

# Neural Network and Deep Learning

## Lecture 1

Yanwei Fu

School of Data Science, Fudan University



# Logistics



# Neural Networks are taking over!

Optional subtitle

- Neural networks have become one of the major thrust areas recently in various pattern recognition, prediction, and analysis problems
- In many problems they have established the state of the art

Often exceeding previous benchmarks by large margins



# Course Information

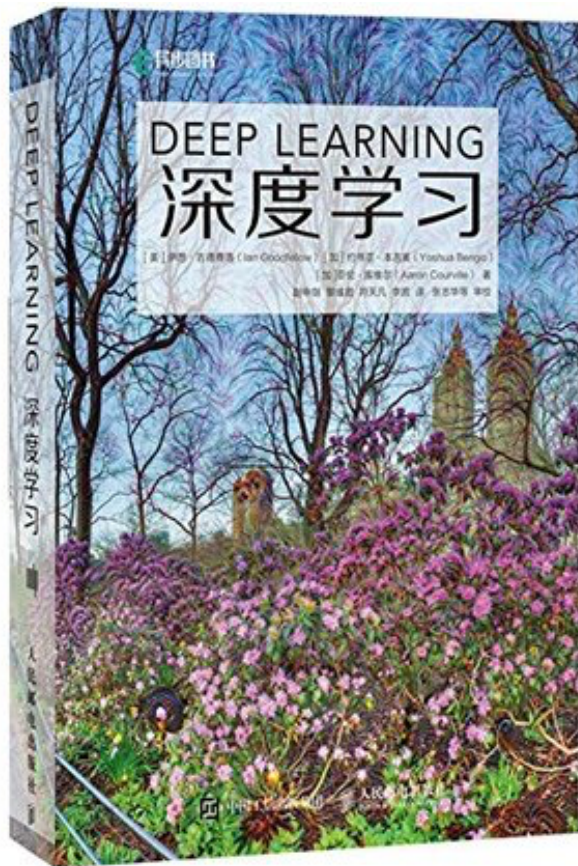
Optional subtitle

- Instructor: 付彦伟
- Email: [yanweifu@fudan.edu.cn](mailto:yanweifu@fudan.edu.cn)
- Course Websites: [http://yanweifu.github.io/courses/Neural\\_network/index.html](http://yanweifu.github.io/courses/Neural_network/index.html)
- Times&Venue:
  - Classroom:
    - H6406 (≡ 6-8 [1-2],[4-5],[7-8],[10-11],[13-14],16)
    - Hxinjinbo-1309 (≡ 6-8 3,6,9,12,15)
- Office Hours: Wed. 4:00-5:30pm, Xin jinbo building, 1506
- TAs: Chengming Xu ( email: [dlcourse.xcm@gmail.com](mailto:dlcourse.xcm@gmail.com) ); Yikai Wang (email: [yikaiwang19@fudan.edu.cn](mailto:yikaiwang19@fudan.edu.cn)); Boyan Jiang (email: )



# Textbooks

Optional subtitle



Ian Goodfellow, Yoshua Bengio, Aaron Courville

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## 《神经网络与深度学习》

Neural Networks and Deep Learning

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邱锡鹏

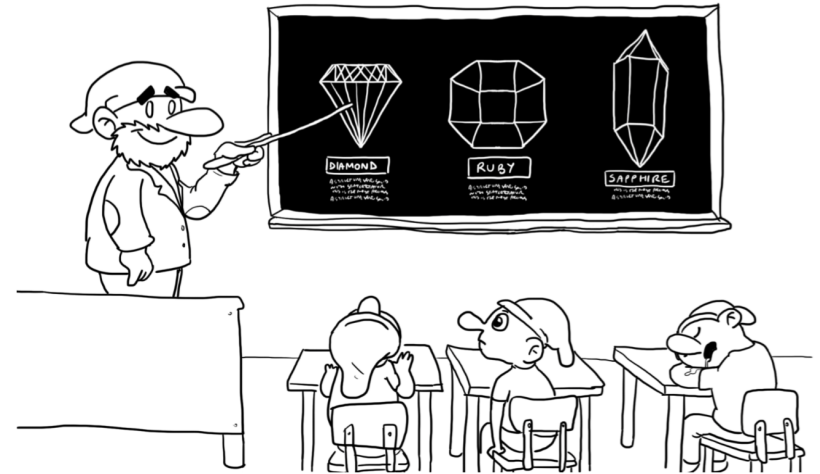
xpqi@fudan.edu.cn

2019 年 11 月 21 日



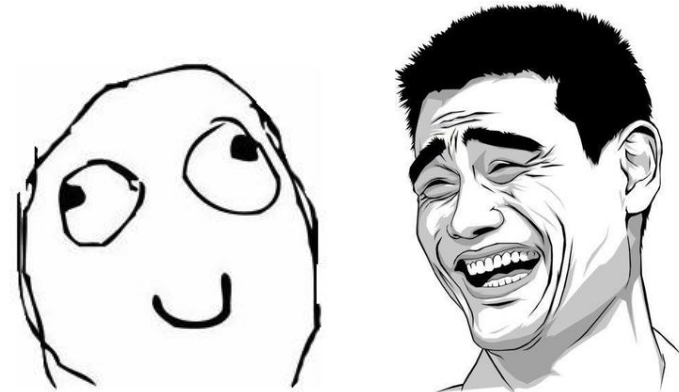
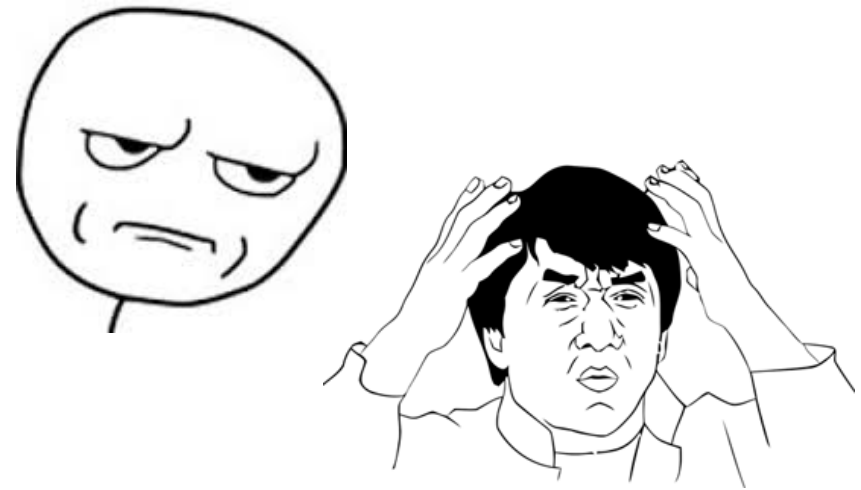
# Course Requirement

- Prerequisites on Math:
  - Basic linear Algebra/Calculus: vectors, matrices, eigenvalues;
  - Probability: conditional probability, expectations;
  - Multivariate calculus: gradients, optima;
- Prerequisites on programming:
  - Data structures: pointers, trees, heaps, hash maps, graphs;
  - Scientific computing: matrix factorisation .



# Course Work

- Final scores=
  - +Class attendance/discussion (10%);
  - +Homework Assignment (10%)
  - +Projects (40%); 3-4; 每个人独立完成
  - +Final Project (40%): hit NeurPIS submissions;  
每个队最多两个人。
- Courses:
  - We will give some tutorial.
- Pain and Happiness
  - Huge efforts to code, debug, read and think;
  - Worth doing it!!



# Academic Integrity (学术诚信)

- **Academic integrity** is the **moral code or ethical policy of academia**. This includes values such as **avoidance of cheating or plagiarism; maintenance of academic standards; honesty and rigor** in research and academic publishing. ([https://en.wikipedia.org/wiki/Academic\\_integrity](https://en.wikipedia.org/wiki/Academic_integrity))
- No cheating and plagiarism,
  - How to define *Plagiarism*? We follow ACM Policy on Plagiarism.
  - 抄袭和被抄袭双方的成绩都将被取消.
  - 作业、报告、期末论文的署名原则：署你名字的工作必须由自己完成；允许讨论，但作业必须独立完成，并在作业中列出所有参与讨论的人。不允许其他任何形式的合作——尤其是与已经完成作业的同学“讨论”。
  - 这是学术底线。





# Questions?

Optional subtitle

Post on eLearning



# Machine Learning & AI

人工智能研究的主要方法：

1, **符号主义方法**：认知是一种符号处理过程，人类思维过程可用符号来描述，思维就是计算，这种思想一度构成了人工智能的基础理论。

代表人物：司马贺（西蒙，Herbert Alexander Simon）和纽厄尔（Allen Newell），物理符号系统，1975 年图灵奖获得者。

2, **联结主义方法**：模拟人的智能要依靠仿生学，特别是需要模拟人脑，建立脑模型。人类思维的基本单元是神经元，而不是符号，智能是相互联结的神经元竞争与协作结果。

代表人物：麦卡洛克（Warren McCulloch), 皮茨（Walter Pitts)提出的神经元的数理模型。

3, **行为主义方法**：模拟人在控制过程中的智能行为和作用，研制所谓的控制论动物。

代表人物：博德（H.W.Bode) 和埃文斯（W.R.Evans)等。



# Resources—Conferences

## Machine Learning:

- Neural Information Processing Systems (NeurIPS)
- International Conference on Machine Learning (ICML)
- European Conference on Machine Learning (ECML)
- Uncertainty in Artificial Intelligence (UAI)
- Computational Learning Theory (COLT)
- International Conference on AI & Statistics (AISTATS)

## AI In general

- AAAI Conference on Artificial Intelligence (AAAI) (AAAI: Association for the Advancement of Artificial Intelligence);
- International Joint Conference on Artificial Intelligence (IJCAI);

## Pattern Recognition&Computer Vision:

- European Conference on Computer Vision (ECCV)
- IEEE Conference on Computer Vision and Pattern Recognition (CVPR)
- IEEE International Conference on Computer Vision (ICCV)
- ICPR/ACCV/BMVC;
- Data Mining:
  - ACM SIGKDD (Knowledge discovery and Data Mining);
  - ACM SIGIR/ICDM;

## Natural Language Processing:

- ACL/EMNLP/COLING



# Resources—Journals

## Machine Learning&AI:

- Journal of Machine Learning Research (JMLR)
- IEEE Trans on Pattern Analysis and Machine Intelligence (TPAMI);
- Artificial Intelligence;
- International Journal of Computer Vision (IJCV);

## Statistics:

- The Annals of Statistics;

## Pattern Recognition &NN:

- Neural Computation
- Neural Networks;
- IEEE Transactions on Neural Networks and Learning System

## Data Mining:

- IEEE Transactions on Knowledge and Data Engineering (TKDE)



# What is Machine Learning?

- **Definition of ML (Mitchell, 1997): WELL-POSED LEARNING PROBLEMS.**
  - A computer program is said to learn from experience  **$E$**  with respect to some class of tasks  **$T$**  and performance measure  **$P$** , if its performance at tasks in  **$T$** , as measured by  **$P$** , improves with experience  **$E$** .
- **Example: A computer program that learns to play checkers**
  - **Task:** playing checkers games;
  - **Experience:** obtained by playing games against itself;
  - **Performance Measure:** percent of games won against opponents



## A handwriting recognition learning problem:

- Task ***T***: recognizing and classifying handwritten words within images
- Performance measure ***P***: percent of words correctly classified
- Training experience ***E***: a database of handwritten words with given classifications

## A robot driving learning problem: an example from (Mitchell, 1997)

- Task ***T***: driving on public four-lane highways using vision sensors;
- Performance measure ***P***: average distance traveled before an error (as judged by human overseer)
- Training experience ***E***: a sequence of images and steering commands recorded while observing a human driver;

## Example: Spam classification

- Task ***T***: determine if emails are Spam or non-Spam.
- Experience ***E***: Incoming emails with human classification
- Performance Measure ***P***: percentage of correct decisions



# Notations, formally

## Task:

$\mathcal{X}$  input variables (from input set), a.k.a., features, predictors, independent variables.

$\mathcal{Y}$  output variables (from output set), a.k.a., response or dependent variable.

$f : \mathcal{X} \rightarrow \mathcal{Y}$  Prediction function,

## Performance:

$l : \mathcal{X} \rightarrow \mathcal{Y}$  Loss function,

$l(y, y')$  is the cost of predicting  $y'$  if  $y$  is correct.

## Experience: task-dependent, many different scenarios

- Supervised Learning, Unsupervised Learning, Reinforcement Learning,
- Semi – supervised Learning, Multiple Instance Learning, Active Learning.

# Supervised Learning

- A labeled training set examples with outputs provided by an expert,

$$\mathcal{D} = \{(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)\} \subset \mathcal{X} \times \mathcal{Y}$$

- Regression Vs. Classification problems,
  - **Regression**:  $Y$  is **quantitative** (e.g price, blood pressure);
  - **Classification**:  $Y$  takes values in a finite, unordered set (survived/died, digit 0-9, cancer class of tissue sample), **qualitative**.

Other problems such as ranking is often formulated as either problem.

## Definition,

- A supervised learning system (or learner),  $L$  is a (computable) function from the set of (finite) training sets to the set of prediction functions:

$$L : \mathbb{P}^{<\infty}(\mathcal{X} \times \mathcal{Y}) \rightarrow \mathcal{Y}^{\mathcal{X}}$$
$$L : \mathcal{D} \mapsto f$$

So if presented with a training set  $\mathcal{D}$ , it provides a decision rule/function

$$f : \mathcal{X} \rightarrow \mathcal{Y}$$

Let  $L$  be a learning system.

- Process of computing is  $f = L(\mathcal{D})$  called training (phase).
- Applying  $f$  to new data is called prediction, or testing. (phase).