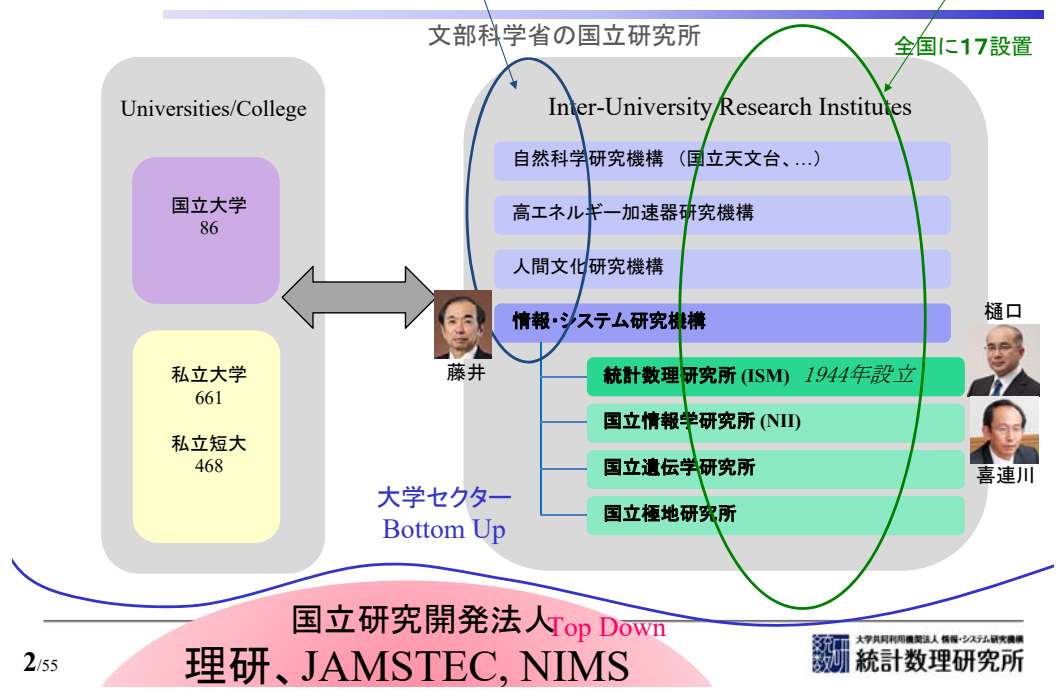


# 人工知能とデータサイエンティストの 役回り

樋口知之 (情報・システム研究機構 統計数理研究所)



# 大学共同利用機関法人と大学共同利用機関



## アウトライン

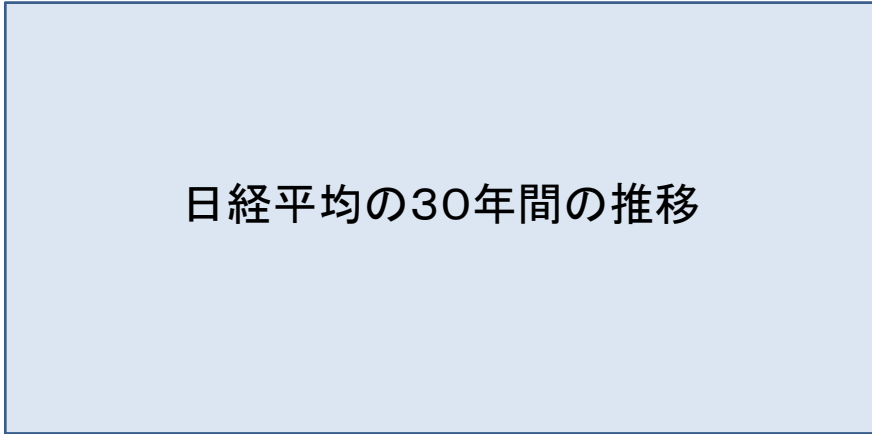
1. 平成の30年間
2. 統計学、機械学習、深層学習
3. 帰納法の弱点
4. 人材育成

## アウトライン

1. 平成の30年間
2. 統計学、機械学習、深層学習
3. 帰納法の弱点
4. 人材育成

# 20年、30年、そして50年

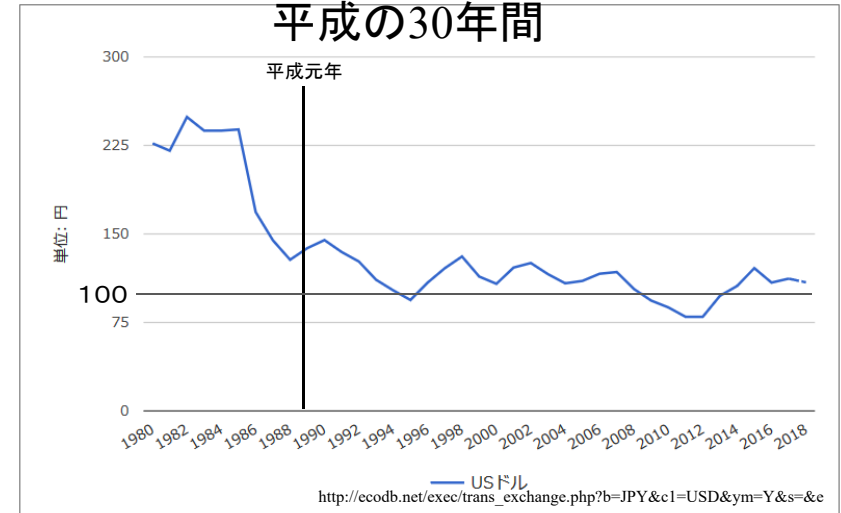
## 平成の30年間



(図のソース: 2017/12/29日経Webニュース)

# 円ドルレートと時価総額

## 平成の30年間



# Manufacturing companies to IT companies

Top of market capitalization of stock price

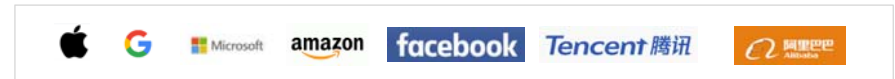
Compared with 10 years ago

	2007, May		2017, May	
469	Exxon Mobil	1	Apple	796 ×7.6
	GE	2	Google	×4.3
	Microsoft	3	Microsoft	×1.8
	Citi Group	4	Amazon	×16.8
	Petro China	5	Facebook	---
	AT&T	6	Berkshire Hathaway	
	Royal Dutch Shell	7	Johnson & Johnson	
	Bank of America	8	Exxon Mobil	×0.7
	Industrial and Commercial Bank of China	9	Tencent 腾讯	×42.5
216	Toyota	10	Alibaba 阿里巴巴	297

billion US\$

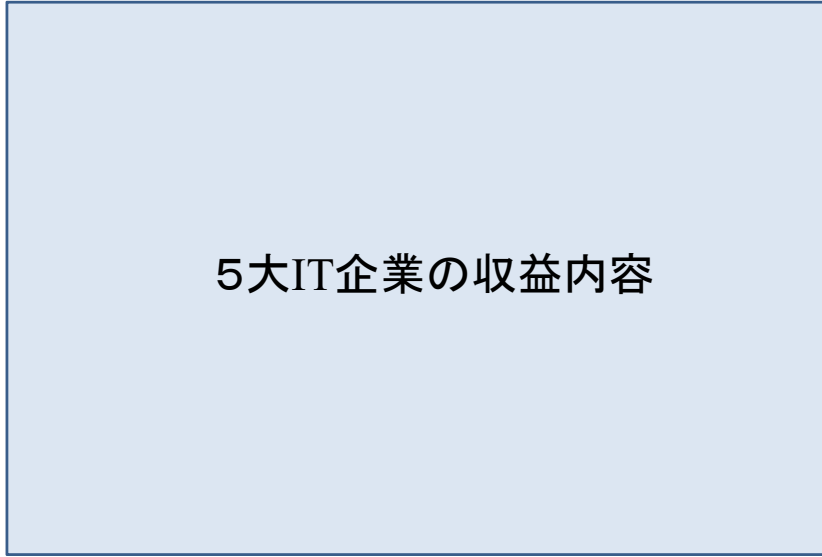
Original News Source: NikkeiWeb2017/June/02

# Modern Seven Sisters



<http://blog.livedoor.jp/sagittariun/archives/11995435.html>

# IT主要5大企業の収益源

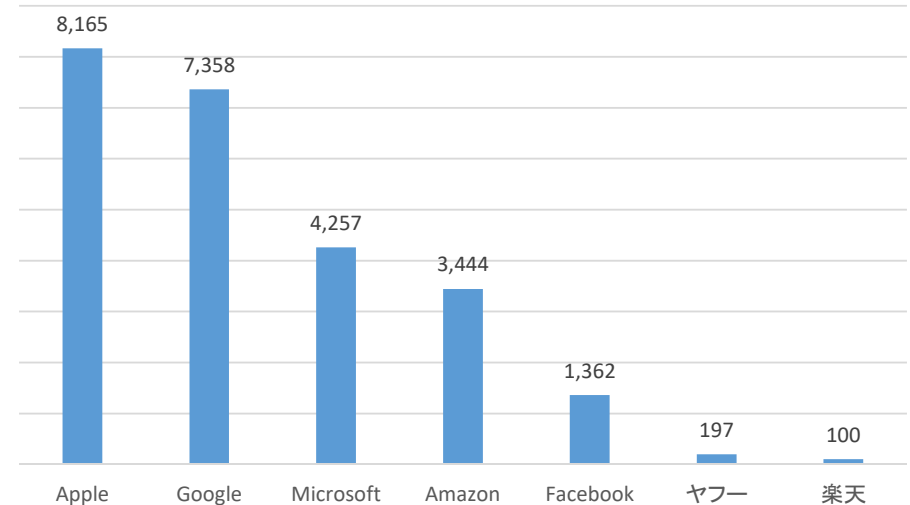


5大IT企業の収益内容

(ソース <https://www.businessinsider.jp/post-33925>  
<http://www.visualcapitalist.com/chart-5-tech-giants-make-billions/>)

# 投資額の格差 (USのAI×データ企業との比較)

設備投資額 (億円/年)



各社のFY2013 IR資料：有形固定資産の取得、100円/\$で換算

安宅和人 「“シン・ニホン”：AI×データ時代における日本の再生と人材育成」を参照  
[http://www.meti.go.jp/committee/sankoushin/shin\\_sangyoukouzou/pdf/013\\_06\\_00.pdf](http://www.meti.go.jp/committee/sankoushin/shin_sangyoukouzou/pdf/013_06_00.pdf)

# 検索＋スマホ世代が変える社会

知識がモノの形で伝達 → 知識が脳内で構造化  
 ↓  
 知識がフラット化

価値観の転換

- Sharing
- エコシステム
- Citizen Science

アントレプレナー  
ブーム

# 人工知能学会の栄枯盛衰 1986年創設

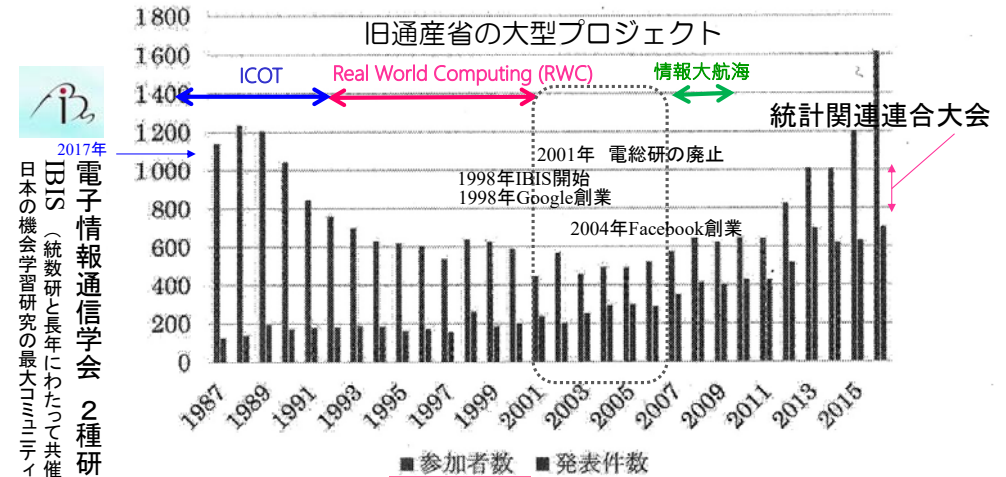
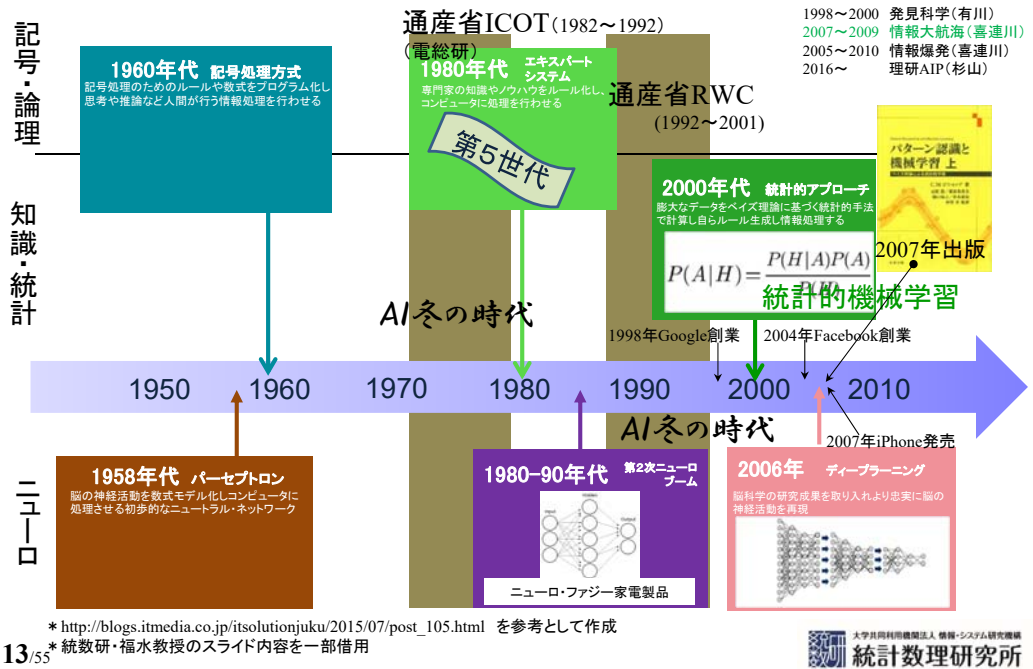


図4 全国大会参加者数と発表件数の推移

2016年10月に、それまで最高であった1992年10月の会員数を超える

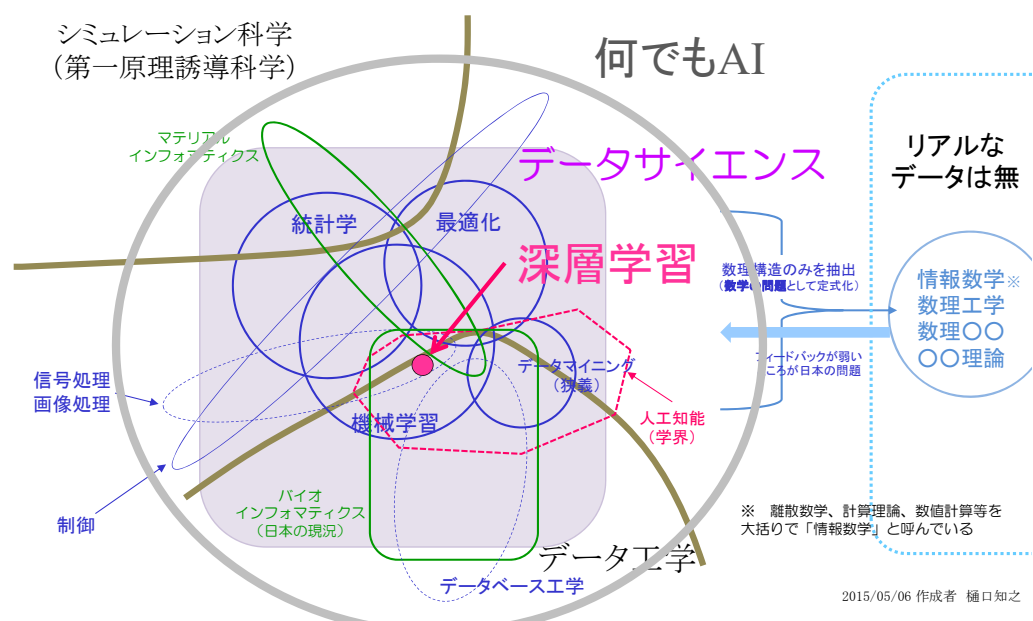
# 人工知能研究および研究開発プロジェクトの歴史



# アウトライン

1. 平成の30年間
2. 統計学、機械学習、深層学習
3. 帰納法の弱点
4. 人材育成

# データに関連した数理分野の俯瞰図



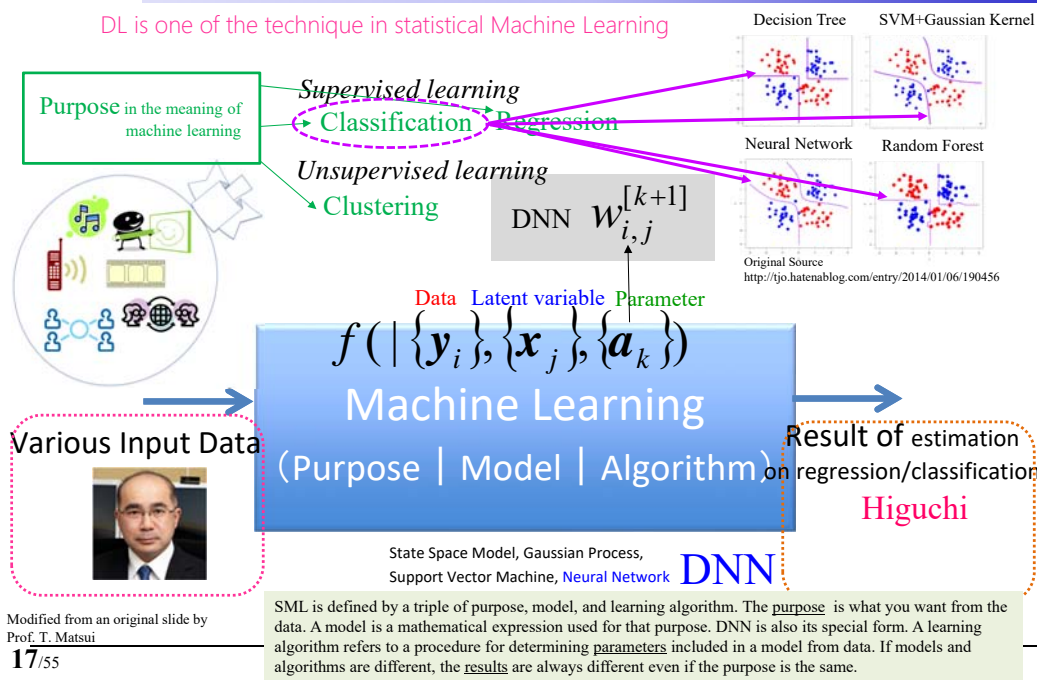
# Prosperity of AI (To be precise, "The Third Neural network Boom")

- In the past seven years, the AI field has developed dramatically
  - IBM Watson (Question Answering System) wins Human Quiz King in 2011.
 

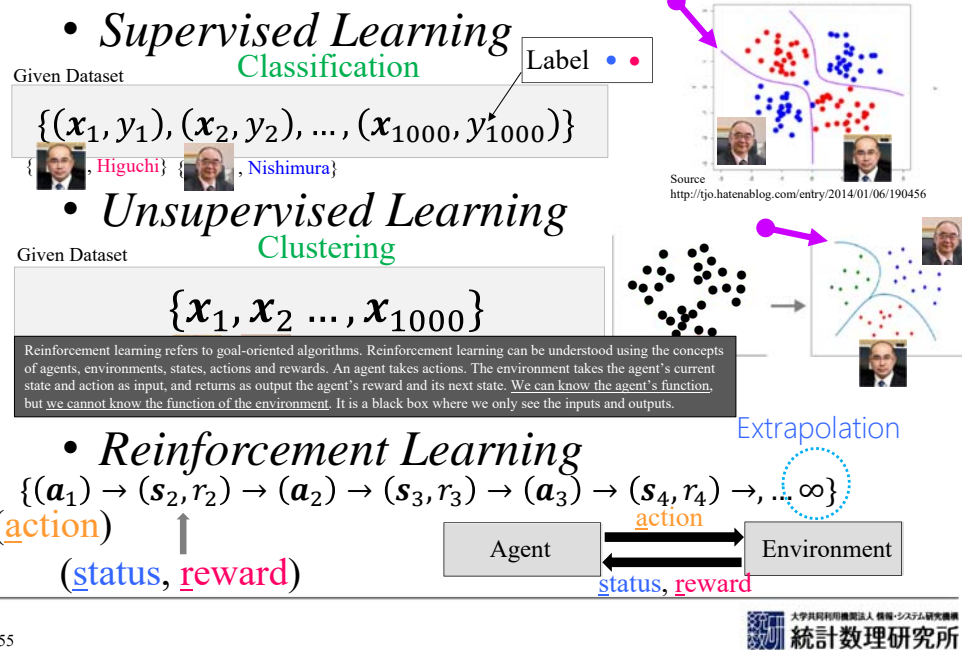
No part of deep learning.  
Knowledge base of Big data analysis result + expert system
  - Overwhelming victory with Deep Learning in many contests, such as **images and sounds** competitions from 2011
    - a. 2011 Image recognition of traffic sign; (erroneous recognition rate 0.56%, second place 1.16%)
    - b. 2012 General object recognition for **ImageNet** (misrecognition rate 15.3%, second place 26.2%)
  - Increased attention to deep learning. IT companies focus on research
    - Google: Google Brain Project, Google Deep mind
    - Facebook: Facebook AI Research
    - Baidu
  - 2018 Google AlphaGO wins the world pro

# Triplet defining statistical machine learning

DL is one of the technique in statistical Machine Learning



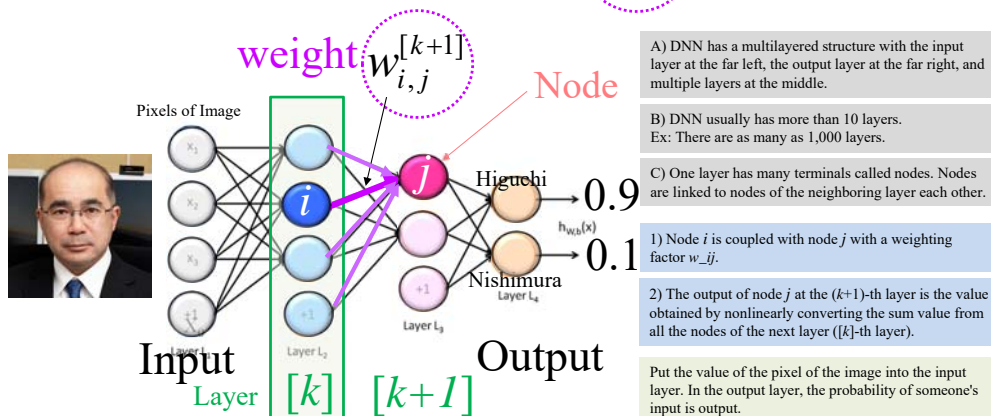
# Three main types of Machine Learning methods



# Multilayer Neural Network and Deep NN

The word of Deep Learning (DL) is used like Deep Neural Network (DNN)

$$f(x_j^{[k+1]}) = \frac{1}{1 + \exp\left\{ \sum_{i=1}^{I([k+1])} w_{i,j}^{[k+1]} x_i^{[k]} + b_j^{[k+1]} \right\}}$$



# 深層学習の強み：人工知能の「度量衡」

長さを計る基準を「度」、体積は「量」、重さは「衡」と定め、...

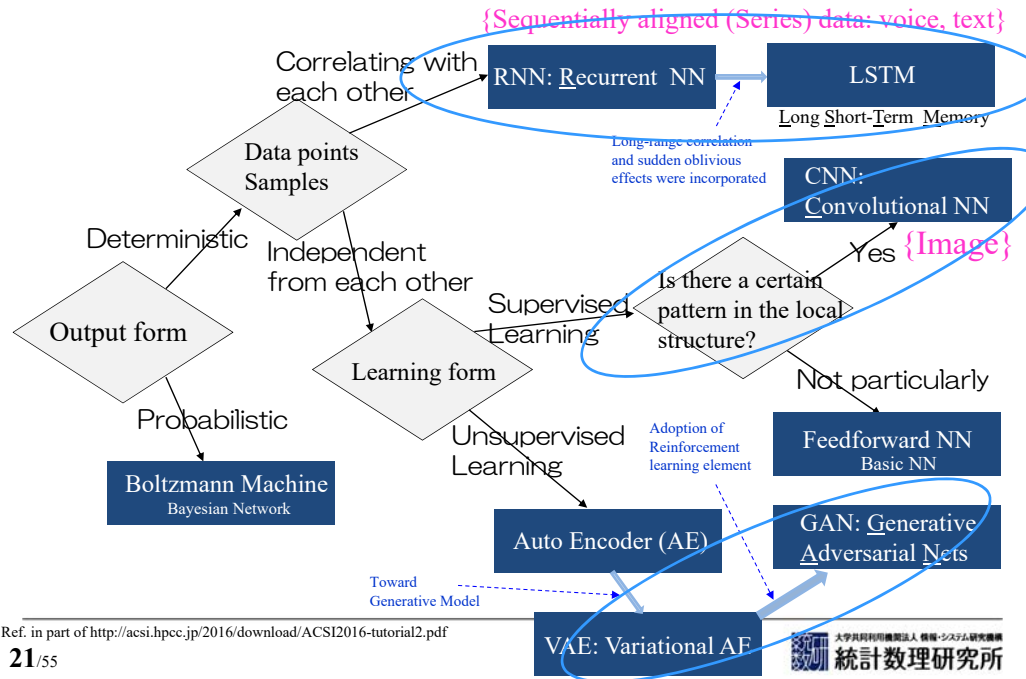
2018年 人工知能学会誌 特集「AIとデータ」に招待論文を寄稿

<http://100percent-magazine.net/2015/09/08/759>

- ① パラメータ学習がバックプロパゲーションと確率的勾配降下法で統一
- ② 計算プラットフォームが汎化
  - a. 専用計算機の整備 (比較的廉価)
    - GPGPU クラスタ
  - b. オープン・無償な開発プラットフォームの整備
    - オープンソース: TensorFlow, Caffe, Keras, Chainer
- ③ オープンかつリアルタイムな成果公開
  - ✓ 査読プロセスをとらず、プレリサーバ(arXiv)にアップ

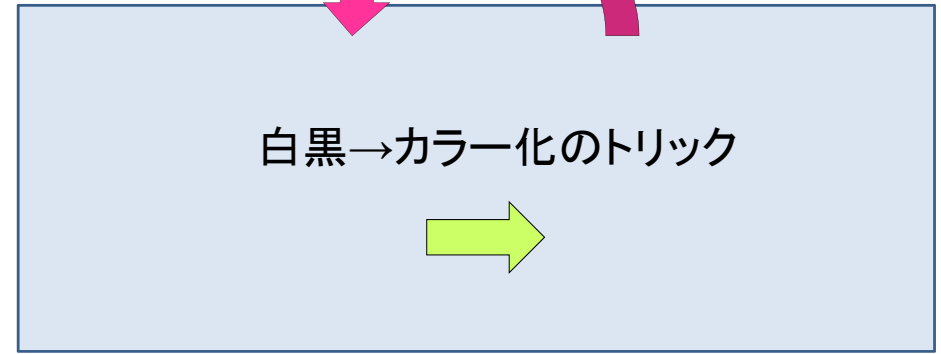
理論的には第二次AIブームから大きな発展はこれから.....

# Classification of DNN



# Automatic coloring (black and white photography → color) tool

(2016, May 31)



## Supervised Learning

Source: <http://www.itmedia.co.jp/news/articles/1605/31/news119.html>

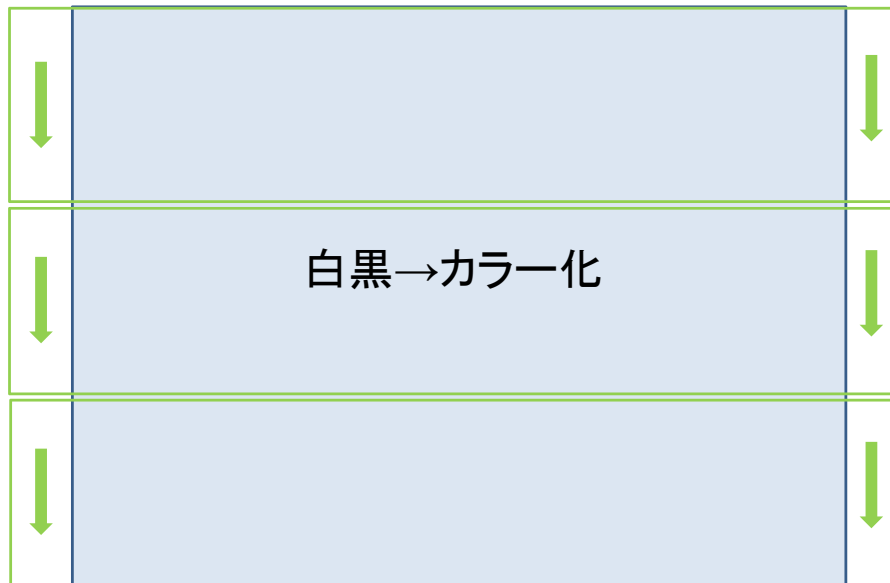


With DL, it is also possible to color black and white photos. The actual procedure is as follows. First, from a large amount of color photographs on the Internet, a black-and-white photograph is created by several methods. By doing this, you can get a huge amount of sets of black-and-white photos and color photos. We assign black-and-white photograph to an input of DL, color photograph to an output, and then perform supervised learning.

22/55

統計数理研究所

# Automatic coloring 2



Source: <http://www.itmedia.co.jp/news/articles/1605/31/news119.html>

23/55

統計数理研究所

# Work at the cash register and Passport Control



The work required at each shop need only take pictures from various angles for one product. Even products that do not have bar codes can be automated.

The task of object recognition by human eyes has been replaced by machines, and that system is deeply penetrated into our social life.

羽田空港・成田空港の  
入国ゲートの様子

24/55

統計数理研究所

# Prevent overlooking by doctors

## 内視鏡画像中のがん細胞のAI同定

Example of polyp detection

Source: <http://www.itmedia.co.jp/news/articles/1707/10/news095.html>

2017, July

- ✓ The aim is to prevent overlooking due to lesions and sites of occurrence difficult to detect with the naked eye, technological disparities of doctors.
- ✓ The established system at the National Cancer Center (NCC), Japan, uses AI employing deep learning to process its own algorithm for an image analysis at high speed.
- ✓ About 5,000 endoscopic images diagnosed at NCC are used for learning parameters in DL.

# The future of medical care depends on medical stakeholders and AI



2017, Feb

## Helping clinicians get patients from test to treatment, faster



The results so far...

We set up DeepMind Health to put the UK's most advanced technology at the service of patients, nurses and doctors.

27 February 2017

In order to make use of the power of this deep learning for the medical field, DeepMind recently launched company DeepMind Health in UK (about 2 years ago). DeepMind is a company that produced Alpha-Go and Alpha-zero, which won the Go professional. UK is very active in using big data concerning health and medical care.

# Amazing advancements of Speech recognition and text processing

## Smart Speaker

<https://smarthacks.jp/mag/23875>

Amazon and Google don't have an intention to make a profit by selling these products. Through these products, the purpose is to store the living space information of various people as a whole big data. And, in order to create the next information service from big data, to dominate the market and gain profits, they are daringly making "manufacturing".

By adopting a deep neural network suitable for processing sequence data with long correlation called LSTM in **November 2016**, the automatic translation function has improved significantly.



## Automatic translation

# The judge becomes a job of AI

## Precog in Minority Reports (2002) Robot Police in Elysium (2013)

- Utilizing AI is not limited to lawyer work, and its introduction is also under way in the hearing process of the trial. There are many mistakes due to misunderstanding of the clerk, etc. in the record to be recorded. It can be expected that the burden will be reduced.
- It is said that AI judges can realize by learning past case data. The judge is a suitable job for AI in that it absorbs the data of enormous cases in the past and *can be designed* to make an objective judgment rather than human.

# Bank work disappears at AI

## 1) Window operations such as deposit transfer

It is almost compatible with ATM already. Virtual banking personnel will now respond to taxpayers who need manual labor.

## 2) Loan and foreign exchange business

The machine automatically judges the analysis and rating of the company to be financed and can also present conditions such as interest rates. AI system for foreign exchange can check documents and exchange money.

## 3) Securities / investment trust sales business

Algorithmic trading that automates stock ordering is mainstream. A choice of investment targets and establishment of investment period by AI are also advanced.

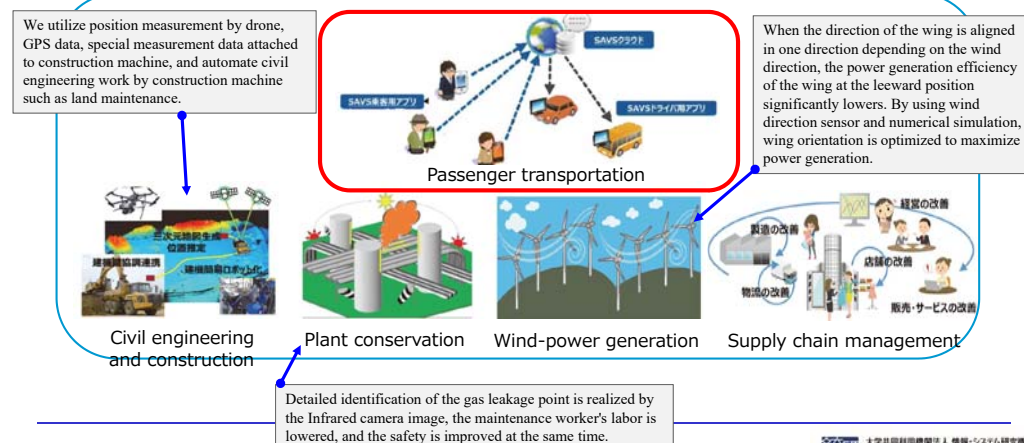
## 4) Trust business

In addition to administrative management departments and operations departments in real estate and securities, advice tasks such as **testamentary trusts and inheritance countermeasures** are replaced with AI.

# NEDO (New Energy and industrial technology Development Organization) R&D project for AI and robotics

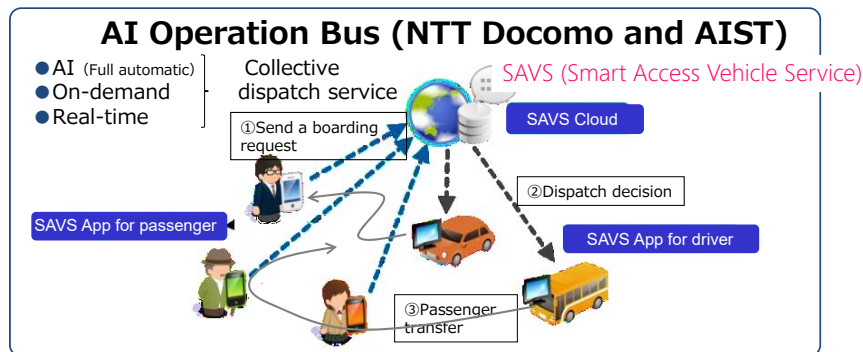
NEDO has started the project "**Integration technology development as the core of next generation AI and robot**" project aiming for early social implementation of AI technology from last year.

## Social implementation of AI technology to 5 areas



## Yokohama MaaS (Mobility as a Service) Demonstration experiment of AI operation bus

- In order to realize on-demand multipart traffic in urban areas, it is necessary to have a huge amount of movement demand unique to urban areas, and data that serves as a basis for allocation/strategy in the large change.
- In the "AI Operation Bus" demonstration experiment, we acquire vehicle dispatch data and data on the user in urban areas, as well as user data such as convenience and comfort. By using the acquired data, we can realize dispatch/operation processing by AI corresponding to the movement demand in urban areas.



## 横浜MaaS実証概要＋輸送数推移

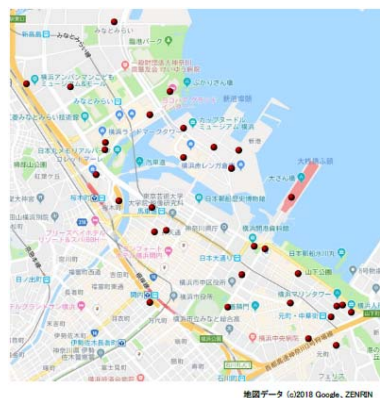
2018年12月10日  
株式会社NTTドコモ  
IoTビジネス部 先進ビジネス推進



「便利な移動手段により商業施設の利用が高まる」  
 「魅力ある商業施設のため交通需要が増加する」 の好循環を促す  
 “便利な移動+サービス”を連携させるMaaSプラットフォームによる実証

【実証実験概要】

- ◆ 実施時期：2018年10月5日～12月10日
- ◆ 利用方法：
  - ・専用スマートフォンアプリ等による配車要求
  - ・協賛する商業施設の情報やクーポンをスマホに配信
- ◆ 輸送能力：同時に50人（10台程度のタクシー車両）
- ◆ モニタ：観光客を中心とする来街者（無料）
- ◆ 実証の特徴
  - ・オンデマンド/ダイナミックルート/シェアリング交通
  - ・500程度の地場商業施設と連携(実証期間中に倍増)

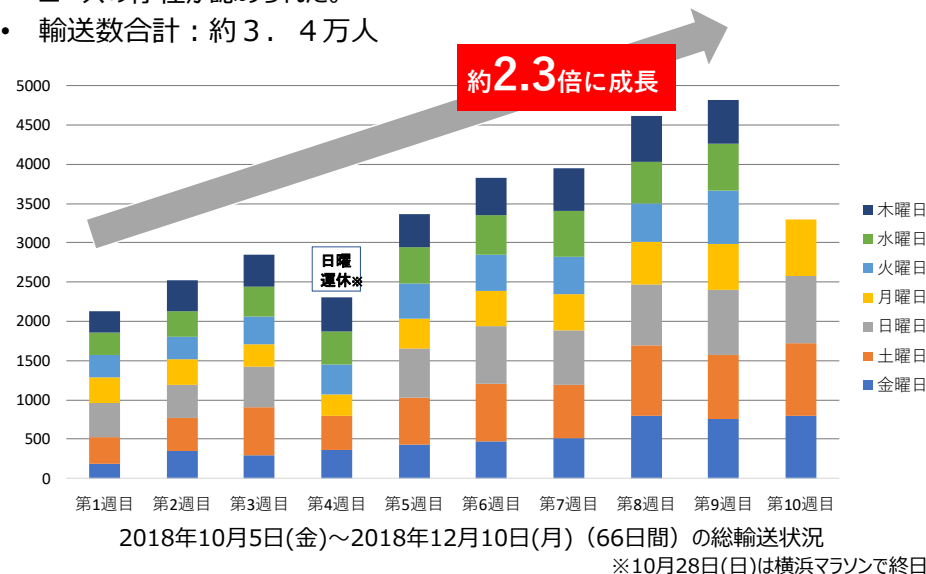


【提供価値（仮説）】

観光客	商業施設	自治体
便利な移動・お得な旅行	来街者に合わせた情報発信	災害時の来街者の影響把握

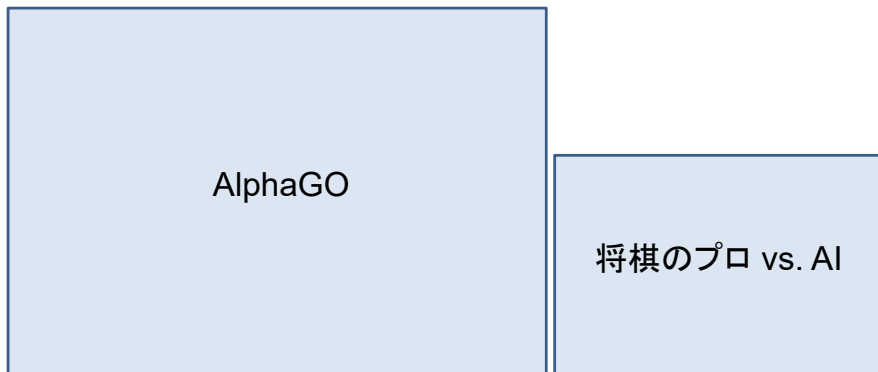
みなとみらい21、関内エリアに31ヶ所の乗降ポイントを設置  
(サイン型予約端末「まちかど端末」は28か所設置)

- 開始から9週間の輸送数は約2.3倍の成長であり、その後も伸び続け、ニーズの存在が認められた。
- 輸送数合計：約3.4万人



Deep Learning with Reinforcement Learning

- Google AlphaGO wins the world pro

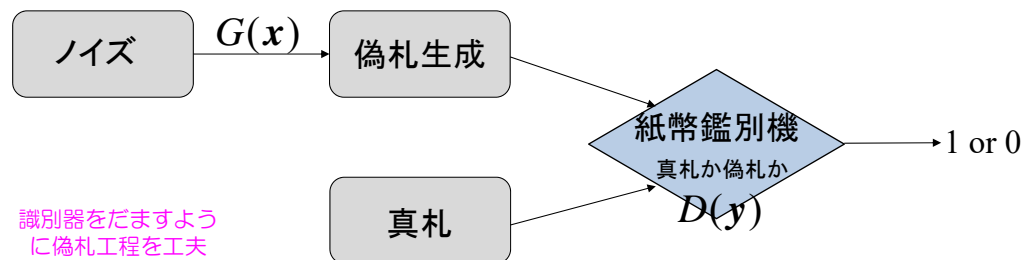


The biggest feature of **Alpha Zero** is that it learned three games with the same program. When it repeat self-game, it will initially show random hands and hands, but gradually learn from the outcome of win and lose and choose more advantageous hands. The time required for learning until reaching a certain strength is about 9 hours for chess, about 12 hours for shogi and 13 days for go. It has become strong until they beat other AIs of the world champions level. Traditional artificial intelligence used **practical game record for learning**.

いたちごっこ：GANの基本アルゴリズムの比喩的解説

$D(y)$ ：真札である確率

$G(x)$ ：偽札を生成するモデル(偽札工程)



識別器をだますように偽札工程を工夫

真札・偽札を正しく識別

学習アルゴリズム

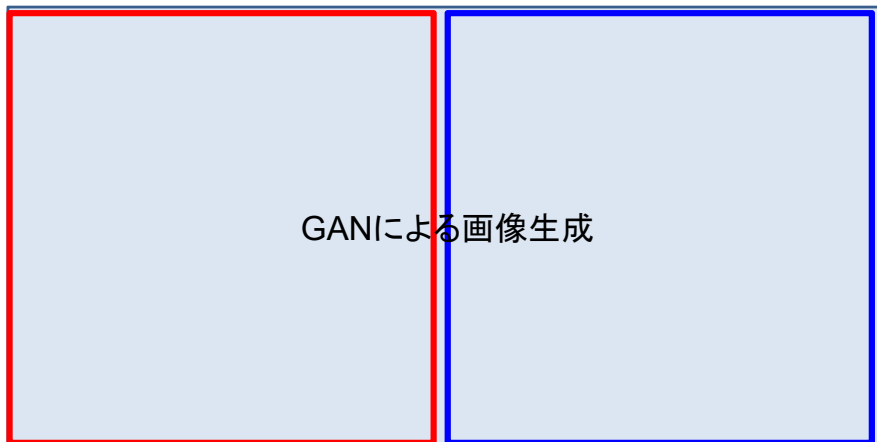
$$\min_G \max_D E_{y \sim p_{\text{data}}(y)} [\ln D(y)] + E_{x \sim p_x(x)} [\ln(1 - D(G(x)))]$$

GANを使った報告は、うまくいった例しか出さないケースが多々あり！

# Deep Convolutional Generative Adversarial Network (DCGAN)

Unsupervised Representation Learning with Deep Convolutional Generative Adversarial Networks  
Alec Radford, Luke Metz, Soumith Chintala,

(2016)

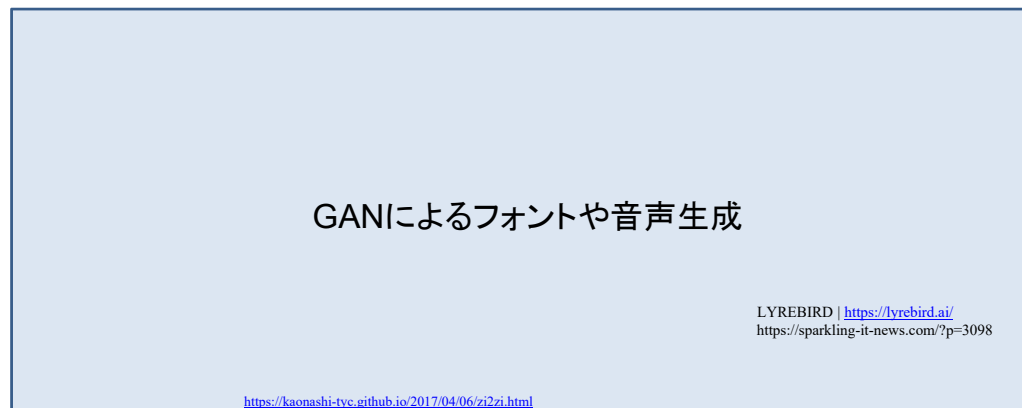


<http://www.whichfaceisreal.com/index.php>

From <https://ishmaelbelghazi.github.io/ALI/>

# GAN的技術の先にあるのは...

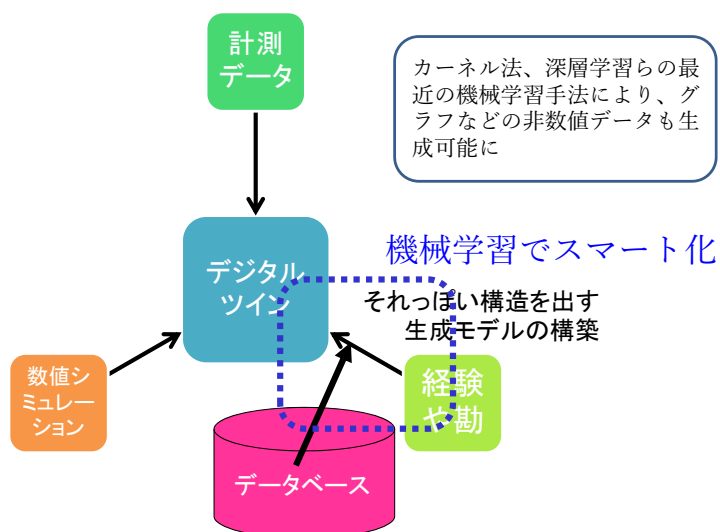
■ 真贋判別の機能を具備する器機には破壊的影響あり



2017年4月6日

2017年4月26日

# 暗黙知（経験と勘）生成モデル

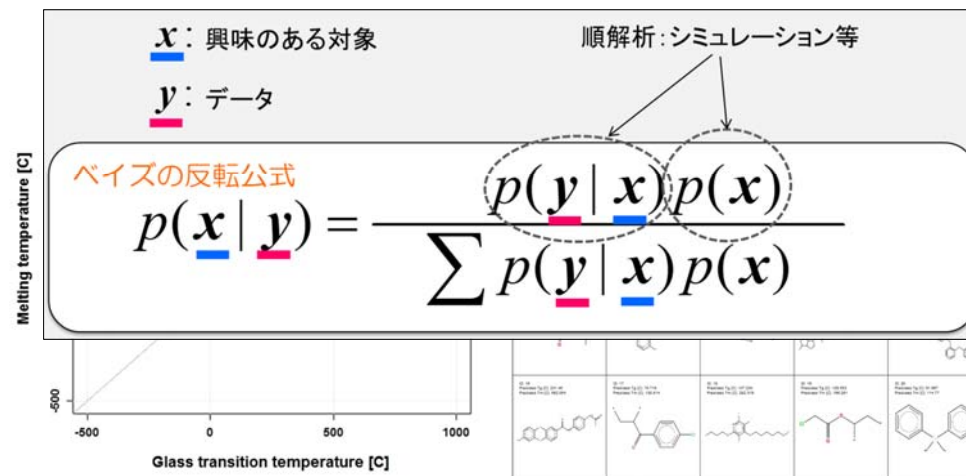


# NEW FUNCTIONAL POLYMERS

Higher glass transition temperatures and higher melting points



吉田亮教授@統計数理研究所

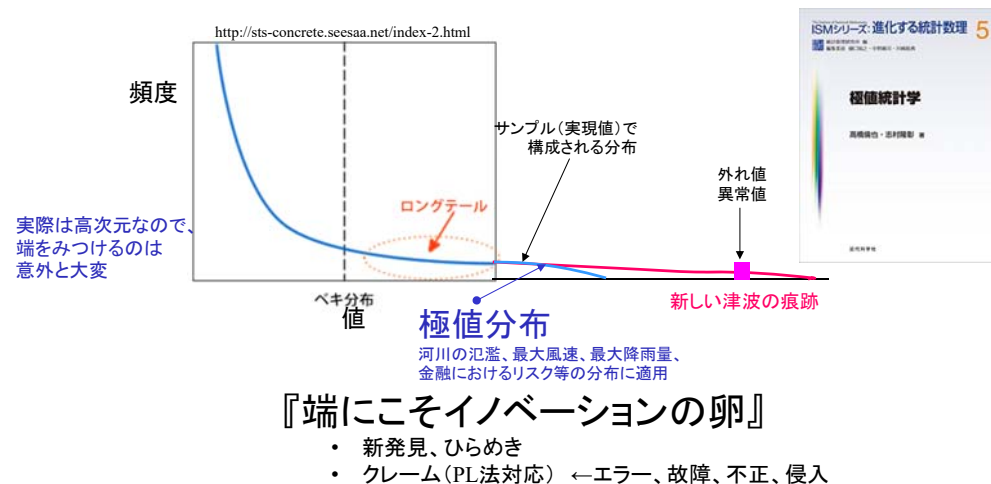


Orange: existing polymers with experimentally measured  $T_g$  and  $T_m$   
Blue: polymers computationally created by *iqspr* (circle size: the magnitude of likelihood)

# アウトライン

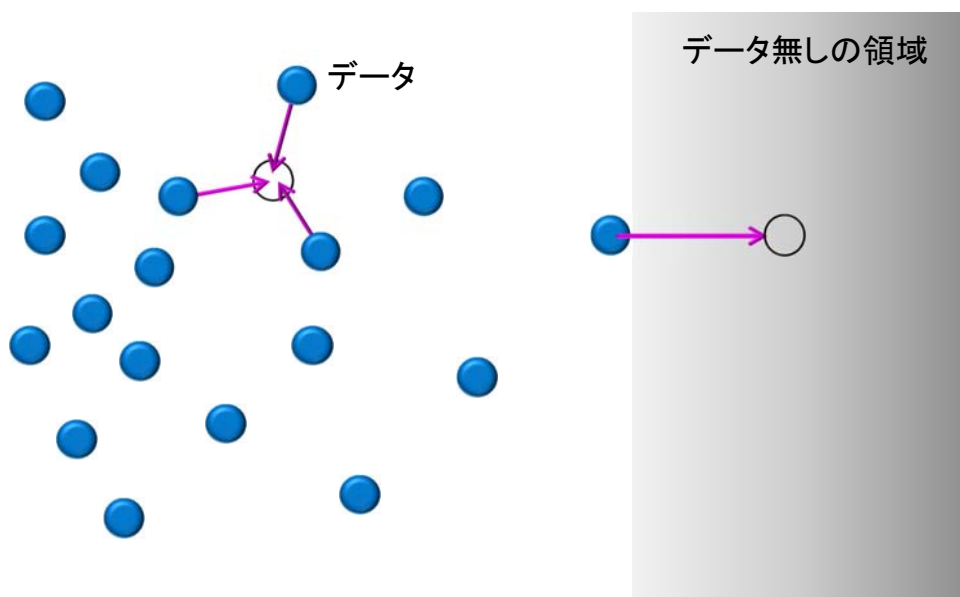
1. 平成の30年間
2. 統計学、機械学習、深層学習
3. 帰納法の弱点
4. 人材育成

# 帰納法の弱点1：不（未）観測とデータの質 サンプリングの問題

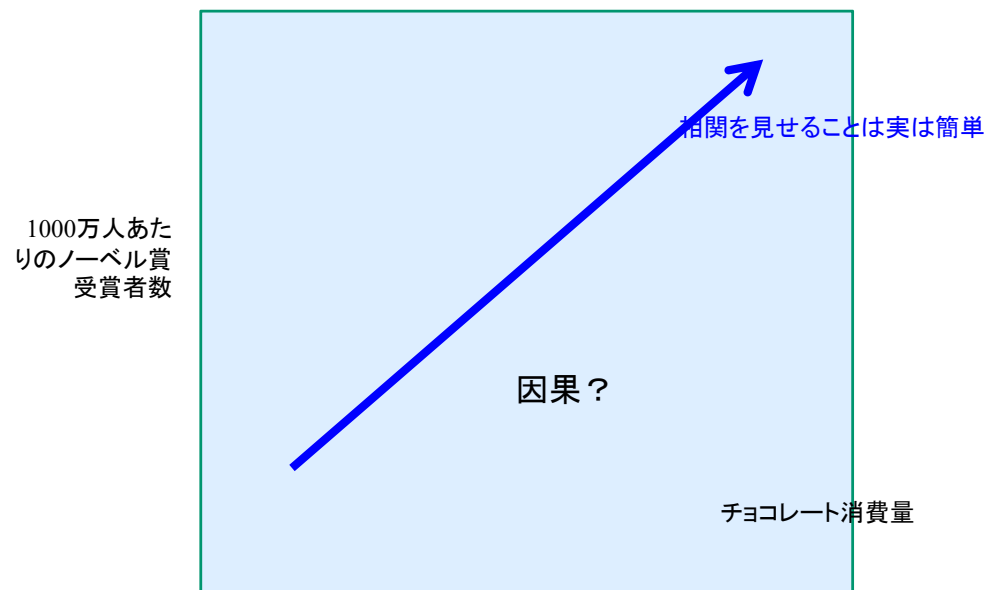


- ✓ データの偏り
- ✓ そもそも観測できていない?

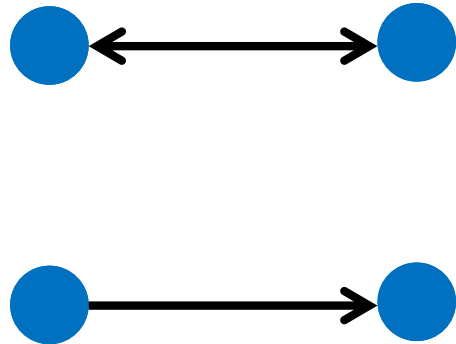
# 内挿と外挿



# 帰納法の弱点2：因果と相関



# 相関と因果



# 帰納法の弱点3: 偽陽性と偽陰性のバランス

画像誤認識

Google の痛い失敗  
2015年 Googleのタグ付け性能

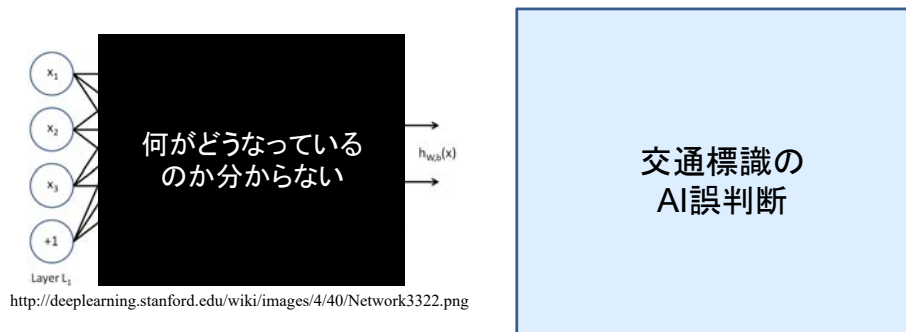
ロンドン警視庁サイバー班の悩み  
ソース 2017/12/15 ネットニュース  
<http://news.livedoor.com/article/detail/14056063/>

テスラ 半自動運転モードで追突  
2018年1月22日(米国時間)、  
ロサンゼルス郡の高速道路405号線

医療診断: 偽陰性を避けるため、偽陽性を甘受  
自動運転: 偽陽性や誤検知を避けるため、情報をあえて無視する制御

データのみからルールを定める帰納法の限界  
事例ベース、(セミ)ブラックボックス

# DNNはブラックボックス → グレイボックスへ



<http://deeplearning.stanford.edu/wiki/images/4/40/Network3322.png>



- ✓ 結果をみても何が特徴なのかわかりづらい。そもそも人海戦術。
- ✓ 途中の計算がうまくいっているのかどうか分からない。

Figure 2. Outline of the DeepFace architecture. The input image is first processed by a convolutional layer, followed by three locally-connected layers and two fully-connected layers. Colors illustrate outputs for each layer. The net includes more than 120 million parameters, where more than 95% come from the local and fully connected layers.

# グレイボックス

要因を読み解く必要性は何か？

- 可読性が大切 → 何が要因で、どんなメカニズム？
- 個別化(ニーズや利用状況にあわせること)が容易

■ブラックボックスでは機能しない

# Fairness

## Bias as a technical matter

- ✓ Interpretability
- ✓ Accountability
- ✓ Reproducibility
- ✓ Transparency

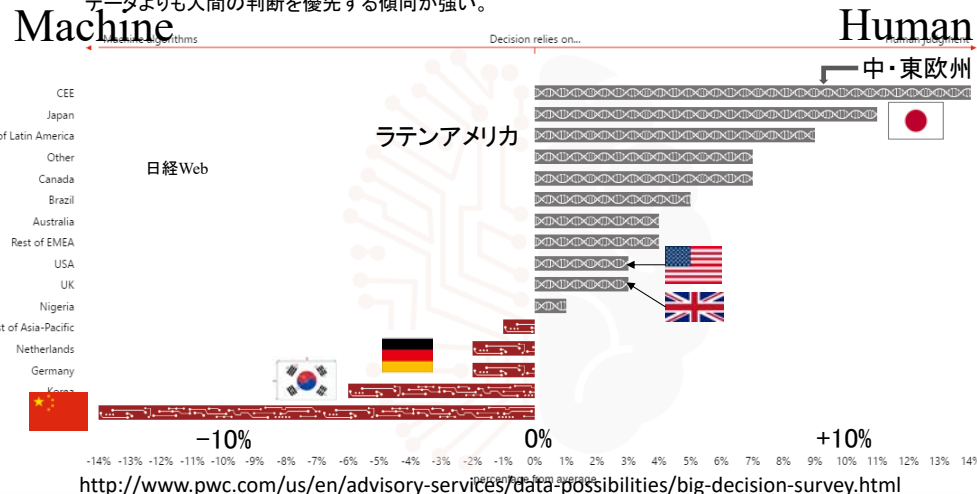
### Foundation of Statistics

- Data imbalance
- Sampling strategy
- Generative model

25日に施行  
「データ保護規則(GDPR)」

# 志向性の格差

- A good mix augments human judgment with machine algorithms to create better outcomes.
- What type of analysis will inform your next big decision?
- データの利活用が最も進んでいるUSでは、経営層は最終的には人間の判断を優先する傾向がある。我が国はデータ利活用人材の教育や人数、雇用されている企業規模が小さいにもかかわらず、データよりも人間の判断を優先する傾向が強い。



<http://www.pwc.com/us/en/advisory-services/data-possibilities/big-decision-survey.html>

# アウトライン

1. 平成の30年間
2. 統計学、機械学習、深層学習
3. 帰納法の弱点
4. 人材育成

# 激しい生き残り競争



## US jobs rankings 2016: actuaries drop to 10th

Latest annual Jobs Related report based the rankings on four core criteria: **environment, income, outlook and stress**

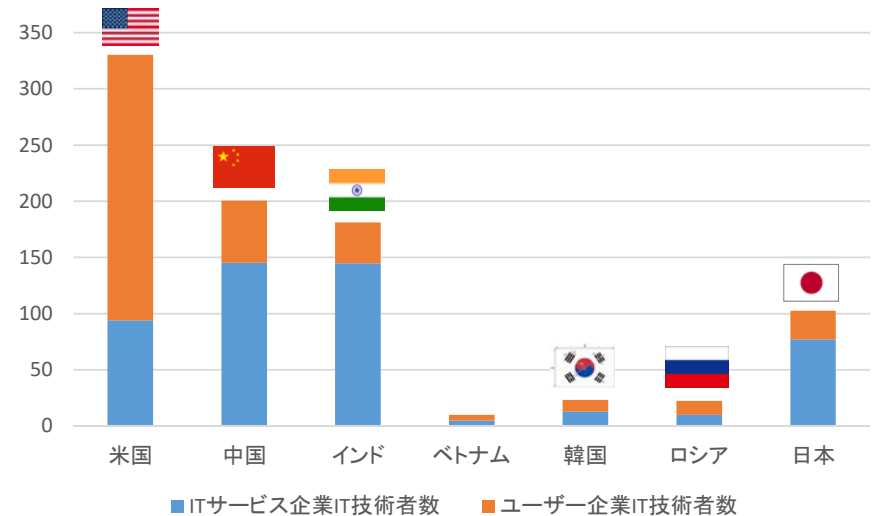
Top 10 best jobs and how they fared in 2015 in brackets, with annual median salary

1. Data scientist (+5) – \$128,240
2. Statistician (+2) – \$79,990
3. Information security analyst (N/A) – \$88,890
4. Audiologist (-2) – \$73,060
5. Diagnostic medical sonographer (N/A) – \$62,540
6. Mathematician (-3) – \$103,720
7. Software engineer (+1) – \$97,990
8. Computer systems analyst (+1) – \$82,710
9. Speech pathologist (+2) – \$71,550
10. Actuary (-9) – \$96,700

- See more at: <http://www.theactuary.com/news/2016/04/us-jobs-rankings-2016-actuaries-drop-to-10th-place/#sthash.bhVnrj14.dpuf>

## 人材数の格差:ベンダー or ユーザー企業

ITエンジニア数 (万人)



IPA (情報処理推進機構) 「グローバル化を支えるIT人材確保・育成施策に関する調査」

(統数研・神谷特任准教授のスライド(2017)から改編)

## データは副産物。そして抽象物。

Feature Vector (Descriptor)の選択  
End-to-end Learning

Data cleansing, Data Editing, Data Curating

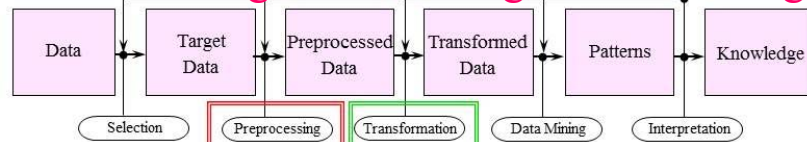
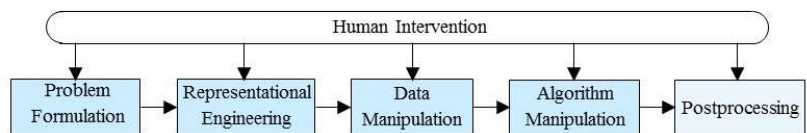


図 1: Fayyad等による知識発見のプロセス



データ処理の流れ 図 2: Langleyによる知識発見のプロセス

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