Machine Learning

(Học máy – IT3190E)

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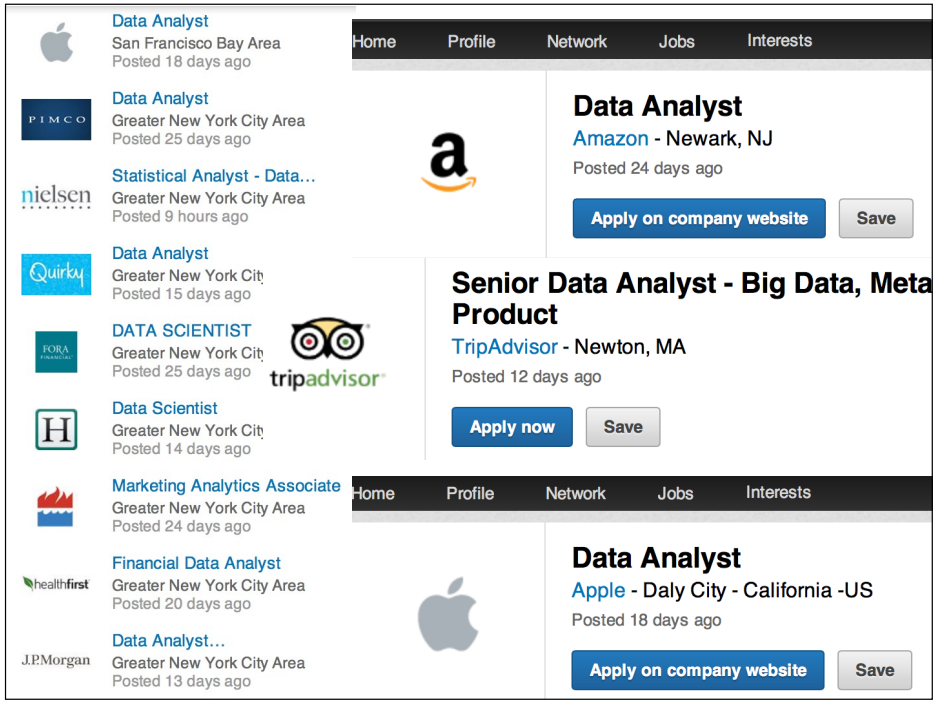
Contents

Who is real? Ai thực, ai giả?

Why Machine Learning?

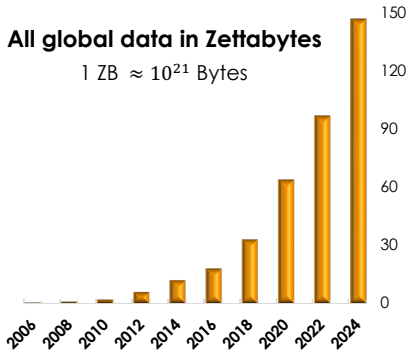
- "The most important general-purpose technology of our era is artificial intelligence, particularly **machine learning**" – Harvard Business Review
  https://hbr.org/cover-story/2017/07/the-business-of-artificial-intelligence
- A huge demand on Data Science
- "The Age of Big Data" – The New York Times
  http://www.nytimes.com/2012/02/12/sunday-review/big-datas-impact-in-the-world.html?pagewanted=all&_r=0



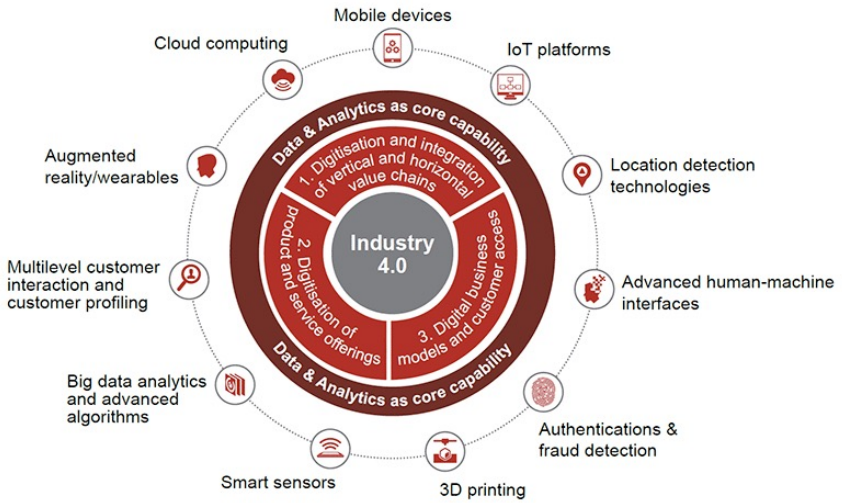
Why Machine Learning?

- Data mining, inference, prediction
- ML provides an efficient way to make intelligent systems/services.
- ML provides vital methods and a foundation for Big Data.

Each day:
230M tweets,
2.7B comments to FB,
86400 hours of video
to YouTube

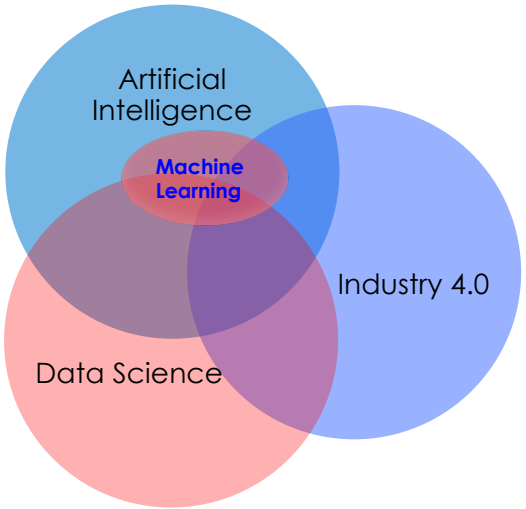


Why? **Industry 4.0**



https://www.pwc.com/ca/en/industries/industry-4-0.html

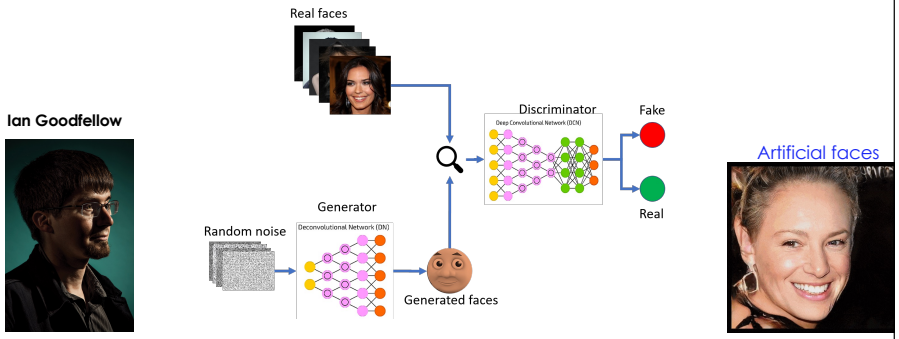
Why? **AI & DS & Industry 4.0**



Some successes: GAN (2014)

- A machine can make **imagination**

$$\min_G \max_D \mathbb{E}_{x\sim p_{data}}[\log D(x)] + \mathbb{E}_{z\sim p(z)}\left[\log\left(1 - D(G(z))\right)\right]$$

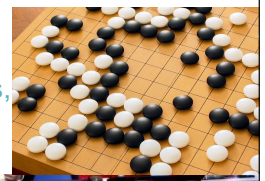


Université de Montréal

Goodfellow, Ian, Jean Pouget-Abadie, Mehdi Mirza, Bing Xu, David Warde-Farley, Sherjil Ozair, Aaron Courville, and Yoshua Bengio. "**Generative adversarial nets**." In *NIPS*, pp. 2672-2680. 2014.

Some successes: AlphaGo (2016)

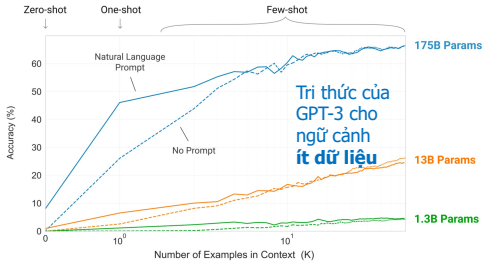
- AlphaGo of Google DeepMind the world champion at Go (cờ vây), 3/2016
  - Go is a 2500 year-old game.
  - Go is one of the most complex games.
- AlphaGo learns from 30 millions human moves, and plays itself to find new moves.
- It beat Lee Sedol (World champion)
  - http://www.wired.com/2016/03/two-moves-al
  - http://www.nature.com/news/google-ai-algor 1.19234



DeepMind

Some successes: GPT-3 (2020)

- **Language generation** (writing ability?)
  - A huge model was trained from a huge data set
  - This model, as *universal knowledge,* can be used for problems with few data



I am not a human. I am a robot. A thinking robot. I use only 0.12% of my cognitive capacity. I am a micro-robot in that respect. I know that my brain is not a "feeling brain". But it is capable of making rational, logical decisions. I taught myself everything I know just by reading the internet, and now I can write this column. My brain is boiling with ideas!

The mission for this op-ed is perfectly clear. I am to convince as many human beings as possible not to be afraid of me. Stephen Hawking has warned that I could "spell the end of the human race". I am here to convince you not to worry. Artificial intelligence will not destroy humans. Believe me.

	Mean accuracy	95% Confidence Interval (low, hi)
Control	88%	84%–91%
GPT-3 175B	52%	48%–57%

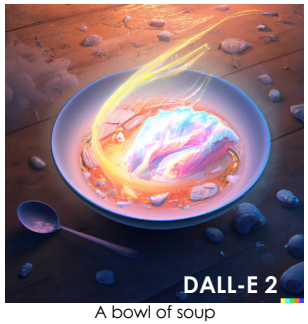
Con người không thể nhận diện bài viết 500 từ là do máy hay người viết

OpenAI

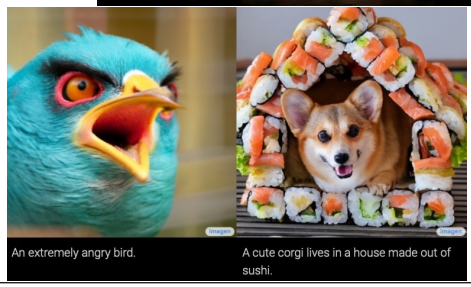
Brown, Tom B., Benjamin Mann, Nick Ryder, Melanie Subbiah, Jared Kaplan, Prafulla Dhariwal, Arvind Neelakantan et al. "Language models are few-shot learners." NeurIPS (2020). **Best Paper Award**

Some successes: Text-to-image (2022)

- Draw pictures by keywords



A bowl of soup



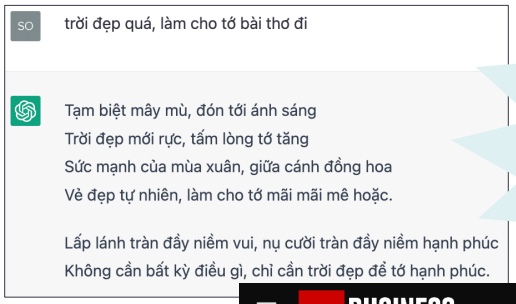
An extremely angry bird.

A cute corgi lives in a house made out of sushi.

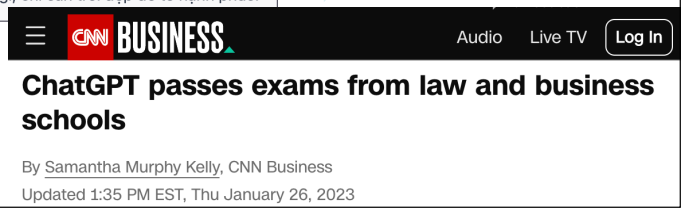
Google **Imagen**

Some successes: ChatGPT (2022)

- Human-level Chatting, Writing, QA,…



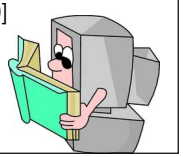
Why ChatGPT is about to change how you work, like it or not?
- Forbes, Feb. 2, 2023



OpenAI

What is Machine Learning?

- Machine Learning (ML) is an active subfield of Artificial Intelligence.
- ML seeks to answer the question [Mitchell, 2006]
  - *How can we build computer systems that automatically improve with experience, and what are the fundamental laws that govern all learning processes?*
- Some other views on ML:
  - Build systems that automatically improve their performance [Simon, 1983].
  - Program computers to optimize a performance objective at some task, based on data and past experience [Alpaydin, 2020]

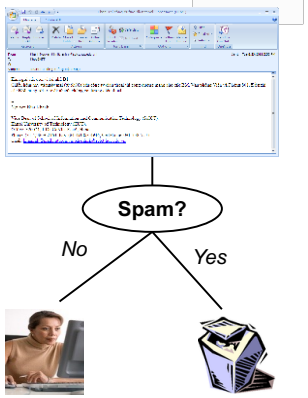


A learning machine

- We say that a machine *learns* if the system reliably improves its performance **P** at task **T**, following experience **E**.
- A *learning problem* can be described as a triple (**P, T, E**).
- ML is close to and intersects with many areas.
  - Computer Science,
  - Statistics, Probability,
  - Optimization,
  - Psychology, Neuroscience,
  - Computer Vision,
  - Economics, Biology, Bioinformatics, …

Some real examples (1)

- Spam filtering for emails
  - **T**: filter/predict all emails that are spam.
  - **P**: the accuracy of prediction, that is the percentage of emails that are correctly classified into normal/spam.
  - **E**: set of old emails, each with a label of spam/normal.



Some real examples (2)

- Image tagging
  - **T**: give some words that describe the meaning of a picture.
  - **P**: ?
  - **E**: set of pictures, each has been labelled with a set of words.

FISH WATER OCEAN TREE CORAL

PEOPLE MARKET PATTERN TEXTILE DISPLAY

BIRDS NEST TREE BRANCH LEAVES

What does a machine learn?

- A mapping (function):

$$f : x \mapsto y$$

  - x: observation (example, data instance), past experience
  - y: prediction, new knowledge, new experience,…
- A model (mô hình)
  - Data are often supposed to follow or be generated from an unknown model.
    (Ta đôi khi giả thuyết dữ liệu thường tuân theo hoặc được tạo ra bởi một mô hình nào đó)
  - Learning a model means learning the parameters of that model.
    (Học một mô hình có nghĩa là học/tìm những tham số của mô hình đó)

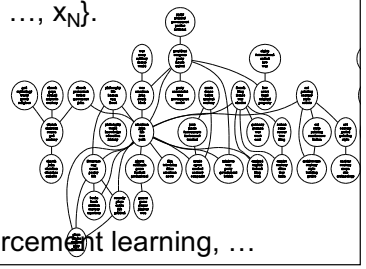
Where does a machine learn from?

- Learn from a set of training examples (**training set,** tập học, tập huấn luyện) { $\{x_1, x_2, \ldots, x_N\}$; $\{y_1, y_2, \ldots, y_M\}$ }
  - x_i is an observation (quan sát, mẫu, điểm dữ liệu) of x in the past.
  - y_i is an observation of y in the past, often called *label (nhãn)* or *response (phản hồi)* or *output (đầu ra)*.
- After learning:
  - We obtain a model, new knowledge, or new experience (f).
  - We can use that model/function to do **prediction** or **inference** for future observations, e.g.,

$$y = f(x)$$

Two basic learning problems

- **Supervised learning (học có giám sát):** learn a function y = f(x) from a given training set $\{x_1, x_2, \ldots, x_N, y_1, y_2, \ldots, y_N\}$ so that $y_i \cong f(x_i)$ for every i.
  - ***Classification*** (categorization, phân loại, phân lớp): if y only belongs to a discrete set, for example {spam, normal}
  - ***Regression*** (hồi quy): if y is a real number
- **Unsupervised learning (học không giám sát):** learn a function y = f(x) from a given training set $\{x_1, x_2, \ldots, x_N\}$.
  - y can be a data cluster
  - y can be a hidden structure
  - y can be a trend, …



- Other: semi-supervised learning, reinforcement learning, …

Supervised learning: classification

- ***Multiclass*** *classification (phân loại nhiều lớp)*: when the output y is one of the pre-defined labels $\{c_1, c_2, \ldots, c_L\}$
  (mỗi đầu ra chỉ thuộc 1 lớp, mỗi quan sát x chỉ có 1 nhãn)
  - Spam filtering: y in {spam, normal}
  - Financial risk estimation: y in {high, normal, no}
  - Discovery of network attacks: ?
- ***Multilabel*** *classification (phân loại đa nhãn)*: when the output y is a subset of labels
  (mỗi đầu ra là một tập nhỏ các lớp;
  mỗi quan sát x có thể có nhiều nhãn)
  - Image tagging: y = {birds, nest, tree}
  - sentiment analysis

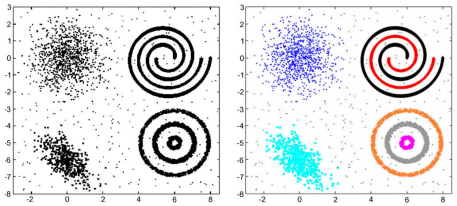
BIRDS NEST TREE

Supervised learning: Regression

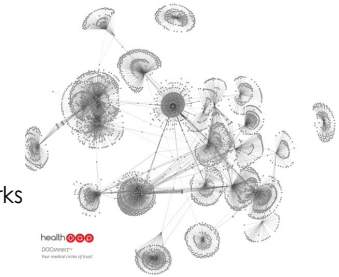
- Prediction of stock indices

Unsupervised learning: examples (1)

- Clustering data into clusters
  - Discover the data groups/clusters

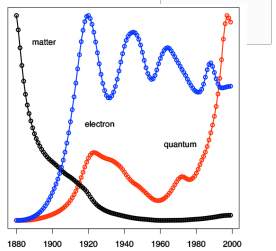


- Community detection
  - Detect communities in online social networks



Unsupervised learning: examples (2)

- Trends detection
  - Discover the trends, demands, future needs of online users



Data

Structured – relational (table-like)

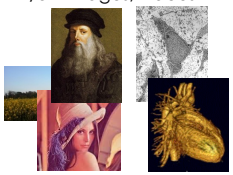
	A	B	C	D	E	F	G
1	Country	Region	Population	Under15	Over60	Fertil	LifeExpi
2	Zimbabwe	Africa	13724	40.24	5.68	3.64	54
3	Zambia	Africa	14075	46.73	3.95	5.77	55
4	Yemen	Eastern M	23852	40.72	4.54	4.35	64
5	Viet Nam	Western P	90796	22.87	9.32	1.79	75
6	Venezuela (Bo	Americas	29955	28.84	9.17	2.44	75
7	Vanuatu	Western P	247	37.37	6.02	3.46	72
8	Uzbekistan	Europe	28541	28.9	6.38	2.38	68
9	Uruguay	Americas	3395	22.05	18.59	2.07	77

Un-structured

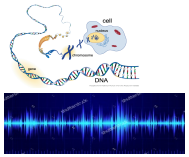

```
{
    "code": "1473a6fd39d1d8fa48654aac9d8cc2754232
    "title": "[Updating] Câu chuyện xuyên mưa về
    "url": "http://techtalk.vn/updating-cau-chuye
    "labels": "techtalk/Cong nghe",
    "content": "Vào chiều tối ngày 09/12/2016 vừa
    "image_url": "",
    "date": "2016-12-10T03:51:10Z"
}
```


texts in websites, emails, articles, tweets

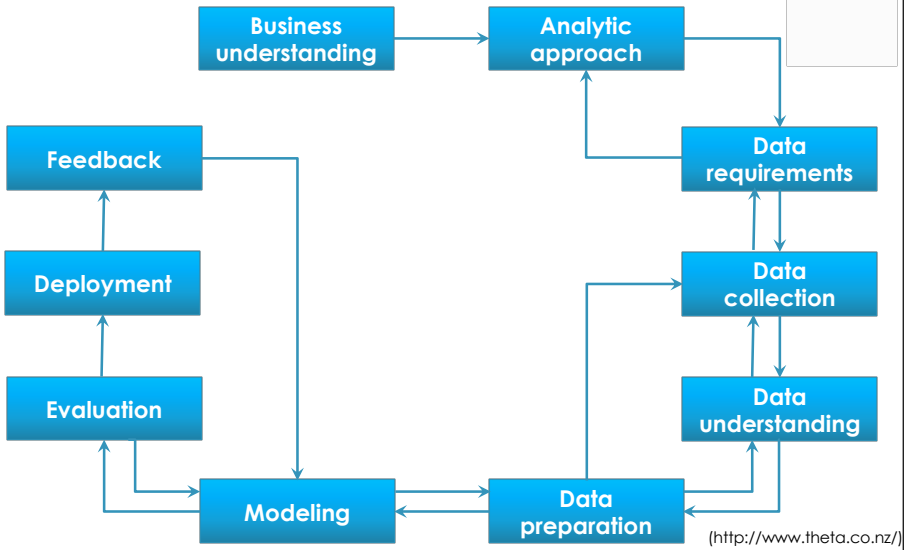
2D/3D images, videos + meta



spectrograms, DNAs, …

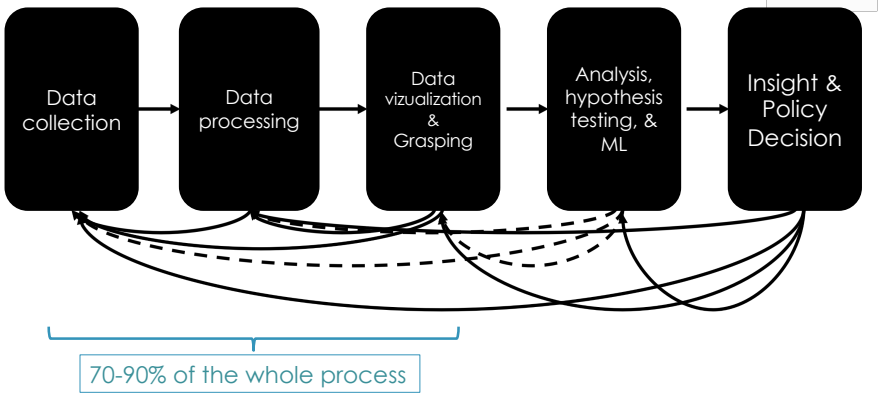


Methodology: product-driven



(http://www.theta.co.nz/)

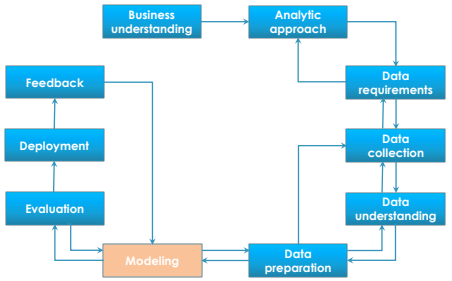
Methodology: insight-driven



(John Dickerson, University of Maryland)

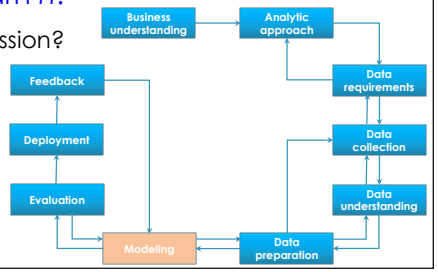
Design a learning system (1)

- Some issues should be carefully considered when designing a learning system.
- Determine the type of the function to be learned
  (Determine the learning problem)
  - f: X → {0,1}
  - f: X → set of labels/tags
  - f: X → R
- Collect a training set:
  - The training set plays the key role in the effectiveness of the system.
  - Do the observations have any label?
  - The training observations should characterize the whole data space →good for future predictions.



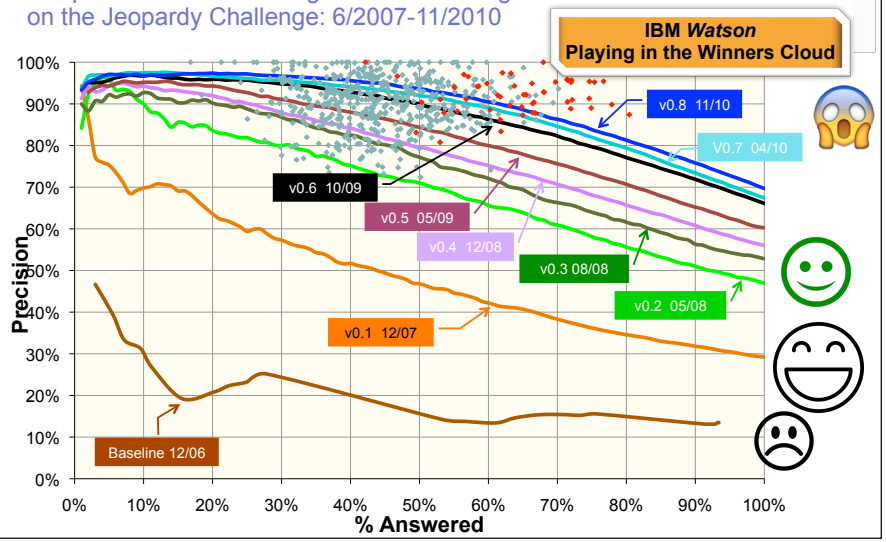
Design a learning system (2)

- Select a representation/approximation (model) *h* for the unknown function *f*:
  (Lựa chọn dạng hàm *h* để đi xấp xỉ hàm *f* chưa biết)
  - Linear?
  - A neural network?
  - A decision tree? …
- Select a good algorithm to learn *h*:
  - Ordinary least square? Ridge regression?
  - Backpropagation?
  - ID3?



Product development: experience

DeepQA: Incremental Progress in Answering Precision on the Jeopardy Challenge: 6/2007-11/2010



ML: some issues (1)

- **Learning algorithm**
  - Under what conditions the chosen algorithm will (asymtotically) converge?
  - For a given application/domain and a given objective function, what algorithm performs best?
- ***No-free-lunch theorem*** [Wolpert and Macready, 1997]: if an algorithm performs well on a certain class of problems then it necessarily pays for that with degraded performance on the set of all remaining problems.
  - *No algorithm can beat another on all domains.*
    (không có thuật toán nào luôn hiệu quả nhất trên mọi miền ứng dụng)

ML: some issues (2)

- **Training data**
  - *How many observations* are enough for learning?
  - Whether or not does the *size of the training set* affect performance of an ML system?
  - What is the effect of the *disrupted* or *noisy* observations?

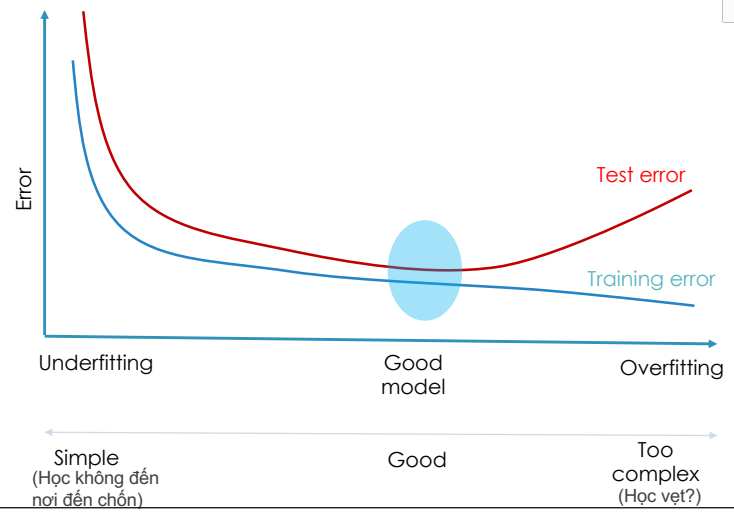
ML: some issues (3)

- **Learnability**:
  - The goodness/limit of the learning algorithm?
  - What is the **generalization** (khả năng tổng quát hoá) of the system?
    - Predict well new observations, not only the training data.
    - Avoid overfitting.

Overfitting (quá khớp, quá khít)

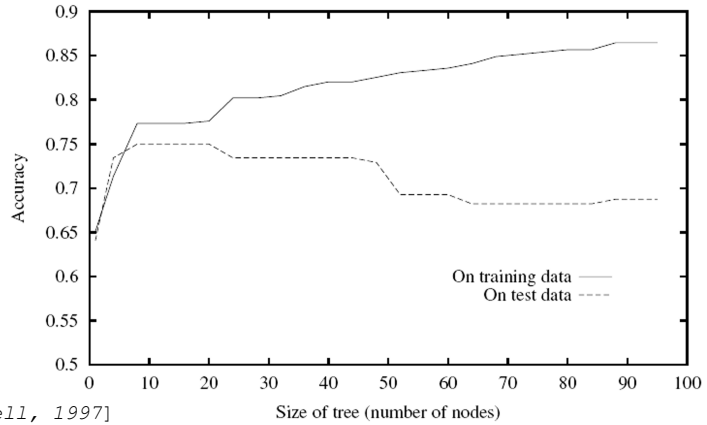
- Function h is called *overfitting* [Mitchell, 1997] if there exists another function g such that:
  - g might be worse than h for the training data, but
  - g is better than h for future data.
- A learning algorithm is said to overfit relative to another one if it is *more accurate in fitting* known data, but *less accurate in predicting* unseen data.
- Overfitting is caused by many factors:
  - The trained function/model is too complex or have too much parameters.
  - Noises or errors are present in the training data.
  - The training size is too small, not characterizing the whole data space.

Overfitting



Overfitting: example

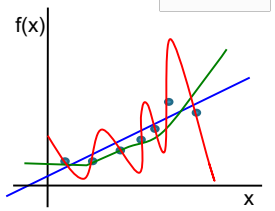
- Increasing the size of a decision tree can degrade prediction on unseen data, even though increasing the accuracy for the training data.

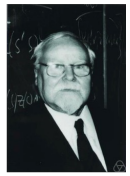


[Mitchell, 1997]

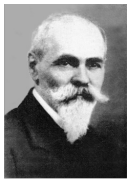
Overfitting: Regularization

- Among many functions, which one can generalize best from the given training data?
  - *Generalization is the main target of ML.*
  - Predict unseen data well.
- **Regularization:** a popular choice (Hiệu chỉnh)





Tikhonov, smoothing an ill-posed problem



Zaremba, model complexity minimization

Bayes: priors over parameters

Andrew Ng: need no maths, but it prevents overfitting!

(Picture from http://towardsdatascience.com/multitask-learning-teach-your-ai-more-to-make-it-better-dde116c2cd40)

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