

Computational Futures & Artificial Intelligence

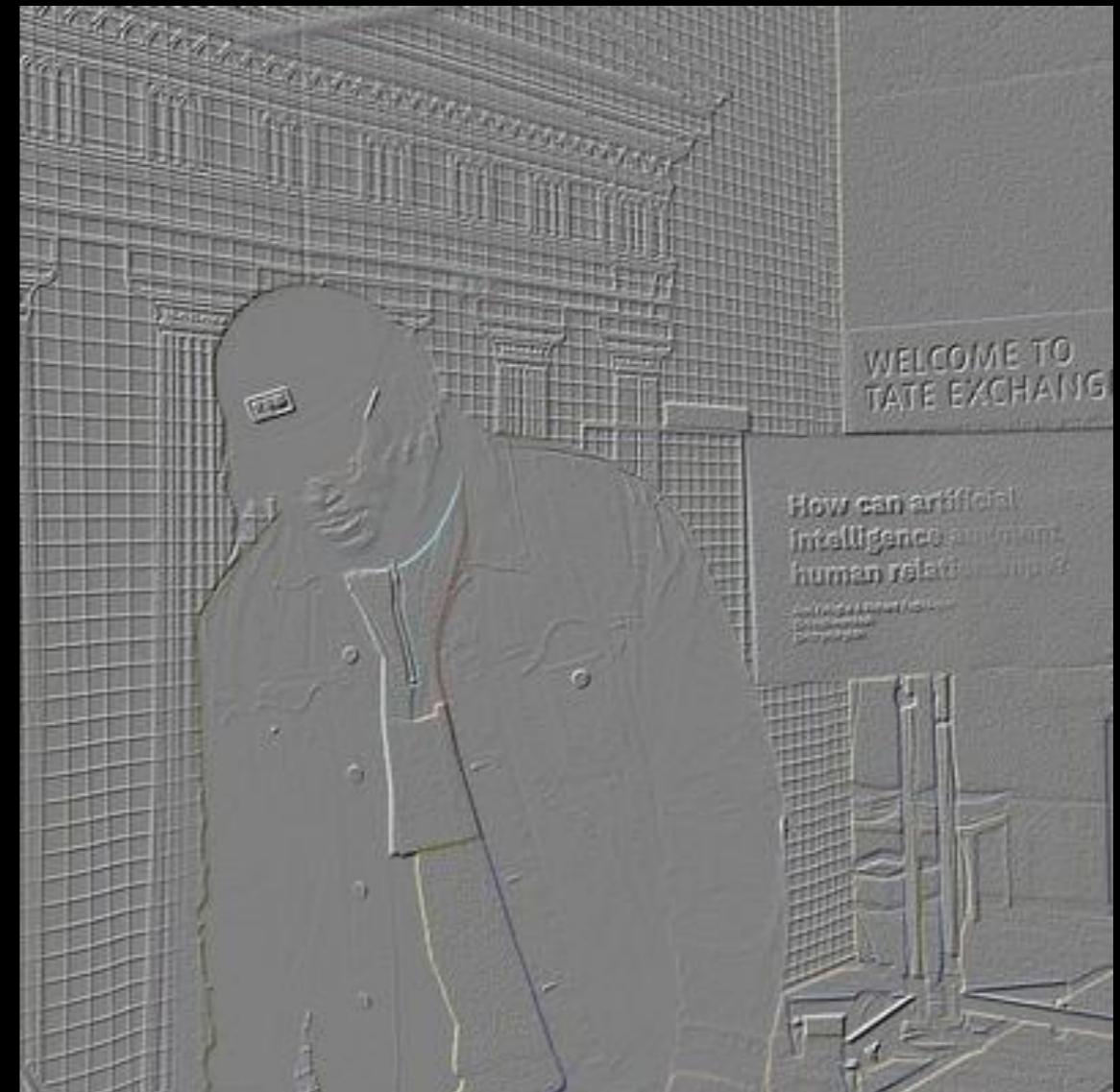
(a introduction lesson)

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@creativetechish

Alex Fefegha
of **COMUZI** - a design studio

Design+Code Person
Digital Interaction ‘Artist’

AI and Creativity
AI and its future impact on
society.



Creative Technologist noun

Definition:

On a day to day, my role alternates between being a design researcher, interaction designer & a maker/hacker creating emerging technology prototypes which demonstrate novel and effective ideas.

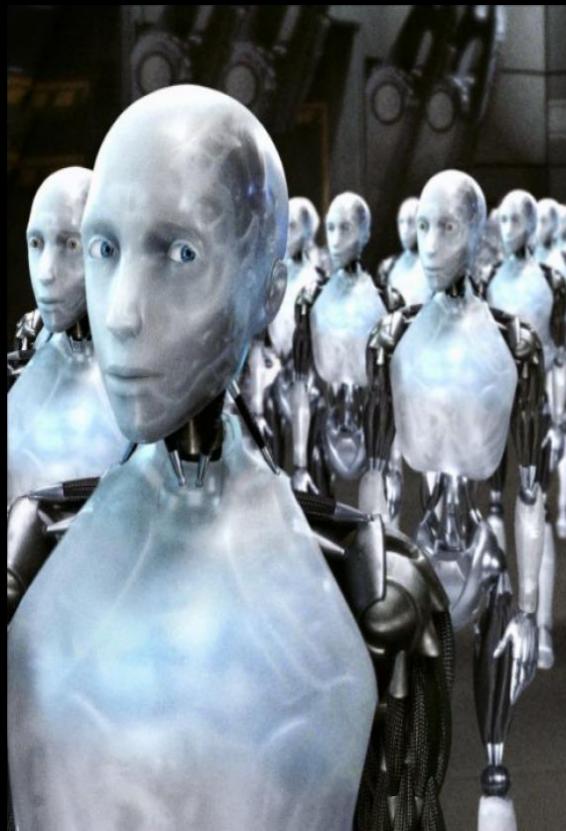
Positive Human Interaction

Reward
Accessibility
Control
Safety
Usability
Benefiance
Autonomy
Eudaimonia

Emerging Technology

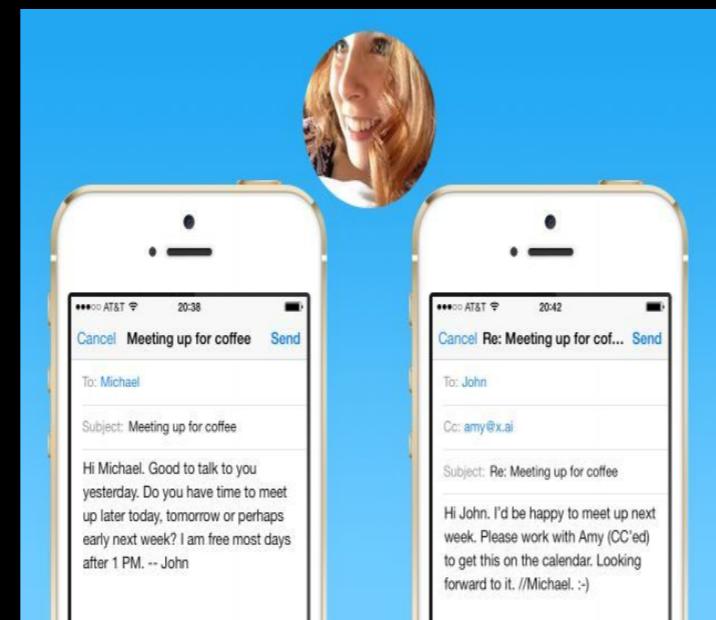
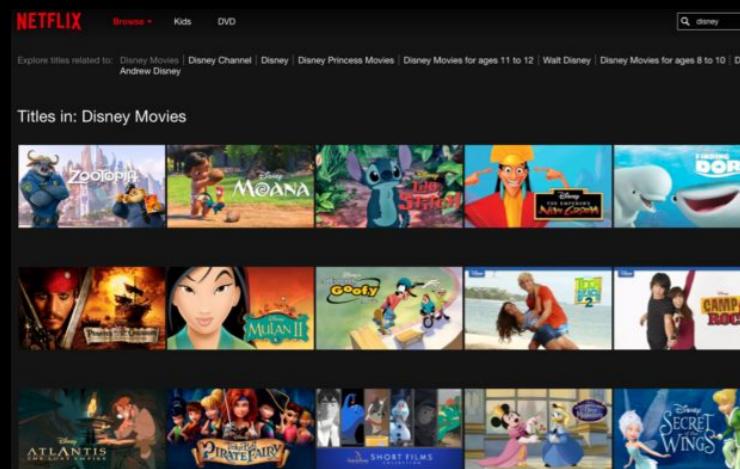
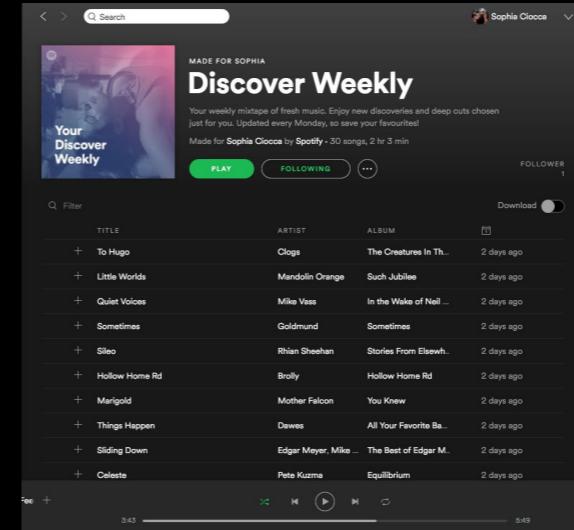
AI & ML
Robotics
AR
VR
Blockchain
IoT
New Interfaces
Mobile & Web

What is Artificial Intelligence?



Yes it is Terminator!

AI is a part of everyday life.



Artificial Intelligence *noun*

Definitions:

The art of how to make computers do things at which humans are currently better.

The process of machines learning to think like humans.

- **Chatbots**

(Natural Language Processing)

- **Personal Assistants (Siri, Alexa)**

(Natural Language Processing, Speech Processing + Machine Learning)

- **Autopilot by Tesla**

(Machine Learning, unstructured Data and situational awareness)

- **Deepdream**

(Machine Learning)

- **AlphaGo**

(Machine Learning)

- **IBM Watson**

(Machine Learning, speech, Structured and unstructured data)

Narrow AI

Basic Tasks

‘AI’

Deep AI

Continuously learning

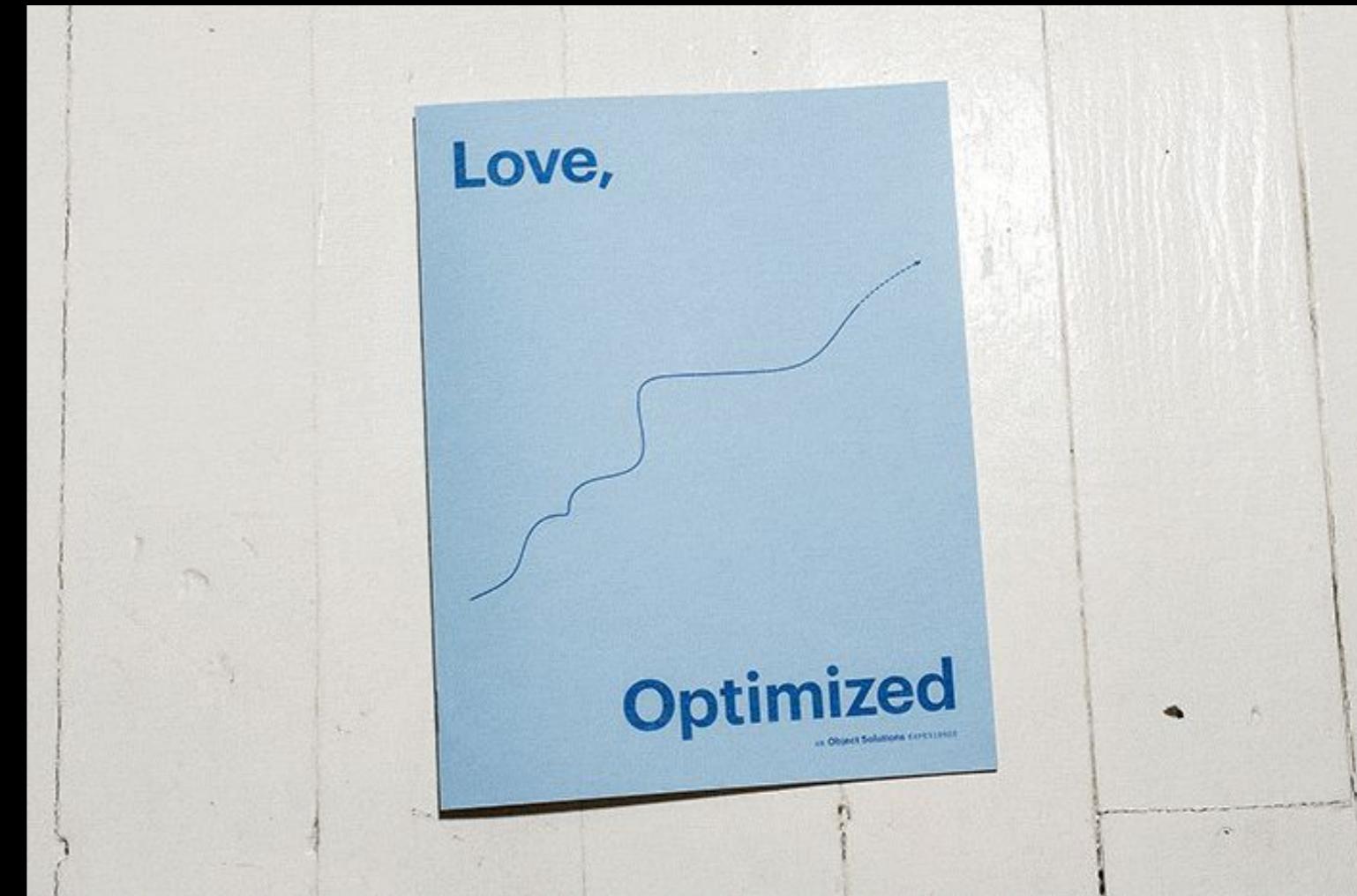
Alex's obsession -
active collaborative
interaction between
humans and machines.

to be fair, I spend
most of my time
building bots.

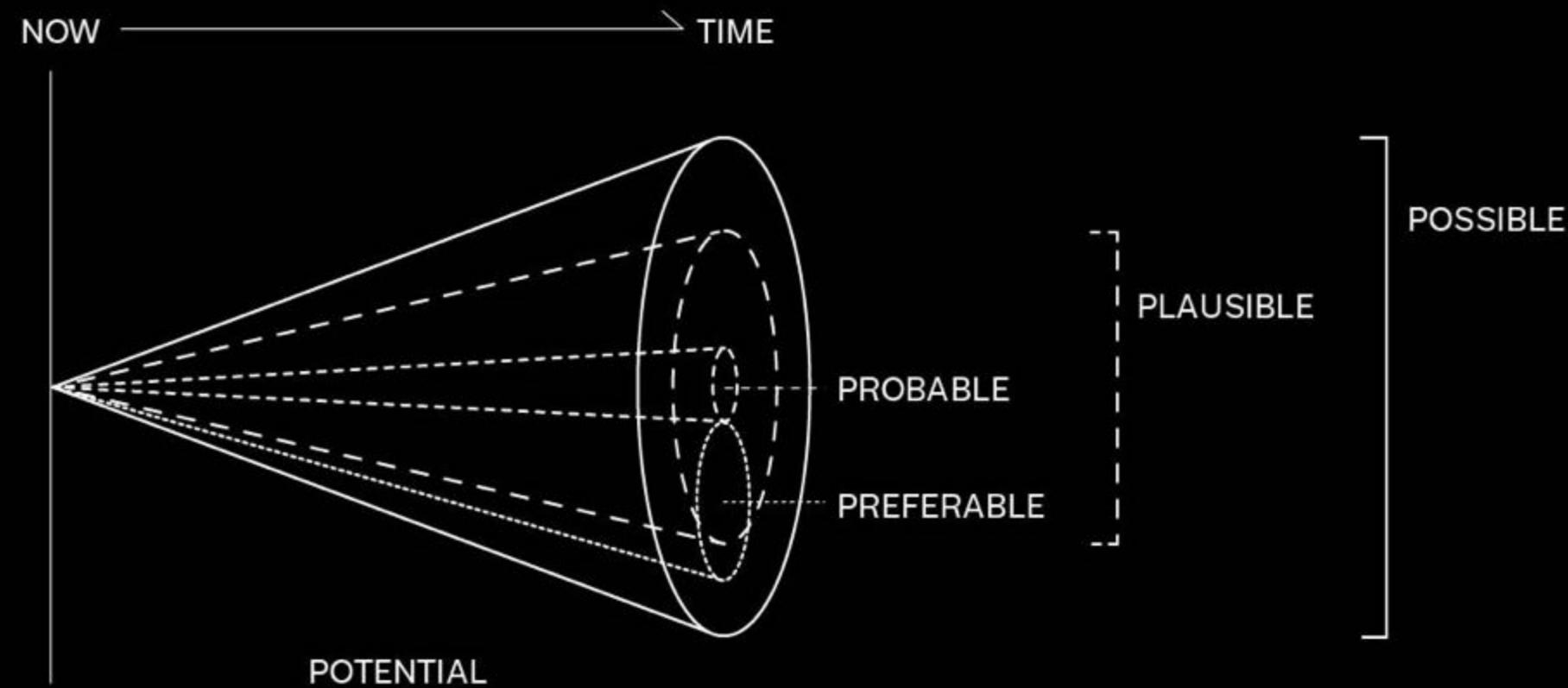
I spend most of our time ‘attempting’ to build prototypes.

(design+code artefacts—digital or physical—whose main goal is to provoke discussion among different types of users and stakeholders.)

Design Fiction



Design being used as a medium to critique social, cultural and ethical implications of emerging technologies and trends.





Mapping Ethical Futures with Artificial Intelligence

Alexander Fefegha

Masters Dissertation

Central Saint Martins, University of the Arts London

MA Innovation Management 2018

Many of the narratives and criticism displayed by design fiction practitioners seem to only apply to the aesthetic concerns of the intellectual northern European middle classes.

so I am gonna need you all create some different futures.

by the way who are you?
where are you from?
what is your human super power?
what do you do?
what made you sign up to the course?
what are your expectations?

WTF is a feminist
conversation?

Could there ever be a
feminist response to ‘Hey
Alexa, what’s the weather
like today?’

Building feminist voice bots



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FEMINIST
INTERNET



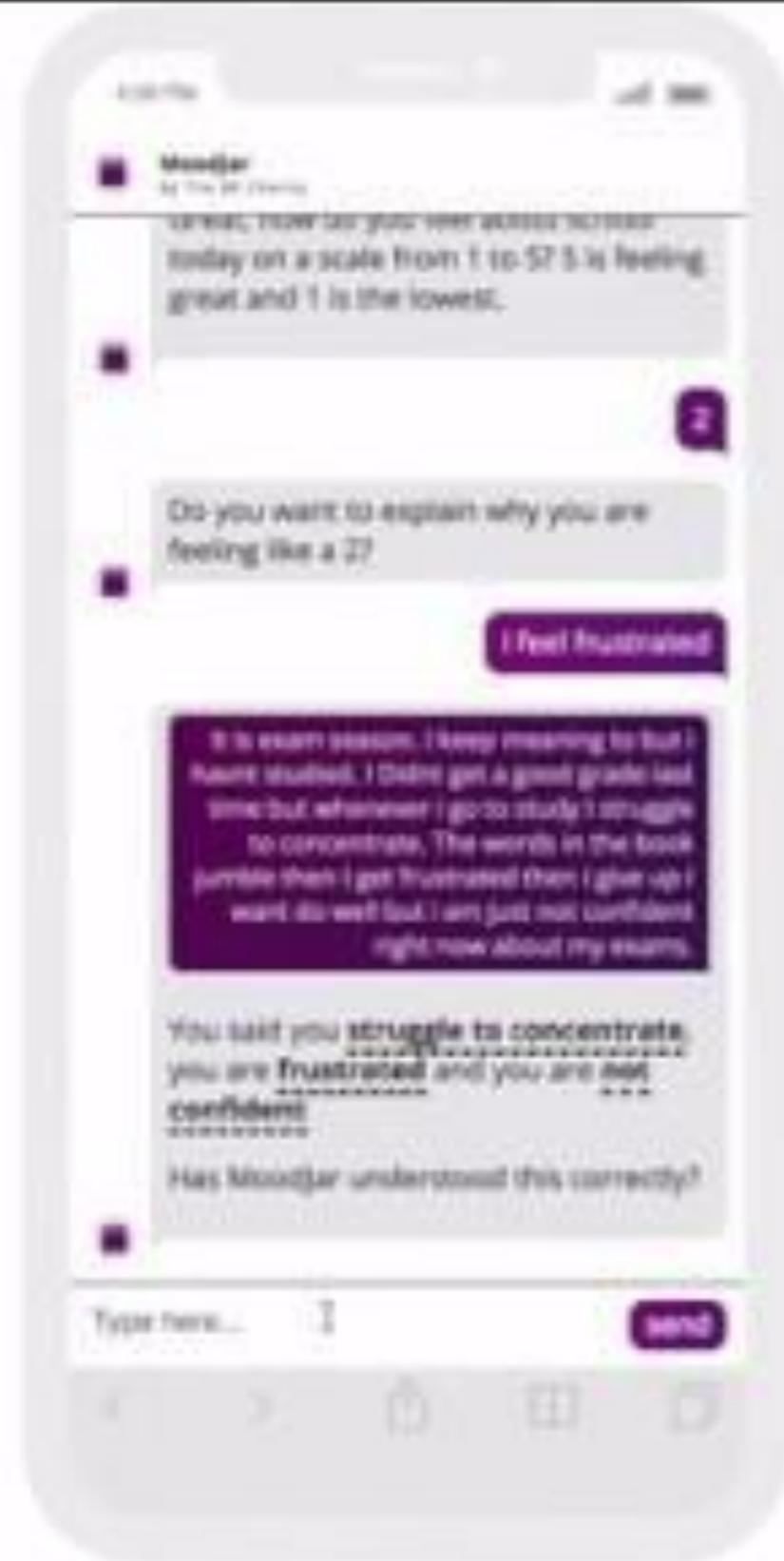
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FEMINIST
INTERNET

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how do you design an
explainable algorithm?

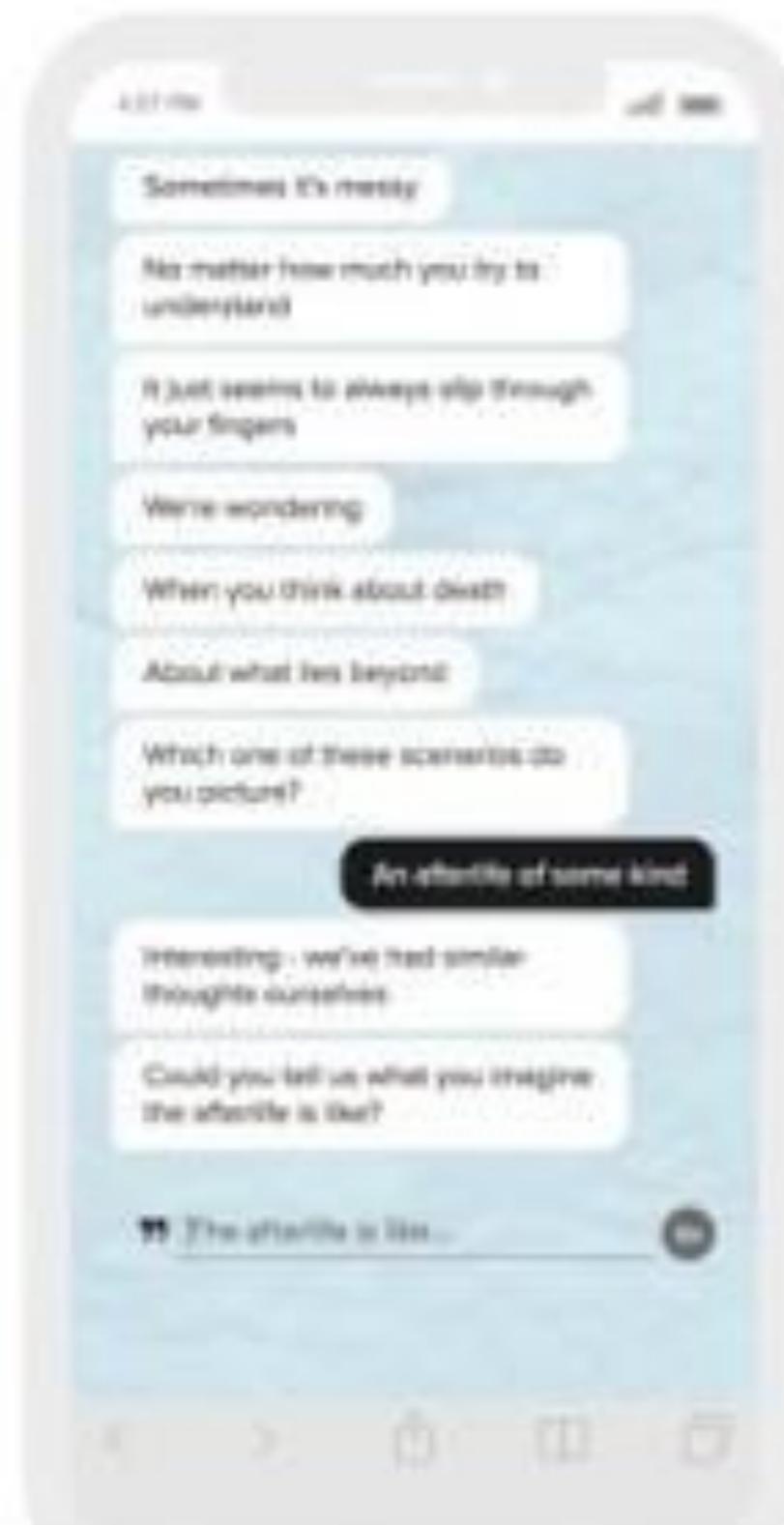
a bot that told you how it made an automated decision.



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what happens to your
digital self when you
die?

a bot that shares and
collect thoughts about
your death + digital rights



what is AI bias and
what can I do about
it?

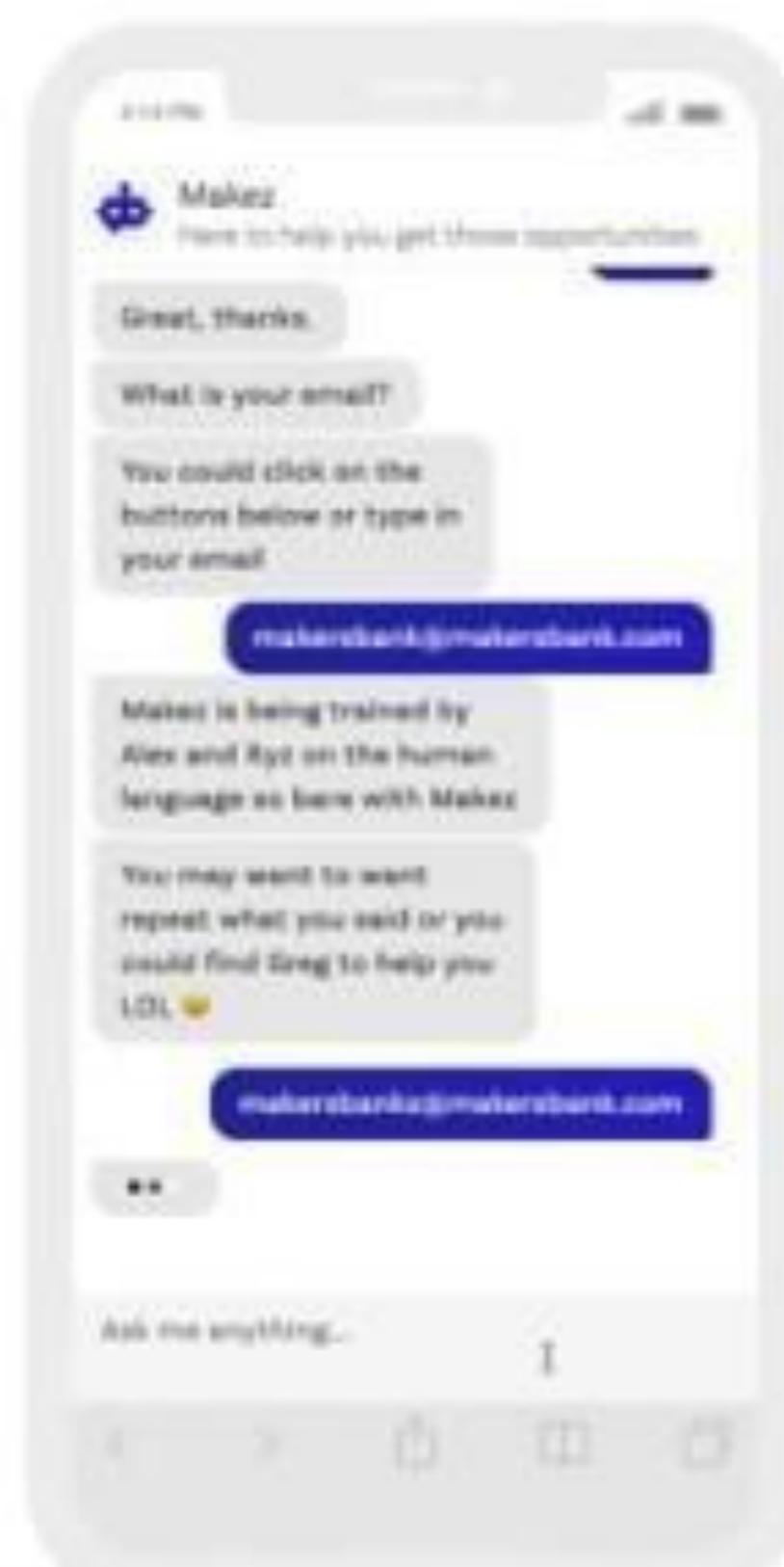
a bot that teaches you how bias in AI systems occur.



CVs, resumes and qualifications really cut off a lot of people of getting opportunity.

Can we guide someone to think about their strengths?

a bot that guides young people on finding jobs.



Cool bot stuff, Alex.
What you trying to do
next?

7) Randomness and Creativity

*Randomness is an artifact
in the R.C. process of*

A fairly attractive and yet clearly incomplete conjecture is that the difference between creative thinking and unimaginative competent thinking lies in the injection of a some randomness. The randomness must be guided by intuition to be efficient. In other words, the educated guess on the hunch include controlled randomness in otherwise orderly thinking.



Artists and Machine Intelligence

AMI is a program at Google that brings artists and engineers together to realize projects using Machine Intelligence. By supporting this emerging form of artistic collaboration we open our research to new ways of thinking about and working with intelligent systems. Follow along on our Medium blog.

Magenta

Magenta is a research project that explores the role of machine learning in the processes of creating art and music.

About Magenta

Magenta was started by some researchers and engineers from the Google Brain team, but many others have contributed significantly to the project. We develop new deep learning and reinforcement learning algorithms for generating songs, images, drawings, and other materials. But it's also an exploration in building smart tools and interfaces that allow artists and musicians to extend (not replace) their processes using these models. We use TensorFlow and release our models and tools in open source on our GitHub.

ml5.js

Friendly Machine Learning for the Web.

ml5.js aims to make machine learning approachable for a broad audience of artists, creative coders, and students. The library provides access to machine learning algorithms and models in the browser, building on top of [TensorFlow.js](#) with no other external dependencies.

The library is supported by code examples, tutorials, and sample datasets with an emphasis on ethical computing. Bias in data, stereotypical harms, and responsible crowdsourcing are part of the documentation around data collection and usage.

[Get Started](#)

TensorFlow.js

A JavaScript library for training and deploying ML models in the browser and on Node.js

Develop ML with JavaScript
Use flexible and intuitive APIs to build and train models from scratch using the low-level JavaScript linear algebra library or the high-level layers API

Run Existing Models
Use TensorFlow.js model converters to run pre-existing TensorFlow models right in the browser or under Node.js

Retrain Existing Models
Retrain pre-existing ML models using sensor data connected to the browser, or other client-side data.

Demos

[EMOJI SCAVENGER HUNT](#)

[TRY IT LIVE!](#) [GITHUB](#)

Can Computers Create Art?

Aaron Hertzmann
Adobe Research*

May 9, 2018

Abstract

This essay discusses whether computers, using Artificial Intelligence (AI), could create art. First, the history of analogies between art and AI is reviewed, and coding, photography and animation. In each case there were initial fears as to the potential of the technology, followed by a blossoming of new creative and professional opportunities for artists. The current hype and reality of Artificial Intelligence (AI) tools for art making is then discussed, together with predictions about how AI tools will be used. It is then speculated about whether it could ever happen that AI systems could be credited with authorship of artwork. It is theorized that art is something created by social agents, and so computers cannot be credited with authorship of art in our current understanding. A few ways that this could change are also hypothesized.

Chapter 15
Group Cognition and Collaborative AI

Check for updates

Janin Koch and Antti Oulasvirta

Abstract

Significant advances in artificial intelligence suggest that we will be using intelligent agents on a regular basis in the near future. This chapter discusses group cognition as a principle for designing collaborative AI. Group cognition is the ability to relate to other group members' decisions, abilities, and beliefs. It thereby allows participants to adapt their understanding and actions to reach common objectives. Hence, it underpins collaboration. We review two concepts in the context of group cognition that could inform the development of AI and automation in pursuit of natural collaboration with humans: conversational grounding and theory of mind. These concepts are somewhat different from those already discussed in AI research. We outline some new implications for collaborative AI, aimed at extending skills and solution spaces and at improving joint cognitive and creative capacity.

Artificial Intelligence for Augmented Creativity.

Runway is a toolkit that adds artificial intelligence capabilities to design and creative platforms.

Learn More [SIGN UP FOR BETA](#)

Heroic versus Collaborative AI for the Arts

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Abstract

This paper considers the kinds of AI systems we want involved in art and art practice. We explore this relationship from three perspectives: as artists interested in expanding and developing our own creative practice; as AI researchers interested in building AI systems that can support the development and deployment of art and art practice; and as audience members interested in experiencing art. We examine the nature of both art practice and experiencing art to ask how AI can contribute. To do so, we review the history of work in intelligent agents which broadly speaking sits in two main categories: (i) systems that can exhibit intelligent behaviour independently in one, and multi-agent systems (systems which interact with other systems in communities of agents) in the other. In this context we consider the nature of the relationship between AI and Art and introduce two opposing concepts: that of "Heroic AI", to describe the situation where the software takes on the role of the artist, and "Collaborative AI" where the system supports, challenges and provokes the creative activity of humans. We then set out what we believe are the main challenges for AI research in understanding its potential relationship to art and art practice.

In this paper we set out to explore the relationship between AI, Art and Art Practice that can help guide AI research in the future. We do this by considering this relationship from three different perspectives. We then go on to discuss the possibility of using AI systems to produce art. (ii) as AI researchers interested in developing novel AI systems that can be used in the practice of producing art, and (iii) as potential audiences for art produced (partly or fully) using AI.

This leads us to the following three broad sets of questions:

- As artists. What kinds of AI systems do artists and musicians want to use in their art making? How might artists want to interact with AI systems? How much autonomy would they like to have in their AI systems?

TRANSACTIONS ON COMPUTATIONAL INTELLIGENCE AND AI IN GAMES

The ANGELINA Videogame Design System, Part II

Michael Cook, Simon Colton, and Jeremy Gow
ccg.doc.gold.ac.uk

Abstract—Procedural content generation is generally viewed as a means to an end – a tool employed by designers to overcome technical problems or achieve a particular design goal. When we move from generating single parts of games to automating the entirety of their design, however, we find ourselves facing a far wider and more interesting set of problems than mere generation. When the designer of a game is a piece of software, we face questions about what it means to be a designer, about Computational Creativity, and about how to assess the growth of these automated game designers and the value of their output. Answering these questions can lead to new ideas in how to generate content procedurally, and produce systems that can further the cutting edge of game design.

This paper describes work done to take an automated game designer and advance it towards being a member of a creative community. We outline extensions made to the system to give it more autonomy and creative independence, in order to strengthen claims that the software is acting creatively. We describe and reflect upon the software's participation in the games community, including entering two game development contests, and show the opportunities and difficulties of such engagement. We consider methods for evaluating automated game designers as creative entities, and underline the need for automated game design to be a major frontier in future games research.

Index Terms—procedural content generation, automated game design, computational creativity

I. INTRODUCTION

[ml4a](#) [guides](#) [demos](#) [classes](#) [code](#) [slack](#) [twitter](#)

Machine Learning for Artists

ml4a is a collection of free educational resources devoted to machine learning for artists.

It contains an in-progress book which is being written by [@genekogan](#) and can be seen in draft form here. An estimated target release date is mid-2018. Two chapters are complete and others are in varying stages of progress or just stubs containing links.

The book is complemented by a set of 30+ instructional guides maintained by collaborators, along with interactive demos and figures, and video lectures.

[about](#) [contribute](#) [github.com/ml4a](#)

ual: creative computing institute

Human-Computer Collaboration

Lubart (2005)

“Man-Computer symbiosis” (Licklider, 1960)

d’Inverno & McCormack (2015)

Previous work in HCI has focused on exploiting human or machine capabilities

Frich et al (2018)

Halpern et al (2011)

Still & d'Inverno (2016)

collaboration and understanding between participants in a creative process helps to extend the available solution space and facilitates the emergence of creative solution and ideas, which is especially relevant for approaching new problem spaces.

Sawyer & DeZutter (2009)

Brief things to think about

creativity and its measurements?

Boden (2004)

Still & d'Inverno (2016)

Frich et al (2018)

Colton (2008)

Gero (2007)

Jordanous (2012)



Where does AI assist in the creative process?

Lubart, 2005; Gabriel et al, 2016; Lisi, 2015; Koch, 2017; Fiebrink et al; 2018; Bernado et al; 2017; Gillies et al; 2016; Yang et al; 2018, Yang et al; 2018b; Fiebrink et al; 2018

Tool – toy scale

- Is the system designed to solve a problem effectively or is it just for entertainment?

Remote control – autonomous scale

- Does the robot require remote control or is it capable of action without direct human influence?

Reactive – dialogue scale

- Does the robot rely on a fixed interaction pattern or is it able to have dialogue — exchange of information — with a human?

Anthropomorphism scale

- Does it have the shape or properties of a human?

INTERFACES

Bartneck & Okada (2001)



What the best interface for AI to express creativity?

- Be able to make context-dependent decisions.
- Change its creative strategies autonomously.
- Adapt to current user behaviour

The Vision for AI as a supportive creative collaborator

REFERENCES

- Koch, Janin, and Antti Oulasvirta (2018). "Group Cognition and Collaborative AI." In *Human and Machine Learning: Visible, Explainable, Trustworthy and Transparent*. Springer International Publishing.
- Colton, S., Lopez de Mantaras, R. & Stock, O. (2009). Computational creativity: Coming of age. *AI Magazine*, 30(3), pp. 11–14.
- Colton, S., & Wiggins, G. (2012). Computational Creativity: The final frontier? In: *Proceedings of the European Conference on Artificial Intelligence*.
- Gray, B. (1989). Collaborating: Finding common ground for multiparty problems
- McCarthy, J., Minsky, M., Rochester, N., Shannon, C.E., (1955) A Proposal for the Dartmouth Summer Research Project on Artificial Intelligence. Available at <http://raysolomonoff.com/dartmouth/boxa/dart564props.pdf> (Accessed: 16th November 2017)
- Sawyer, R. Keith, and Stacy DeZutter. (2009)."Distributed Creativity: How Collective Creations Emerge from Collaboration." *Psychology of Aesthetics, Creativity, and the Arts* 3 (2): 81
- Licklider, J. C. (1960) Man-computer symbiosis. *IRE Trans. Hum. Factors Electron.* 4–11.
- d'Inverno, M. & McCormack, J. (2015). Heroic versus collaborative AI for the arts. In Yang, Q., and Wooldridge, M., eds., *IJCAI International Joint Conference on Artificial Intelligence* , pp. 2438–2444. AAAI Press.
- Lubart, T. (2005). How can computers be partners in the creative process: Classification and commentary on the Special Issue. *International Journal of Human-Computer Studies*, 63(4-5), 365–369.
- Tensorflow (2018). Move Mirror: An AI Experiment with Pose Estimation in the Browser using TensorFlow.js . Available at: <https://medium.com/tensorflow/move-mirror-an-ai-experiment-with-pose-estimation-in-the-browser-using-tensorflow-js-2f7b769f9b23> [Accessed 20 Aug. 2018].
- Hertzmann, A. (2018). Can Computers Create Art? *Arts*, 7(2), pp. 18.
- Colton, S., Cook, M. & Gow, J. (2016). The ANGELINA videogame design system, parts I and II. *IEEE Trans. Comp. Intell. AI Games* .
- Kogan, G. (2018). Machine Learning for Artists . Available at: <https://ml4a.github.io/ml4a/> [Accessed 20 Aug. 2018].
- Valenzuela, C. & Germanidis, A. (2018). Runway: Artificial Intelligence for Augmented Creativity . Available at: <https://runwayml.com/about/> [Accessed 20 Aug. 2018].
- Boden, M. (2004). *The Creative Mind: Myths and Mechanisms* . London: Routledge.
- Bartneck, C., & Okada, M. (2001). Robotic User Interfaces. *Proceedings of the Human and Computer Conference (HC2001)*, Aizu pp. 130-140
- Colton, S. (2008). Creativity Versus the Perception of Creativity. In: *Computational Systems*. In *Proceedings of the AAAI Spring Symposium on Creative Systems* , pp. 14–20.
- Still, A. & d'Inverno, M. (2016). A history of creativity for future AI research. In Pachet, F.; Cardoso, A.; Corruble, V.; and Ghedini, F., eds., Title: *Proceedings of the 7th Computational Creativity Conference (ICCC 2016)* . Universite Pierre et Marie Curie.
- Jordanous, A. (2012). A Standardised Procedure for Evaluating Creative Systems: Computational Creativity Evaluation Based on What it is to be Creative. *Cognitive Computation*, 4(3), pp. 246–279.
- Michael Karlesky and Katherine Isbister. 2013. Designing for the Physical Margins of Digital Workspaces: Fidget Widgets in Support of Productivity and Creativity. In *Proceedings of the 8th International Conference on Tangible, Embedded and Embodied Interaction (TEI '14)*, 13–20.
- Frich, J., Mose Biskjaer, M., & Dalsgaard, P. (2018). Twenty Years of Creativity Research in Human-Computer Interaction. *Proceedings of the 2018 on Designing Interactive Systems Conference 2018 - DIS '18*.
- Gero, J.S. (2007). AI EDAM at 20: Artificial intelligence in designing. *AI EDAM* , 21(1), pp. 17–18.
- Megan K. Halpern, Jakob Tholander, Max Evjen, Stuart Davis, Andrew Ehrlich, Kyle Schustak, Eric P.S. Baumer, and Geri Gay. 2011. MoBoogie: Creative Expression Through Whole Body Musical Interaction. In *Proceedings of the SIGCHI Conference on Human Factors in Computing Systems (CHI '11)*, 557–560.

Thank you.

Questions?

Share these bots yo

f-xa.co

deathbotbeta.art

creativetechish.wtf

Break

Activity time

Rules

Our focus

our focus is to engage
in critical creative
practice with
technology.

our focus is for us to
be philosophers who
don't write down ideas,
but instead make
objects that embodied
them

our focus is to be
human.

our focus is on rapid
exploration and
experimentation.

our focus is to explore
futures and shape
alternative futures that
is different from the
norm.

there is no right or
wrong.

we are here to learn
and have fun.

collaboration is
everything. people are
cool.

challenge each other in
nice ways :)

Activity time

*'what does man-computer
symbiosis mean to you?'*

The work describes something of Lickliders' vision for a complementary (symbiotic) relationship between humans and computers at a potential time of the future.

Licklider envisioned a future time when machine cognition (cerebration) would surpass and become independent of human direction, as a basic stage of development within human evolution.

Read

Reflect

Ideate

Present

REFERENCES

Koch, Janin, and Antti Oulasvirta (2018). "Group Cognition and Collaborative AI." In Human and Machine Learning: Visible, Explainable, Trustworthy and Transparent. Springer International Publishing.

Sawyer, R. Keith, and Stacy DeZutter. (2009). "Distributed Creativity: How Collective Creations Emerge from Collaboration." *Psychology of Aesthetics, Creativity, and the Arts* 3 (2): 81

Lubart, T. (2005). How can computers be partners in the creative process: Classification and commentary on the Special Issue. *International Journal of Human-Computer Studies*, 63(4-5), 365–369.

Licklider, J. C. (1960) Man-computer symbiosis. *IRE Trans. Hum. Factors Electron.* 4–11.

Hertzmann, A. (2018). Can Computers Create Art? *Arts*, 7(2), pp. 18.

Colton, S., Cook, M. & Gow, J. (2016). The ANGELINA videogame design system, parts I and II. *IEEE Trans. Comp. Intell. AI Games* .

Valenzuela, C. & Germanidis, A. (2018). Runway: Artificial Intelligence for Augmented Creativity . Available at: <https://runwayml.com/about/> [Accessed 20 Aug. 2018].

Boden, M. (2004). The Creative Mind: Myths and Mechanisms . London: Routledge.

Kogan, G. (2018). Machine Learning for Artists . Available at: <https://ml4a.github.io/ml4a/> [Accessed 20 Aug. 2018].

Gabriel, A., Monticolo, D., Camargo, M. & Bourgault, M. (2016) Creativity support systems: A systematic mapping study. *Thinking Skills and Creativity* .

Colton, S. (2008). Creativity Versus the Perception of Creativity. In: Computational Systems. In *Proceedings of the AAAI Spring Symposium on Creative Systems* , pp. 14–20.

Still, A. & d'Inverno, M. (2016). A history of creativity for future AI research. In Pachet, F.; Cardoso, A.; Corruble, V.; and Ghedini, F., eds., Title: *Proceedings of the 7th Computational Creativity Conference (ICCC 2016)* . Universite Pierre et Marie Curie.

Colton, S., Lopez de Mantaras, R. & Stock, O. (2009). Computational creativity: Coming of age. *AI Magazine*, 30(3), pp. 11–14.

Colton, S., & Wiggins, G. (2012). Computational Creativity: The final frontier? In: *Proceedings of the European Conference on Artificial Intelligence* .

Gray, B. (1989). Collaborating: Finding common ground for multiparty problems

d'Inverno, M. & McCormack, J. (2015). Heroic versus collaborative AI for the arts. In Yang, Q., and Wooldridge, M., eds., *IJCAI International Joint Conference on Artificial Intelligence* , pp. 2438–2444. AAAI Press.

Lisi, F.A. (2015). Will AI ever Support Design Thinking? *AI&DE@AI*IA* .

Bernardo, F., Zbyszynski, M., Fiebrink, R. & Grierson, M. (2016). Interactive Machine Learning for End-User Innovation. In: *Designing the User Experience of Machine Learning Systems* . Palo Alto, California, United States.

Gillies, M., Lee, B., D'Alessandro, N., Tilmanne, J., Kulesza, T., Caramiaux, B., Fiebrink, R., Tanaka, A., Garcia, J., Bevilacqua, F., Heloir, A., Nunnari, A., Mackay, W. & Saleema Amershi. (2016). Human-Centred Machine Learning. In: *Proceedings of the 2016 CHI Conference Extended Abstracts on Human Factors in Computing Systems - CHI EA '16*, pp. 3558–3565. ACM Press, New York, New York, USA,

REFERENCES

Tensorflow (2018). Move Mirror: An AI Experiment with Pose Estimation in the Browser using TensorFlow.js . Available at: <https://medium.com/tensorflow/move-mirror-an-ai-experiment-with-pose-estimation-in-the-browser-using-tensorflow-js-2f7b769f9b23> [Accessed 20 Aug. 2018].

McCarthy, J., Minsky, M., Rochester, N., Shannon, C.E., (1955) A Proposal for the Dartmouth Summer Research Project on Artificial Intelligence. Available at <http://raysolomonoff.com/dartmouth/boxa/dart564props.pdf> (Accessed: 16th November 2017)

Agüera y Arcas, B. (2017) Art in the Age of Machine Intelligence. *Arts*, 6(4), pp. 18.

Fiebrink, R., Bernardo, F. & Grierson, M. (2018). User-Centred Design Actions for Lightweight Evaluation of an Interactive Machine Learning Toolkit. *Journal Of Science And Technology Of The Arts* , 10(2), pp. 25-38.

Bartneck, C., & Okada, M. (2001). Robotic User Interfaces. Proceedings of the Human and Computer Conference (HC2001), Aizu pp. 130-140

Michael Karlesky and Katherine Isbister. 2013. Designing for the Physical Margins of Digital Workspaces: Fidget Widgets in Support of Productivity and Creativity. In Proceedings of the 8th International Conference on Tangible, Embedded and Embodied Interaction (TEI '14), 13–20.

Frich, J., Mose Biskjaer, M., & Dalsgaard, P. (2018). Twenty Years of Creativity Research in Human-Computer Interaction. Proceedings of the 2018 on Designing Interactive Systems Conference 2018 - DIS '18.

Megan K. Halpern, Jakob Tholander, Max Evjen, Stuart Davis, Andrew Ehrlich, Kyle Schustak, Eric P.S. Baumer, and Geri Gay. 2011. MoBoogie: Creative Expression Through Whole Body Musical Interaction. In Proceedings of the SIGCHI Conference on Human Factors in Computing Systems (CHI '11), 557–560.

Yang, Q., Scuito, A., Zimmerman, J., Forlizzi, J. & Steinfeld, A. (2018). Investigating How Experienced UX Designers Effectively Work with Machine Learning. In: Proceedings of the 2018 Designing Interactive Systems Conference (DIS '18) , pp. 585-596. ACM, New York, NY, USA.

Yang, Q., Banovic, N. & Zimmerman, J. (2018). Mapping Machine Learning Advances from HCI Research to Reveal Starting Places for Design Innovation. In: Proceedings of the 2018 CHI Conference on Human Factors in Computing Systems (CHI '18) . ACM, New York, NY, USA.

Koch, J. (2017). Design implications for Designing with a Collaborative AI. In: The AAAI 2017 Spring Symposium on Designing the User Experience of Machine Learning Systems Technical Report SS-17-04.

Gillies, M., Lee, B., D'Alessandro, N., Tilmanne, J., Kulesza, T., Caramiaux, B., Fiebrink, R., Tanaka, A., Garcia, J., Bevilacqua, F., Heloir, A., Nunnari, A., Mackay, W. & Saleema Amershi. (2016). Human-Centred Machine Learning. In: Proceedings of the 2016 CHI Conference Extended Abstracts on Human Factors in Computing Systems - CHI EA '16, pp . 3558–3565. ACM Press, New York, New York, USA,

Gero, J.S. (2007). AI EDAM at 20: Artificial intelligence in designing. *AI EDAM* , 21(1), pp. 17–18.