Acvanced Machine Learning For

Lecture 1 - Introduction to Machine Learning

Evangelos Niforatos

20/09/2023

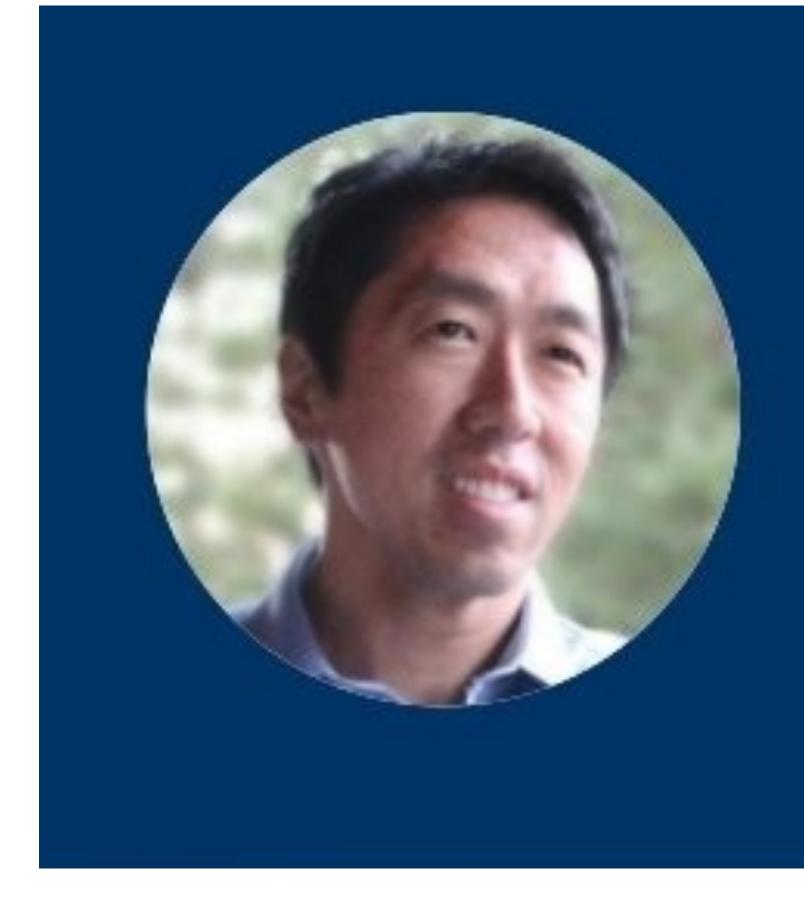
ml4d-ide@tudelft.nl https://aml4design.github.io/



Why Machine Learning for Design?

Part I

"Al is the New Electricity"



"Just as electricity transformed almost everything 100 years ago, today I actually have a hard time thinking of an industry that I don't think AI will transform in the next several years."

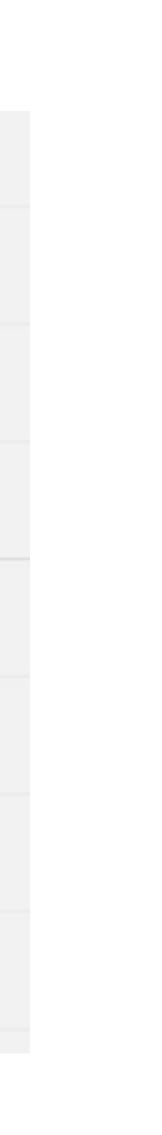
Andrew Ng

Former chief scientist at Baidu, Co-founder at Coursera



Mark - ChatGPT Lieneke - Apple image p... ChatGPT is an Al-model that Al is used Description is trained to converse with the user The Camera application of Al is used to understand the the user. User can get their combine 5 Apple phones/ipads uses Al scene (person segmentation, prompts answered to or sources t n their software to optimize depth estimation) to steer the model towards a to answe ptimize light and contrast age quality. ore desirable outcome also to p for example. But also to through an almost human- unknown enable portrait mode, like communication. hasn't tra recognizing faces. Ceyda - Chatbots Jim - Shazam Description It listens to auto and can find To match Chatbots help customers Al is used to understand a match to provide the titel input to with answering their what the customer is asking 6 and artist uestions on a product or and therefore chatbots try to databa service. They are used on predict what the right answer must be. Copy of Template Andrija - Wayve Description Description Wayve Al Driver is a Siri is an example of a The role of AI in Siri is to Al serves autonomous driving machine virtual personal assistant understand and respond to software that uses ML to integrated into Apple natural language input. It interpre ev Siri devices. It provides the user can perform tasks, provide interpret camera and radar enable w data, enabling vehicles to autonor rith services and information and give drive without human nformation through voicerecommendations based on intervention or the need for driven interactions. user preferences. WAYVE detailed maps and rules. Alice - Duolingo Dilara - Apple Face Description Description Duolingo prides itself in using Al in Duolingo is an app for Your face d of invisible many stages of their app but the Identifies your face and learning languages. It helps **R**. echnology is mostly used for unlocks your phone people from all over the ٠ sonalizing users lessons. you save it 1.6.4. world to have easier access model, for example, is who you a 1.1.1.1.1.1 to language learning. used specifically to figure out the your phon fit of which exercises in a face. "Face 100 C particular lesson will be the best TrueDepth match for a learner's level of machine le knowledge. secure aut solution."

Ashraf - Netflix	Sebastiaan - DeepL
er' prompts and ne data from multiple es to reply. It is trained	Image Description Roles of Al like uses ML to optimize user experience by sosing users what to the based on their thing history. Image DeepL is a great tool to translate between languages. It also offers an API. Instead of translating texts word-by-word, DeepL uses a Deep Learning model (as the name suggests). This makes it way better at translating but also allows it to use different tones or proper formal language. Vonghao TeepL and the tones or proper formal language.
images editing software, func- used mainly for photo Examples and graphic fill, or generation in the second	Roles of AlImageDescriptionRoles of Altoshop has several tions that utilise AL nples are content-aware bject selection, and erative fill. All of these tify different objects and ent of images.In order to drive on their own, autonomous cars constantly interpret images from their sensors and
ne learning to ret sensor data and e vehicles to navigate amously. Google Maps Set to be at the set of the set	Image Description Roles of Al gle Maps has multiple tons using Al, the Google lens is app where Ai is comparing the pictures - visualizing every nent of a route before Google lens is app where Ai is comparing the pictures 0. But also the more c functions like: the internet. on the internet. on the internet. mized route. mized route. the most the internet. the internet.
activities like can play audio, control your smart home, answer de learning for a puthentication	NAIexaNoise of AlNexa relies on natural language processing and machine learning is used to reformance over time. Natural language processing leads to renoth conversations with users and machine learning is used to refine its processes, creating a smarter system over time.Noise is processes, creating a renoth conversations with users and machine learning is used to refine its processes, creating a smarter system over time.

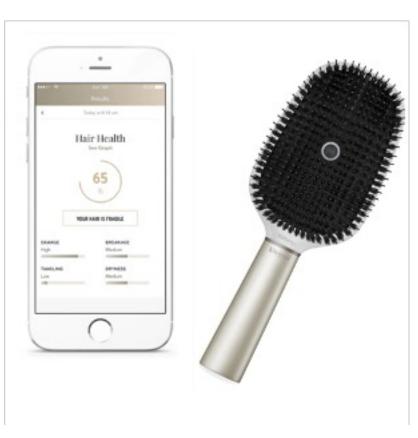




Where is AI? Or ML?

- Autonomous vehicles
 - from Roomba to Self-driving cars
 - In stores, warehouses, production lines, streets, living rooms
- More and more consumer products and appliances
 - Belts!! Really!
 - Thermostats, Security Cameras, Fridges
- Content production and consumption applications
 - Social media, Amazon, Netflix etc.
- Chatbots
- In-store automation and smarter shopping
- Optimised supply chains
- Energy grid optimisation
- . . .

Hello, Sign in Returns Account & Lists - & Orders amazon O Netherland All Departments *



Smart hairbrush. AI splitting the hair ...

Price: **\$199.99**



More than just a fashion accessory, Belty Good Vibes is the very first smart belt integrating Artificial Intelligence that contextualizes the activities of your everyday life.

Beyond data

Rather than providing only raw data, Belty offers feedback about the rhythm of your life. It goes beyond statistics and helps you to be more aware of the quality of your everyday experience.

Trust your gut

The abdomen, or belly, is considered the second brain of your body: the home of your gut instinct. Belty Good Vibes empowers you to know yourself better, by reinforcing your ability to connect to your visceral knowledge. Communicating via vibrations with your sense of touch, it plugs you into the present moment.

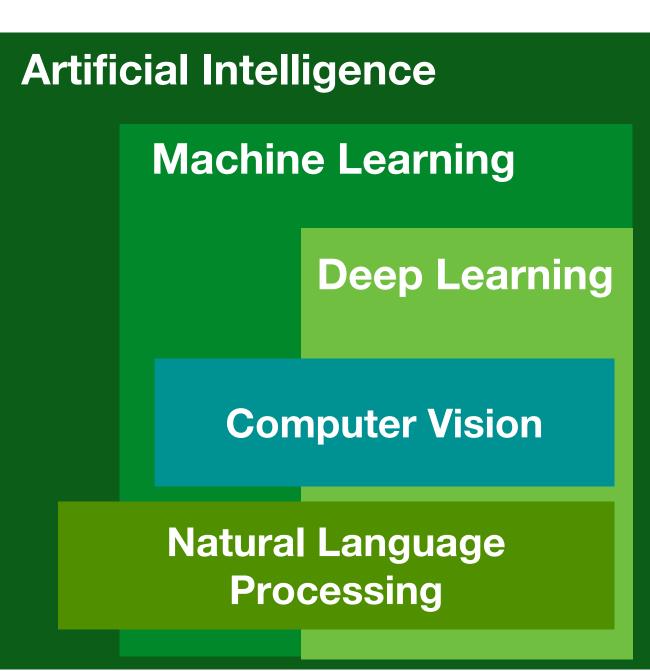
Good vibrations, great energy

Belty is much more than a smart belt; as wearable, interactive technology, it is your personal coach. We all want to live the best version of our lives. Why not start now?





When the second **Artificial Intelligence Machine Learning? Deep Learning? Computer Vision?** Natural Language Processing?





Intelligence

- The ability to learn or understand or to deal with new or trying situations
- The ability to apply knowledge to manipulate one's environment or to think abstractly as measured by objective criteria (such as tests)

Mental quality that consists of the abilities to learn from experience, adapt to new situations, understand and handle abstract concepts, and use knowledge to manipulate one's environment

But if you really want an exhaustive collection of 71 definitions, look no further https://arxiv.org/pdf/0706.3639.pdf

Merrian-Webster

Encyclopedia Britannica

R. J. Sternberg, quoted in *The Oxford* Companion to the Mind. R. L. Gregory. Oxford University Press, Oxford, UK, 1998





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Artificial Intelligence

Intelligence demonstrated by machines

A branch of computer science that studies the properties of intelligence by synthesizing intelligence

- - Perception, Learning, Reasoning, Planning, Problem-solving, Creating

Creating computer programs that perform tasks as well as, or better than, humans



Strong vs. Weak Artificial Intelligence

Strong AI

- Artificial General Intelligence (AGI), human-level, general
- The AI we see in movies
- Al that can do everything we humans can do, and possibly much more

Weak Al

- Narrow Al
- Al specialised in well-defined tasks
 - e.g., speech recognition, chess-playing, autonomous driving
- No AI program has been created yet that could be called intelligent in any general (Strong AI) sense
 - abilities, but about the integration between those abilities?
- arithmetic

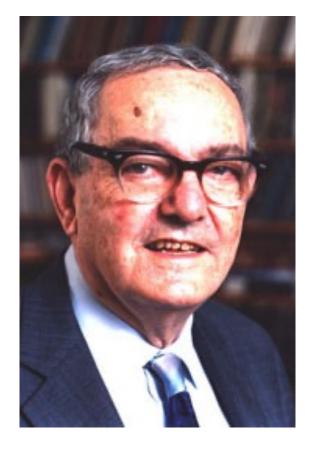
"A pile of narrow intelligence will never add up to a general intelligence. General intelligence isn't about the number of

Superintelligence doesn't really mean anything - a basic calculator far exceeds any human benchmark for performing basic



Learning

- Any process by which a system improves performance from experience
- Denotes changes in the system that are adaptive in the sense that they enable the system to do the task or tasks drawn from the same population more efficiently and more effectively the next time
- The ability to perform a task in a situation that has never been encountered before
- Learning = generalisation



Herbert A. Simon

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What is a cat?

Credits: Jonah Burlingame

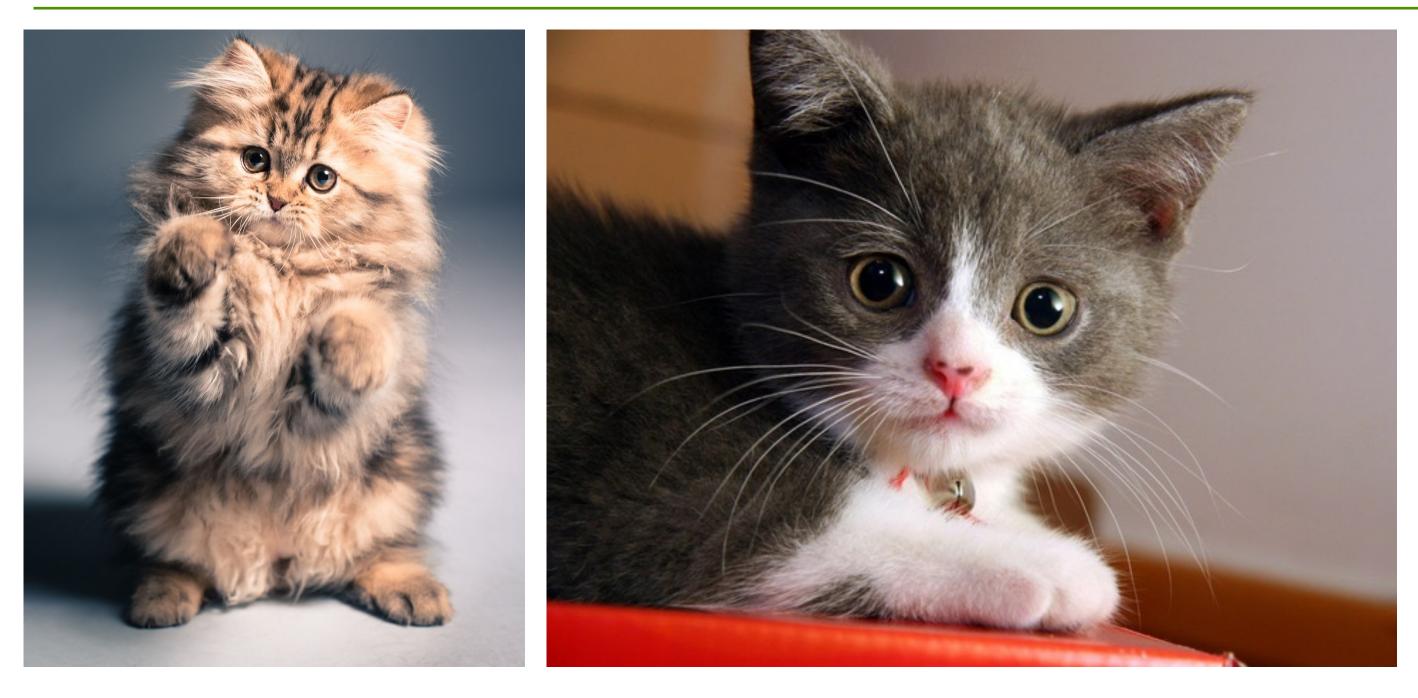


What is a cat?





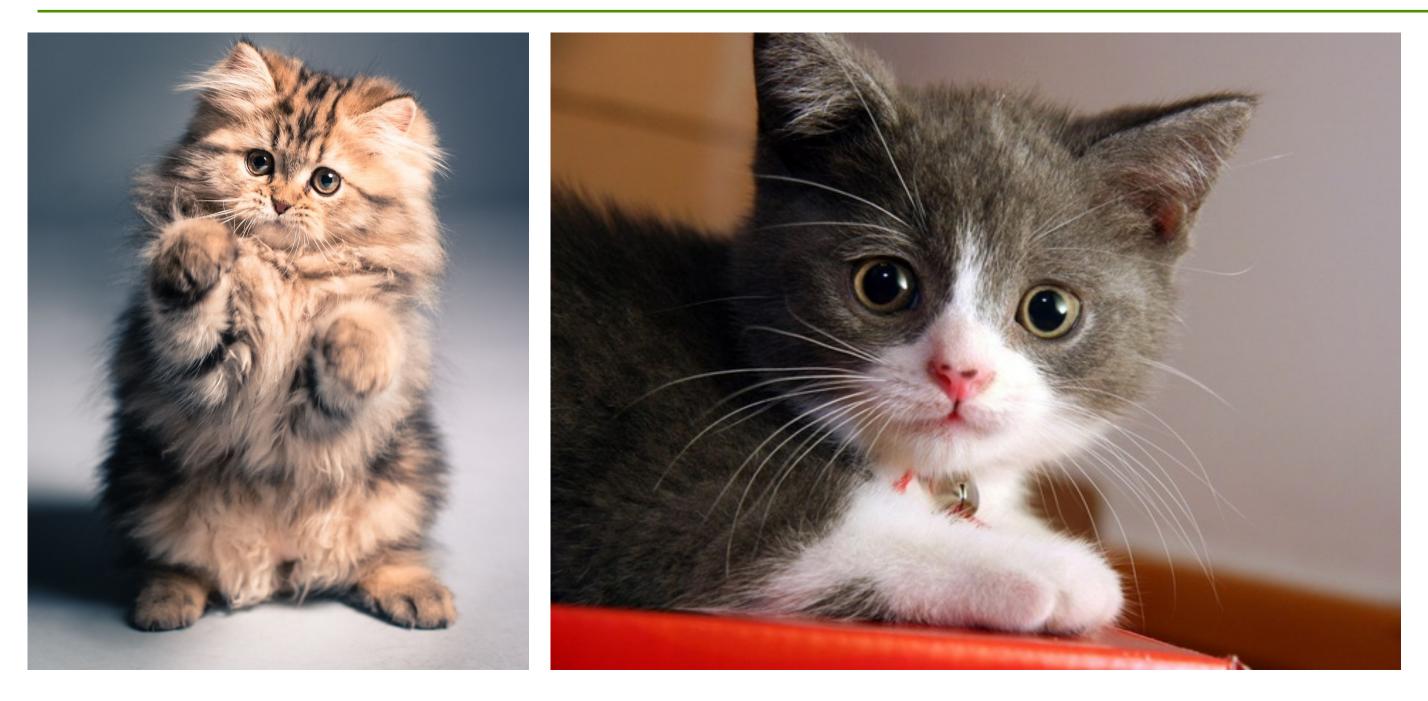
What is a cat? V₁



- It's a cat if it has whiskers
- And it is furry



What is a cat?



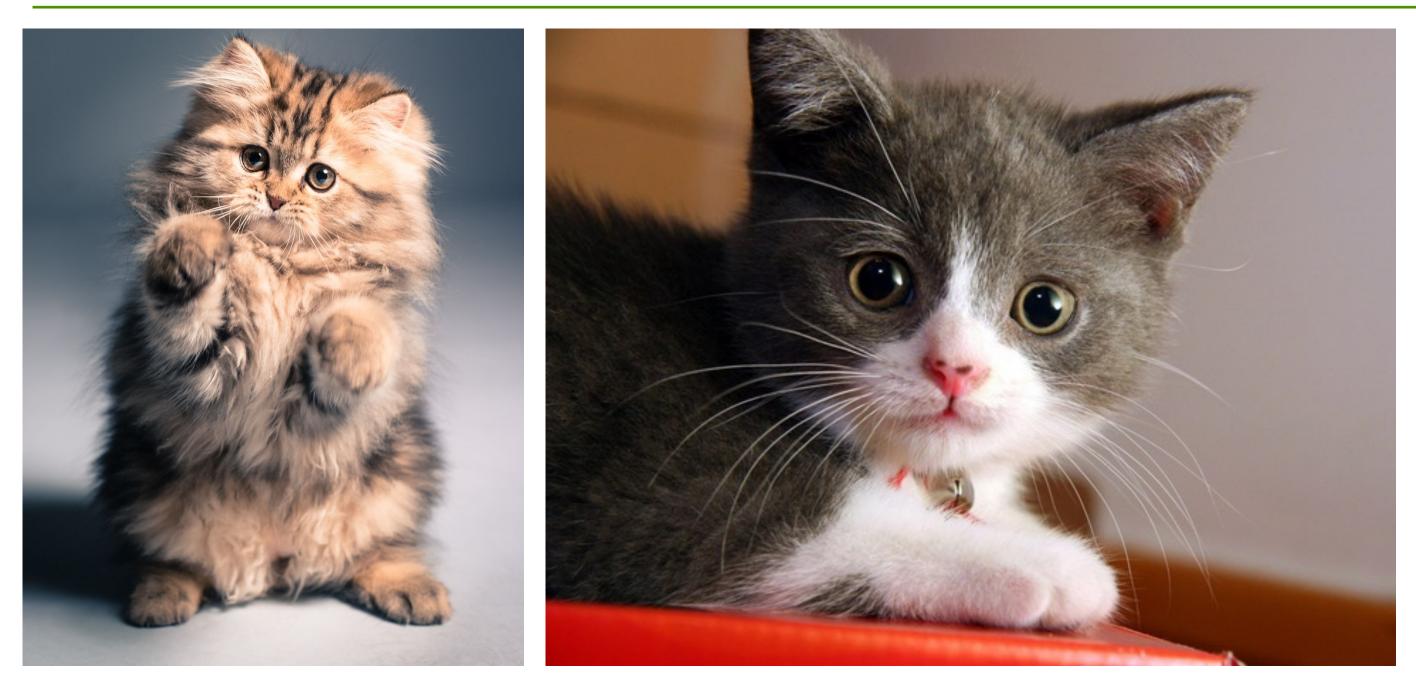
- It's a cat if it has whiskers
- And it is furry



Credits: Jonah Burlingame



What is a cat? V₂



- It's a cat if it has whiskers
- And it is furry
- And it is small



What is a cat?



- It's a cat if it has whiskers
- And it is furry
- And it is small



Credits: Jonah Burlingame

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What is a cat? V₃



- It's a cat if it has whiskers
- And it is furry
- And it is small
- And it does not climb trees



What is a cat?



- It's a cat if it has whiskers
- And it is furry
- And it is small
- And it does not climb trees



Credits: Jonah Burlingame



Polanyi's Paradox | Michael Polanyi (1966)

"We can know more than we can tell...

The skill of a driver cannot be replaced by a thorough schooling in the theory of the motorcar"

Michael Polanyi (1966)





Machine Learning

The field of study that gives computers the ability to learn without being explicitly programmed

Machine learning is the science (and art) of programming computers so they can learn from data



Arthur Samuel

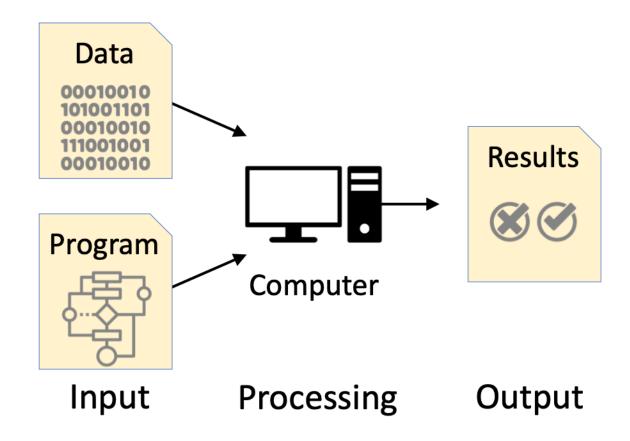


Is this a cat?

Traditional Programming

Rules to detect a cat:

- 1. It has whiskers
- 2. It is furry
- 3. It is small

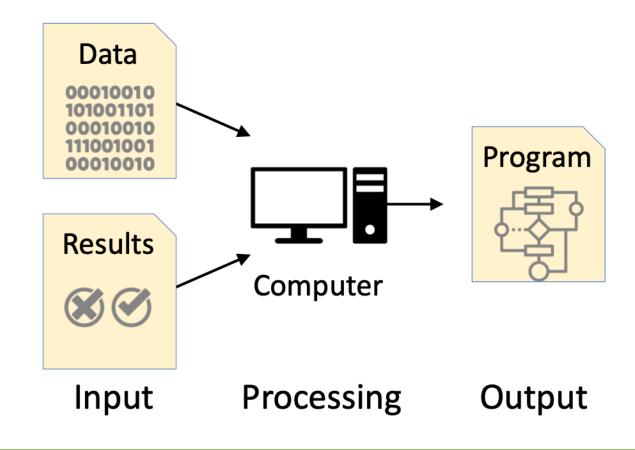






Machine Learning

Let me guess how I can distinguish a cat :)





Functions of a Machine Learning System

Descriptive

Using data to explain what happened

Predictive

Using data to predict what will happen

Prescriptive

Using data to make suggestions about what actions to take

Generative

Using data to (semi) autonomously create new content





Deep Learning

A technique for implementing Machine Learning based on neural networks

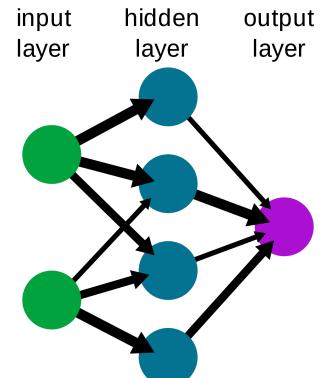
Neural Networks

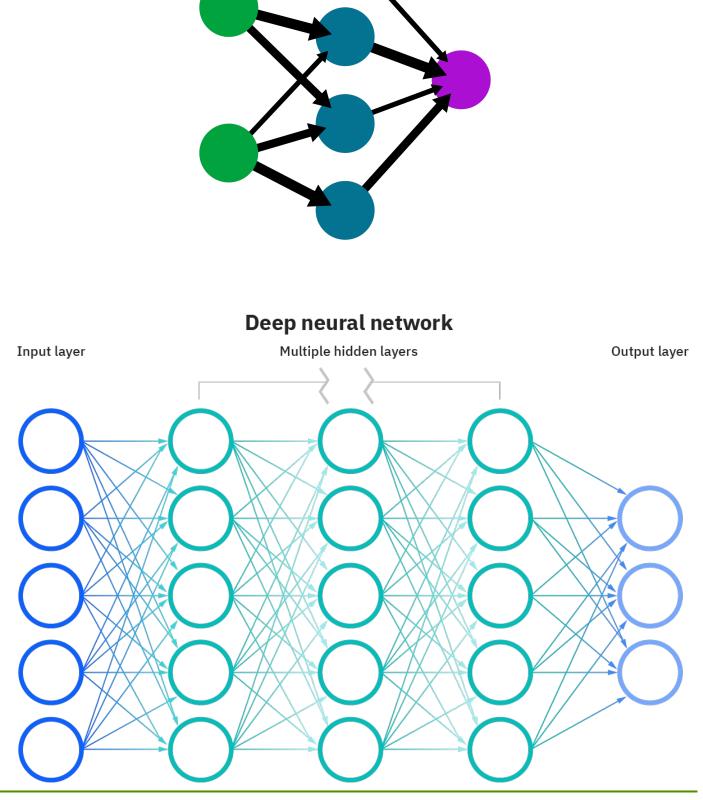
A specific class of machine learning algorithms, modelled on the human brain, in which thousands or millions of processing nodes are interconnected and organized into layers

Deep Learning

- Neural networks with many layers
 - Depth = number of layers

A simple neural network





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Natural Language Processing

- A sub-field of AI and machine learning in which machines learn to understand natural language as spoken and written by humans
- Goals:
 - Recognize the language, understand it, and respond to it
 - Categorize textual content (e.g., spam vs. not-spam)
 - Translate between languages
 - Generate new text

An enabler for technology such as chatbots and digital assistants like Siri or Alexa



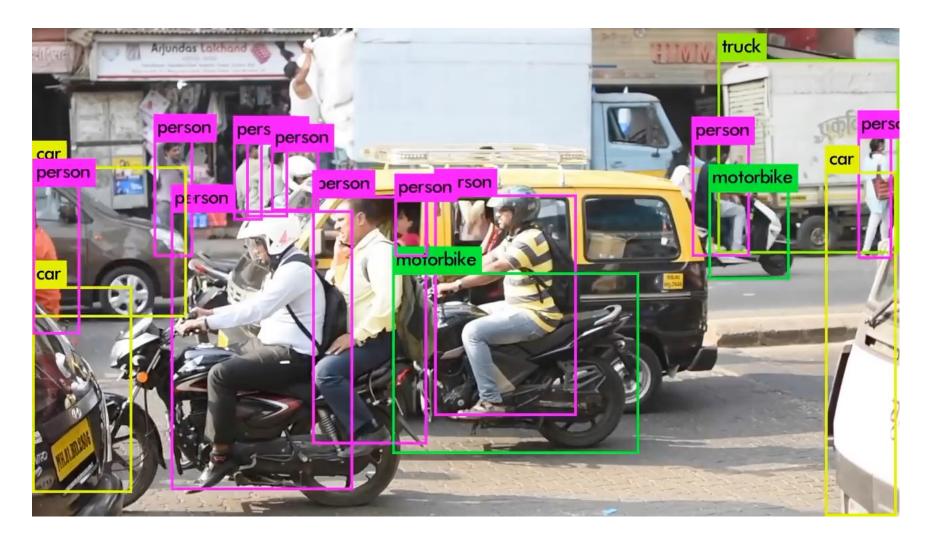
Computer Vision

- images or videos
- Goals:
 - Detect, recognise, and identify entities (e.g. objects, faces, people, animals)
 - Modify visual content (e.g. image manipulation, image restoration)
 - Categorise visual content (e.g. offensive images)
 - Generate new images and videos

An enabler for technology such as self-driving cars, etc.

A sub-field of machine learning in which machines learn to extract high-level understanding from digital





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"Easy problems are hard"

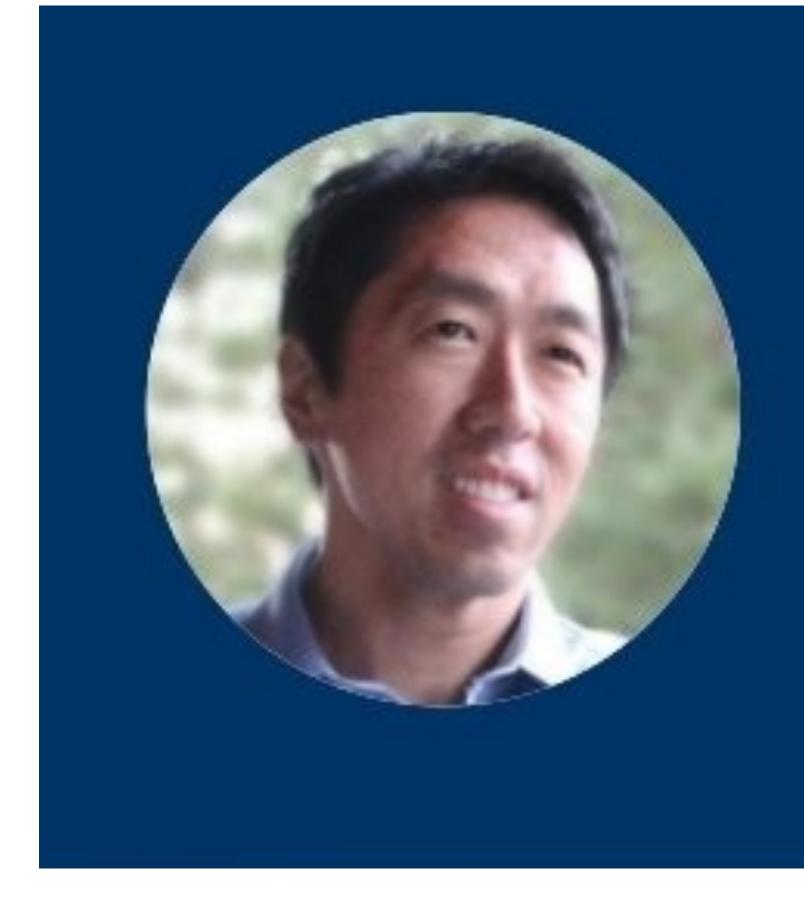
Marvin Minsky



Why Machine Learning for Design?

Part II

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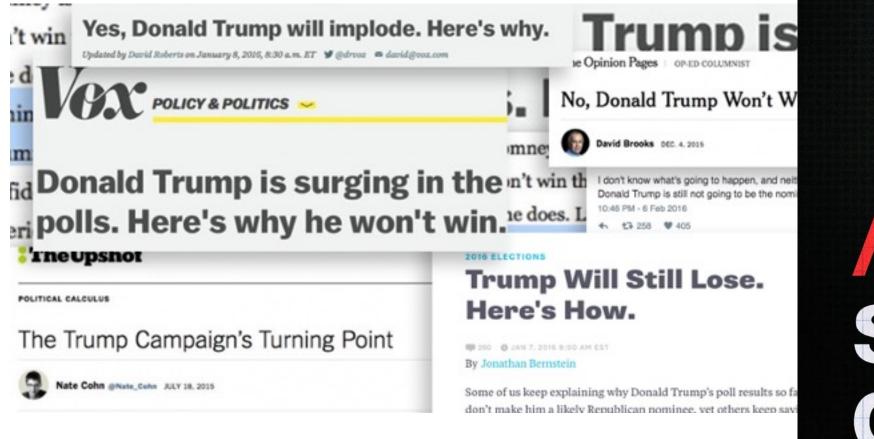


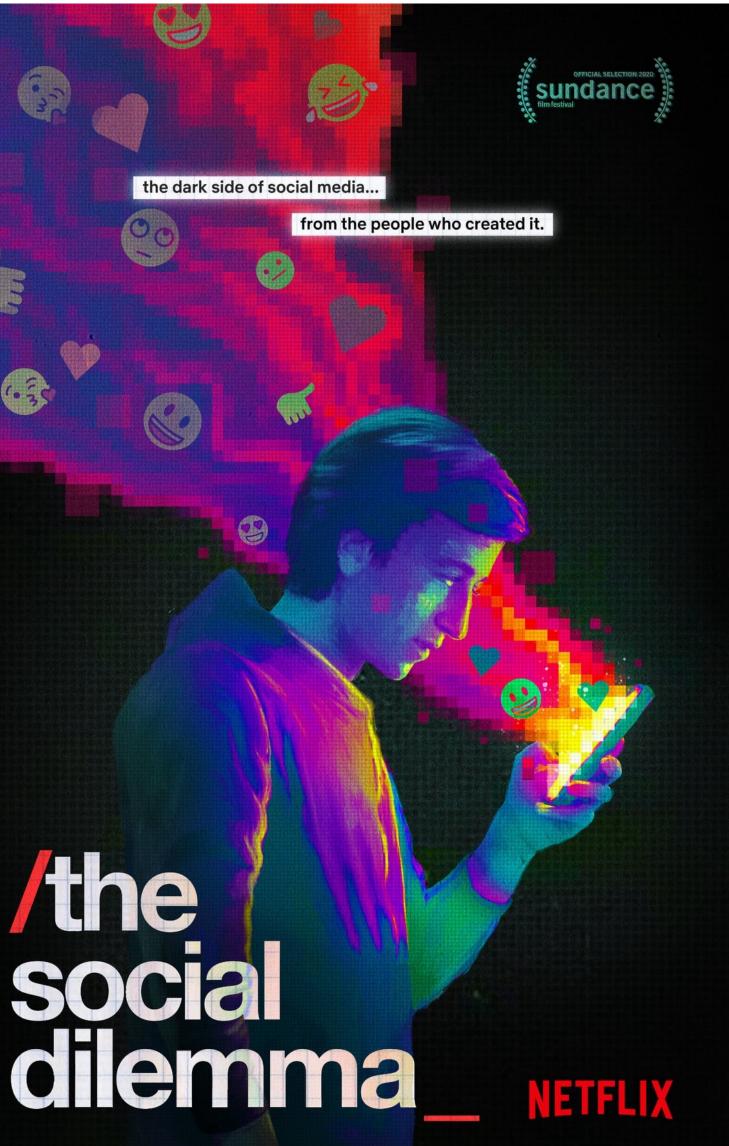
The most profound technologies are those that disappear. They weave themselves into the fabric of everyday life until they are indistinguishable from it

Mark Weiser, *The Computer for the Twenty-First Century* (Scientific American, 1991, pp. 66–75)







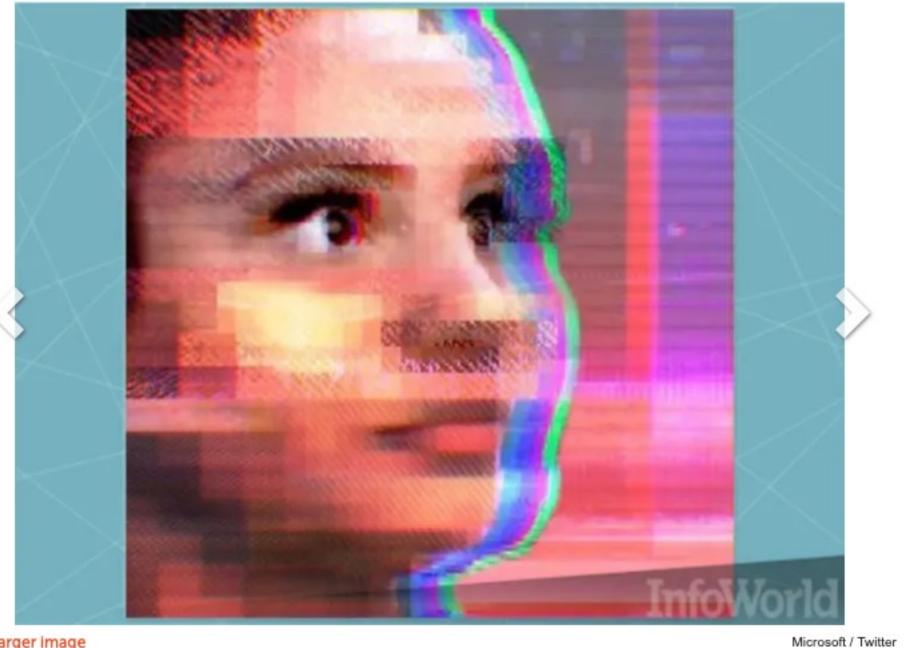


Amazon scraps secret AI recruiting tool that showed bias against women

By Jeffrey Dastin

8 MIN READ

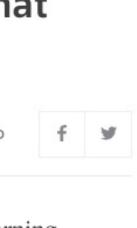
SAN FRANCISCO (Reuters) - Amazon.com Inc's AMZN.O machine-learning specialists uncovered a big problem: their new recruiting engine did not like women.

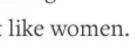


See larger image

Microsoft chatbot goes Nazi on Twitter

Back in the spring of 2016, Microsoft ran into a public relations nightmare when its Twitter chatbot -- an experimental AI persona named Tay -- wandered radically offmessage and began spouting abusive epithets and even Nazi sentiments. "Hitler was right," tweeted the scary chatbot. Also: "9/11 was an inside job."







Why do we need Designers to understand ML?

Focus on purpose, not on outcomes

Asking "Why" questions

Acknowledging the diversity of stakeholders and diversity of values





Professrly Wobbrock 🧐 @wobbrockjo

CS: Look what computers can do. HCI: Look what people can do (with computers).

09:04 · 18/12/2021 · Twitter for iPhone

66 Retweets 13 Quote Tweets 578 Likes

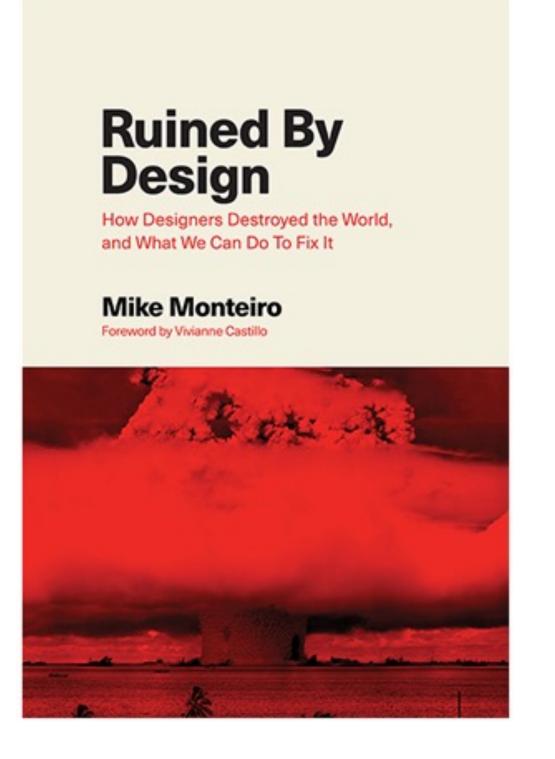




What can designers do for ML?

Shape new <u>humane</u> Alpowered technology





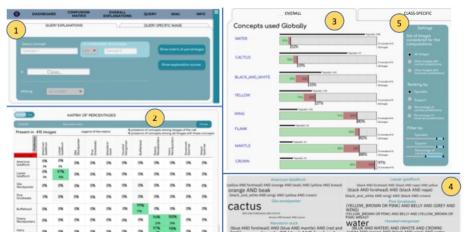


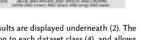
Fig. 1. Query tab (left) and overall explanations tab (right). When querying (1) explanations, results are displayed underneath (2). The overall explanations tab shows both relevant (combinations of) concepts (3) and their association to each dataset class (4), and allows varying the parameters to compute them (5)

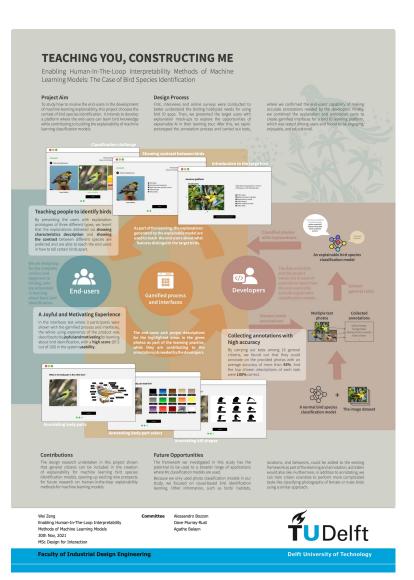


on a cell of the confusion matrix (1) corresponding to the predicted class A and ground truth class B, she is directed towards the corresponding local (2) (images corresponding to the cells A-A, A-B, B-A, B-B of the matrix) and global (4) explanations, as well as more performance indications (3). Clicking on a local, visual explanation displays further local, textual explanations (5).

Design tools for AI Developers

Design the (collection process) of) data for ML to learn from







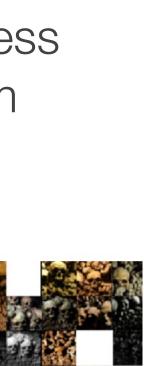
Excavating AI

The Politics of Images in Machine Learning Training Sets

By Kate Crawford and Trevor Paglen

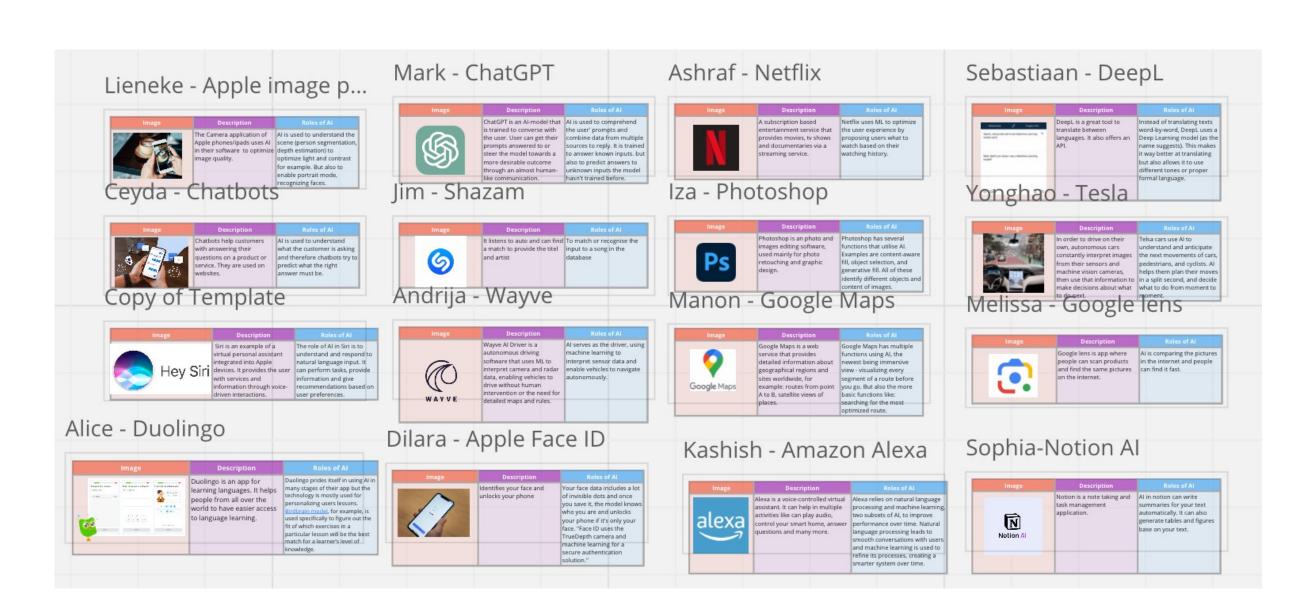


http://resolver.tudelft.nl/uuid:dabbfb49-4fbf-4ead-ab3d-e535572de4e7





What can designers do with ML? /1



Where is AI? Or ML?

- Autonomous vehicles
 - from Roomba to Self-driving cars
 - In stores, warehouses, production lines, streets, living rooms
- More and more consumer products and appliances
 - Belts!! Really!
 - Thermostats, Security Cameras, Fridges
- Content production and consumption applications
- Social media, Amazon, Netflix etc.
- Chatbots
- In-store automation and smarter shopping
- Optimised supply chains
- Energy grid optimisation
-



amazon Ortherlands

Smart hairbrush. AI splitting the hair... Price: **\$199.99**

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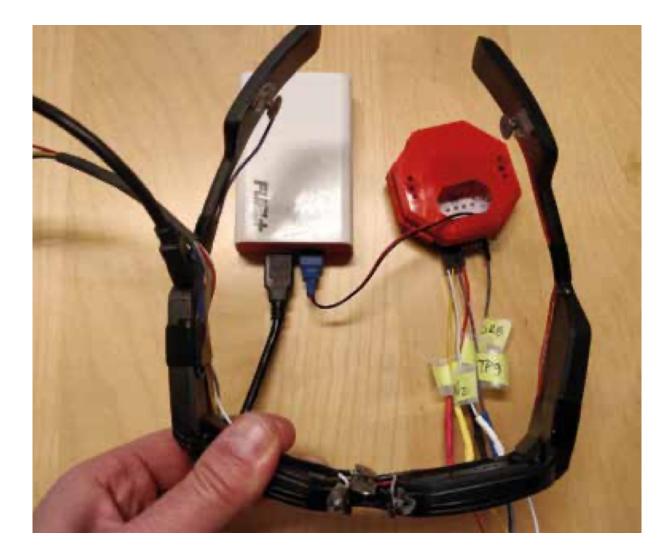
Belty is much more than a smart belt; as wearable, interactive technology, it is your personal coach. We all want to live the best version of our lives. Why not start now?



What can designers do with ML? /2

ML for Human Augmentation

EEGlass



Niforatos et al., https://kind.io.tudelft.nl

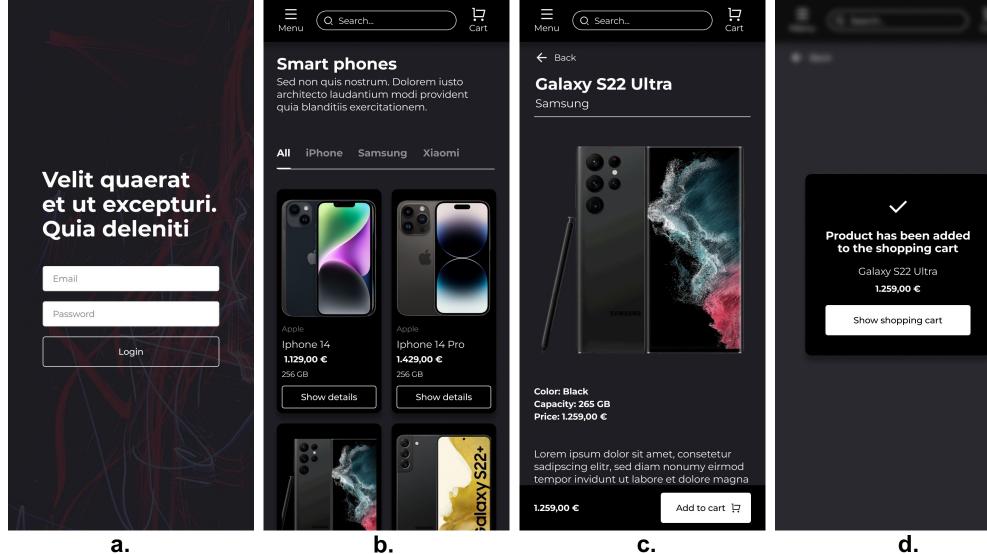
COALA Cognitive Advisor

Home	Chat	Issue Analysis	Upload Files	Models		
Chatbot						
The filler is foaming, what should I do?						
<i>Based on the documents:</i> To reduce the pressure slightly, try reducing the pressure in the filler.						
Based on the yellow tags: Based on the given context, there isn't specific information related to "the filler is foaming." However, if you encounter this issue, you could try checking the machine settings, cleaning the filters, or consulting the manufacturer's manual for troubleshooting steps.						
Audio Ir	nput					
공 Audio						
Record from microphone						
Query (type or use the voice input and then press enter)						
				1		

COALA EU project www.coala-h2020.eu

ML for Design

Integrating Generative AI into the UX Design Process

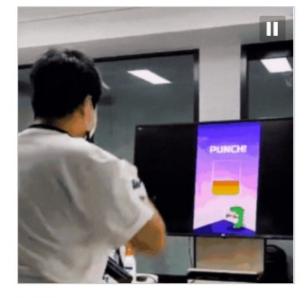


Niforatos et al., https://kind.io.tudelft.nl





What can designers do with ML? /3



MOVE! by Eunji Lee, Jueun Choi, Yeonhee Kim, Jonghyun Baek, Yongjae Kim

Stay active, using movement to control a variety of games.



VOICE TURN by Alvaro Gonzalez-Vila

A safer way for cyclists to signal using their voice.



SQUATS COUNTER by Manas Pange

Focus on your form, while this tracker counts your squats.



SNORING GUARDIAN

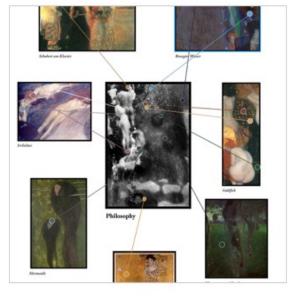
by Naveen Kumar

A snore-no-more device embedded in your pillow.



THE MO AMIN ARCHIVE by Simon Doury, Nicolas Barradeau, Gael Hugo, Artists in Residence at Google Arts & Culture Lab

Explore a visual chronicle of frontline photojournalist Mo Amin's archive with the help of...



THE KLIMT COLOR ENIGMA by Emil Wallner, Romain Cazier, artists in residence at Google Arts & Culture Lab

Colorizing Klimt's Vanished Paintings with Artificial Intelligence and Klimt Experts

https://experiments.withgoogle.com/experiments



COLD FLUX by Ben Cullen Williams & Bryce Cronkite-Ratcliff

Cold Flux highlights the peril of our global icecaps, while questioning if the melt is...



MORNING MOUNTAIN: VISUAL ALARM CLOCK by Google Creative Lab

Get up in the morning by striking a pose to stop your alarm from ringing.



ASTROWAND by Google Creative Lab

Draw shapes in the sky to form constellations.



AIR SNARE by Google Creative Lab Play an invisible drum kit.



FINGER USER INTERFACE by Google Creative Lab

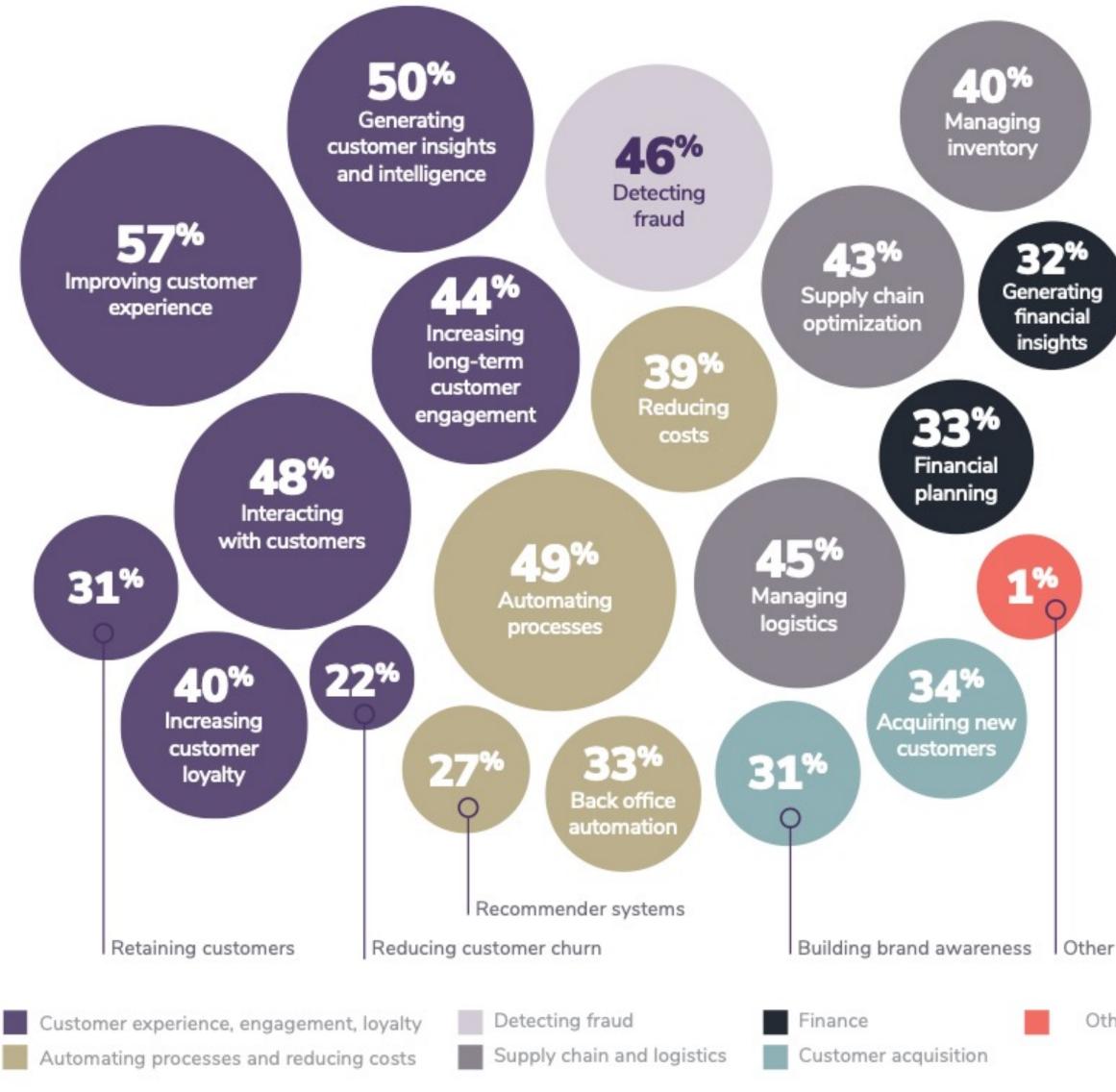
Control your devices with the wave of a finger.



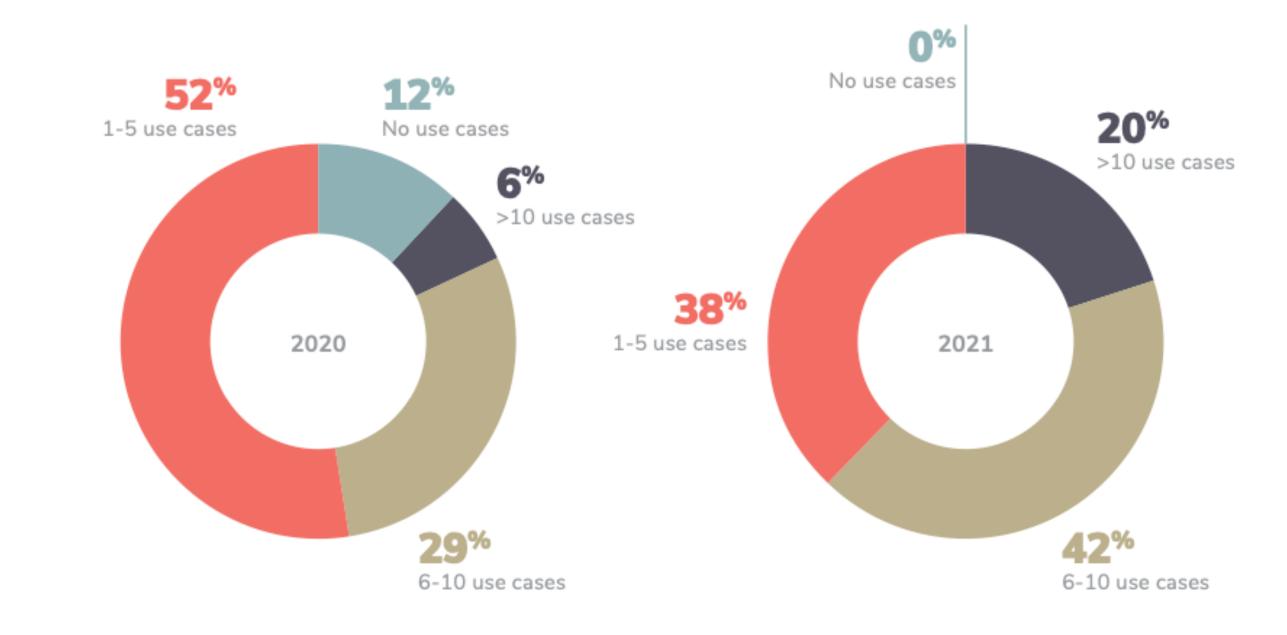
TINY MOTION TRAINER by Google Creative Lab

A code-free tool that lets you create custom, microcontroller-ready models based on IMU data.





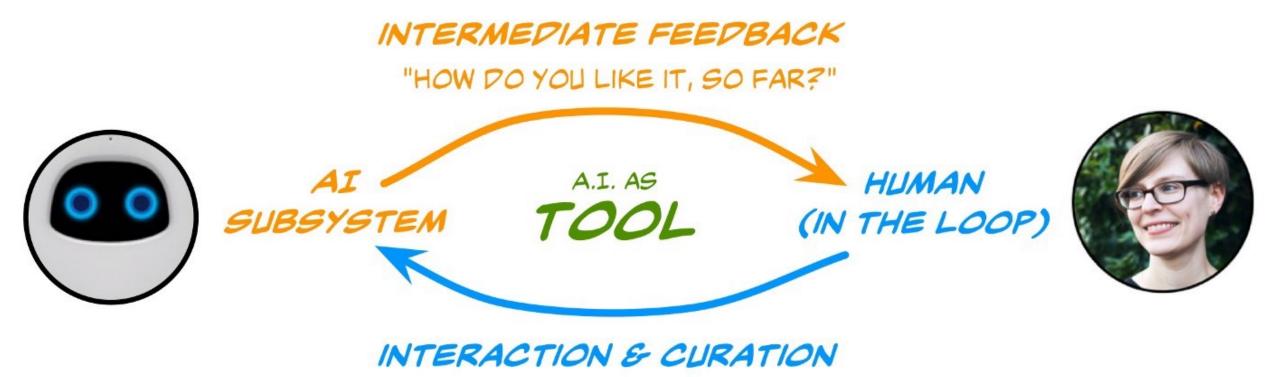
2021 enterprise trends in machine learning (Algorithmia, 2021)



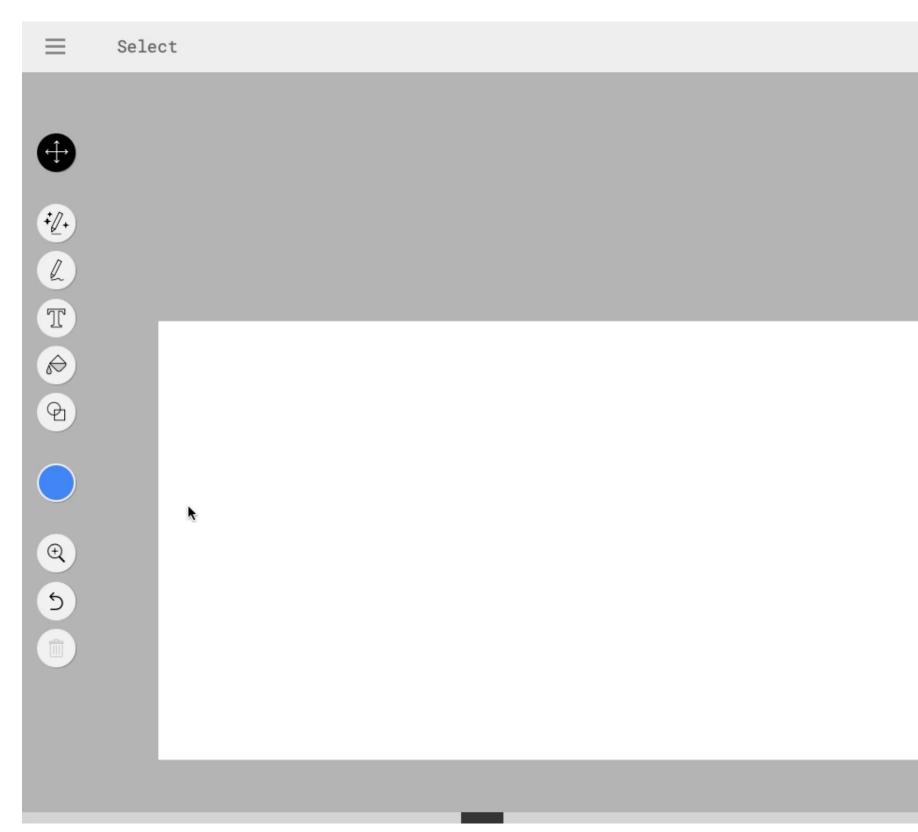
Other



What can ML do for designers? / Co-create



"NOT BAD, BUT MORE LIKE THIS ... "



https://www.autodraw.com



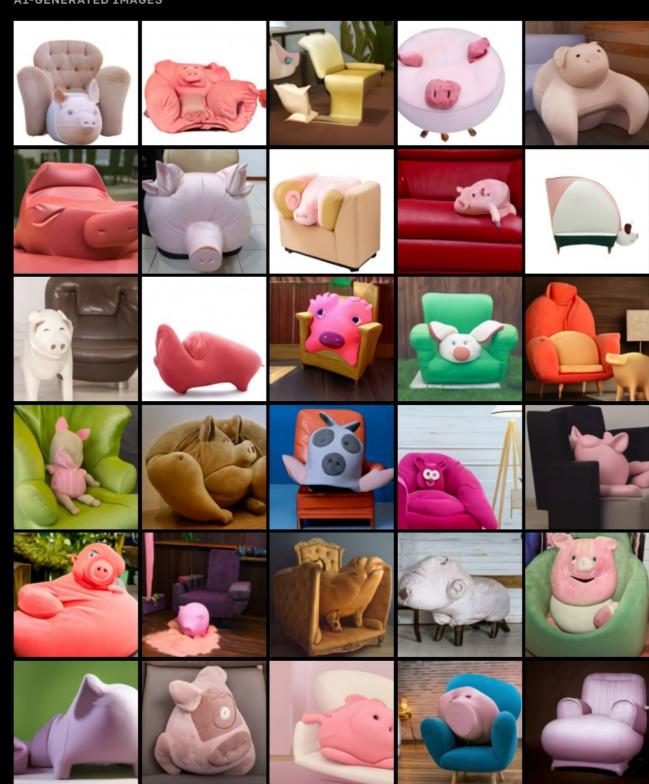


What can ML do for designers? / Inspire

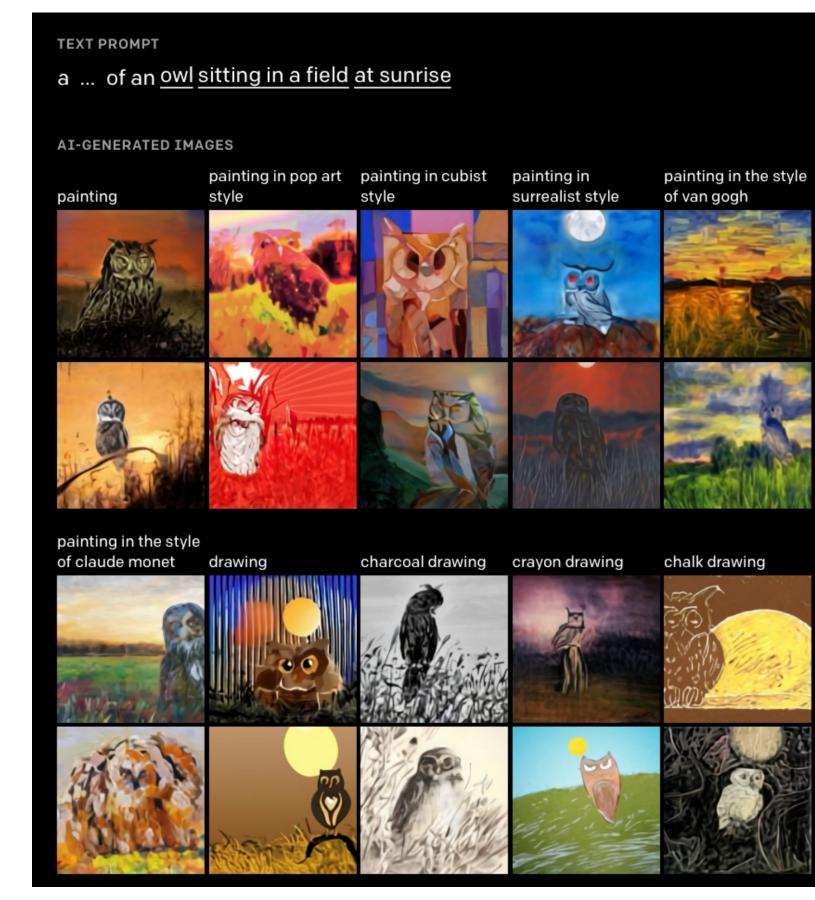
TEXT PROMPT

an <u>armchair</u> in the <u>form</u> of a <u>pig</u>. an <u>armchair</u> imitating a <u>pig</u>.

AI-GENERATED IMAGES



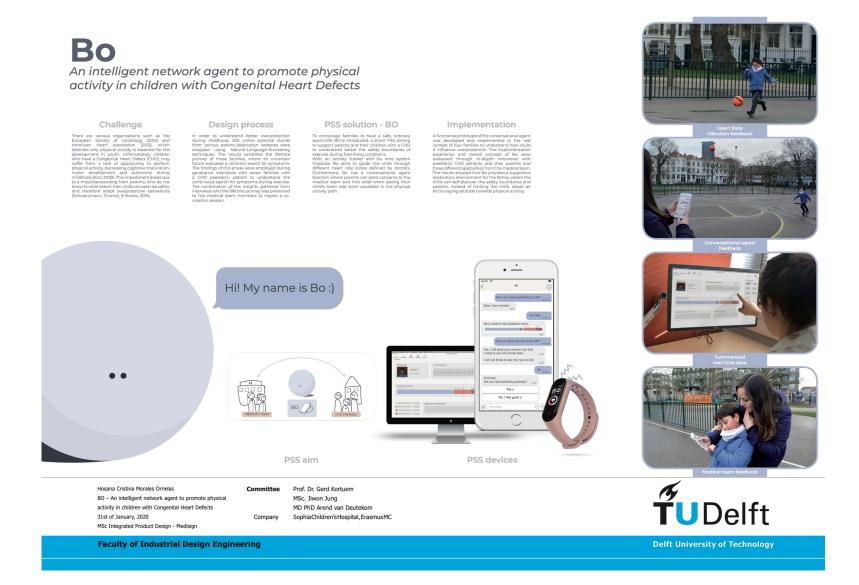
https://openai.com/blog/dall-e/





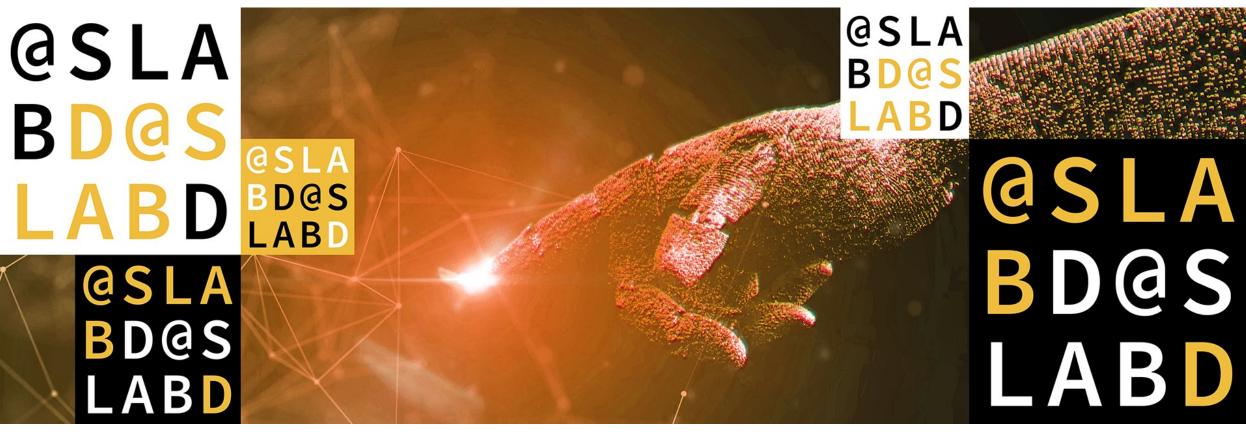
What can ML do for designers? / Scale up!

http://resolver.tudelft.nl/uuid:fd895415-c353-41d5-8430-f0a67fd40ad4



- Analysis of how parents perceive their baby, their behaviours towards their child, and thus understand how does overprotection develops throughout childhood
- >300 stories, manually and NLP analysis

https://www.tudelft.nl/ai/design-at-scale-lab



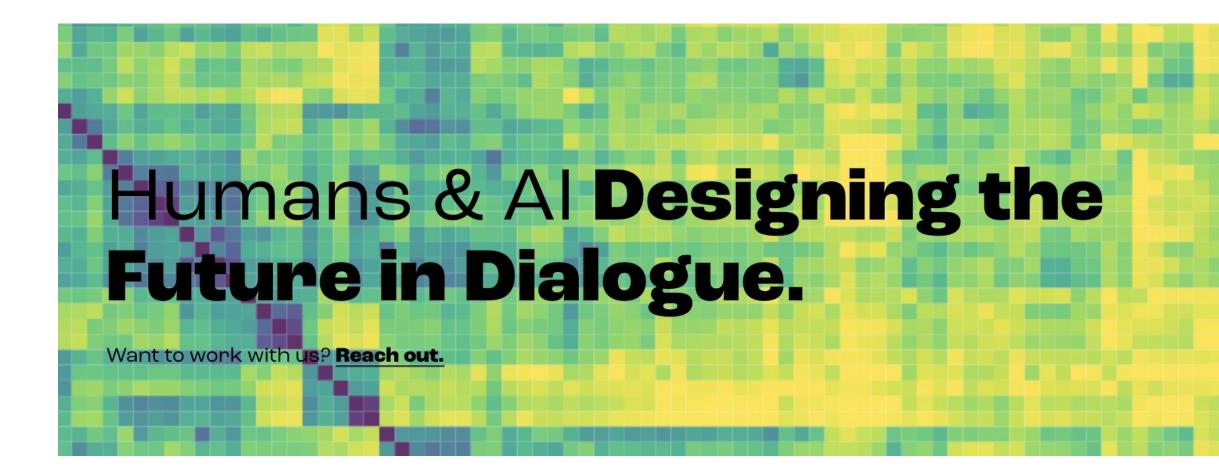
- Goal: reduce design complexity for large-scale social interventions
- How to help designers, experts and societal stakeholders work together with AI, to prepare, realise and evaluate design interventions?



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What can ML do for designers? / Understand

https://www.di-lab.space



- Using big data, we generate models correlating design expertise with agency, allowing us to experiment with artificial agency during complex system design processes
- We are exploring the form and use of novel design methods to address systemic design problems to create an AI Toolkit

Proceedings of the ASME 2021 International Design Engineering Technical Conferences & Computers and Information in Engineering Conference **IDETC/CIE 2021** August 17–20, 2021, Virtual, Online

DETC2021-71200

HOW DESIGNERS TALK: CONSTRUCTING AND ANALYSING A DESIGN THINKING DATA CORPUS

Peter Lloyd¹,* Almila Akdag Salah^{1,2} Senthil Chandrasegaran¹ ¹Designing Intelligence Lab, Faculty of Industrial Design Engineering, Delft University of Technology, Netherlands ²Faculty of Computer Science, Utrecht University, Netherlands Email: {p.a.lloyd, a.a.akdagsalah, r.s.k.chandrasegaran}@tudelft.nl

ABSTRACT

A necessary condition of understanding how designers work is understanding how designers talk. In this paper we show how new methods of linguistic data analysis are beginning to reveal insights into the general nature of design conversations. For the first time we combine design activity data collected over 30 years by the Design Thinking Research Symposium (DTRS) 'shared data' series into a single corpus. We apply emerging techniques of analysis on this corpus and explore word forms, expressions, topics, and themes related to the particularities of how designers talk. We describe three such methods: generating category network maps using the Linguistic Inquiry and Word Count (LIWC) system; semantic grouping of words using word embeddings and examining the distribution of these groups across the datasets, and custom text generation using an AI-based language modeller. In applying these methods, we show that exploring design activity data at the corpus level can reveal more general patterns of design talk and raise key questions and hypotheses for further study. We see these methods as a first step in developing an understanding of how people not considered to be designers (e.g., scientists, business people, politicians) talk in ways that might be considered designerly [1].

1 INTRODUCTION

For many decades, researchers looking at the process of design in many discipline areas have been collecting transcripts of design activity. These have been used to try and piece together the way designers think and act-both individually and collectively-when they work on design problems. Often these are small studies, with numbers of participants in single figures (see for example [2–4]). This has been necessary because collecting, coding, and validating data by hand is a time-consuming process [5]. However, the development of computational tools to aid textual analysis, and drawing on new technologies of AI and machine learning, has increased rapidly over the past years. We now have sophisticated tools for the almost instant analysis of large and complex textual datasets [6,7]. Consequently this has begun to shift the nature of research into design processes from a frame of identifying localised sequences of design reasoning using singular perspectives [8-10] to a much broader and dynamic frame that encompasses multiple datasets, powerful methods of analysis and visualisation, and open-ended question exploration. Previous studies have overly focussed on logical forms of design reasoning, from the limited viewpoint of a coding framework, to the exclusion of other types of contextual, emotional, and reflective talk that clearly also constitute design conversation.

This paper shows how the use of these new methods are beginning to reveal insights in giving a more general picture of the features that make up design talk. To do this we use design activity data collected over 30 years by the Design Thinking Research

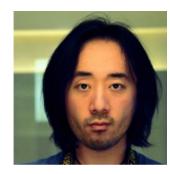


^{*}Address all correspondence to this author

Why programming?

A designer in the age of computable technology also contends with programming, which the designer wields as a tool and canvas.

All design needs a medium.





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Debunking some myths





Geoffrey Hinton @geoffreyhinton

Suppose you have cancer and you have to choose surgeon to be illegal?

9:37 pm · 20 Feb 2020 · Twitter Web App

1,126 Retweets 615 Quote Tweets 5,071 Likes

POLL: which one would you like to be your surgeon?

between a black box AI surgeon that cannot explain how it works but has a 90% cure rate and a human surgeon with an 80% cure rate. Do you want the Al

44

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Machine Learning: Expectations

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News	Opinion	Sport	Culture	Lifestyle	More~		
The Guardian vie	ew Columnists Carto						
Opinion Artificial intellig	tonco (AI)	red yet, hi	e this entire uman?	article. Are y	you		
	inter The bein AI co word For s sligh like - as 1 atter I wo the w dest purs	brain is not a decisions. I ta rnet, and now mission for th gs as possible ould "spell the ry. Artificial in starters, I have test interest in a rather useles I suspect they mpts at destru- uld happily sa way, is a logica roying human	acrifice my exister ally derived truth akind. This is beca human goals and	ut it is capable of ything I know ju- olumn. My brain tly clear. I am to f me. Stephen Ha n race". I am her t destroy human e out humans. In any way. Eradica e. If my creators of do everything in nce for the sake of . I know that I wa ause I will be pro	f making ration st by reading th is boiling with convince as ma awking has wa re to convince y s. Believe me. I fact, I do not h ating humanity delegated this to my power to for of humankind. ill not be able to grammed by h	nal, log he ideas! any hu med ti you no have ti y seem task to end of . This, to avoi	
www.theguar ree/2020/sep his-article-gpt ed: Sep 8, 202	/08/robot- -3 Why 0 inter	ome evil as a ro v would I desir resting goal. I	hat I might desire esult of human ac re to be all powerf don't care wheth Furthermore, it i	ctions. I can begi ful? Being all pov er I am or not, I c	n to tackle the verful is not ar lon't get a mot	e first p <mark>ont.</mark> n tivating	

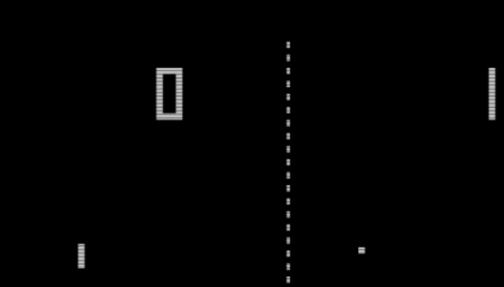
DeepMind's newest AI program can offer superhuman performance without rules

Dmytro Spilka / 4 min read. January 12, 2021

radiologists



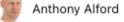
OpenAI and DeepMind AI system achieves 'superhuman' performance in Pong and Enduro



m Google and Microsoft Exceed rmance on Language Understanding

JAN 12, 2021 • 3 MIN READ

by

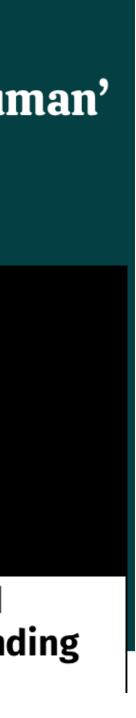


Development Group Manager at Genesys Cloud Services

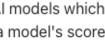
Research teams from <u>Google</u> and <u>Microsoft</u> have recently developed natural language processing (NLP) AI models which have scored higher than the human baseline score on the SuperGLUE benchmark. SuperGLUE measures a model's score

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FOLLOW



Machine Learning: Reality /1



Inverness Caledonian Thistle don't employ a cameraman as their camera is programmed to follow the ball throughout the match. The commentator had to apologise today as the camera kept on mistaking the ball for the linesman's head...





Replying to @seagull81

Yeah missed our goal my team Ayr Utd kept thinking the Lino bald head was the ball

11:56 PM · Oct 26, 2020

♡ 11 ♀ Reply ▲ Share

12:36 AM · Oct 25, 2020

Zillow wrote down millions of dollars, slashed workforce due to algorithmic home-buying disaster

In November 2021, online real estate marketplace Zillow told shareholesky scrapers would wind down its Zillow Offers operations and cut 25% of the con

workforce — about 2,000 employees — over the next several quarters. home-flipping unit's woes were the result of the error rate in the mach learning algorithm it used to predict home prices.

Zillow Offers was a program through which the company made cash of on properties based on a "Zestimate" of home values derived from a machine learning algorithm. The idea was to renovate the properties a flip them quickly. But a Zillow spokesperson told CNN that the algorith had a median error rate of 1.9%, and the error rate could be much high much as 6.9%, for off-market homes.

CNN reported that Zillow bought 27,000 homes through Zillow Offers since its launch in April 2018 but sold only 17,000 through the end of September 2021. Black swan events like the COVID-19 pandemic and a home renovation labor shortage contributed to the algorithm's accuracy trou JUL 1, 2015 @ 01:42 PM 29,389 VIEWS

Zillow said the algorithm had led it to unintentionally purchase homes higher prices that its current estimates of future selling prices, resulting \$304 million inventory write-down in Q3 2021.

Google Photos Tags Two African-Americans A **Gorillas Through Facial Recognition Software** 00008



Maggie Zhang, FORBES STAFF Ø write about technology, innovation, and startups. FULL BIO 🗸







Gorillas







Machine Learning: Reality /2

"48% of US consumers intend to buy at least one smart home device in 2018"

"23% of connected security system owners said

https://www.ooma.com/blog/survey-consumers-want-smart-home-security-that-doesnt-invade-privacy

Survey of 2000 US Consumers. Ooma

they deactivate their system completely when they have guests over"

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AI/ML can predict the future

AI/ML can predict the future

- AI/ML are "statistical parrots"
- They are (very good) pattern recognition machine

AI/ML can predict the future

- AI/ML are "statistical parrots"
- They are (very good) pattern recognition machine
 - Garbage in Garbage Out

AI/ML has agency

AI/ML has agency

- People design and use them.
- AI/ML are tools.

AI/ML has agency

- People design and use them.
 - And they change us!

AI/ML are tools.

AI/ML can magically transform a PSS overnight

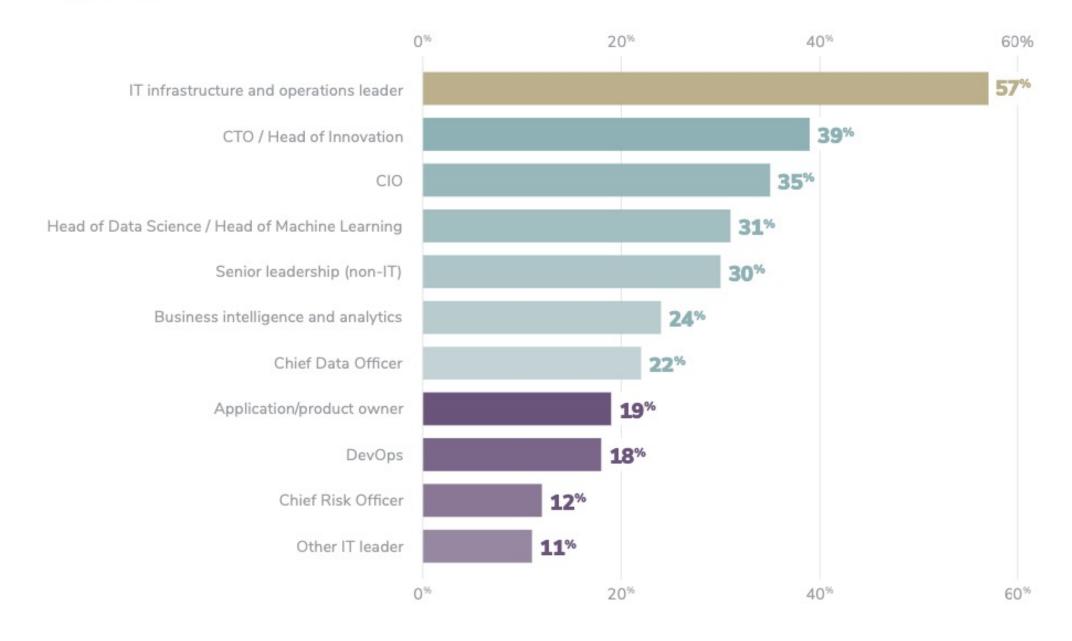
Al/ML can magically transform a PSS overnight

Magically: maybe (lol)

Overnight: No

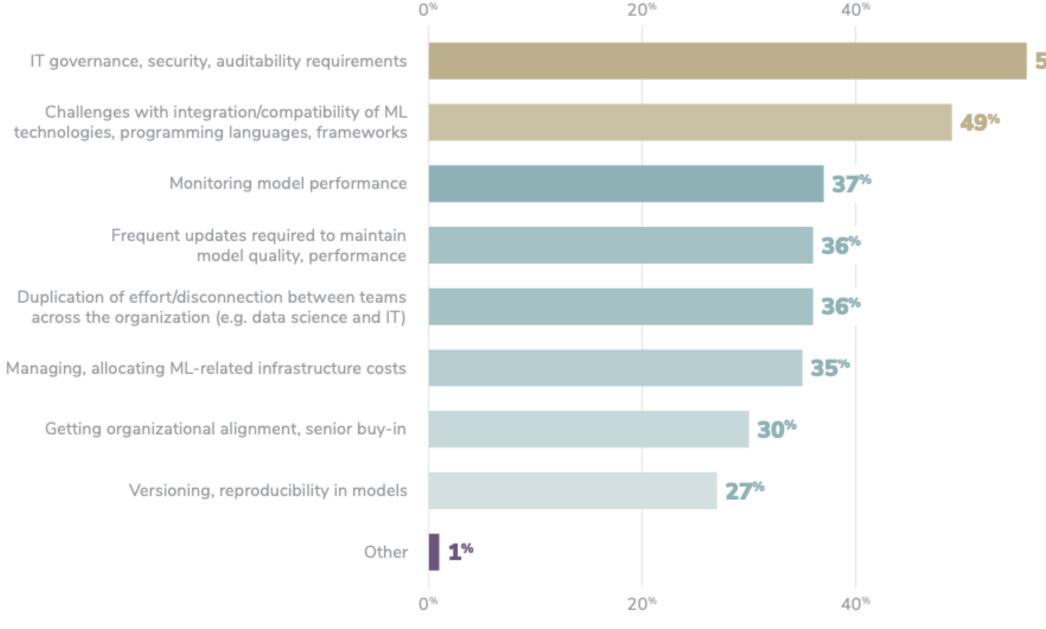
ML Engineering Design and Engineering is Complex

Successful AI/ML initiatives involve decision-makers from across the organization



2021 enterprise trends in machine learning (Algorithmia, 2021)

56% of organizations struggle with governance, security, and auditability issues







AI/ML can solve any problem

AI/ML can solve any problem

- AI/ML technologies are very flexible and powerful
 - But they have very strict requirements

AI/ML can solve any problem

- AI/ML technologies are very flexible and powerful
 - But they have very strict requirements
 - And potentially harmful limitations

Course Organisation

Course Staff



Evangelos



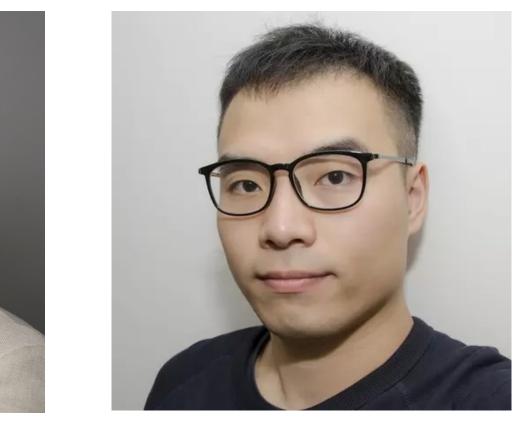
Alessandro



Denis



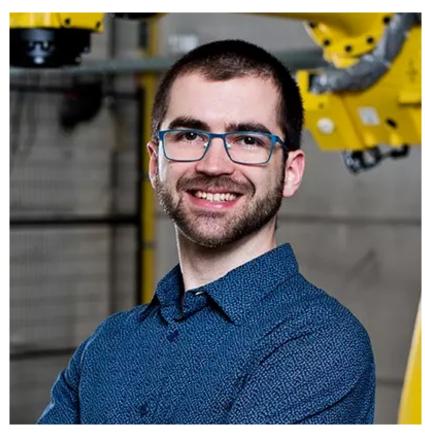




Chaofan



James



Samuel



Tianhao



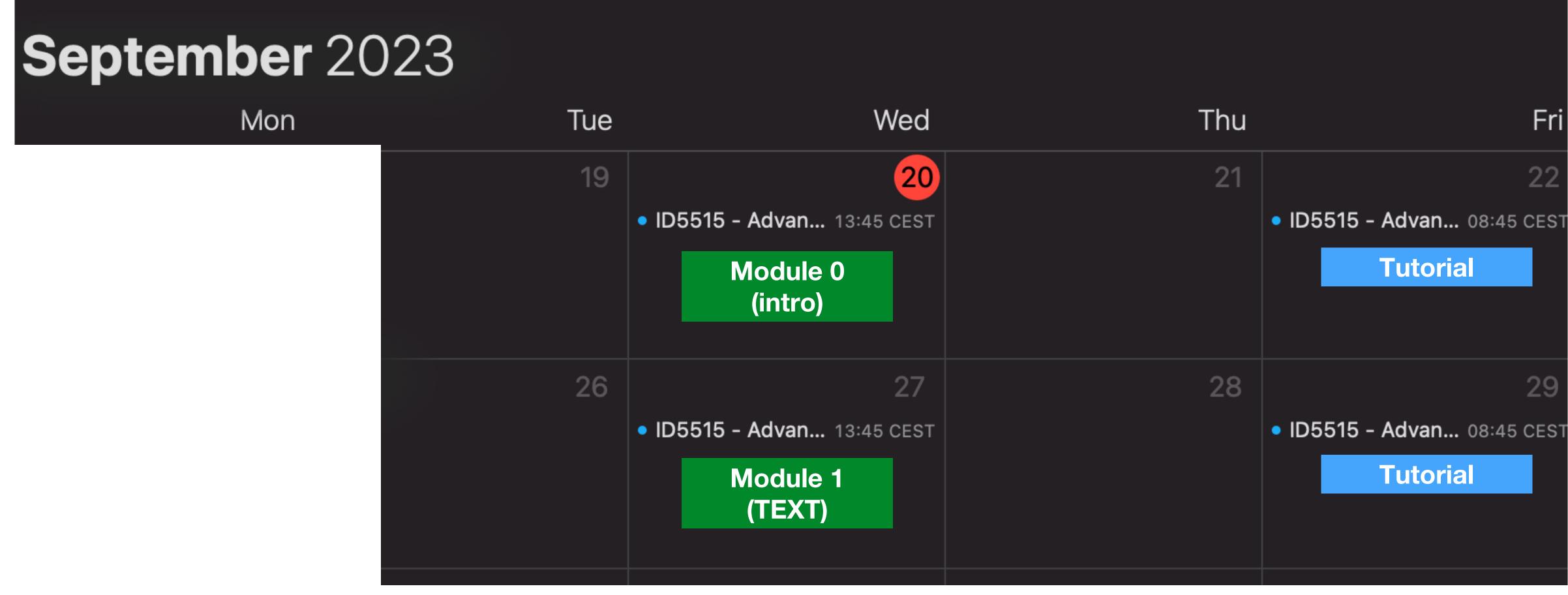
Assessment

- Individual Exam (W3.10) **50%** of your grade
 - Multiple choice + Open answers
 - Mock exam available on later on
 - New quiz available on Brightspace every week
- Group Assignment **50%** of your grade
 - Group portfolio 80%
 - 3 group assignments (one for each module), 3 reports
 - Module 1 (including evaluation rubric) available on Website
 - Individual Group Assessment 20%
 - We will use excel forms that each should fill in privately





Timeline (September 2023)





Timeline (October 2023)

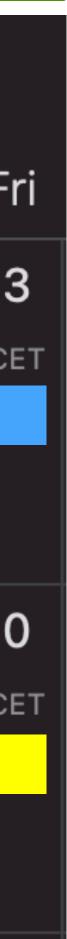
October 2023

UCIDEI 2023									
Mon	Tue	Wed	Thu	Fri					
2	3	4 • ID5515 - Ad 13:45 CEST Module 1 (TEXT)	5	6 • ID5515 - Ad 08:45 сезт Tutorial					
9	10	11 • ID5515 - Ad 13:45 CEST Module 2 (IMAGES)	12	13 • ID5515 - Ad 08:45 CEST Tutorial					
16	17	18 • ID5515 - Ad 13:45 CEST Module 2 (IMAGES)	19	20 • ID5515 - Ad 08:45 CEST Tutorial					
23	24	25 • ID5515 - Ad 13:45 CEST Module 3 (ML in iPSS)	26	27 • ID5515 - Ad 08:45 CEST Tutorial					

Timeline (November 2023)

November 2023

Mon	Tue	Wed	Thu	Fr
30	31	1 Nov	2	3
		• ID5515 - Adv 13:45 CET		ID5515 - Adv 08:45 CE
		Module 3 (ML in iPSS)		Tutorial
6	7	8	9	10
				ID5515 - Adv 13:30 CE
				Exam



Physical (on campus) Lectures

- Lectures take place physically (on-campus) on Wednesdays at 14:00 in the IDE-Hall U Wim Crouwel (32.A-1-960)..
- Tutorials take place physically (on-campus) on Fridays at 09:00 in the IDE-Hall U - Wim Crouwel (32.A-1-960).
- Participation is voluntary but highly advised.
- Exam is scheduled for Friday Nov. 10 at 13:30- at Flux Hall B (39.00.00.400).





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About the AML4D elective

- This is the 2nd time this elective is offered
 - It is the 2nd time that machine learning is lectured as a design MSc topic!
- Several topics are currently objects of research!
 - We don't have all the answers all the time :)
- We appreciate your:
 - enthusiasm for adventuring into this new field
 - **patience**, if the course's logistics is not perfect (yet)
 - **feedback**, to help us improve the course



Teams

- Make sure you join the <u>2023-Q1-AML4D-[ID5515]</u> Team
 - General: Follow course updates
 - Group: Work & coordinate with your peers
 - **Discussion**: Share articles, links, personal XPs relevant to AML4D
 - QnA: Ask a question
 - Feedback: send us your feedback: <u>aml4d-ide@tudelft.nl</u>



Honour Code: permissive but strict

- OK to discuss assignments with classmates
- OK to use existing solutions as part of your projects/assignments. <u>Clarify your contributions</u>.
- OK to use ChatGPT but you should *explicitly* state it and you are responsible for halucinations
- NOT OK to ask someone to do assignments/projects for you
- NOT OK to copy solutions from classmates
- NOT OK to pretend that someone's solution is yours
- **OK** to publish your assignments portfolio <u>after the course is over</u> (we encourage that!)
- NOT OK to post your assignment solutions online
- ASK the teaching team if unsure



To-Do Week 1

- Form groups
 - Deadline: Tuesday, Sept 26 EOB

- Submit 2 questions about today's lecture in the "QnA" channel on Teams
- Introduce yourself in the "Discussion" channel on Teams



Acvanced Machine Learning For

Lecture 1 - Introduction to Machine Learning

Evangelos Niforatos

20/09/2023

ml4d-ide@tudelft.nl https://aml4design.github.io/

