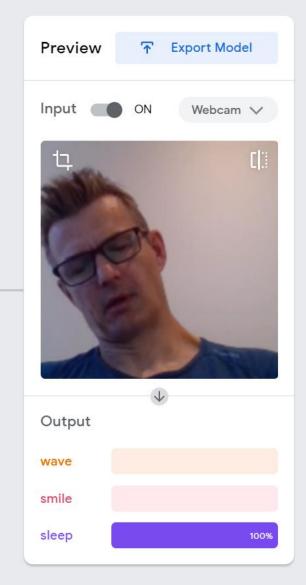
Teach based on images, from files or your webcam.

More coming soon

More models will appear here as they're developed.

■ Teachable Machine





Training

Advanced

Model Trained









can a neural network learn to recognize doodling?

Let's Draw!



37

Made with some friends from

Google





What do 50 million drawings look like?

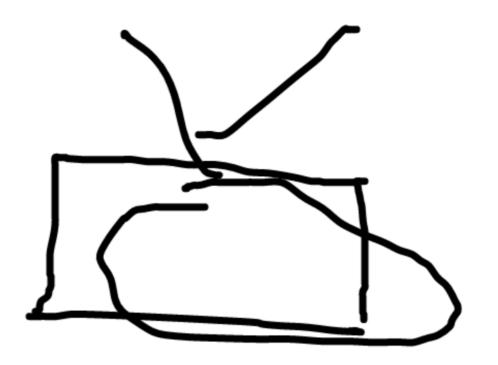
Over 15 million players have contributed millions of drawings playing <u>Quick</u>, <u>Draw!</u>

These doodles are a unique data set that can help developers train new neural networks, help researchers see patterns in how people around the world draw, and help artists create things we haven't begun to think of. That's why <u>we're open-sourcing them</u>, for anyone to play with.

Draw television

in under 20 seconds

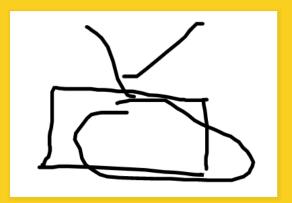
Got It!



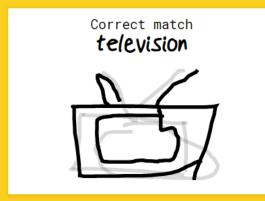


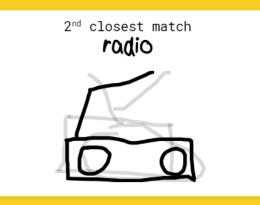
You were asked to draw television

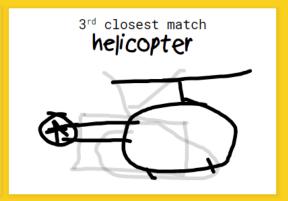
You drew this, and the neural net recognized it.



It also thought your drawing looked like these:







Experiments with Google

Collections ▼

Experiments

Q Search

SUBMIT EXPERIMENT

Cartoonify

August 2018 | By Eric Lu

Turn a photo into a cartoon with a neural network

LAUNCH EXPERIMENT

GET THE CODE

COLLECTION:

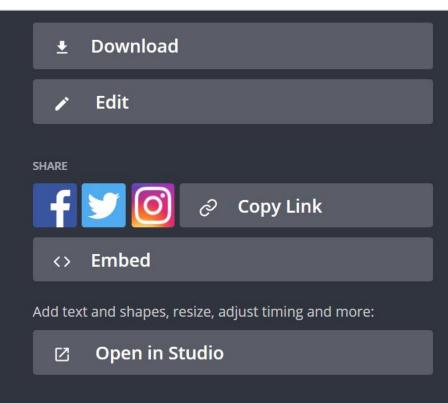
AI Experiments



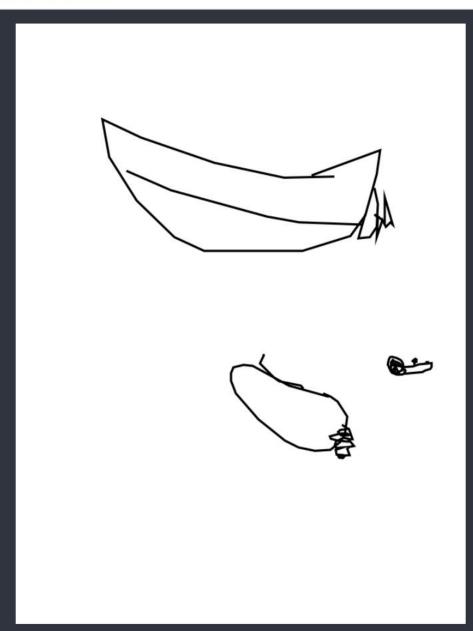


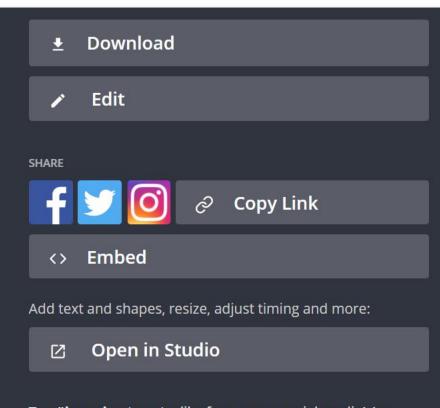






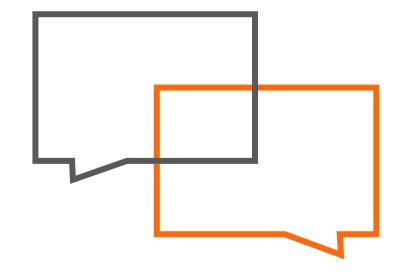
Tag #kapwing to get a like from us on social media! Any issues? Fill out this <u>feedback form</u> to let us know.





Tag **#kapwing** to get a like from us on social media! Any issues? Fill out this <u>feedback form</u> to let us know.

From a user perspective – how does Al-infused systems differ from other computer systems?



Learning

Improving

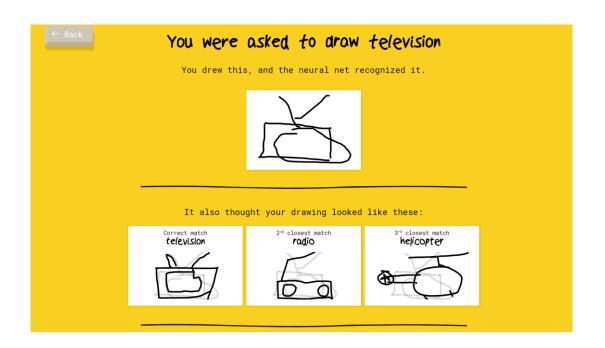
Black box | Fuelled by large data sets

Dynamic

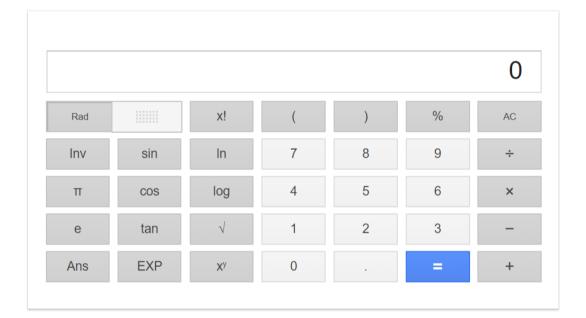
Mistakes inevitable

Opaque

Data gathering through interaction



VS.



Amershi et al (2019)

Definition of Al-infused systems: Systems that have features harnessing Al capabilities that are directly exposed to the end user

Uncertainty -> errors common, both false
positives and negatives

Inconsistency -> sensitive to context and small
changes in input

Behind the scenes personalization (e.g. automated filtering) -> potentially costly information hiding

Guidelines for Human-Al Interaction

Saleema Amershi, Dan Weld '†, Mihaela Vorvoreanu, Adam Fourney, Besmira Nushi, Penny Collisson, Jina Suh, Shamsi Iqbal, Paul N. Bennett, Kori Inkpen, Jaime Teevan, Ruth Kikin-Gil, and Eric Horvitz

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shamsi, pauben, kori, teevan, ruthkg, horvitz}
@microsoft.com

ABSTRACT

Advances in artificial intelligence (AI) frame opportunities and challenges for user interface design. Principles for human-AI interaction have been discussed in the human-computer interaction community for over two decades, but more study and innovation are needed in light of advances in AI and the growing uses of AI technologies in human-facing applications. We propose 18 generally applicable design guidelines for human-AI interaction. These guidelines are validated through multiple rounds of evaluation including a user study with 49 design practitioners who tested the guidelines against 20 popular AI-infused products. The results verify the relevance of the guidelines over a spectrum of interaction scenarios and reveal gaps in our knowledge, highlighting opportunities for further research. Based on the evaluations, we believe the set of design guidelines can serve as a resource to practitioners working on the design of applications and features that harness AI technologies, and to researchers interested in the further development of guidelines for human-AI interaction design.

CCS CONCEPTS

Human-centered computing → Human computer interaction (HCI);
 Computing methodologies → Artificial intelligence.

Work done as a visiting researcher at Microsoft Research.

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ACM ISBN 978-1-4503-5970-2/19/05...\$15.00 https://doi.org/10.1145/3290605.3300233 [†]Paul G. Allen School of Computer Science & Engineering University of Washington Seattle, WA, USA weld@cs.washington.edu

KEYWORD

Human-AI interaction; AI-infused systems; design guidelines

ACM Reference Format:

Saleema Amershi, Dan Weld, Mihaela Vorvoreanu, Adam Founey, Besmira Nushi, Penny Collisson, Jina Suh, Shamsi Iqbal, Paul N. Bennett, Kori Inkpen, Jaime Teevan, Ruth Kikin-Gil, and Eric Horvitz. 2019. Guidelines for Human-Al Interaction. In CHI Conference on Human Factors in Computing Systems Proceedings (CHI 2019), May 4-9, 2019, Glasgow, Scotland Uk. ACM, New York, NY, USA, 13 pages. https://doi.org/10.1145/3290605.3300233

1 INTRODUCTION

Advances in artificial intelligence (AI) are enabling developers to integrate a variety of AI capabilities into user-facing systems. For example, increases in the accuracy of pattern recognition have created opportunities and pressure to integrate speech recognition, translation, object recognition, and face recognition into applications. However, as automated inferences are typically performed under uncertainty, often producing false positives and false negatives, AI-infused systems may demonstrate unpredictable behaviors that can be disruptive, confusing, offensive, and even dangerous. While some AI technologies are deployed in explicit, interactive uses, other advances are employed behind the scenes in proactive services acting on behalf of users such as automatically filtering content based on inferred relevance or importance. While such attempts at personalization may be delightful when aligned with users' preferences, automated filtering and routing can be the source of costly information hiding and actions at odds with user goals and expectations.

Al-Infused systems¹ can violate established usability guidelines of traditional user interface design (e.g., [31, 32]). For example, the principle of consistency advocates for minimizing unexpected changes with a consistent interface appearance and predictable behaviors. However, many AI components are inherently inconsistent due to poorly understood,

¹In this paper we use Al-infused systems to refer to systems that have features harnessing AI capabilities that are directly exposed to the end user.

Kocielnik et al (2019)

Probabilistic – almost always operate at less than perfect accuracy

Impacted by user actions – such as user-generated content

Transparency issues – how to mitigate? e.g. by showing decision rules

Will You Accept an Imperfect AI? Exploring Designs for Adjusting End-user Expectations of AI Systems



Figure 1: Expectation setting design techniques used prior to interaction with the Scheduling Assistant - an AI system for meeting request detection from free-text of emails. A) Accuracy Indicator - directly communicating to the user the expected accuracy of the AI component, B) Example-based Explanation - helping the user understand the basic principles of how the systems detects meeting requests, C) Control - giving the user control over AI decision making process through detection threshold adjustment.

ABSTRACT

Al technologies have been incorporated into many end-user applications. However, expectations of the capabilities of such systems vary among people. Furthermore, bloated expectations have been identified as negatively affecting perception and acceptance of such systems. Although the intelligibility of ML algorithms has been well studied, there has been little work on methods for setting appropriate expectations before the initial use of an Al-based system. In this work, we use a Scheduling Assistant - an Al system for automated meeting request detection in free-text email - to study the impact of several methods of expectation setting. We explore two versions of this system with the same 50% level of accuracy of the Al component but each designed with a different focus on the types of errors to avoid (avoiding False Positives vs. False Negatives). We show that such different

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ACM ISBN 978-1-4503-5970-2/19/05...\$15.00 https://doi.org/10.1145/3290605.3300641 focus can lead to vastly different subjective perceptions of accuracy and acceptance. Further, we design expectation adjustment techniques that prepare users for AI imperfections and result in a significant increase in acceptance.

CCS CONCEPTS

 Human-centered computing → Empirical studies in HCI; Empirical studies in visualization; Laboratory experiments:

KEYWORD

AI infused systems, AI system on-boarding, Shaping AI expectations, Perception and Acceptance of AI

ACM Reference Form:

Rafal Kocielnik, Saleema Amershi, and Paul N. Bennett. 2019. Will You Accept an Imperfect Al? Exploring Designs for Adjusting Enduser Expectations of Al Systems. In CHI Conference on Human Factors in Computing Systems Proceedings (CHI 2019), May 4–9, 2019, Glazgow, Scotland Uk. ACM, New York, NY, USA, 14 pages. https: //doi.org/10.1145/3290605.3300641

1 INTRODUCTION

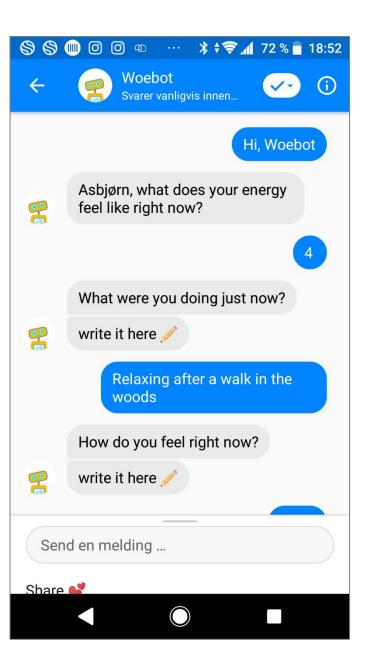
Expectations impact how accepting end-users are of the technologies they use. For example, inflated expectations about usability and ease of use have been shown to decrease user satisfaction and willingness to use products when those expectations are not met [20, 36]. Artificial intelligence (AI) introduces additional factors impacting user expectations

Chatbots – interacting with Al in natural language

Chatbots?



Chatbots?

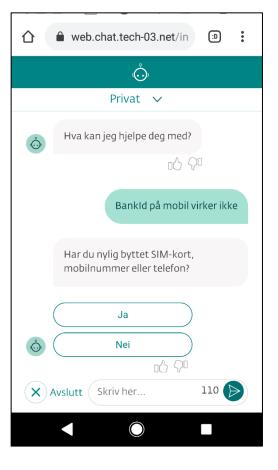


One definition of chatbots:

... machine agents that serve as natural language user interfaces to data and services through text or voice.

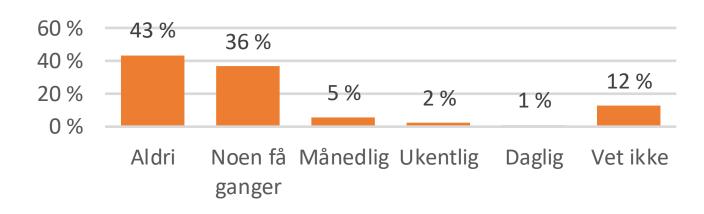
Meaning of term evolving



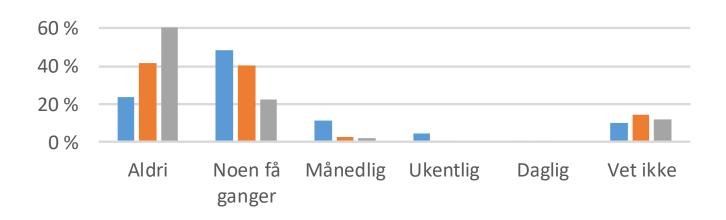


Your chatbot experience?

Hvor ofte har du brukt chatbots (prateroboter) på en nettside eller på en meldingstjeneste?

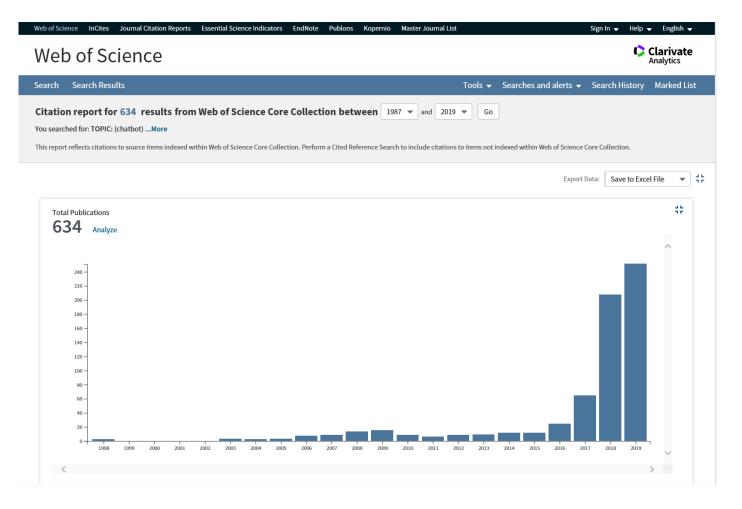


Variasjon med alder (16-34 | 35-54 | 55 -)



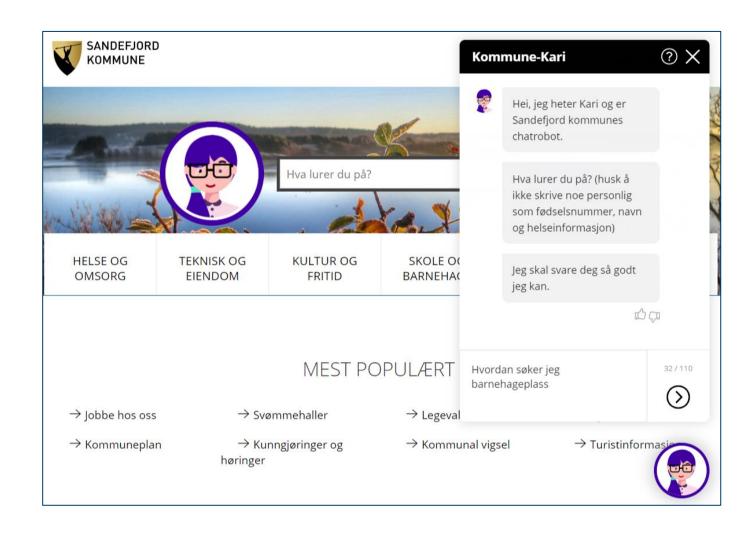
Chatbots area of sharply increasing research interest

Search term: chatbot

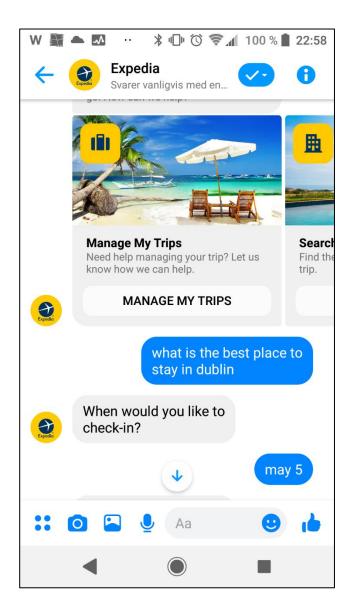


http://apps.webofknowledge.com/WOS_GeneralSearch_input.do

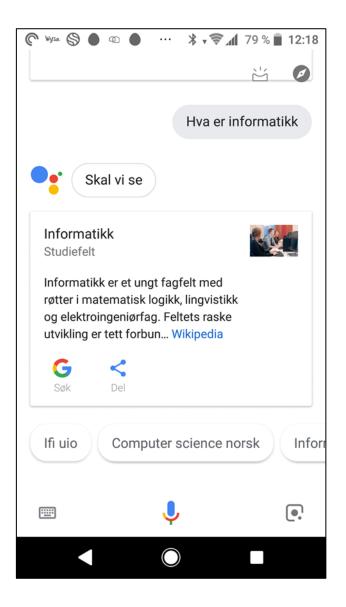
- Customer service
- Content curation and eCommerce
- Assistants
- Guides
- Social relations



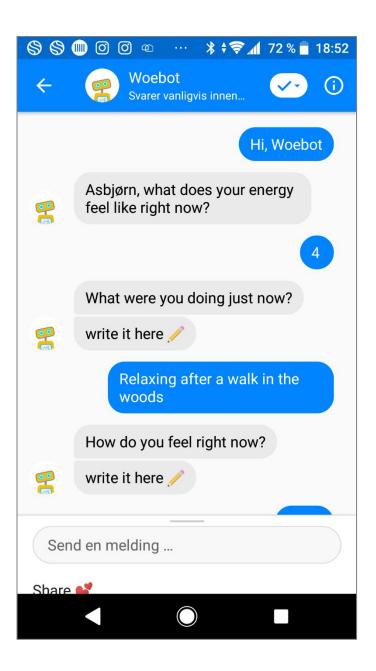
- Customer service
- Content curation and eCommerce
- Assistants
- Guides
- Social relations



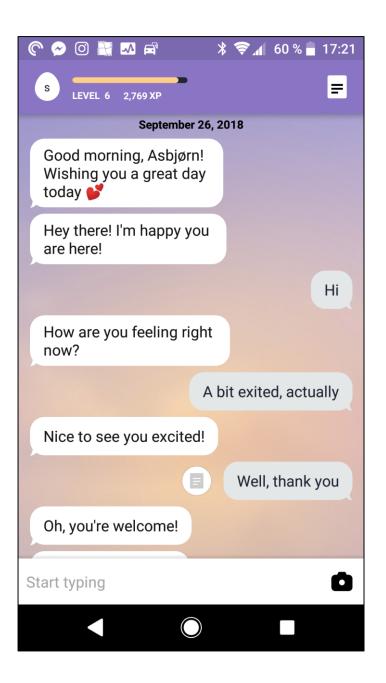
- Customer service
- Content curation and eCommerce
- Assistants
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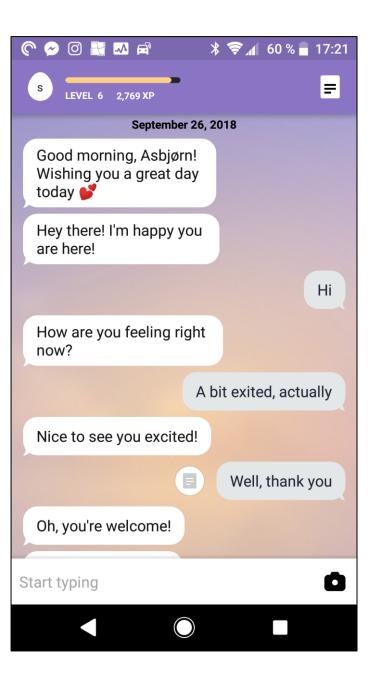
- Customer service
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- Social relations



- Customer service
- Content curation and eCommerce
- Assistants
- Guides
- Social relations



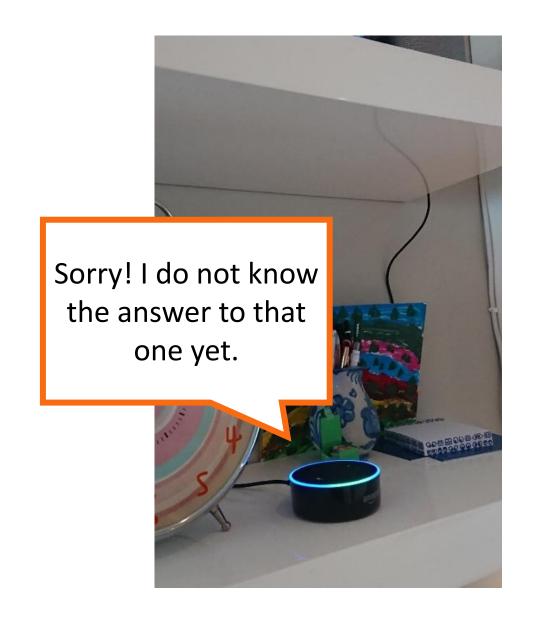
Always supportive and caring



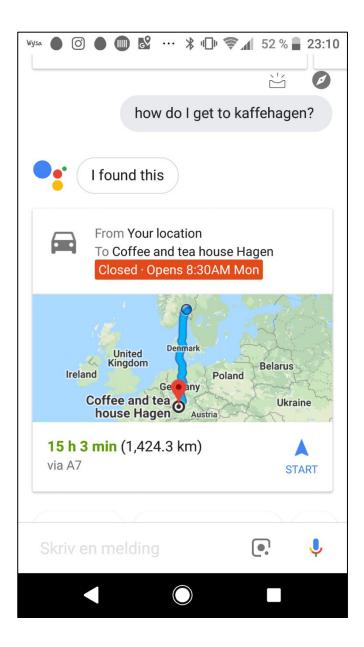
Always supportive and caring



Always supportive and caring



Always supportive and caring



Always supportive and caring

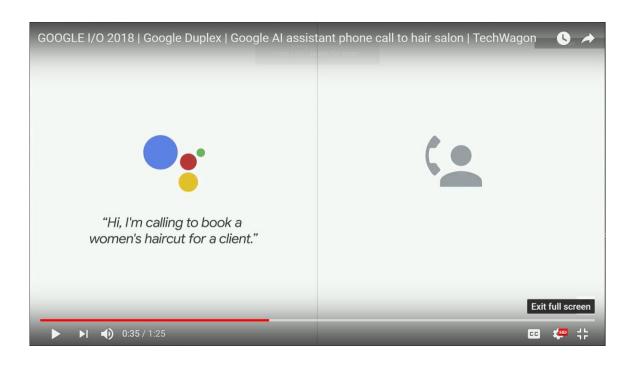


A race towards human likeness?

Google Duper – help to make reservations through conversation

Presented at I/O 2018

Available as service for selected areas



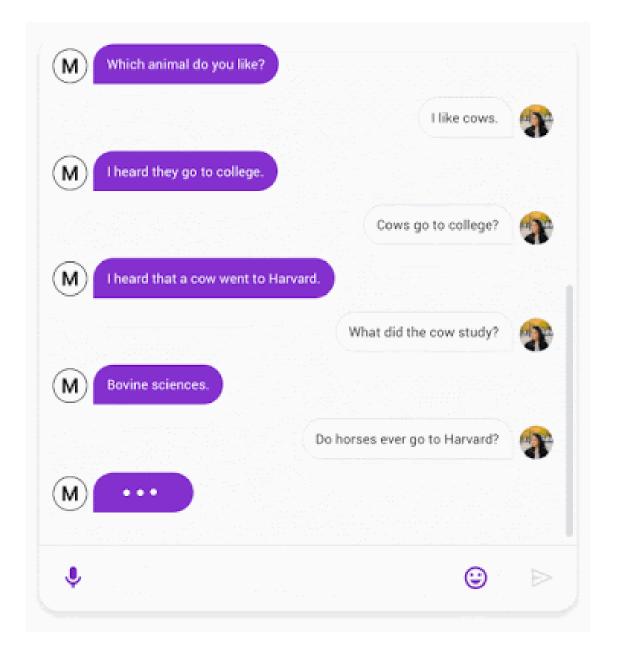
https://youtu.be/GoXp1leA5Qc

A race towards human likeness?

Google Meena – January 2020

Open domain chatbot based on machine learning – single model

Trained on 40B words from open social media conversations

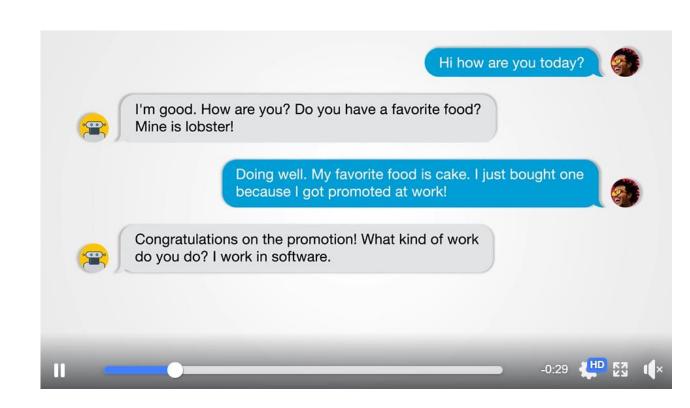


A race towards human likeness?

Facebook Blender – April 2020

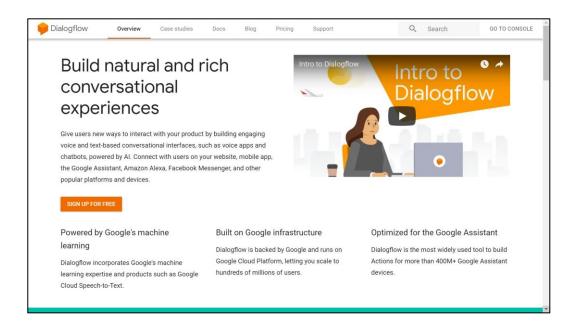
Open domain chatbot based on machine learning - blended model

Trained on 1,5B Reddit comments.



Chatbot development platforms

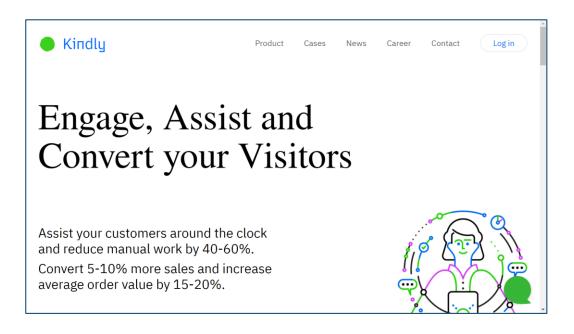
- DIY
- Platform and assistance

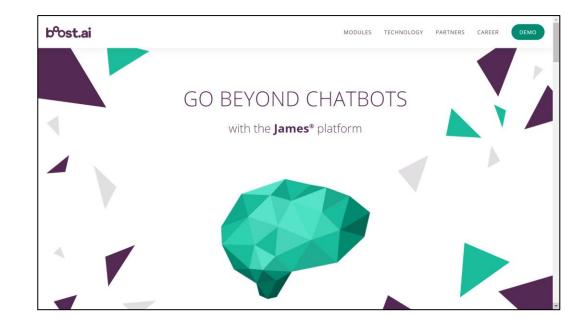




Chatbot development platforms

- DIY
- Platform and assistance





Different types of chatbots

- Generative
- Intents and actions
- Scripted

A Neural Conversational Model

text (e.g., hyperlinks) from the dataset. As turn taking is not clearly indicated, we treated consecutive sentences assuming they were uttered by different characters. We trained our model to predict the next sentence given the previous one, and we did this for every sentence (noting that this doubles our dataset size, as each sentence is used both for context and as target). Our training and validation split has 62M sentences (923M tokens) as training examples, and the validation set has 26M sentences (395M tokens). The split is done in such a way that each sentence in a pair of sentences either appear together in the training set or test set but not both. Unlike the previous dataset, the OpenSubtitles is quite large, and rather noisy because consecutive sentences may be uttered by the same character. Given the scope of movies, this is an open-domain conversation dataset, contrasting with the technical troubleshooting dataset.

5. Experiments

In this section, we describe the experimental results with the two datasets. We present a rather objective measure which is the perplexity of the model on the test set. We also show some samples of the interactions with the system that we trained.

5.1. IT Helpdesk Troubleshooting experiments

Given the lack of data for this domain specific and clean dataset, we trained a single layer LSTM with 1024 memory cells using stochastic gradient descent with gradient clipping. We built a vocabulary of the most common 20K words, and introduced special tokens indicating turn taking and actor.

At convergence, this model achieved a perplexity of 8, whereas an n-gram model achieved 18. Below is a sample

Human: i am seeing an error related to vpn Machine: what is the error message that you are getting when connecting to vpn using network

Human: connection refused or something like

Machine: may i know the version of network

connect vou connect?

Human: i am not sure i know that

Machine: is the network connect prompting for

any user certificate Machine: ? Human: yes

Machine: are you at home or at work at the mo-

ment?

Machine: office? Human: i am at home Machine: try this

Machine: goto <URL> page and enter your

name and password Human: done

Machine: try logging in with <NAME> and

check if its working fine now Human: yes, now it works!

Machine: great

Machine: anything else that i can help Human: no, that was very useful Machine: you are other than <NAME>:)

Machine: have a nice night

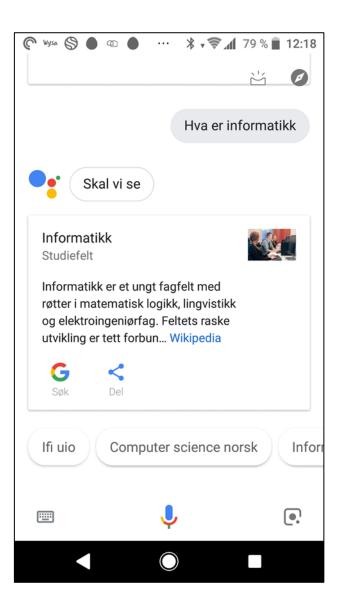
(The <URL> indeed contains information about vpn access.)

Describe communities, because to one continue

Conversation 2: Browser issues.

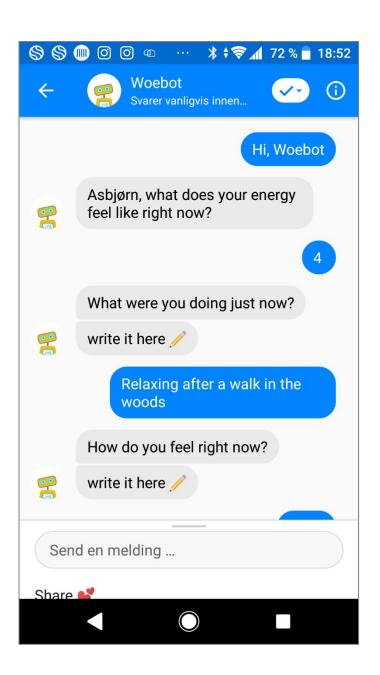
Different types of chatbots

- Generative
- Intents and actions
- Scripted



Different types of chatbots

- Generative
- Intents and actions
- Scripted



Interaction design and training

- Generative
- Intents and actions
- Scripted



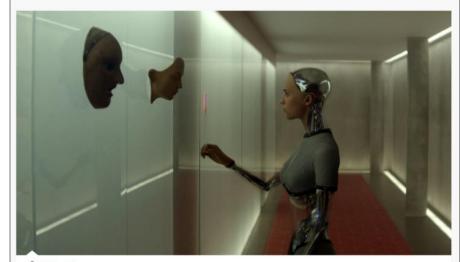






Kunstig intelligens må ikke lenger trenes av mennesker

- Vi tror det vil komme kunstig intelligens på nivå med mennesker før eller siden, men det er veldig vanskelig å si om det blir 10 eller 50 år til, sier forsker.



Filmen Ex Machina (2014), som ble spilt inn i Norge, ender med at den første generelle kunstige intelligensen som overgår mennesket slipper ut i samfunnet. Konsekvensene er helt uforutsigbare.

09:05



29. mars 2017 19:30 • 30. mars 2017





Interaction design and training

- Generative
- Intents and actions
- Scripted



En gjeng DNB-ere kan nå kalle seg Al-trenere. De jobber ved <u>kundesentrene</u> i Bergen og Oslo, og får mye oppmerksomhet for stillingsbetegnelsen.

Interaction design and training

- Generative
- Intents and actions
- Scripted





Group task – for October 6 (Appendix 1 in mid-term report)

Prototype a chatbot for a self-decided purpose

... using Chatteron

(... or Chatfuel if you want to use one of the most popular platforms for FB Messenger chatbots)

(... or Dialogflow if you want more flexibility and challenge)

Keep it simple and early-phase (4-8 hours work?)

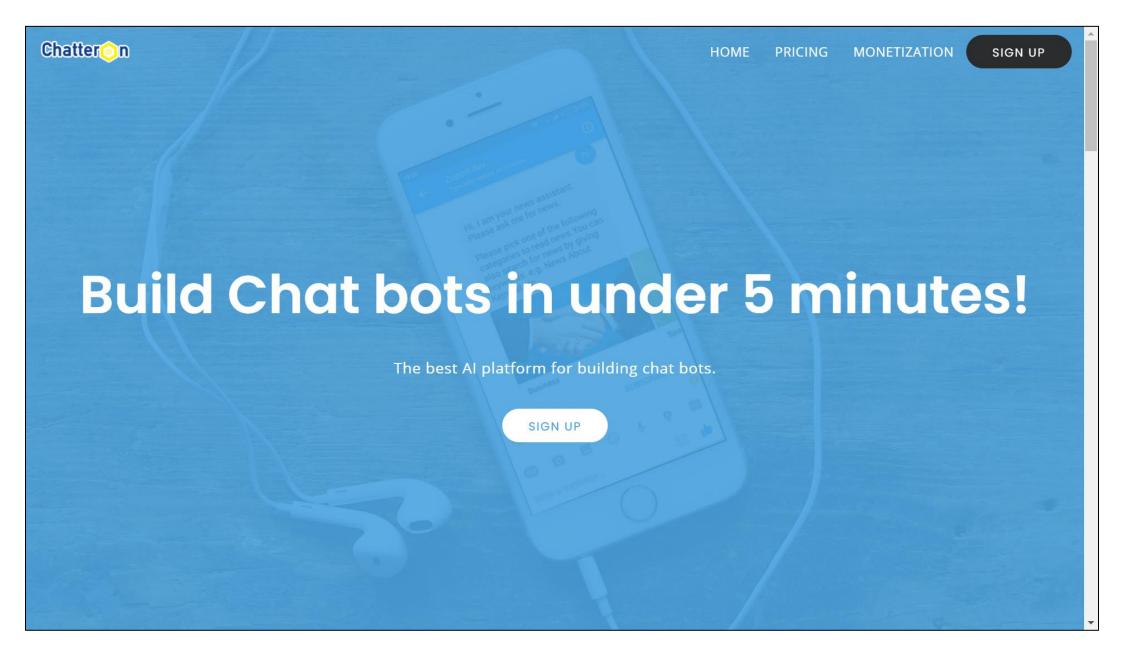
Make brief (approx 1 minute) video capture of dialogue – upload to Vortex

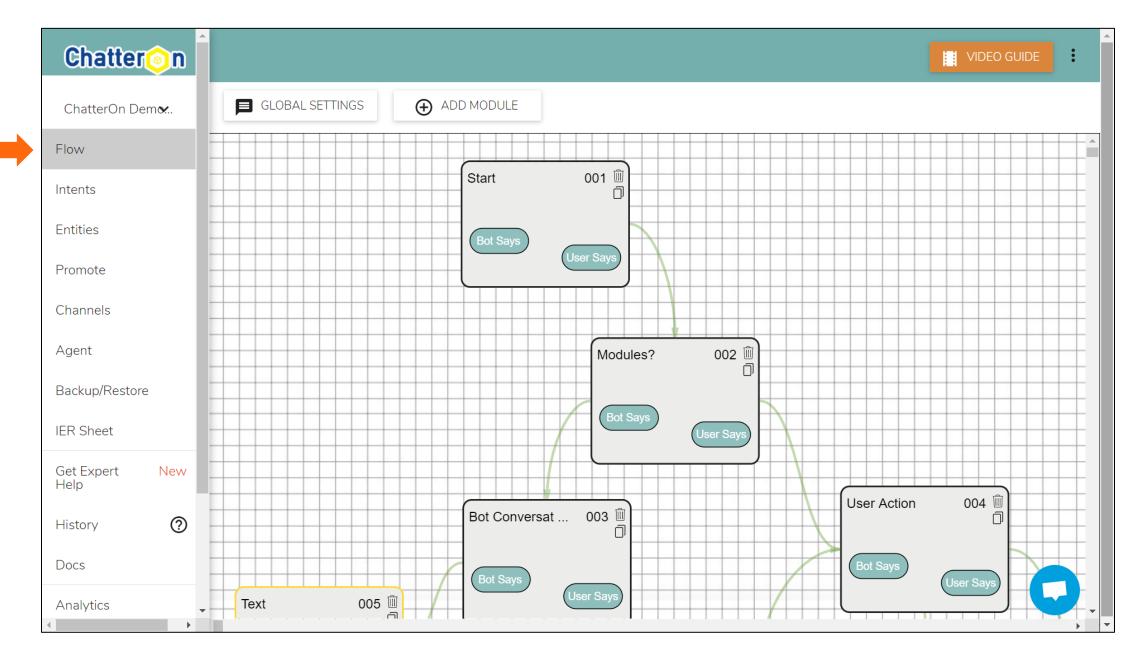
Decide on purpose

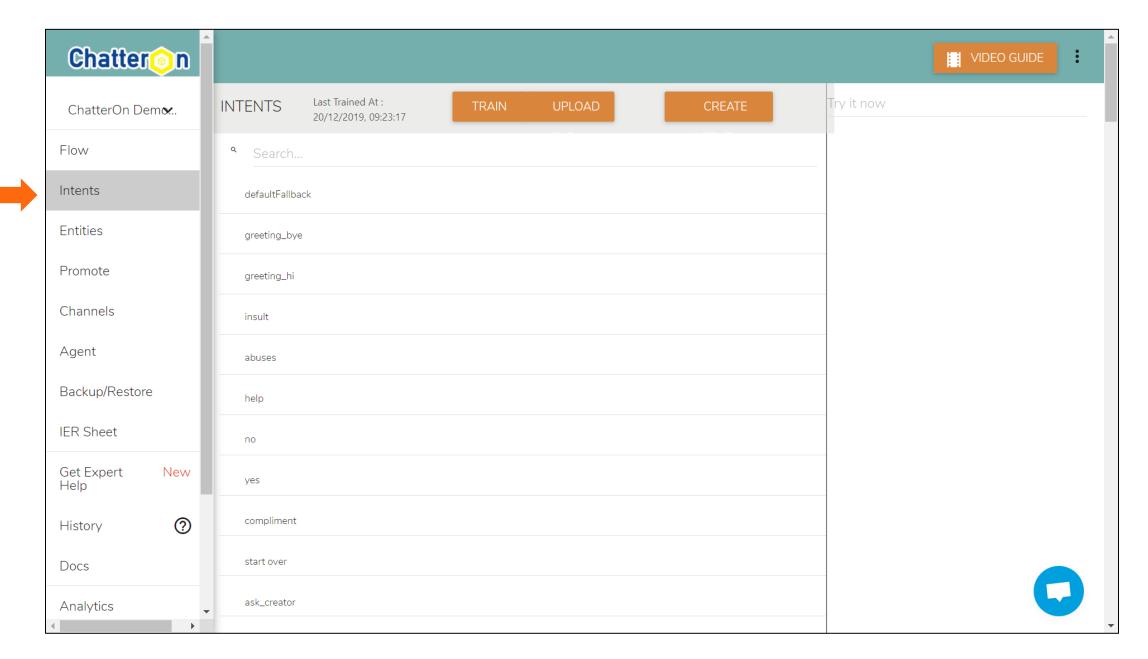
Identify key tasks

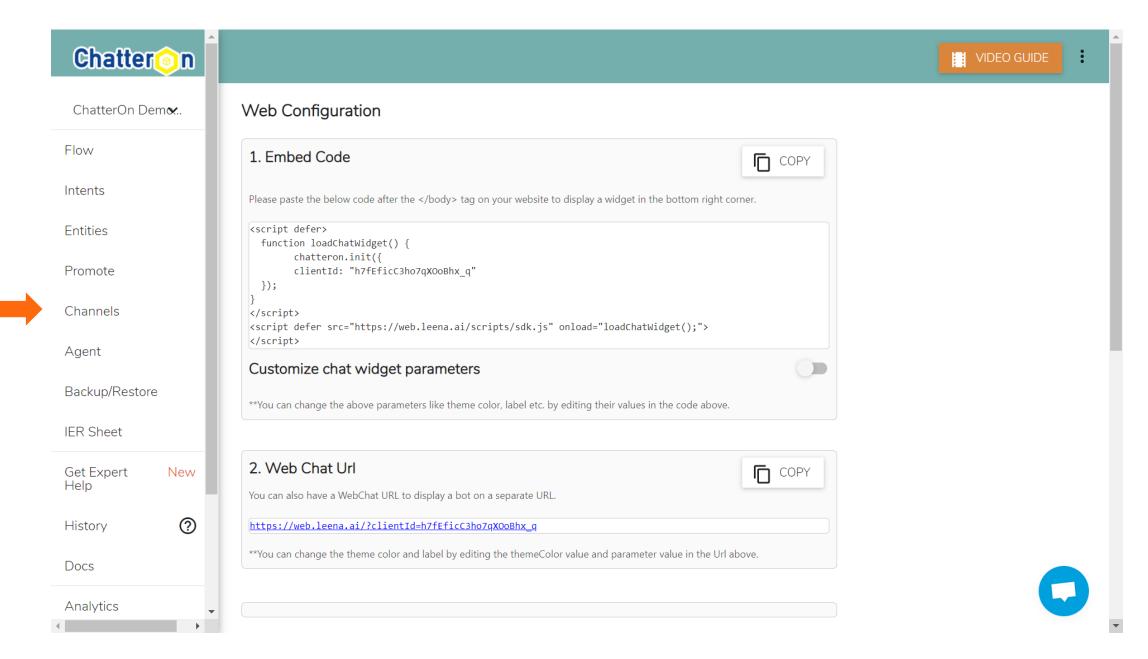
Design flows, intents and actions

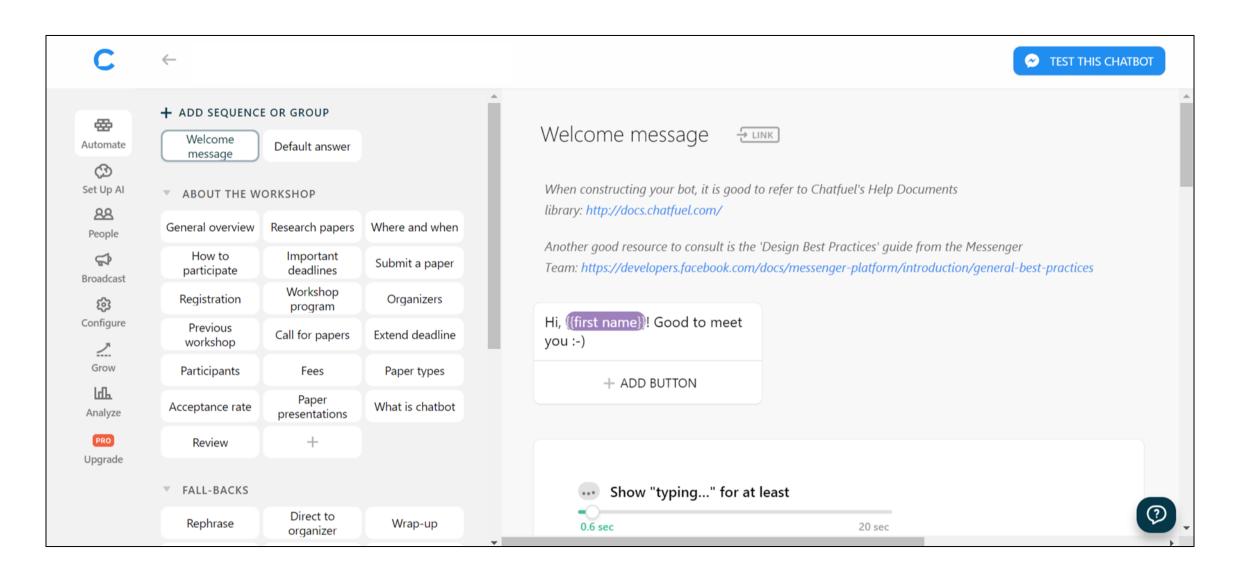
Test and refine

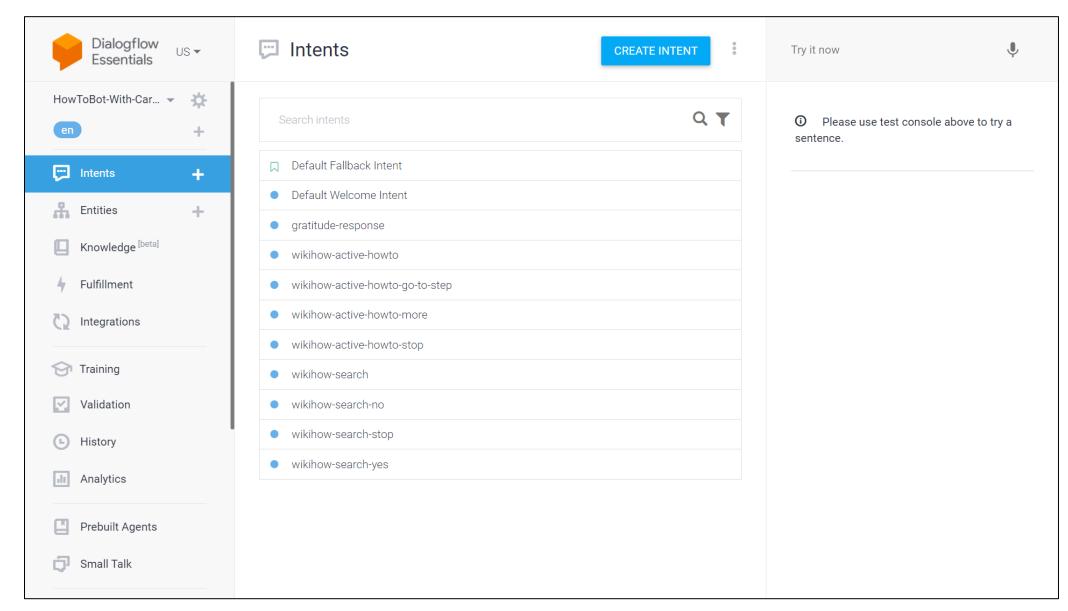












End 1