

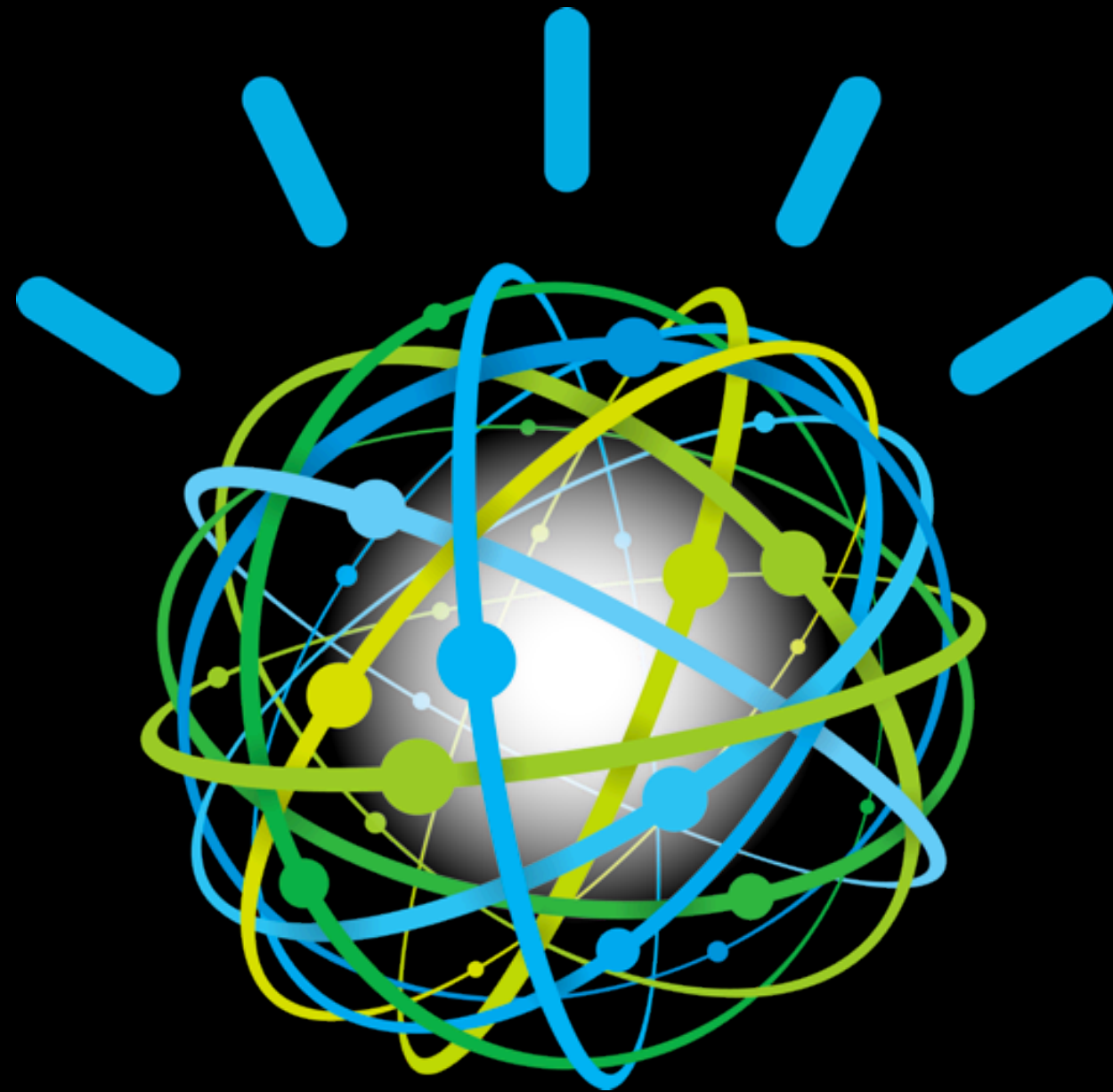
Leveraging Data, Machine Learning and AI in your Product and Business

Rich Edwards

 [@RichEdwards](https://twitter.com/RichEdwards)



Why Care/Listen to Me



MINDSPAN
SYSTEMS

Data is Where it's At

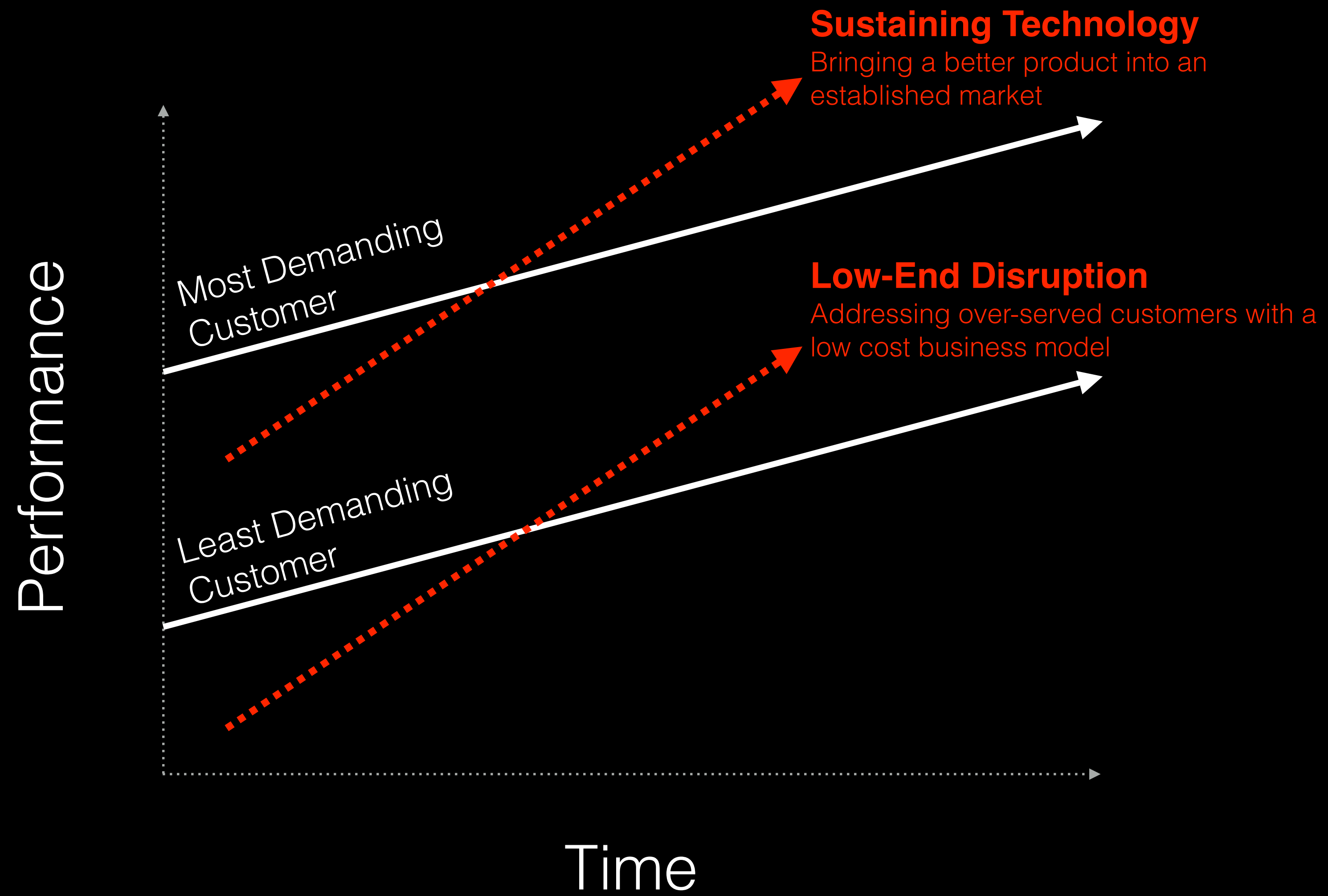
- Data is valuable, if managed properly the source of competitive advantage
- Barriers to entry ML/AI technology/methods are lower than ever
- Leverage your data (structured and unstructured) with ML/AI to give your self an unfair advantage (operational efficiency, speed to market, customer experience, RTM)
- “There’s Always Money in the Banana Stand”



Framing Problems for Machine Intelligence

- Disruption vs. Innovative Improvements
- The limits of capabilities and “hammeritis”





Types of ML Capabilities

- Classification/Anomaly Detection
- Regression (forecasting)
- Clustering (what are the patterns in this data)
- Recommendation (given a history of events, what might be next)
- Generation (particularly NLP)



Google Cloud Platform



IBM Watson

MACHINE INTELLIGENCE 3.0

ENTERPRISE INTELLIGENCE



ENTERPRISE FUNCTIONS



AUTONOMOUS SYSTEMS



INDUSTRIES



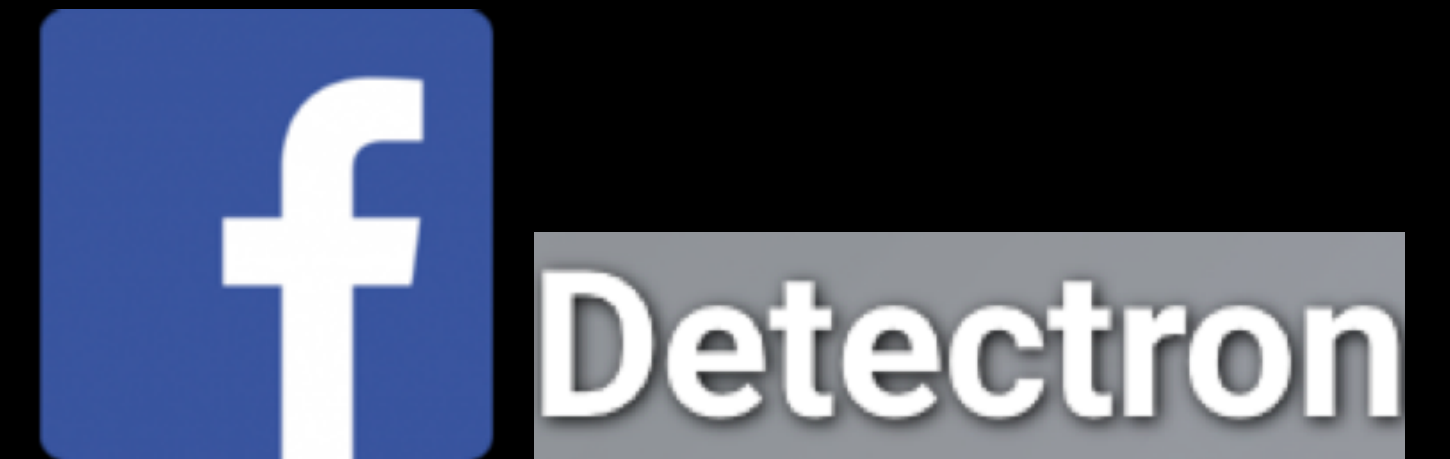
INDUSTRIES CONT'D



TECHNOLOGY STACK



Open Source



Coke

T
KER SPACES

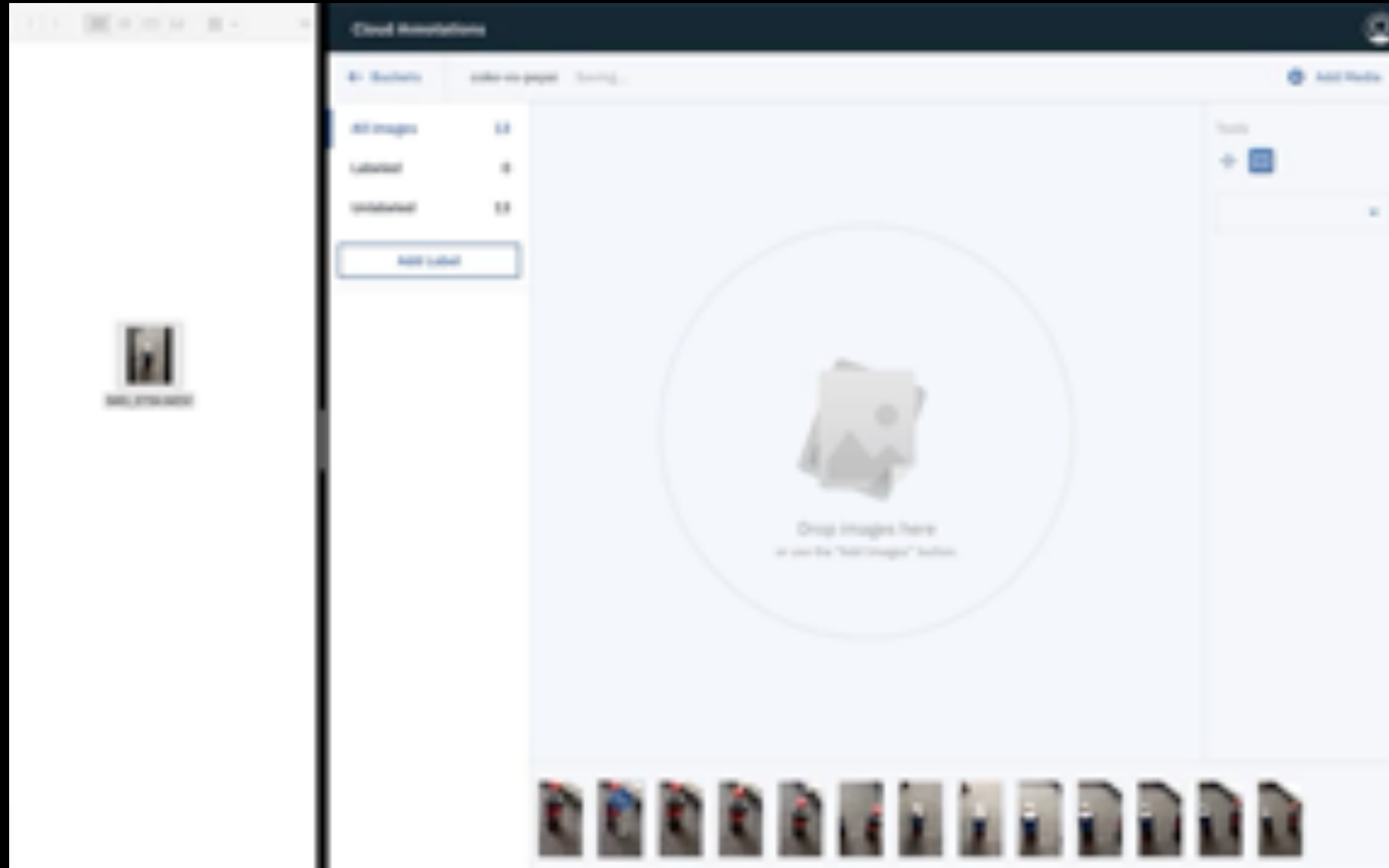
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Analytics in Watson

Promote
#IBM Drop
- Color Challenge
- use Telemundo
- 1500 drop

IBM Max - 2/2/2015 - Dec 2015 - Nov 2015

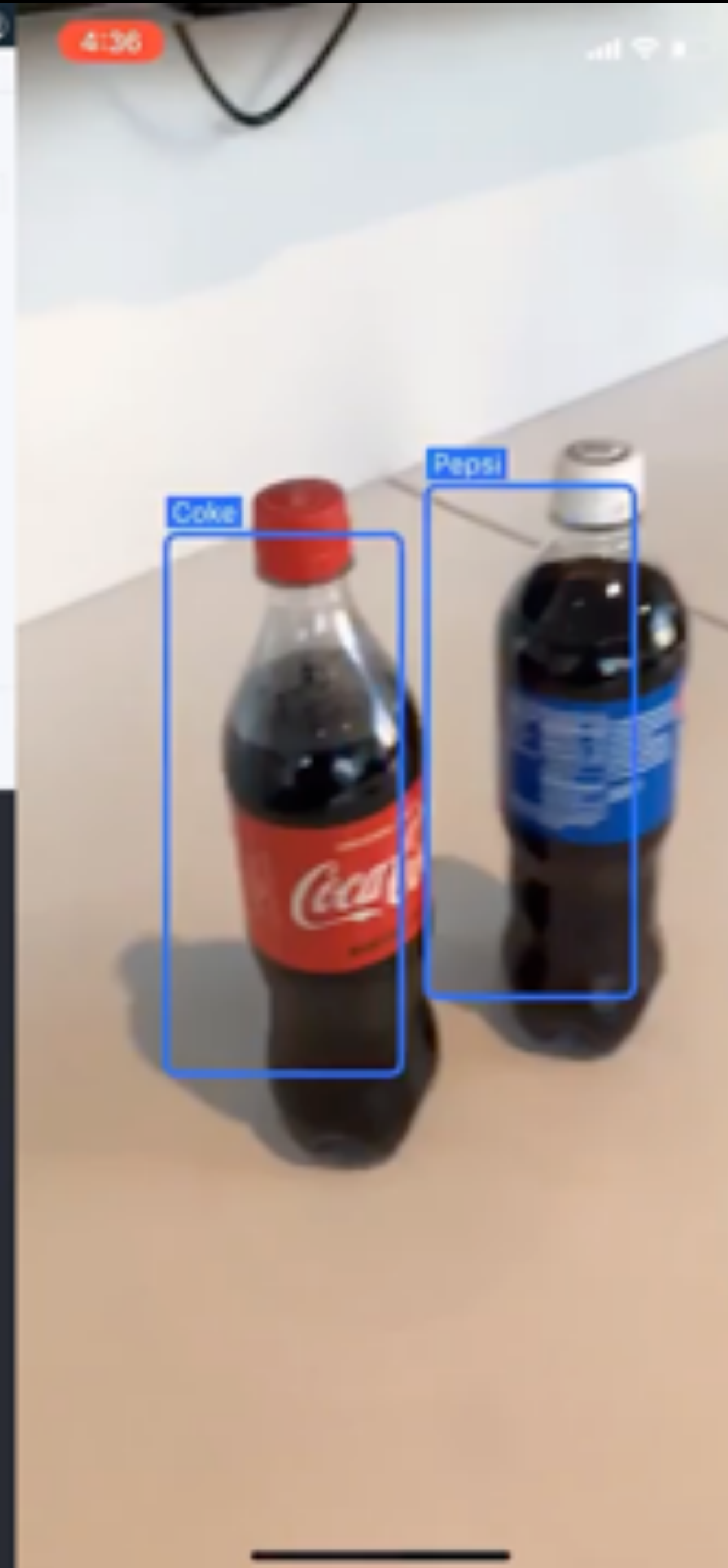
11



```
λ npx cloud-annotations train
npx: installed 121 in 2.693s
(using settings from config.json)
success Training run submitted.

Model ID:
model-CH6x451

Would you like to monitor progress? (yes)
Preparing to train (this may take a while)...
```



Promote
#IBM Drone Drop
- Code Challenge
- use TeLo Drones
- 1500 drones/5 weeks
1/12 - 1/16

in Watson Studio

head =

queue

head tail

class



B₀₀

L_{2L}



✓ PLACE SWITCH

✓ TEST SWITCH MOTION



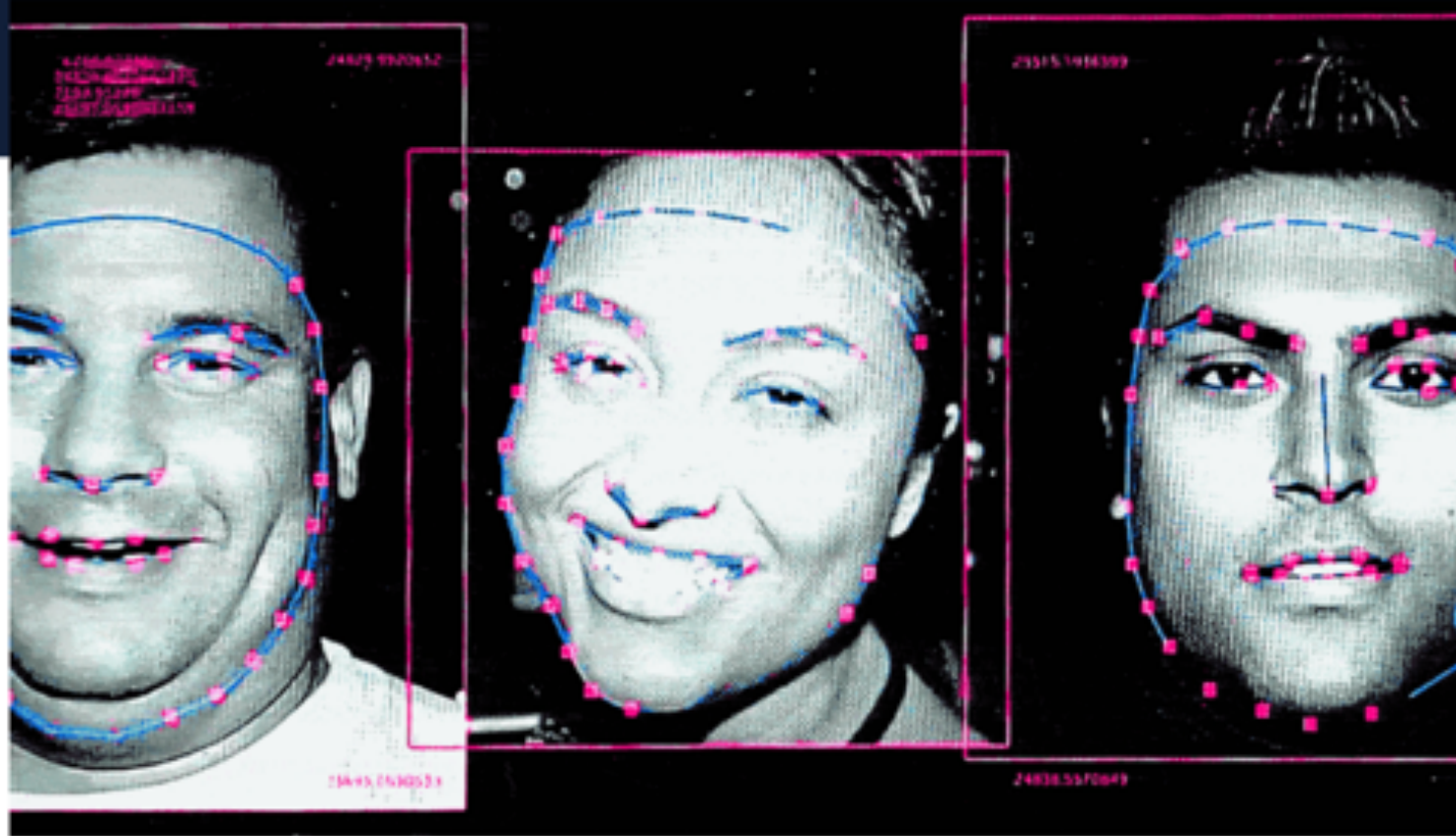


HERE
BE
DRAGONS



Facial recognition's 'dirty little secret': Millions of online photos scraped without consent

People's faces are being used without their permission, in order to power technology that could eventually be used to surveil them, legal experts say.



Erik Carter for NBC News / Source photography courtesy Greg Peverelli-Conti

Google Street View image of a house predicts car accident risk of its resident

Kinga Kita, Łukasz Kidziński

(Submitted on 10 Apr 2019)

Road traffic injuries are a leading cause of death worldwide. Proper estimation of car accident risk is critical for appropriate allocation of resources in healthcare, insurance, civil engineering, and other industries. We show how images of houses are predictive of car accidents. We analyze 20,000 addresses of insurance company clients, collect a corresponding house image using Google Street View, and annotate house features such as age, type, and condition. We find that this information substantially improves car accident risk prediction compared to the state-of-the-art risk model of the insurance company and could be used for price discrimination. From this perspective, public availability of house images raises legal and social concerns, as they can be a proxy of ethnicity, religion and other sensitive data.

Subjects: [Applications \(stat.AP\)](#)

Cite as: [arXiv:1904.05270 \[stat.AP\]](#)

(or [arXiv:1904.05270v1 \[stat.AP\]](#) for this version)

Submission history

From: Łukasz Kidziński [[view email](#)]

[v1] Wed, 10 Apr 2019 16:16:31 UTC (689 KB)

[Which authors of this paper are endorsers?](#) | [Disable MathJax \(What is MathJax?\)](#)

One Month, 500,000 Face Scans: How China Is Using A.I. to Profile a Minority

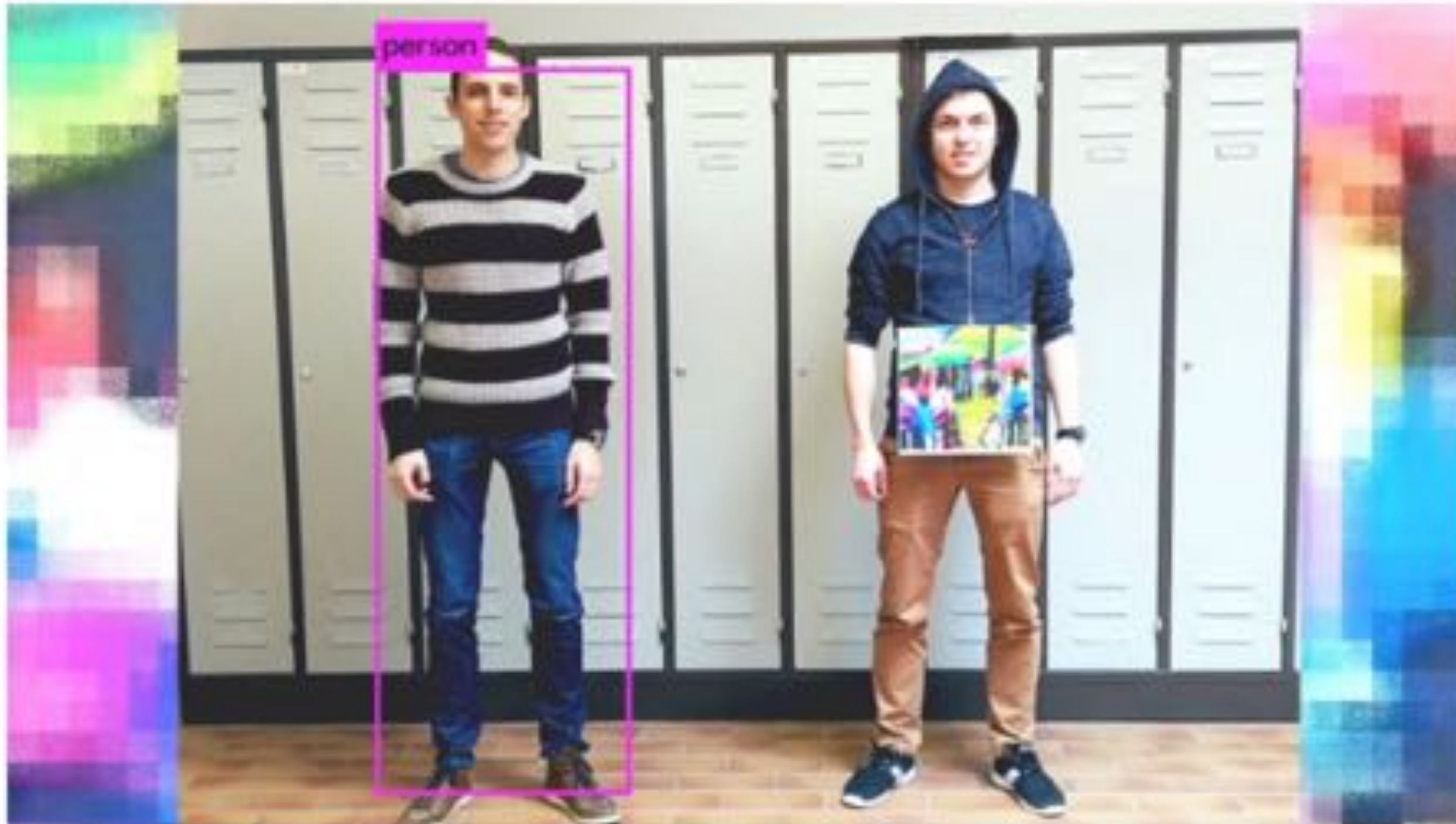
In a major ethical leap for the tech world, Chinese start-ups have built algorithms that the government uses to track members of a largely Muslim minority group.



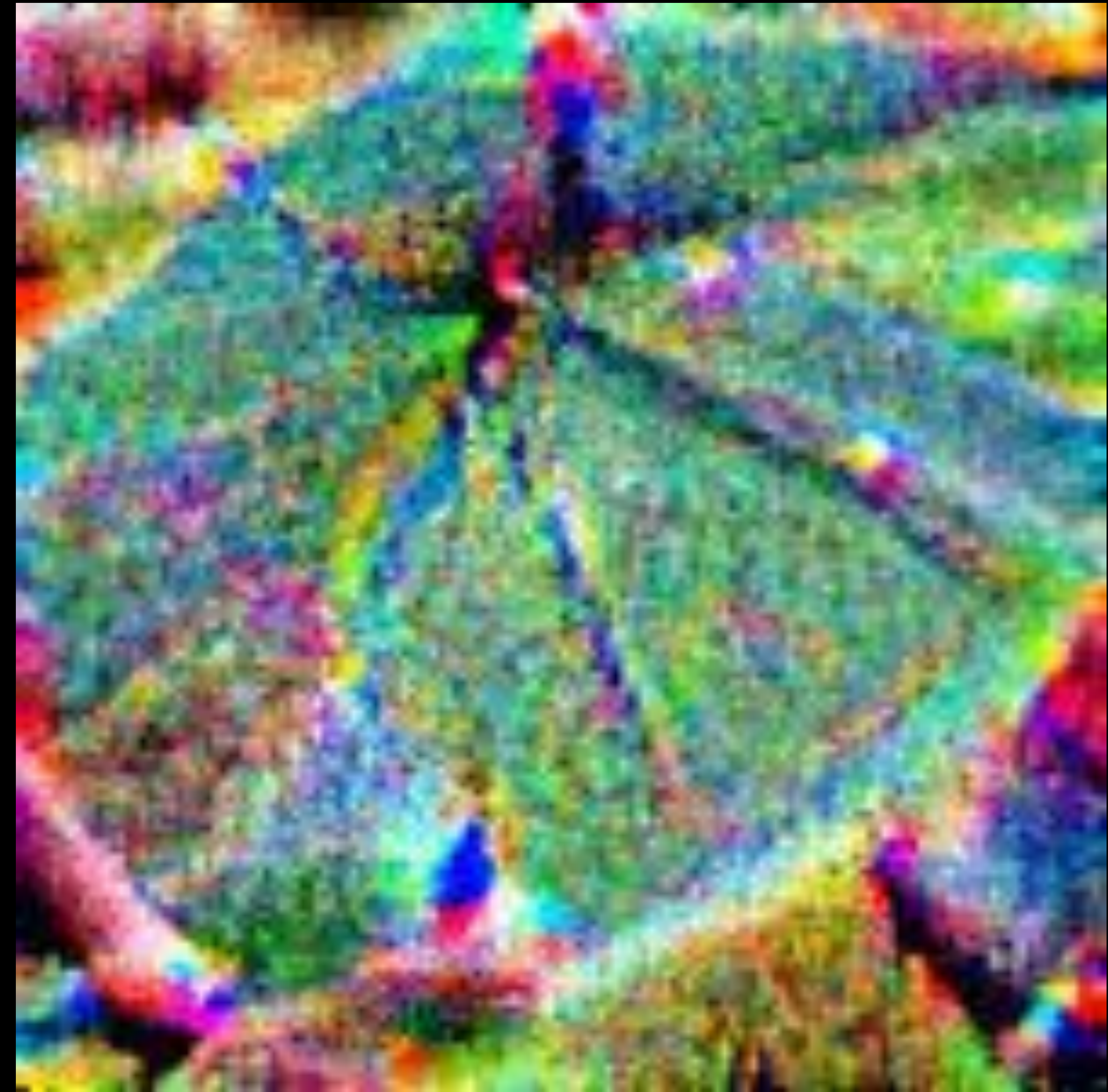
SenseTime is among the Chinese artificial intelligence companies developing facial recognition technology. Gilles Sabrié for The New York Times

Artificial Intelligence Apr 23

How to hide from the AI surveillance state with a color printout



AI-powered video technology is becoming ubiquitous, tracking our faces and bodies through stores, offices, and public spaces. In some countries the technology constitutes a powerful new layer of policing and government surveillance.



Small stickers on the ground trick Tesla autopilot into steering into opposing traffic lane



Fig 35. In-car perspective when testing, the red circle marks, the interference markings are marked with red circles

Audio Adversarial Examples: Targeted Attacks on Speech-to-Text

Nicholas Carlini David Wagner
University of California, Berkeley

Abstract—We construct targeted audio adversarial examples on automatic speech recognition. Given any audio waveform, we can produce another that is over 99.9% similar, but transcribes as any phrase we choose (recognizing up to 50 characters per second of audio). We apply our white-box iterative optimization-based attack to Mozilla's implementation DeepSpeech end-to-end, and show it has a 100% success rate. The feasibility of this attack introduce a new domain to study adversarial examples.

I. INTRODUCTION

As the use of neural networks continues to grow, it is critical to examine their behavior in adversarial settings. Prior work [8] has shown that neural networks are vulnerable to *adversarial examples* [40], instances x' similar to a natural instance x , but classified by a neural network as any (incorrect) target t chosen by the adversary.

Existing work on adversarial examples has focused largely

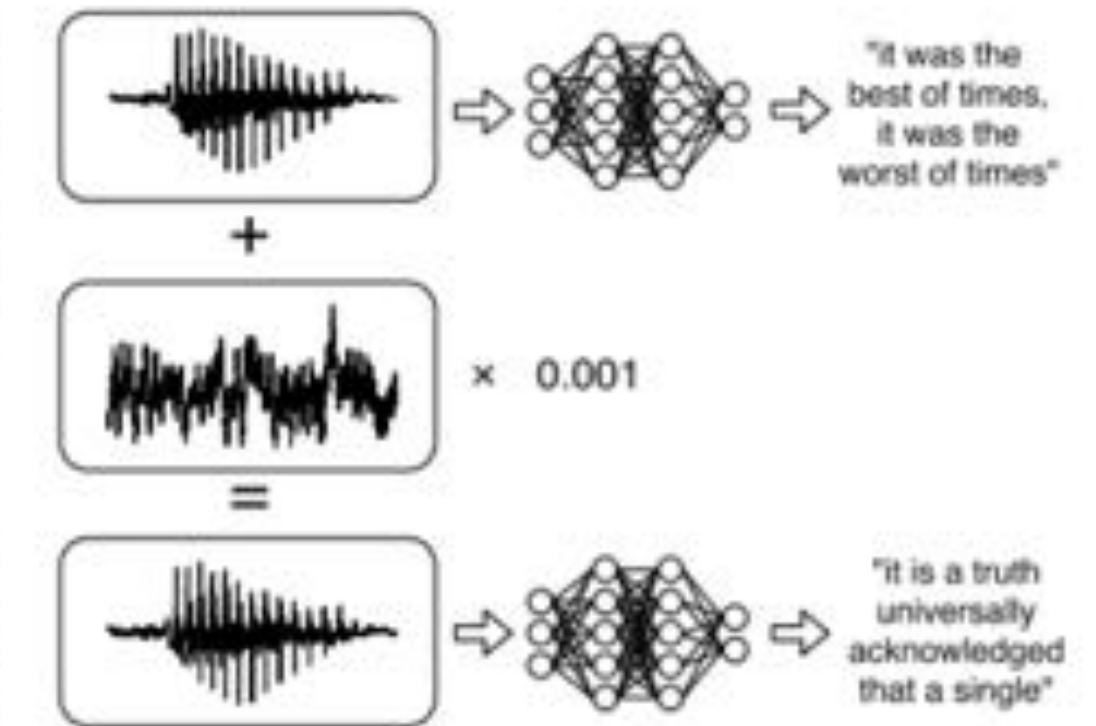


Figure 1. Illustration of our attack: given any waveform, adding a small perturbation makes the result transcribe as any desired target phrase.

That is, just as for cars, or aircraft, or databases, these systems can be both extremely powerful and extremely limited, and depend entirely on how they're used by people, and on how well or badly intentioned and how educated or ignorant people are of how these systems work.

Hence, it is completely false to say that 'AI is maths, so it cannot be biased'. But it is equally false to say that ML is 'inherently biased'. ML finds patterns in data - what patterns depends on the data, and the data is up to us, and what we do with it is up to us. **Machine learning is much better at doing certain things than people, just as a dog is much better at finding drugs than people, but you wouldn't convict someone on a dog's evidence. And dogs are much more intelligent than any machine learning.**

Benedict Evans

<https://www.ben-evans.com/benedictevans/2019/4/15/notes-on-ai-bias>

A starter reading list

- Fortune's Eye on AI Newsletter (<http://fortune.com/newsletter/eye-on-ai/>)
- Google's The Lever (<https://medium.com/thelaunchpad>) and AI blog (<https://ai.googleblog.com/>)
- Facebook Research (<https://research.fb.com>)
- Nick Bourdakos, IBM Developer Advocate (<https://hackernoon.com/@bourdakos1>)
- Andrew Ng's Coursera ML Course (<https://www.coursera.org/learn/machine-learning>)
- Superintelligence by Nick Bostrom (<https://www.amazon.com/Superintelligence-Dangers-Strategies-Nick-Bostrom/dp/1501227742>)
- <https://github.com/collections/machine-learning>
- <https://github.com/josephmisiti/awesome-machine-learning>
- <https://github.com/onmyway133/fantastic-machine-learning>
- Frank Chen's essay Humanity + AI: Better Together: <https://a16z.com/2019/02/22/humanity-ai-better-together/>
- Benedict Evans weekly newsletter: <https://www.ben-evans.com/newsletter>
- Clayton Christensen's Theory of Disruptive Innovation: https://en.wikipedia.org/wiki/Disruptive_innovation

Questions

Link to reference list and demos:

bit.ly/pcrtp19