The background image shows a coastal landscape with snow-capped mountains, several wind turbines on a rocky outcrop, a large white ship in the water, and an offshore oil platform. The sky is blue with some clouds and a small airplane flying. The overall scene is a mix of natural and industrial elements.

# WHO IS WHO? *DIGITALIZATION, MACHINE LEARNING, DIGITAL TWIN, ETC.* KORT OM MODELLTYPER, METODER OG BEGREPER

Thomas Welte, SINTEF Energi

Digitalisering i vattenkraften, 9. mai 2019, Arlanda

# Digitalization



1

## To digitize

The process of converting analogue to digital data



2

## Digitalization

The process when an organization digitizes central parts of its business



3

## Digital transformation

The effect of digitalization on an organization. New value chains and services.



## Bild 2

---

- LB1**      **Replace shutterstuck picture**  
Lukas Bach; 2018-09-16
- LB2**      **Hvad er historien her?**  
Lukas Bach; 2018-09-16
- AMR1**     **nye forretningsmodeller, papiravis til nettavis, rubrikka nnonser til finn.no**  
Anne Marthine Rustad; 2018-09-16

# Digitalization

---

There is no digitalization and no digital transformation without digitization (of paper and processes)

Digitalization leads to digital business, digital transformation requires digital business and digitization

<https://www.i-scoop.eu/digitization-digitalization-digital-transformation-disruption/>

- Digitization

- "Digitization is creating a digital (bits and bytes) version of analog/physical things such as paper documents, microfilm images, photographs, sounds and more. So, it's simply converting and/or representing something non-digital (other examples include signals, health records, location data, identity cards, etc.) into a digital format which then can be used by a computing system for numerous possible reasons."

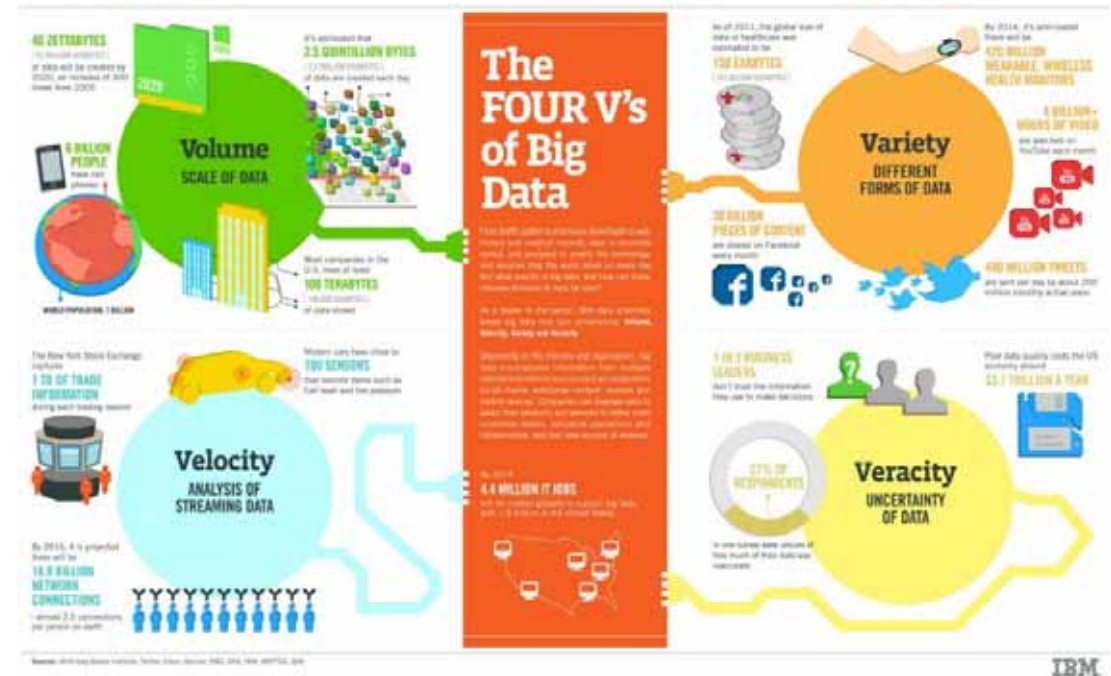
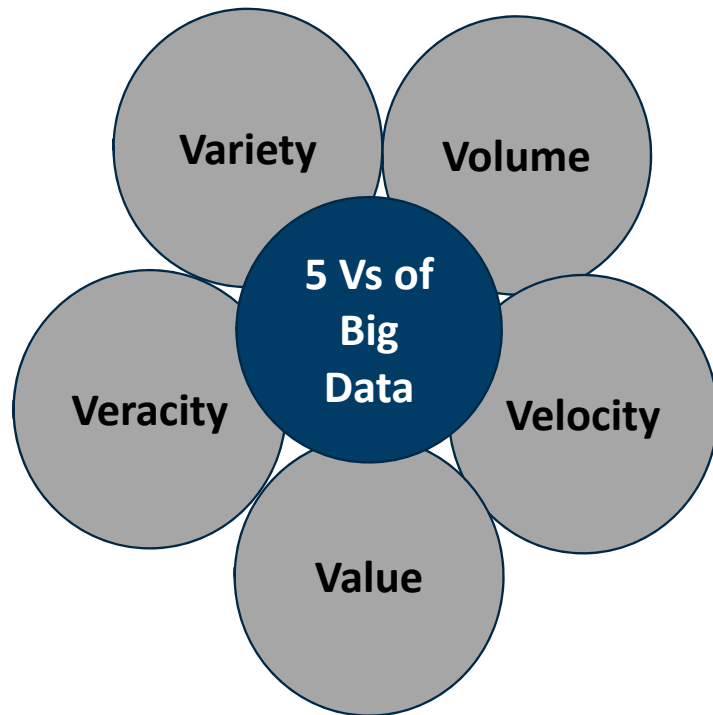
- Digitalization

- "Enabling, improving and/or transforming business operations and/or business functions and/or business models/processes and/or activities, by leveraging digital technologies and a broader use and context of digitized data, turned into actionable, knowledge, with a specific benefit in mind."

- Digital transformation

- "Digital transformation [...] is broader than digitalization as a way to move to digital business."

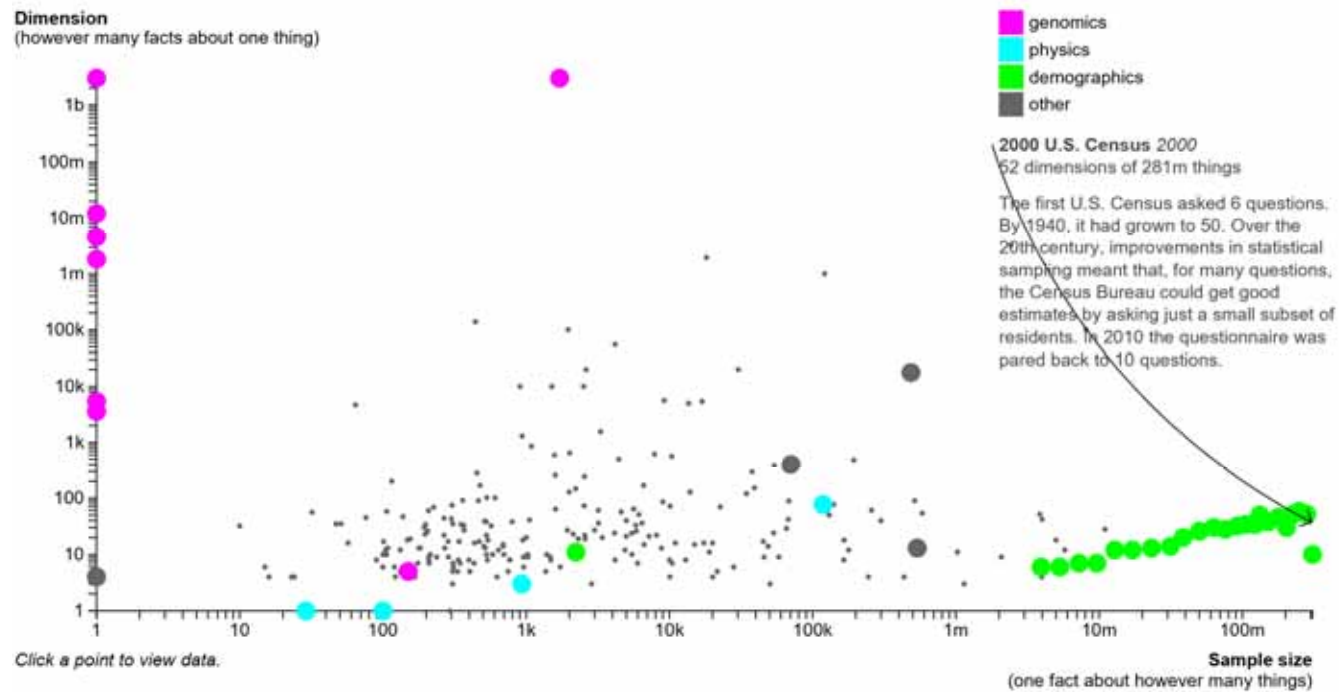
# Big data – the 5 "V"s



- **Volume:** the vast amount of data
- **Velocity:** the speed at which new data is generated and the speed at which data moves around
- **Value:** the ability to use the data to generate value
- **Veracity:** the messiness or trustworthiness of the data
- **Variety:** the different types of data we can now use

[Toph Tucker, "Interactive graphic: How big is big data?", <https://www.bloomberg.com/graphics/2014-fix-this-big-data/>, 2014]

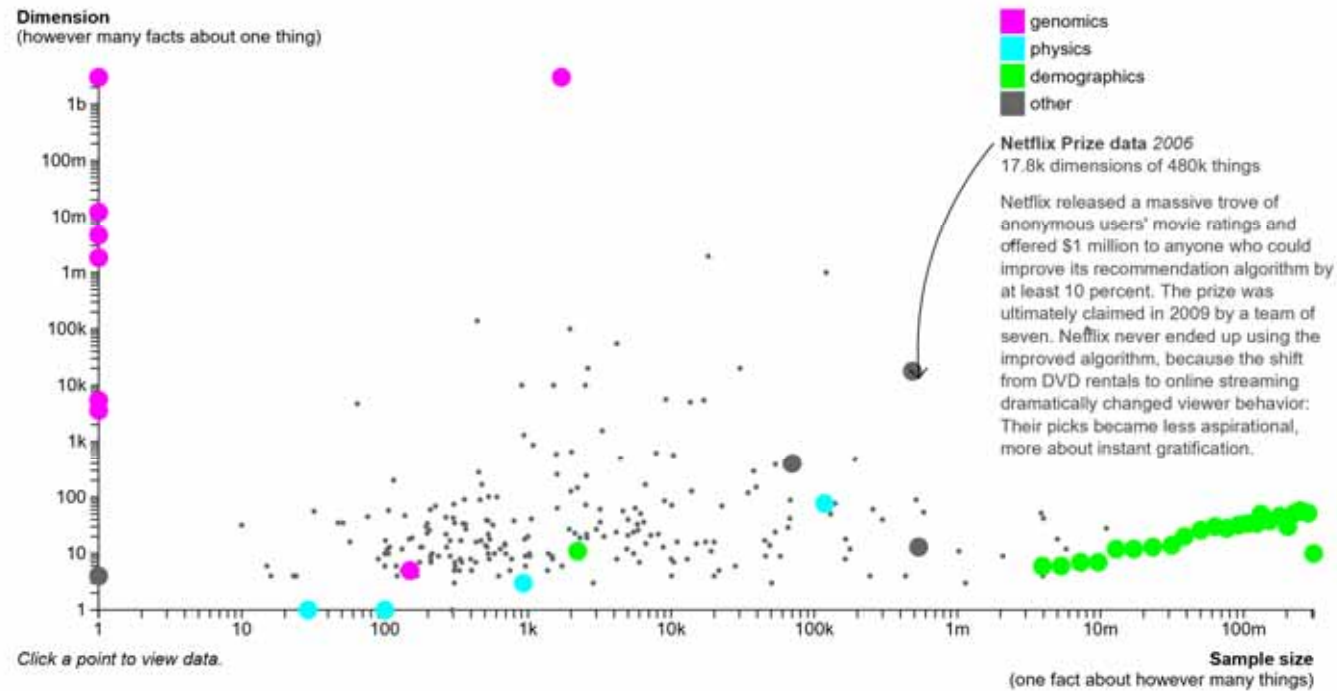
# Big Data





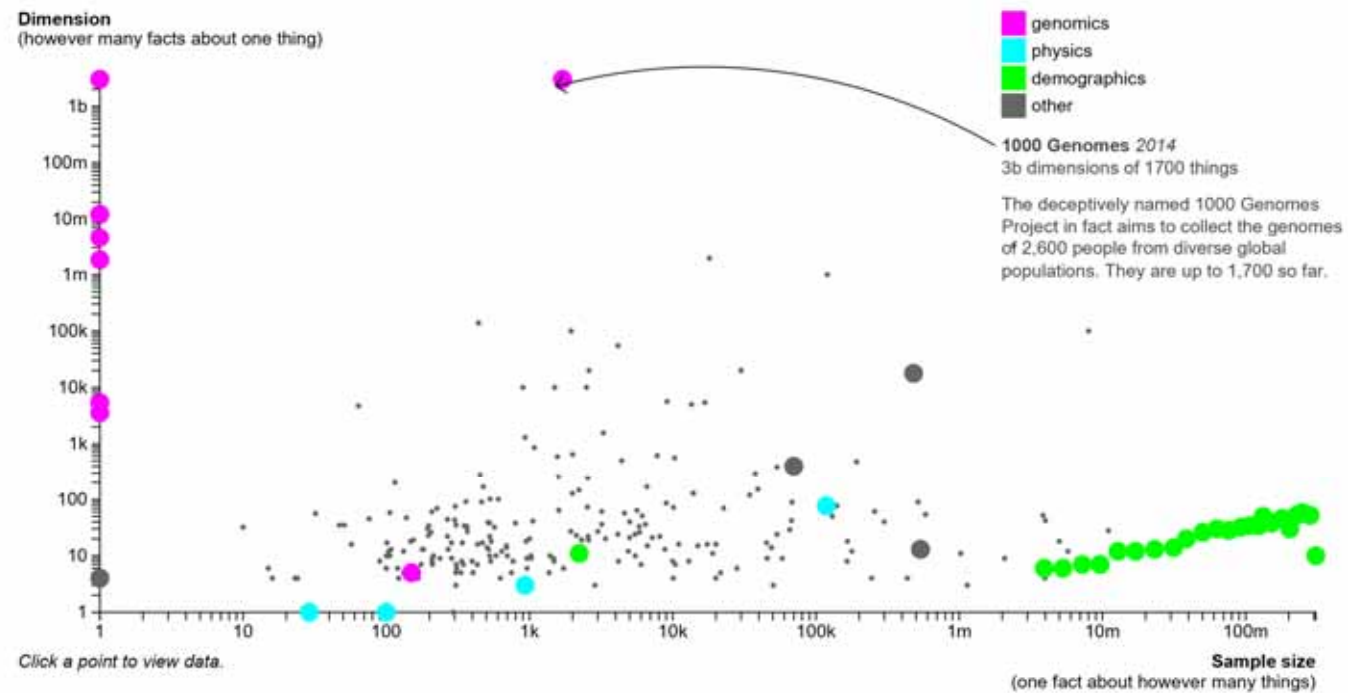
[Toph Tucker, "Interactive graphic: How big is big data?", <https://www.bloomberg.com/graphics/2014-fix-this-big-data/>, 2014]

# Big Data



[Toph Tucker, "Interactive graphic: How big is big data?", <https://www.bloomberg.com/graphics/2014-fix-this-big-data/>, 2014]

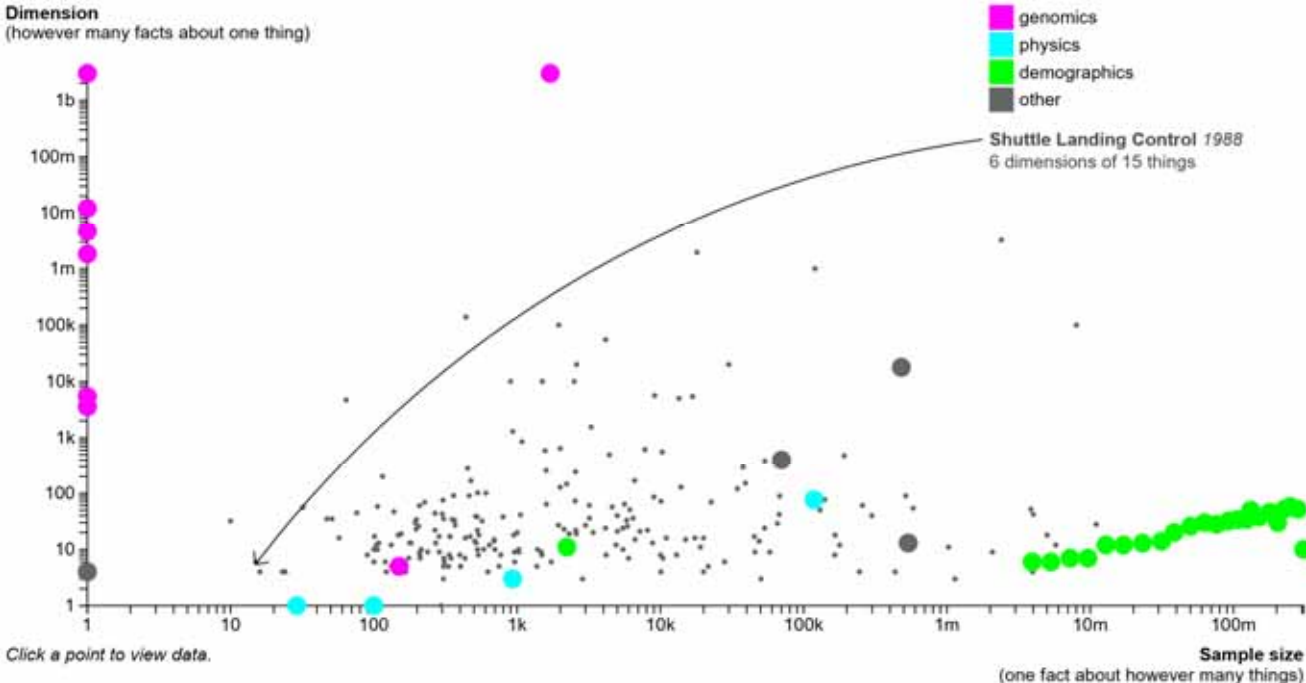
# Big Data





[Toph Tucker, "Interactive graphic: How big is big data?", <https://www.bloomberg.com/graphics/2014-fix-this-big-data/>, 2014]

# Big Data



# CPS, IoT, IoS

---

- Cyber-physical system (CPS)
  - "Cyber-Physical Systems (CPS) comprise interacting digital, analog, physical, and human components engineered for function through integrated physics and logic. [...]" [National Institute of Standards and Technology (NIST), U.S. Department of Commerce]
  - "Cyber-Physical Systems (CPS) are **integrations of computation, networking, and physical processes.**" [<https://ptolemy.berkeley.edu/projects/cps/>]
- Internet of things (IoT)
  - "Internet of things (IoT) is the **extension of Internet connectivity into physical devices and everyday objects.**" [[https://en.wikipedia.org/wiki/Internet\\_of\\_things](https://en.wikipedia.org/wiki/Internet_of_things)]
- Internet of Services (IoS)
  - Access to services available on the web

**Soon, IoT will transition into Internet of Service**



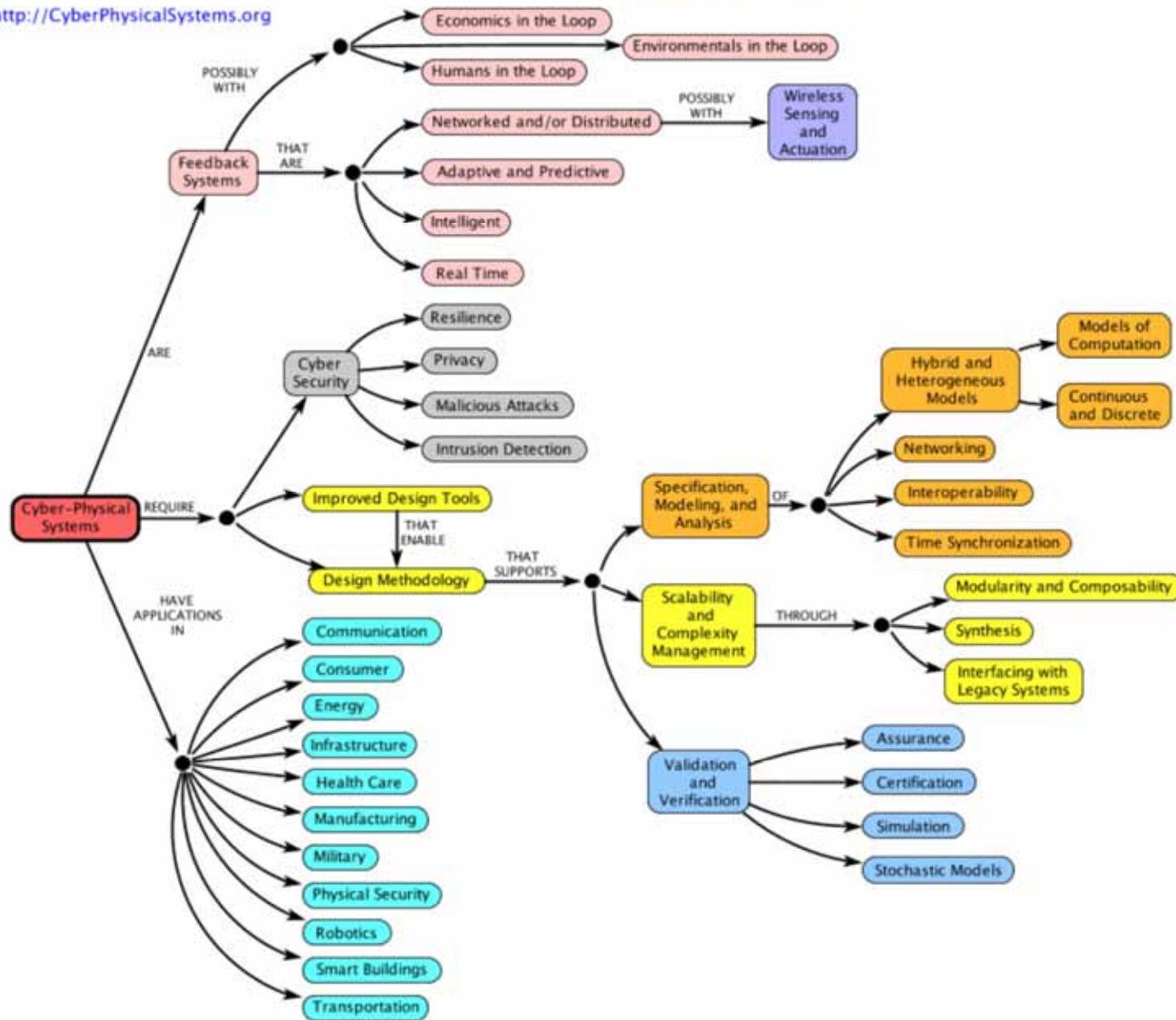
S S Iyengar and Jerry Miller, DEC 20 2016, 01:02AM IST | UPDATED: DEC 20 2016, 11:53AM IST

# Cyber-Physical Systems - a Concept Map

See authors and contributors.

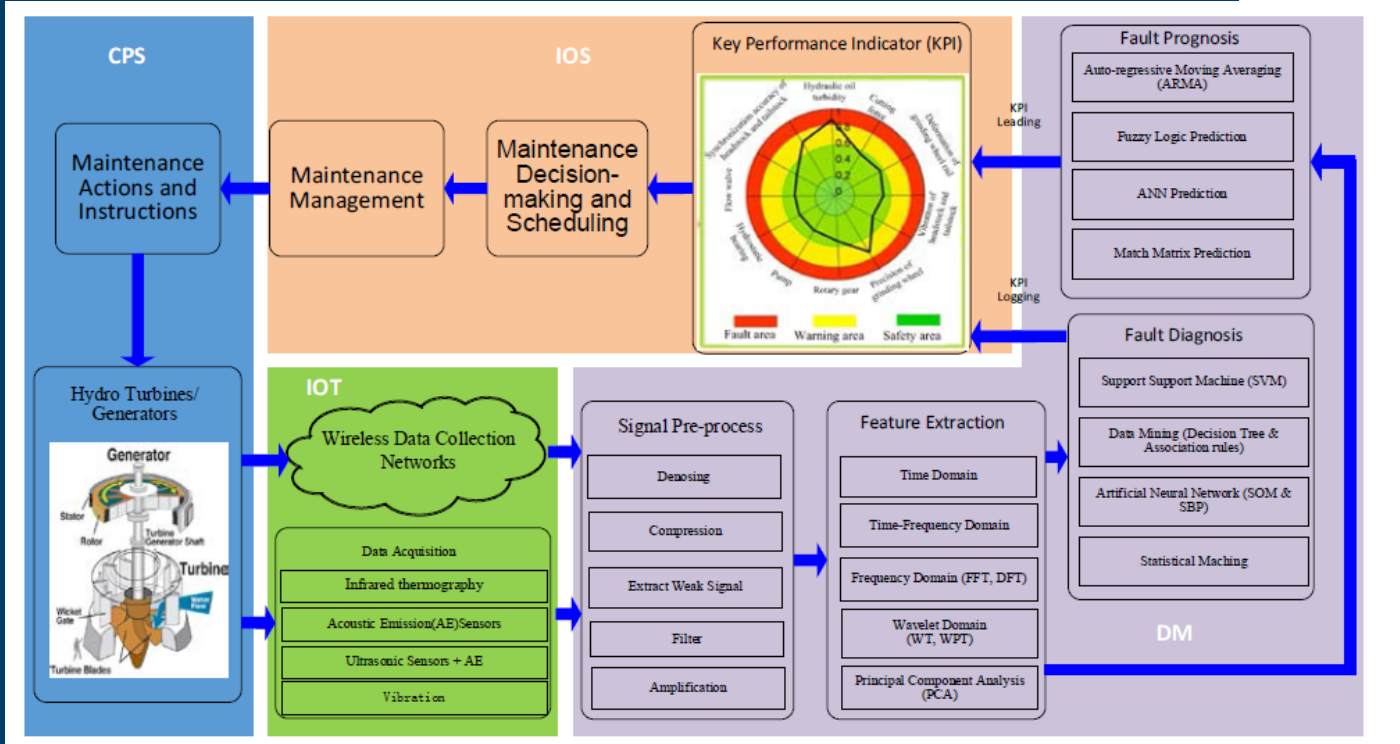
<http://CyberPhysicalSystems.org>

<http://cyberphysicalsystems.org>



# Framework for data-driven predictive maintenance

Where can concepts like cyber-physical systems (CPS), internet of things (IOT), data mining (DM) and internet of services (IOS) be used for predictive maintenance of hydropower plants?



# Data mining

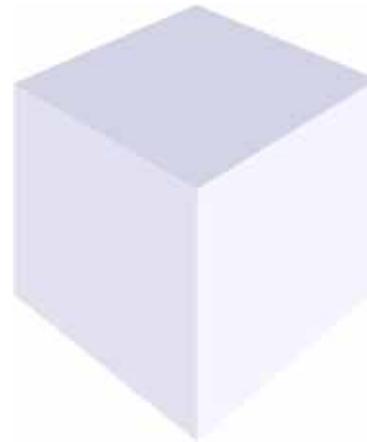
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- Data mining
  - "Process of discovering patterns in large data sets"  
[[https://en.wikipedia.org/wiki/Data\\_mining](https://en.wikipedia.org/wiki/Data_mining)]
  - Analyse large amount of data and generate new information

# Type of models (model classes)

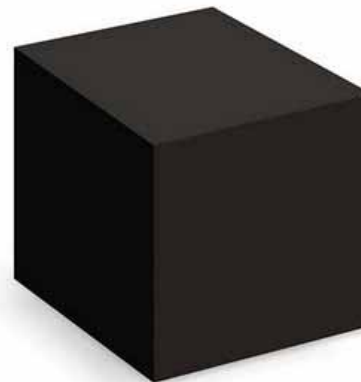
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- Physical models (PM)



$$Y = f(X1, X2) = X1 + X2$$

- Stochastic models (SM)

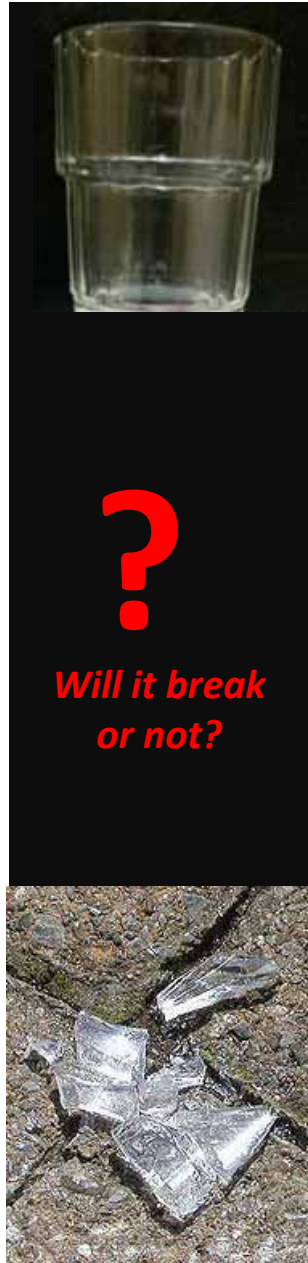


$$Y = f(X1, X2) = ?$$

- Machine learning (ML)

# Model classes – Main differences

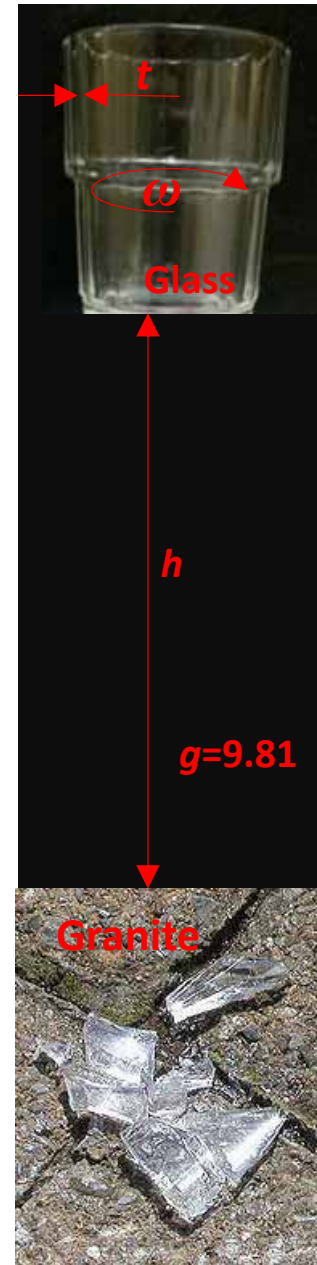
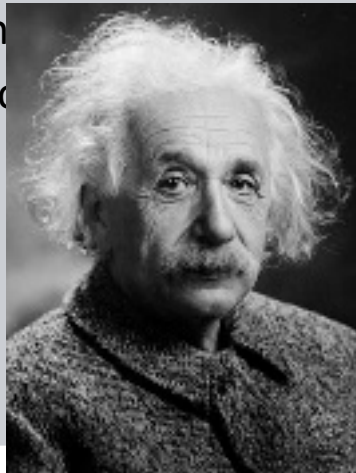
PM	SL	ML
<ul style="list-style-type: none"><li>+ White-box model</li><li>+ Clear meaning of model parameters</li><li>- Problem-specific model</li><li>- Challenging if good model is not available</li></ul>	<ul style="list-style-type: none"><li>+ General model</li><li>+ Takes uncertainty into account</li><li>+ Requires a group consisting of comparable items</li><li>- Parameter estimation requires observations related to lifetime/reliability</li></ul>	<ul style="list-style-type: none"><li>+ General model</li><li>+ Large data sets</li><li>+ Short-term predictions</li><li>+ Identification of faults</li><li>+ Fault prediction and diagnosis</li><li>- Often black-box model</li><li>- Requirements of quality data</li><li>- Few examples on lifetime prediction and reliability estimation</li></ul>





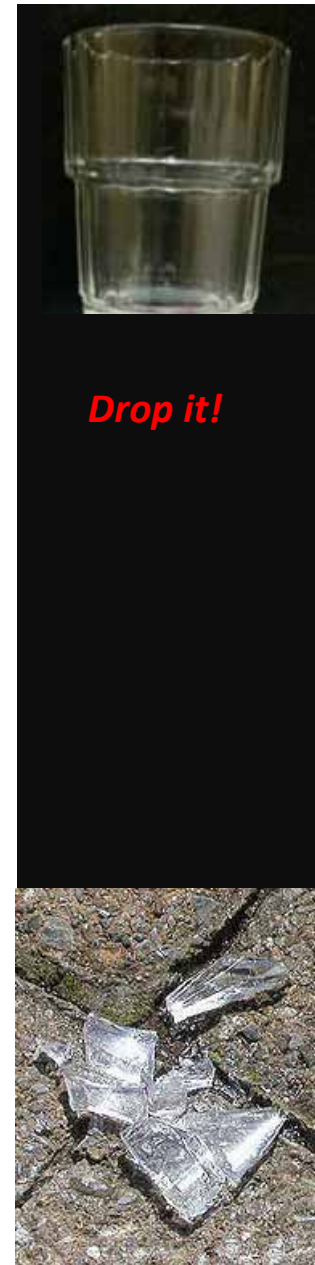
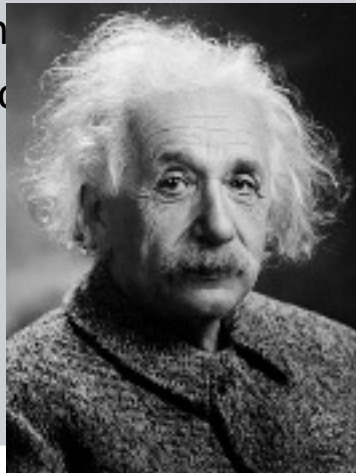
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PM	SL	ML
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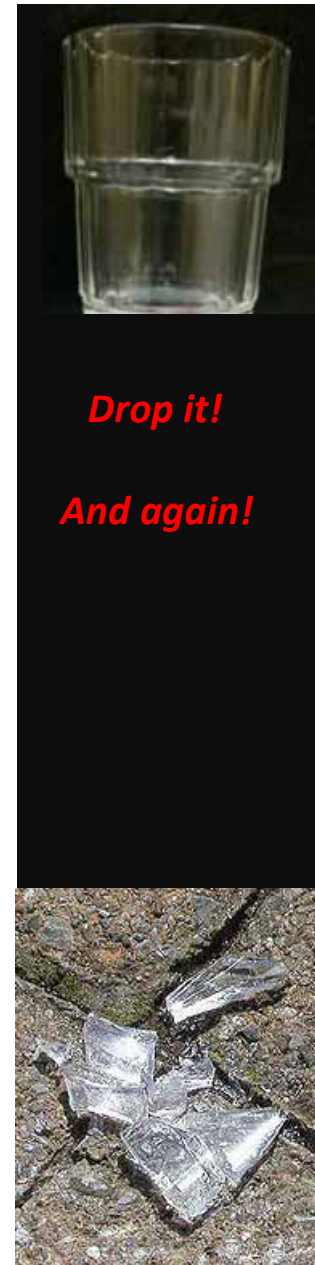
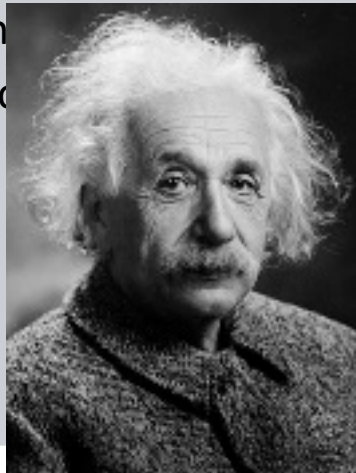
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# Model classes – Main differences

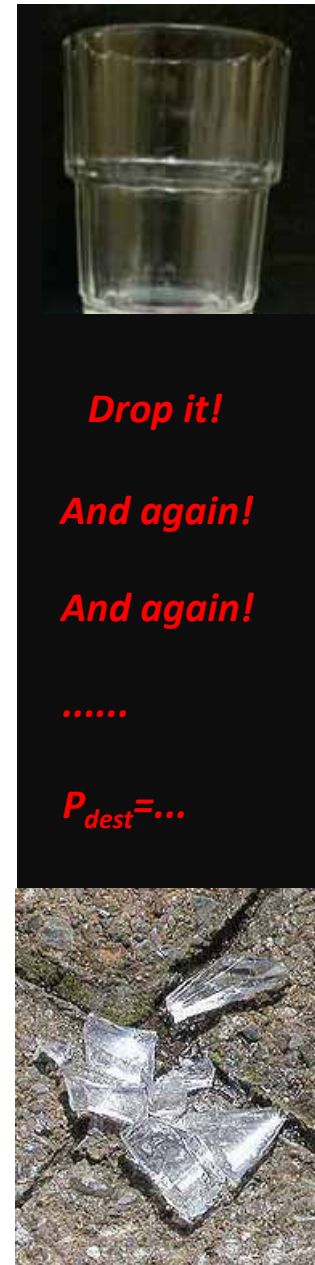
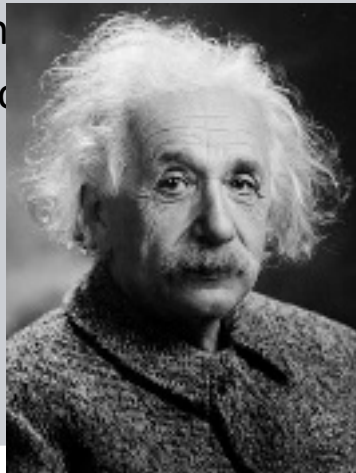
PM	SL	ML
<ul style="list-style-type: none"> <li>+ White-box model</li> <li>+ Clear meaning of model parameters</li> <li>- Problem-specific model</li> <li>- Challenging to build</li> </ul>	<ul style="list-style-type: none"> <li>+ General model</li> <li>+ Takes uncertainty into account</li> <li>+ Requires a group consisting of considerable experts</li> <li>- Parameters are often hard to estimate</li> <li>- Few examples on lifetime prediction and reliability estimation</li> </ul>	<ul style="list-style-type: none"> <li>+ General model</li> <li>+ Large data sets</li> <li>+ Short-term predictions</li> <li>+ Identification of faults</li> <li>+ Fault prediction and diagnosis</li> <li>- Often black-box model</li> <li>- Requirements of quality data</li> <li>- Few examples on lifetime prediction and reliability estimation</li> </ul>

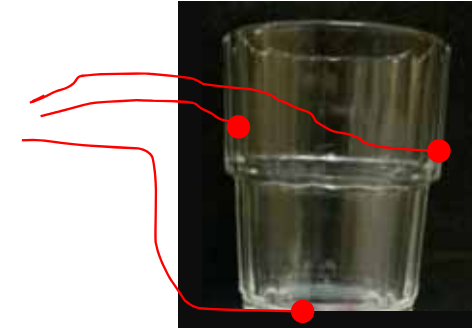


*Drop it!*  
*And again!*

# Model classes – Main differences

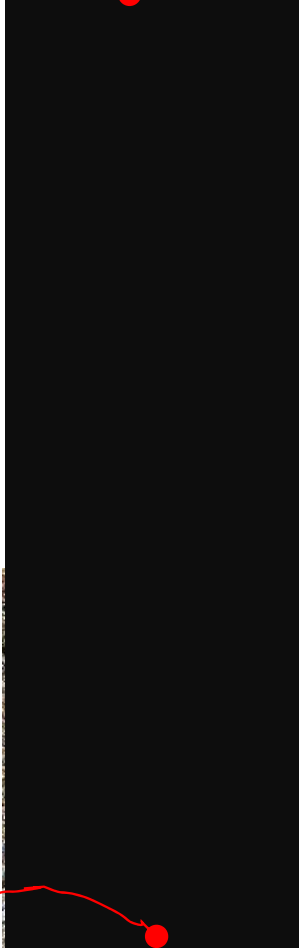
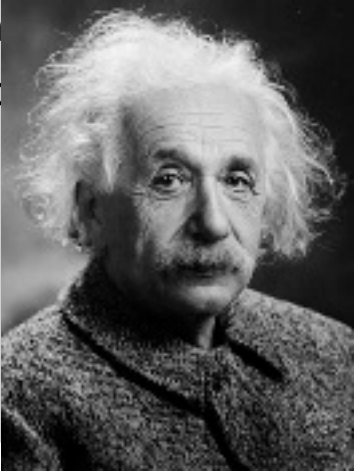
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# Model classes – Main differences

PM	SL	ML
<ul style="list-style-type: none"><li>+ White-box model</li><li>+ Clear meaning of model parameters</li><li>- Problem-specific model</li><li>- Challenging to build</li></ul>	<ul style="list-style-type: none"><li>+ General model</li><li>+ Takes uncertainty into account</li><li>+ Requires a group consisting of considerable experts</li><li>- Parameter estimation is difficult</li><li>- Requires a lot of data</li></ul>	<ul style="list-style-type: none"><li>+ General model</li><li>+ Large data sets</li><li>+ Short-term predictions</li><li>+ Identification of faults</li><li>+ Fault diagnosis</li><li>- Often requires a lot of data</li><li>- Requires a lot of data</li><li>- Few experts</li><li>- Prediction accuracy is often low</li></ul>

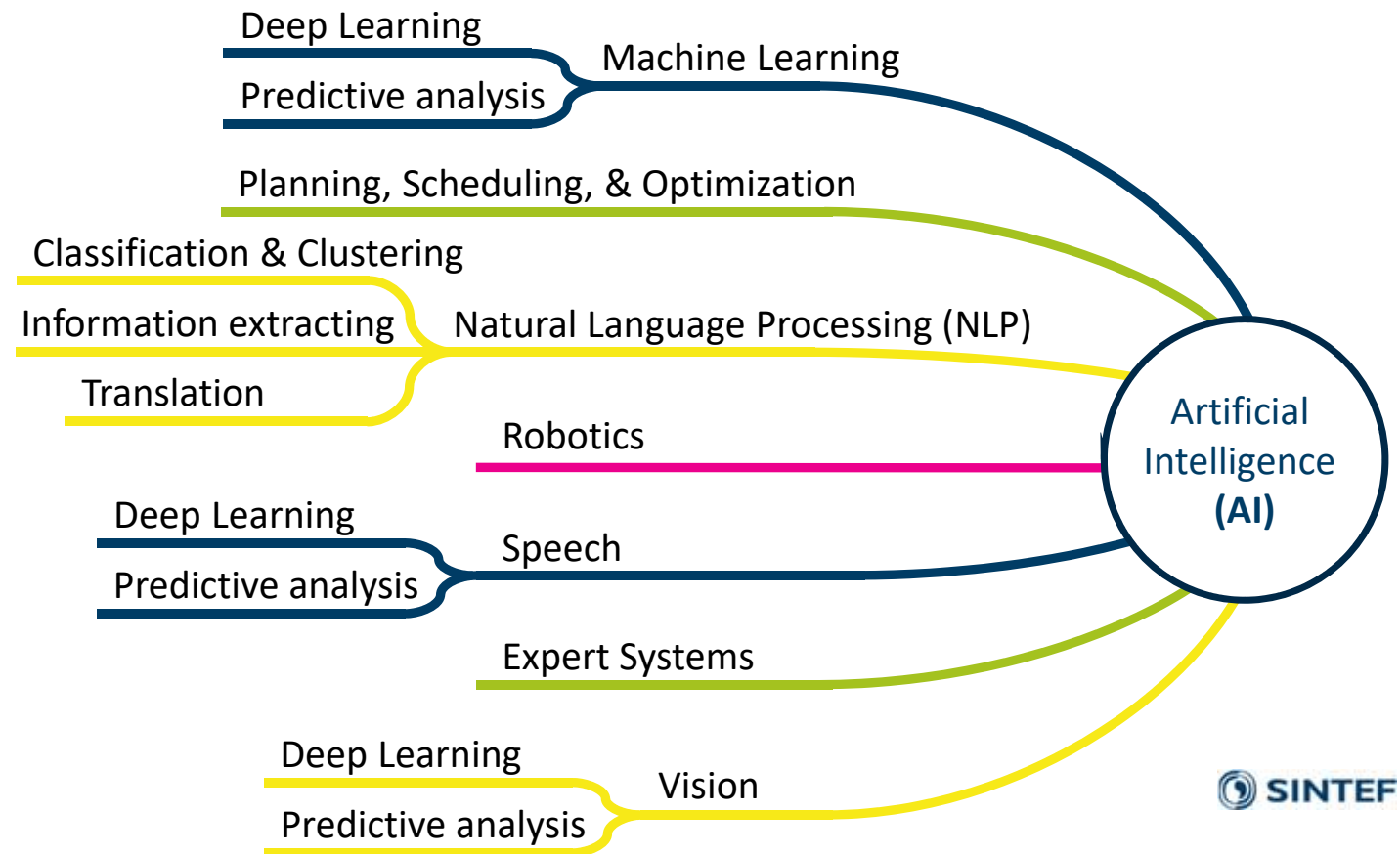




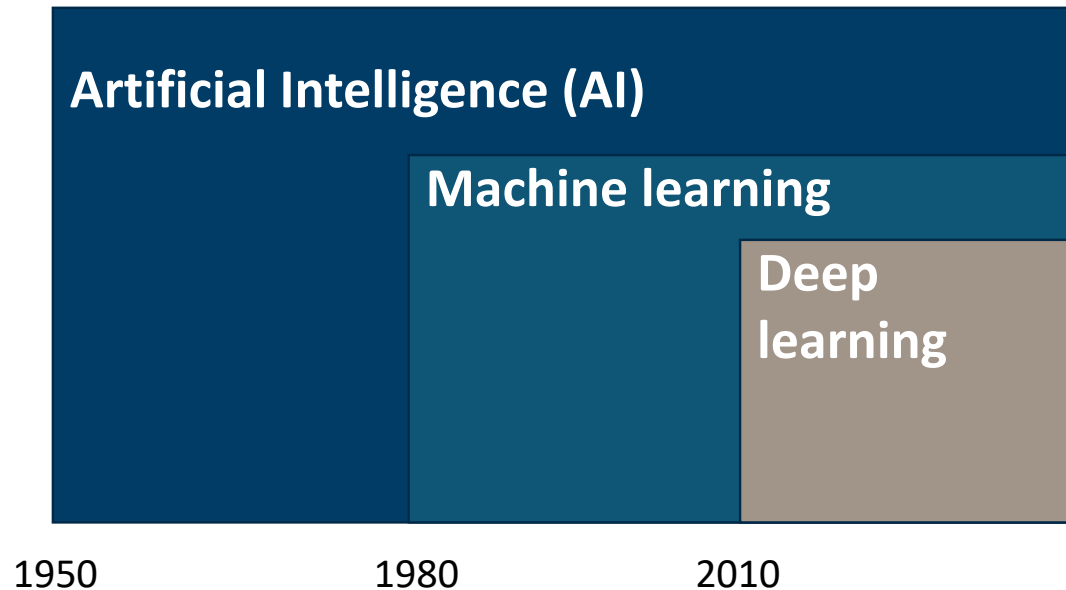
# AI is a crossroad of multiple fields..

# .. and there are several types of AI

- Computer science
- Mathematics
  - Logics
  - Optimisation
  - Analysis
  - Probabilities
  - Linear algebra
- Cognitive sciences
- ....
- Domain expertise



# AI - Machine learning – Deep learning



## Artificial Intelligence (AI)

- A program that can sense, decide, act and adapt

## Machine Learning

- A way of achieving AI
- Algorithms whose performance improve as they are exposed to more data

## Deep Learning

- Subset of machine learning
- Multi-layered neural networks learn from vast amounts of data

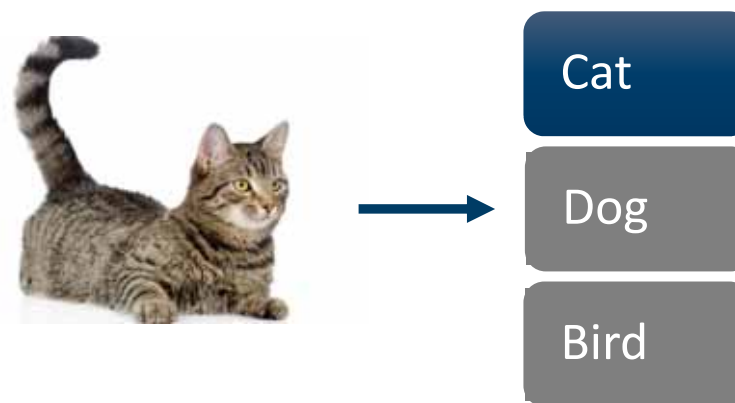


# Machine learning

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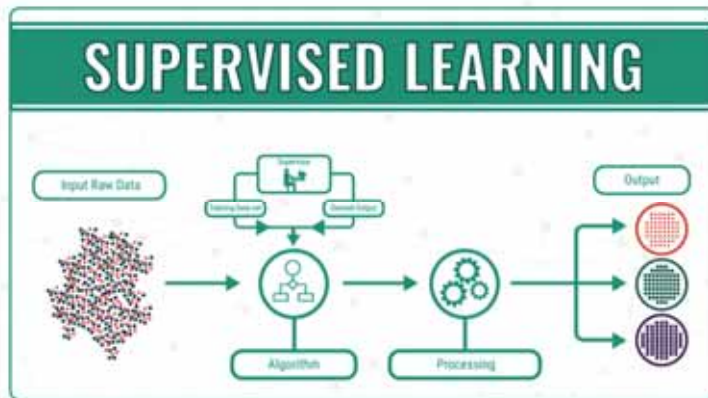
All machine learning have a two step approach:

- 1. Learning phase:** Using input data to identify parameters best describing task at hand
- 2. Inference phase:** Take learned parameters as input to perform task

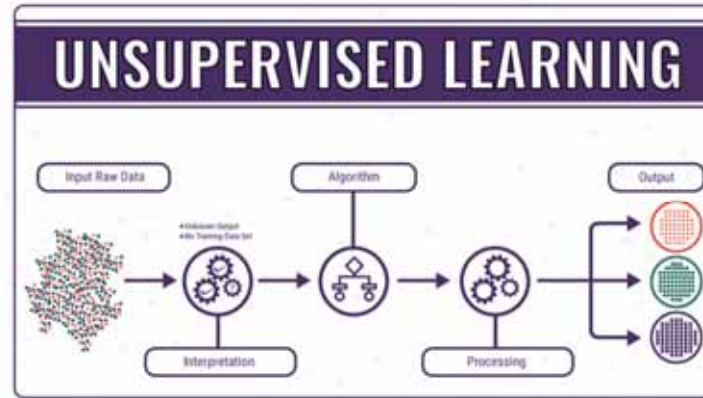


# Main types of machine learning

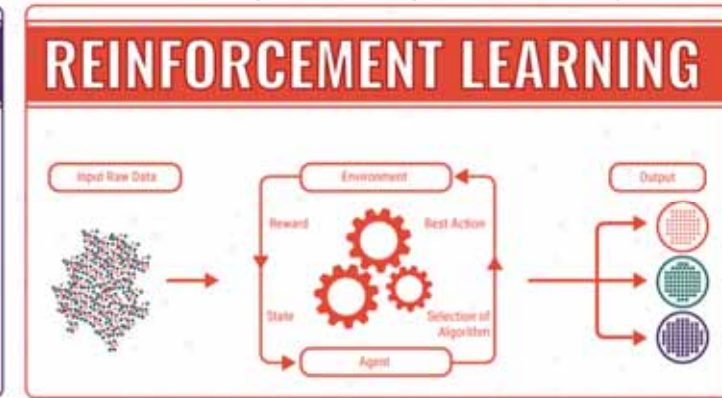
Image credit: bigdata-madesimple.com



- Trained to learn mapping from input to output
- Tagged training set



- Only given input
- No "correct" answer provided
- No tagged data



- Learning through trial and error with rewards and punishments
- Hybrid between supervised and unsupervised

## Bild 23

---

SR2

I'm thinking something along these lines. Reinforcement learning can then be illustrated similarly to unsupervised, but with a reward whenever the ducks ends up in the same category, and e.g. get ducklings (delayed reward). Health example could be e.g. tumor/no tumor classification (supervised), tumor type clustering (unsupervised), and how will the tumor react to treatment? (Re-inforcement learning, delayed reward)

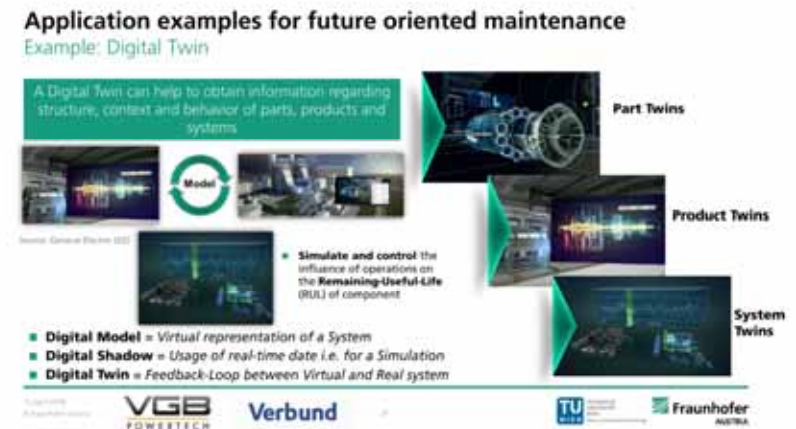
Signe Riemer-Sørensen; 2018-09-21

# Digital twin

- Digital twin
  - "a digital replica of a living or non-living physical entity" [[https://en.wikipedia.org/wiki/Digital\\_twin](https://en.wikipedia.org/wiki/Digital_twin)]
  - Properties:
    - "Connection between the physical model and the corresponding virtual model or virtual"
    - "connection is established by generating real time data using sensors"

- Digital model: "Virtual representation of a system"
- Digital shadow: "Usage of real-time data"
- Digital twin: "Feedback-loop between virtual and real system"

[R. Glawar: "Maintenance trends and tools in the age of digitalization",  
VGB Workshop *Digitalization in Hydropower*, 12 April 2018, Vienna]



# Reality: mixed/augmented, virtual and simulated

- Mixed and augmented: Combination of computer generated and true reality
- Virtual: Computer generated
- Simulated: Computer generated/virtual reality indistinguishable from "true" reality

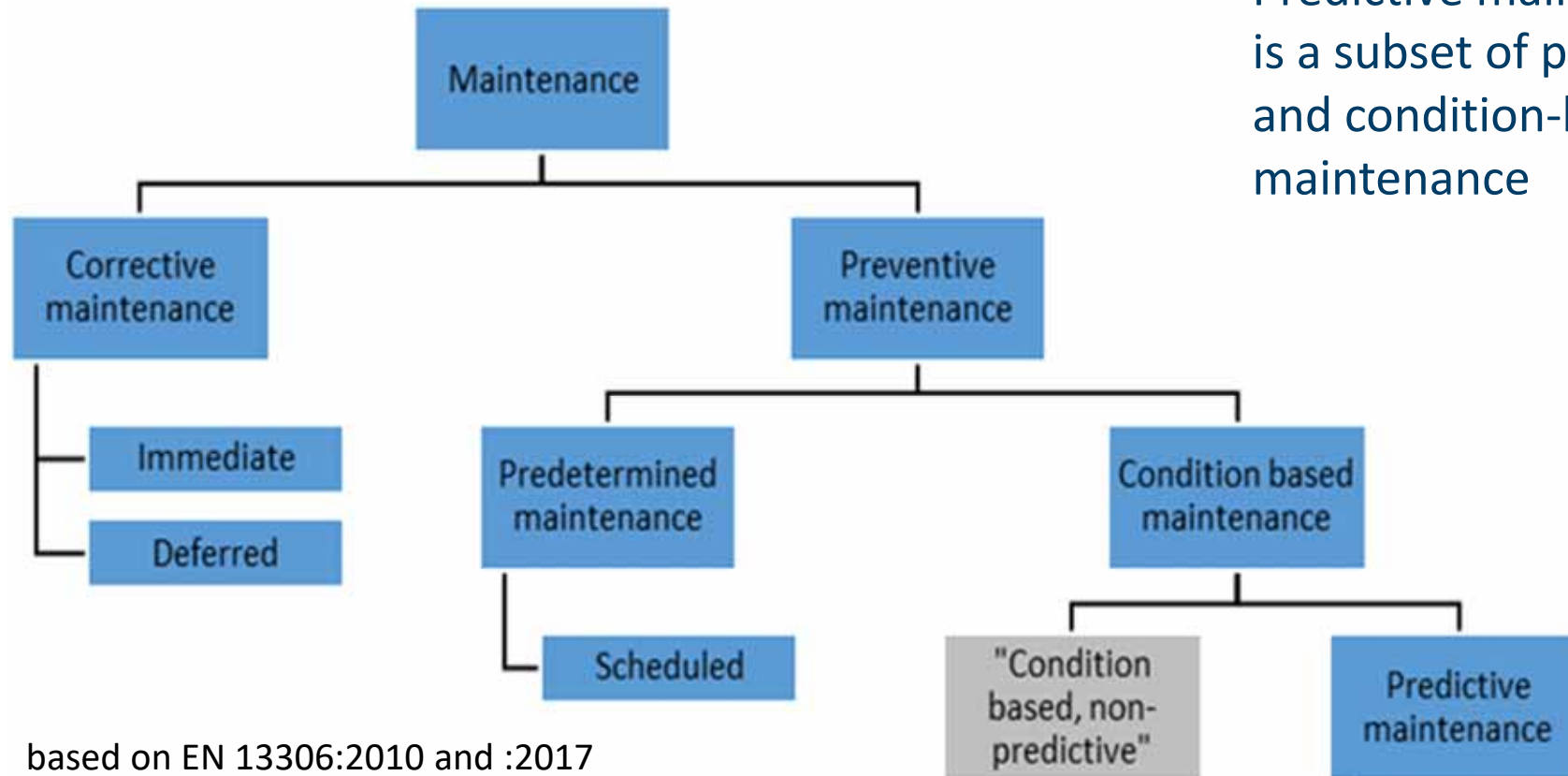


[https://sv.wikipedia.org/wiki/F%C3%B6rst%C3%A4rkt\\_verkighet#/media/File:Navit\\_Reality\\_View\\_next\\_to\\_reality.jpg](https://sv.wikipedia.org/wiki/F%C3%B6rst%C3%A4rkt_verkighet#/media/File:Navit_Reality_View_next_to_reality.jpg)



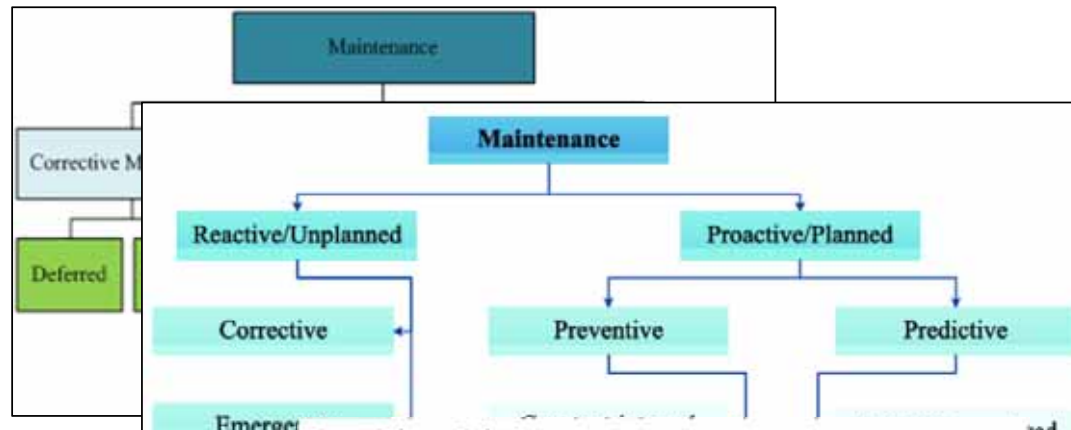
[https://en.wikipedia.org/wiki/Virtual\\_reality#/media/File:Reality\\_check\\_ESA384313.jpg](https://en.wikipedia.org/wiki/Virtual_reality#/media/File:Reality_check_ESA384313.jpg)

# Types of maintenance



Predictive maintenance is a subset of preventive and condition-based maintenance

# Types of maintenance – Other definitions



Wang, K. S., 2018  
backlash error  
ANNs." Advan

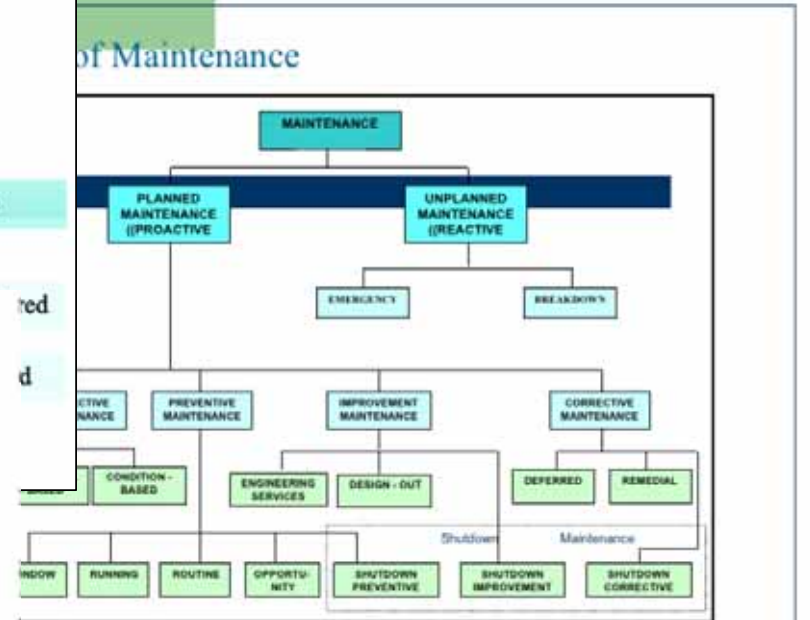
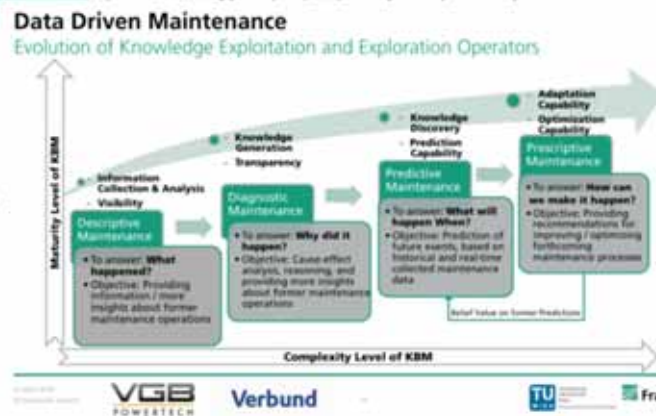


Figure 2.4 Maintenance Types





Teknologi for et bedre samfunn