

### AI: The New Electricity

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### AI: the New Electricity

"Al is the new electricity.

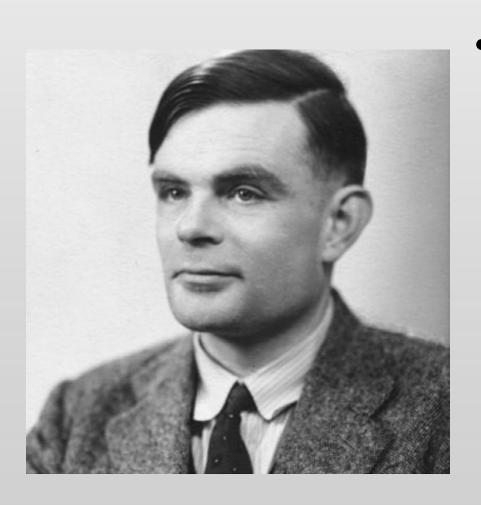
Just as electricity transformed industry after industry 100 years ago,

I think Al will do the same."



Andrew Ng, Stanford, Baidu, Coursera

Al



- "I believe that at the end of the century the use of words and general educated opinion will have altered so much that one will be able to speak of machines thinking without expecting to be contradicted."
  - Alan Turing,
     Computing Machinery
     and Intelligence (1950)

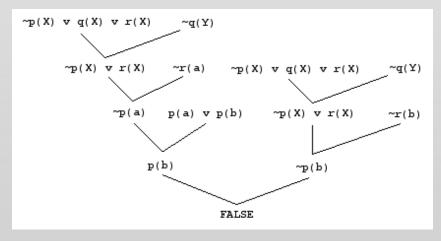


Every aspect of learning or any other feature of intelligence can in principle be so precisely described that a machine can be made to simulate it. An attempt will be made to find how to make machines use language, form abstractions and concepts, solve kinds of problems now reserved for humans, and improve themselves. We think that a significant advance can be made in one or more of these problems if a carefully selected group of scientists work on it together for a summer.

John McCarthy, Dartmouth Workshop 1956

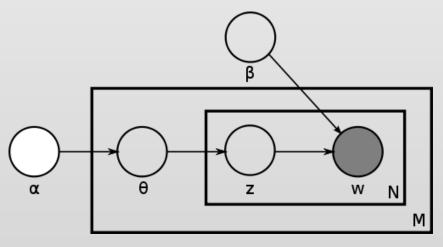
### GOFAI (1960-90)

- Knowledge Representation
- First Order Logic and Theorem Proving



- at(restaurant,Alice)
- at(restaurant,Bob)
- at(restaurant,Carol)
- works\_at(restaurant,Carol)
- has\_job(restaurant,waitress,Carol)
- orders(Bob,pizza)
- orders(Alice,sushi)
- forall X,Y,Z. orders(X,Y) and has\_job(restaurant,waitress,Z) -> serves(Z,X,Y)
- serves(Carol, Bob, pizza) serves(Carol, Alice, sushi)

### Statistical Machine Learning 1990-



Density

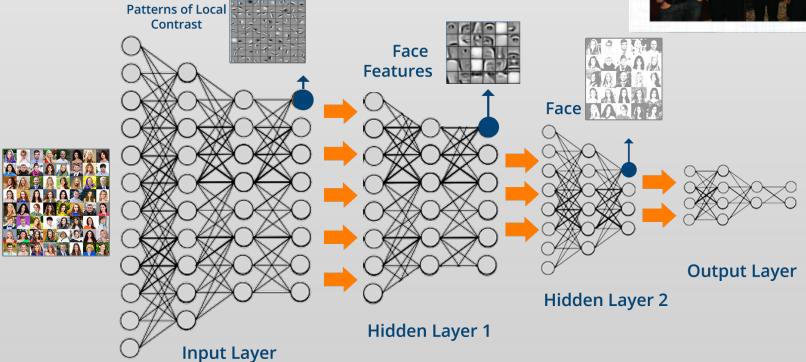
- Mixture components (weighted)

- Probabilistic models
- Statistical learning and inference
- Data driven (no hardcoded rules)

### Deep Learning 2005-

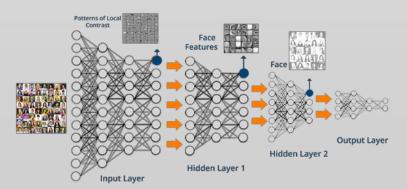






# Why Now? Convergence of Technologies

- Data sensing, acquisition revolution
- Rapid increase in computing power
- Novel algorithms
- Software frameworks



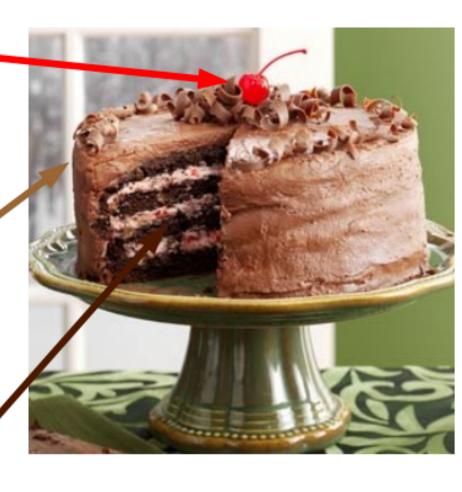




- "Pure" Reinforcement Learning (cherry)
  - The machine predicts a scalar reward given once in a while.
  - A few bits for some samples
- Supervised Learning (icing)
  - The machine predicts a category or a few numbers for each input
  - Predicting human-supplied data
  - **▶** 10→10,000 bits per sample
- Unsupervised/Predictive Learning (cake)
  - The machine predicts any part of its input for any observed part.
  - Predicts future frames in videos
  - Millions of bits per sample

(Yes, I know, this picture is slightly offensive to RL folks. But I'll make it up)

Yann LeCun, NIPS 2016





### Data-driven Secure Business Intelligence





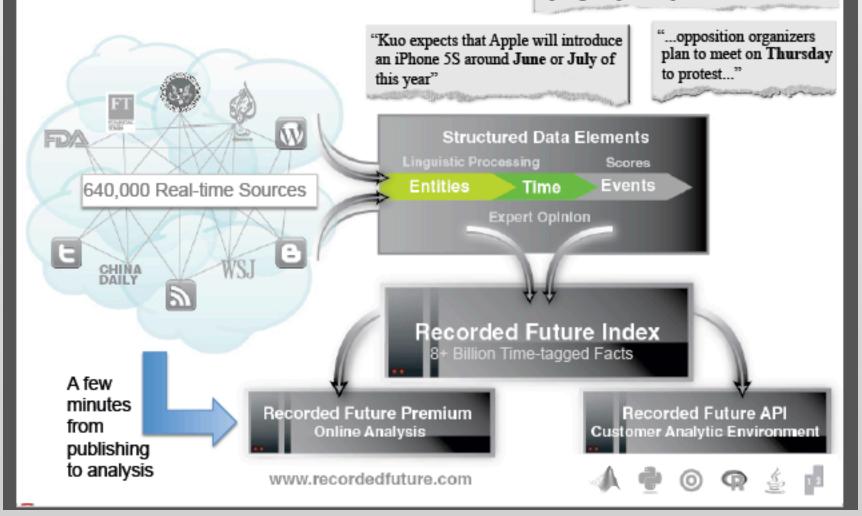
David Sands, Peter Damaschke, Devdatt Dubhashi, Gerardo Schneider, Andrei Sabelfeld



# Inside the Web Intelligence Machine



Drought and malnutrition hinder next spring's expansion plans in Kabul...

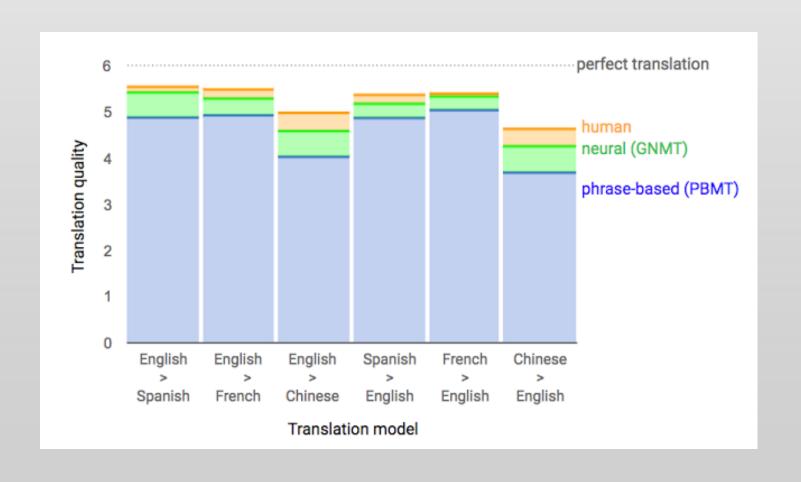


### **Google Translate**

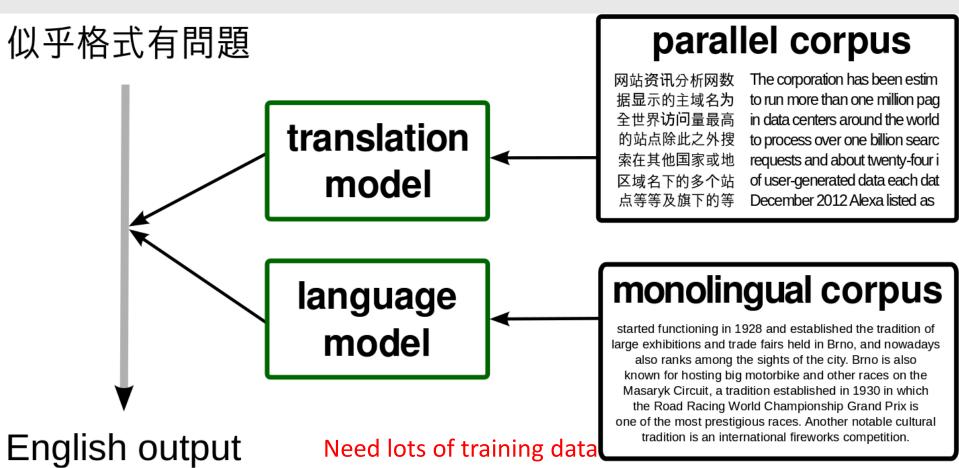
| Input sentence:                                           | Translation (PBMT):                                                                                                                                                                        | Translation (GNMT):                                                                                                                                                      | Translation (human):                                                                                                                                                              |
|-----------------------------------------------------------|--------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|--------------------------------------------------------------------------------------------------------------------------------------------------------------------------|-----------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| 李克強此行將啟動中加<br>總理年度對話機制,與<br>加拿大總理杜魯多舉行<br>兩國總理首次年度對<br>話。 | Li Keqiang premier<br>added this line to start<br>the annual dialogue<br>mechanism with the<br>Canadian Prime Minister<br>Trudeau two prime<br>ministers held its first<br>annual session. | Li Keqiang will start the<br>annual dialogue<br>mechanism with Prime<br>Minister Trudeau of<br>Canada and hold the first<br>annual dialogue between<br>the two premiers. | Li Keqiang will initiate the annual dialogue mechanism between premiers of China and Canada during this visit, and hold the first annual dialogue with Premier Trudeau of Canada. |

reduce translation errors across its Google Translate service by between 55 percent and 85 percent

### Al Revolution in NLP

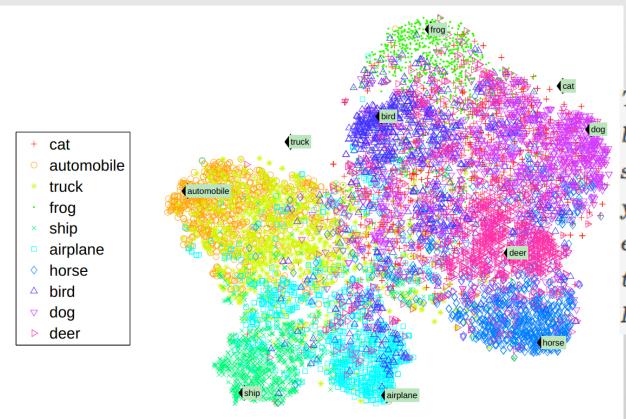


## Supervised Learning



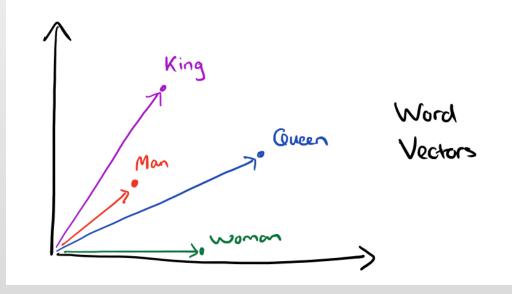
- EU Parliament documents in multiple languages
- Bibles in multiple languages

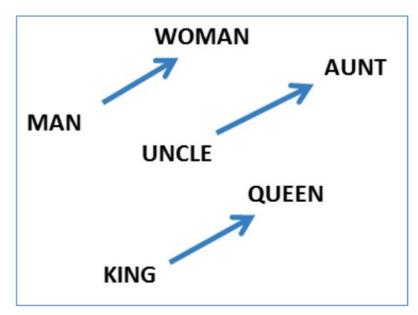
### Word Embeddings

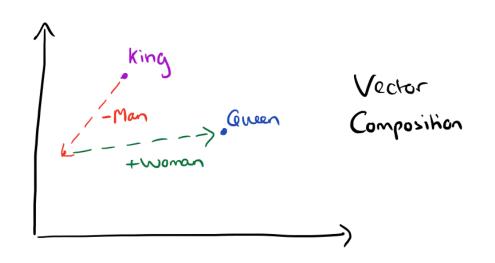


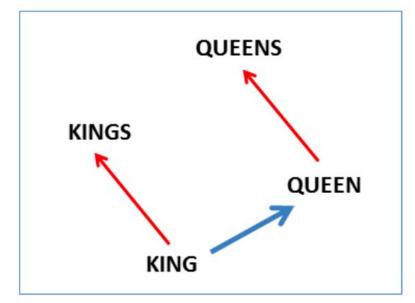
 $W: \mathrm{words} o \mathbb{R}^n$ 

The use of word representation become a key "secret sauce" success of many NLP systems is years, across tasks including entity recognition, part-otagging, parsing, and semant labeling. (Luong et al. (2013))

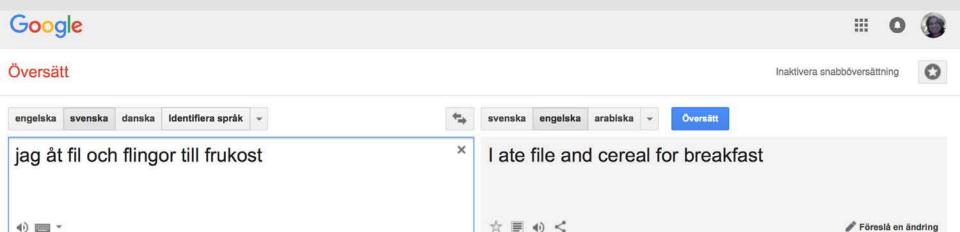


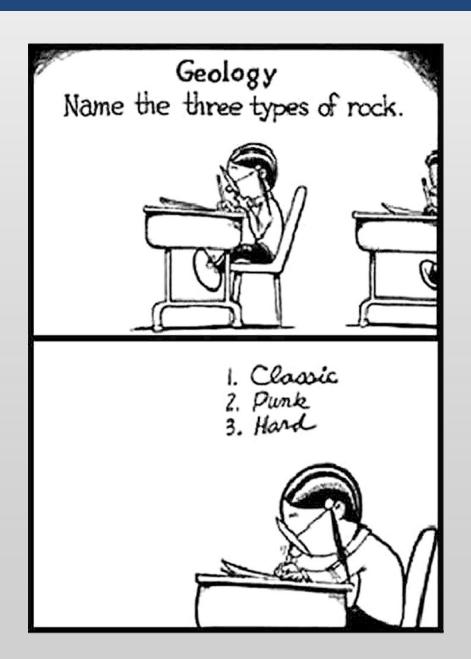






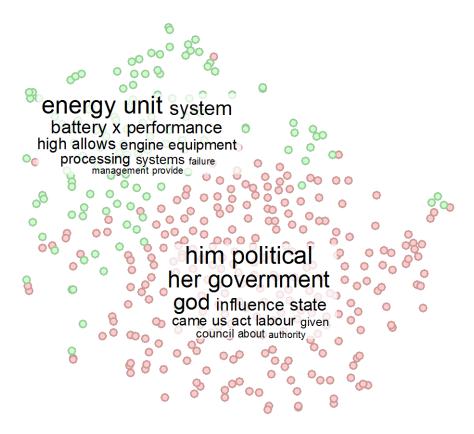
### Word senses and Machine Translation





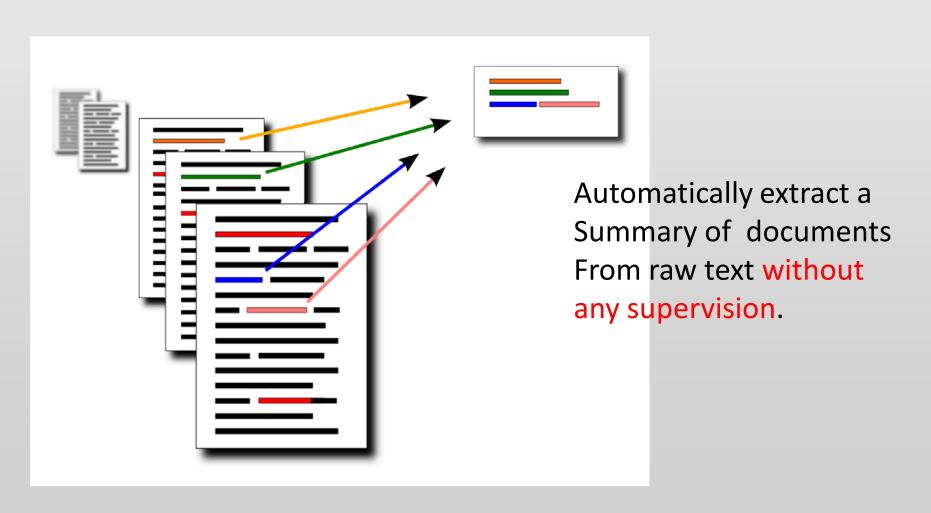
### Unsupervised Learning of Word senses

Instance cloud for: 'power'



Learn the different senses of a word from raw text without any training data

### Document summarization



### **Document Summarization**



**FINDWISE** 

### **Summaries**

(Approx. 40 words)

# [Multiple Kernel Learning]

The report said Andreas Lubitz repeatedly set the plane for an unauthorised descent earlier that day. He had locked the flight captain out of the cockpit. Five minutes on the Duesseldorf-Barcelona flight 07:21:10 - Plane told to descend to 21,000ft

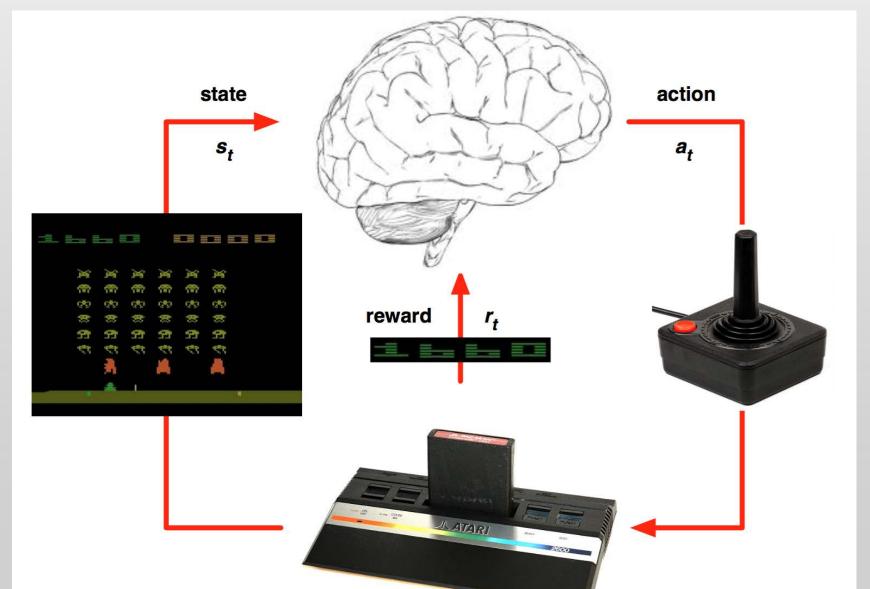
### [TextRank]

The co-pilot of the Germanwings plane that

### Original Text

SEARCH DRIVEN SOLUTION ness Tech Science Magazine Entertainment & Arts Health Pictures World selected Africa Asia Australia Europe selected Latin America Middle East US & Canada [Germanwings crash: Co-pilot Lubitz 'practised rapid descent'] 21 minutes ago From the section Europe [Germanwings co-pilot Andreas Lubitz is known to have suffered depression in the past] [Alps plane crash] What drives people to murder-suicide? The victims of the Germanwings plane crash Germanwings: Unanswered questions Flight 4U 9525: The final 30 minutes [The co-pilot of the Germanwings plane that crashed in the French Alps in March appears to have practised a rapid descent on a previous flight, a report by French investigators says.]] [The report said Andreas Lubitz repeatedly set the plane for an unauthorised descent earlier that day.] Lubitz is suspected of deliberately crashing the Airbus 320, killing all 150 people on board. [He had locked the flight captain out of the cockpit. 1 The plane had

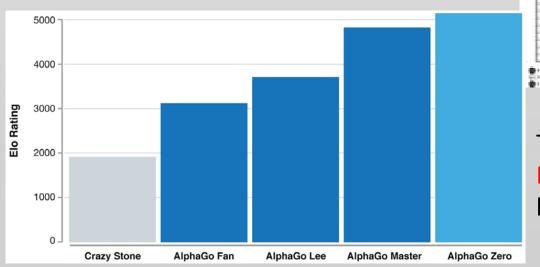
# Reinforcement Learning

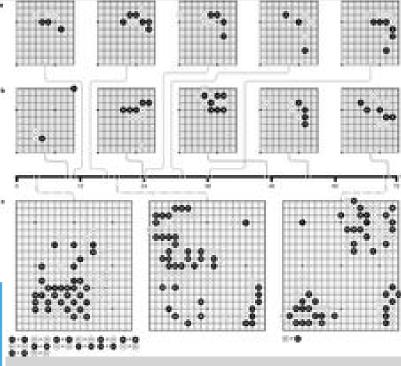




### AlphaGoZero: Al Tabula Rasa







Trained from scratch without any Human input only for 36 hours and beat the previous version 100-0!

### Learning to Communicate



- Agents interact sending messages to solve a common task.
- Invent a language grounded in real world objects

E, Jorge, M. Kageback and E. Gustavsson, "Learning to play *Guess Who?* And inventing a Grounded Language as a Consequence", NIPS Deep RL Workshop (2016)

**AI: NEW ELECTRICITY** 

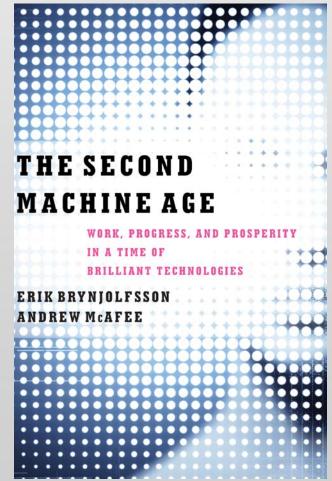
### Electricity and AI as General Purpose

Wide scope for Technologies improvement and

elaboration

 Application across a wide range of uses

- Potential for use in a wide variety of products and processes
- Strong complementarities with existing or potential new technologies



### Easy to Use and Improve

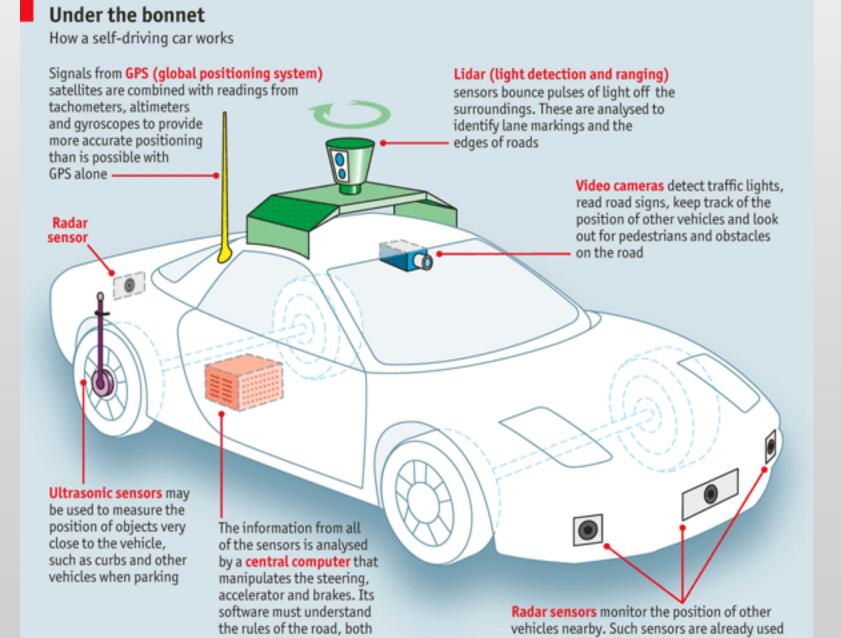




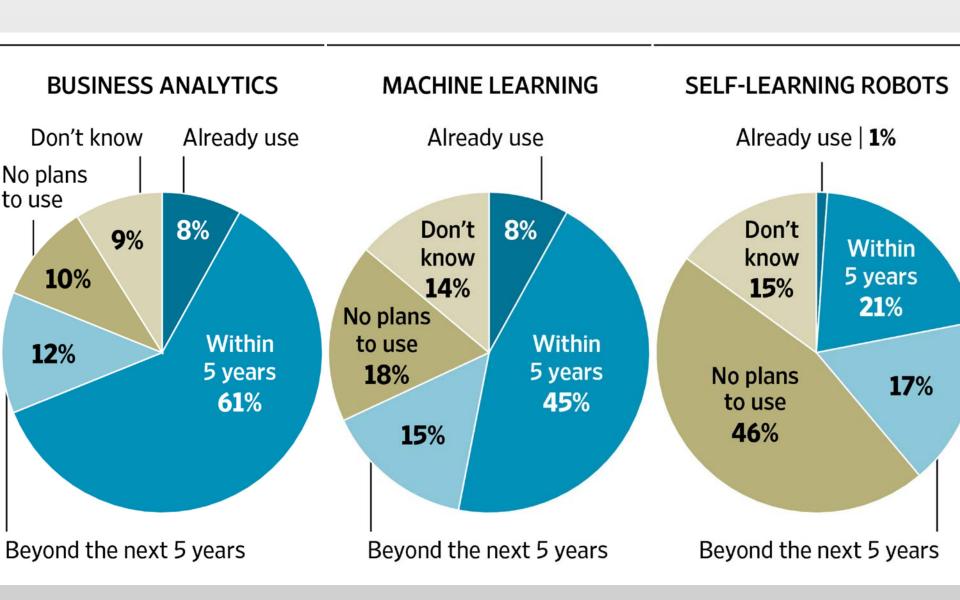




#### Data is the New Oil



#### MACHINE INTELLIGENCE 3.0 TECHNOLOGY STACK -AGENT ENABLERS ENTERPRISE INTELLIGENCE -(I) OCTANE.AI howdy. Maluub\( \lambda \in \text{KITT-AI} SENSOR -INTERNAL DATA -OpenAl Gym Kasisto OUTOMAT Orbital Insight Planet (F) Gridspace Tolkio PREDIX GOOT MAANA mattermark Quid semanticmachines PRIMER WATSON clarifa A DEEPVISION nexidia(b) ( twilio Tatafox PREMISE DATA SCIENCE -® Cycorp Q Palantir ARIMO\* Sentenai @ PLANET OS cortica @lgocion CAPIO Expect Labs Bottlenose MOTIVA SDOMING STARKBEYOND ( rapidminer UPTAKE WIMUBIT Preferred Alation @sapho Outlier enigma CBINSIGHTS SPACE\_KNOW Captricity Clover Mobvoi Alluvium kaggle DataRobot Vhat AYASDI Digital Reasoning OTracxn predata netra deepomatic Ourious.Al seldon **yseop** big MACHINE LEARNING ENTERPRISE FUNCTIONS CognitiveScale GoogleML Context relevant CUSTOMER SUPPORT— BCYLANCE CARKTRACE Otextio entelo MINTIGO Lattice RADIUS collective[i] **5sense** Digital Genius Kasisto SCALED SPARKCOGNITION CONTROL COMETRIC ₩ade & Wendy hi fuse|machines LELOQUENT Wiseio deepsense.io reac ive skymind 🦓 bonsai salesforce INSIDE SALES .COM unifive / SpringRole ACTIONIO Prendesk GIGSTER HIRE Vue NATURAL LANGUAG Preact @CLARABRIDGE X+AIYLIEN LEXALYTICS spaCy () LUMINOSO Narrative / **AUTONOMOUS** GENTS -Science > O cortical.io MonkeyLearn PROFESSIONAL -GROUND NAVIGATION -DEVELOPMENT drive M @ AdasWorks SKYDIO SHI Dbutterai Pogo SKIPFLAG **I**SIGOPT HyperOpt fuzzy<sup>i0</sup> okite CLEARPATH wfetch Cortana Allo ZOOX MOBILEVE Airware LIII - LILY 🔘 clara 💢 x.ai 🔅 slack rainforest **lobe** Anodot OUBER GOOD TIESLA facebook 🔏 7 DroneDeploy KINDRED Signifai i AYER 6 \* 🤲 bonsai talla Zoom sudo pilotai 🙏 SKYCATCH OnuTanamy Auro Robotics IN HARVEST Siri 0 @ Replika DATA CAPTURE CrowdFlower & diffbot CrowdAl import INDUSTRIES Paxata DATASIFT amazon mechanical turk enigma INVESTMENT -LEGAL -LOGISTICS -AGRICULTURE -EDUCATION -WorkFusion DATALOGUE TRIFACTA parsehub blue J BEAGLE BLUEØRIVER MAVYX KNEWTON Volley Bloomberg sentient M NAUTO Acerta **OPEN SOURCE LIBRARIES** tule ATRACE Pivot gradescope Everlaw RAVEL PRETECKT CA ISENTIUM KENSHO Keras Chainer CNTK TensorFlow ( **VCTI** coursera Sseal ROSS ATETTAVION AGRI-DATA # alphasense Dottoming Routific clearmetal H2O DEEPLEARNING4J C-CEREBELLUM Quandl UUDACITY alt school Descartes Descartes Objection LEGAL ROBOT MARBLE PITSTOP DSSTNE Scikit-learn AzureML nneon MXNet DMTK Spork PaddlePaddle WEKA INDUSTRIES CONT'D -HARDWARE -KNUPATH TENSTORRENT CIrrascale MATERIALS -**RETAIL FINANCE** -(intel nervana Movidius PULSE CareSkore NIDIA zymergen Citrine BUTTERFLY 3SCAN TALA zestfinance TICarbonX color GRAIL Eigen Innovations tensilica GoogleTPU \$\infty 1026 Labs Qualcomm ARTERYS @ enlitic deep genomics @RECURSION Lendo earnest (0) SIGHT MACHINE NLUMINIST Numerate BAYLABS @ imagia Cerebras Isosemi Affirm /// MIRADOR Oncord @sentrian GINKGO Manotronics Atomwise verily ON HIOME RESEARCH -Atomwise Numerate Coogle DeepMind wealthfront Betterment CALCULARIO OpenAl Processes ELEMENT' Vicarious KNOGGIN ANumenta Kimera Systems Cogital shivonzilis.com/MACHINEINTELLIGENCE · Bloomberg BETA

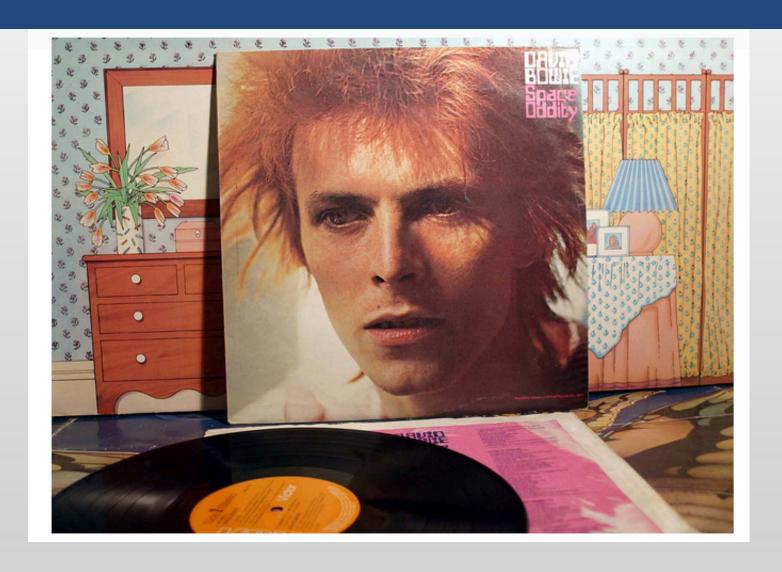


### **Complementary Technologies**

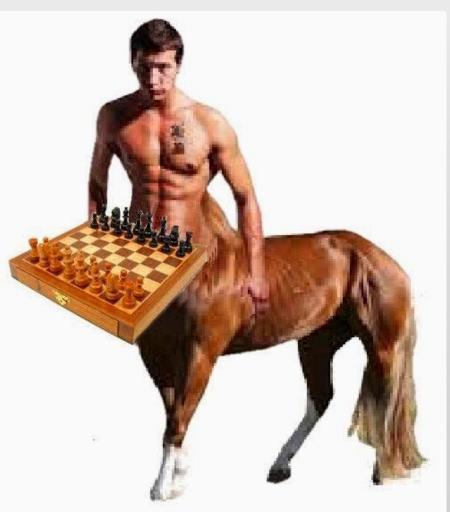


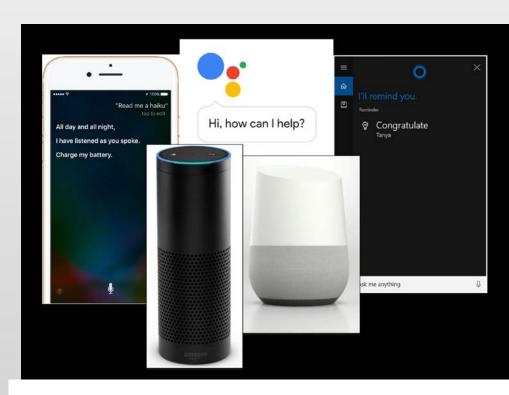
CRISPR/CAS9 Gene Editing for Life Sciences

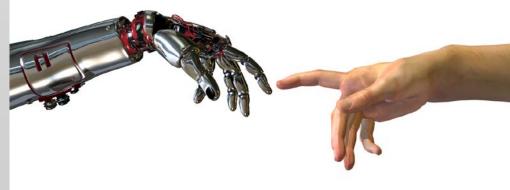
**Additive Manufacturing** 



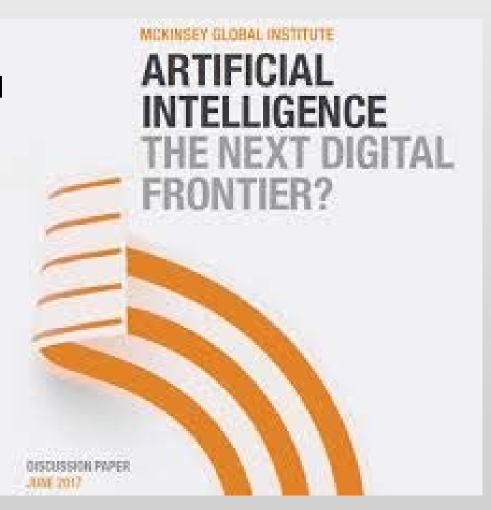
Music itself is going to become like running water or electricity.



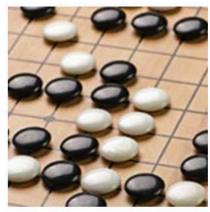




- Al will contribute as much as \$15.7 trillion to the world economy by 2030 (PwC)
- \$6.6 trillion from increased productivity as businesses automate processes and augment with new Al technology, and \$9.1 trillion from consumption sideeffects as shoppers snap up personalized and higherquality goods



### BigData@Chalmers Seminars



10 November 2017 - BigData@Chalmers seminar

Artificial Intelligence Starting From a Blank Slate

Speakers: Devdatt Dubhashi, Mikael Kågebäck and others,

Department of Computer Science and Engineering, Chalmers

More information >>



27 October 2017 - BigData@Chalmers seminar **Privacy in the hands of the data analyzer** Speaker: Kobbi Nissim, Georgetown University **More information** >>

# Data Science: Algorithms

| LP 1                            | LP 2                                                                                         | LP 3                                                                                                | LP 4                                                 |
|---------------------------------|----------------------------------------------------------------------------------------------|-----------------------------------------------------------------------------------------------------|------------------------------------------------------|
| Introduction to Data<br>Science | Stochastic<br>processes and<br>Bayesian statistics                                           | DAT340 - Applied machine learning                                                                   | DAT345 -<br>Techniques for<br>large-scale data       |
| Nonlinear optimization          | TDA507 Computational methods in bioinformatics (eller) DAT246 Empirical software engineering | TDA206 Discrete optimization  (eller)  CIU187 Information visualization  (eller)  TIN093 Algorithms | TDA231 Algorithms for Machine Learning and Inference |

| LP 1                            | LP 2                  | LP 3            | LP 4 |
|---------------------------------|-----------------------|-----------------|------|
| TIN093 Algorithms               | Data science project  | Master's thesis |      |
| SSY340 Deep<br>Machine Learning | TDA251<br>Algorithms, |                 |      |

## Data Science: Large Scale Systems

| LP 1                            | LP 2                                               | LP3                                                                           | LP 4                                                 |
|---------------------------------|----------------------------------------------------|-------------------------------------------------------------------------------|------------------------------------------------------|
| Introduction to Data<br>Science | Stochastic<br>processes and<br>Bayesian statistics | DAT340 - Applied machine learning                                             | DAT345 -<br>Techniques for<br>large-scale data       |
| Nonlinear<br>optimization       | TDA596<br>Distributed<br>systems                   | EDA263 Computer<br>Security<br>(eller)<br>TDA297<br>Distributed<br>systems II | TDA231 Algorithms for Machine Learning and Inference |

| LP 1                                                              | LP 2                              | LP3             | LP 4 |
|-------------------------------------------------------------------|-----------------------------------|-----------------|------|
| DAT300<br>Data-driven<br>support for<br>cyber-physical<br>systems | Data science project              | Master's thesis |      |
| SSY340 Deep<br>Machine Learning                                   | DAT295 Autonomous and Cooperative |                 |      |

## Data Science: Optimization

| LP 1                            | LP 2                                               | LP 3                                                                                                           | LP 4                                                          |
|---------------------------------|----------------------------------------------------|----------------------------------------------------------------------------------------------------------------|---------------------------------------------------------------|
| Introduction to Data<br>Science | Stochastic<br>processes and<br>Bayesian statistics | DAT340 - Applied machine learning                                                                              | DAT345 -<br>Techniques for<br>large-scale data                |
| Nonlinear<br>optimization       | MVE190 Linear<br>statistical models<br>[MV]        | TDA206 Discrete optimization (eller) SSY097 - Image analysis [EE] (eller) TMA521 Large-scale optimization [MV] | TDA231<br>Algorithms for<br>Machine Learning<br>and Inference |

| LP 1                                                                                                                      | LP 2                                                    | LP 3            | LP 4 |
|---------------------------------------------------------------------------------------------------------------------------|---------------------------------------------------------|-----------------|------|
| TMA265 Numerical linear algebra [MV] (eller) TMA881 High performance computing [MV] (eller) RRY025 Image processing [SEE] | Data science project                                    | Master's thesis |      |
| SSY340 Deep<br>Machine Learning                                                                                           | TDA507<br>Computational<br>methods in<br>bioinformatics |                 |      |

#### Tillvalskurser

Nedan följer en lista med blandade tillvalskurser på olika institutioner utöver de som nämnts ovan.

#### LP1

- TMA265 Numerical linear algebra (Math)
- MVE187 Computational methods for Bayesian statistics (Math)
- FFR105 Stochastic optimization algorithms (Physics)
- TIF160 Humanoid robotics (Physics)
- RRY025 Image processing (Earth and Space Sciences)
- FFR135 Artificial neural networks (Physics)

#### LP2

- DAT246 Empirical software engineering (CSE)
- TDA507 Computational methods in bioinformatics (CSE)
- TIF155 Dynamical systems (Physics)
- FFR120 Simulation of complex systems (Physics)
- SSY130 Applied signal processing (E2)
- TDA357 Databases (CSE) (\*\* First cycle course)

#### LP3

- TMA521 Large-scale optimization (Math) (\*\* Given every other year)
- TIF150 Information theory for complex systems (Physics)
- FFR110 Computational biology A (Physics)
- SSY097 Image analysis(S2/Elec)
- VVT105 Geographical Information Systems (Architecture)

#### LP4

- TMS016 Statistical image analysis (Math)
- MVE220 Financial risk (Math)
- TMS087 Financial Time Series (Math)
- MVE440 Statistical Learning for Big Data (Math)
- SSY115 eHealth (E2)
- SSY315 Bayesian statistics (E2)

