Lecture 1: Course Introduction CS480/680 Intro to Machine Learning

2023-1-10

Pascal Poupart David R. Cheriton School of Computer Science





- Introduction to Machine Learning
- Course website and logistics



Instructor

- Pascal Poupart (Professor and CIFAR AI Chair)
 - 20+ years experience in Machine Learning







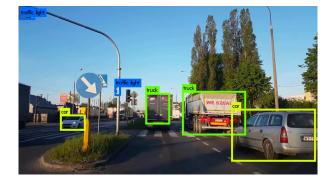


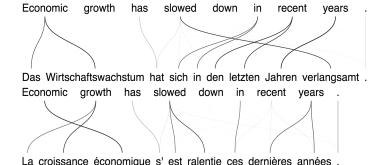
Machine Learning

- Traditional computer science
 - Program computer for every task



- New paradigm
 - Provide examples to machine
 - Machine learns to accomplish tasks based on examples







Definitions

• Arthur Samuel (1959): **Machine learning** is the field of study that gives computers the ability to learn without being explicitly programmed.

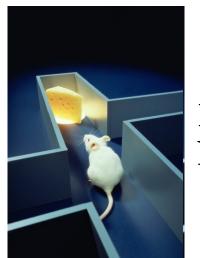
 Tom Mitchell (1998): A computer program is said to learn from experience E with respect to some class of tasks T and performance measure P, if its performance at tasks in T, as measured by P, improves with experience E.



Three Categories



Supervised learning



Reinforcement learning

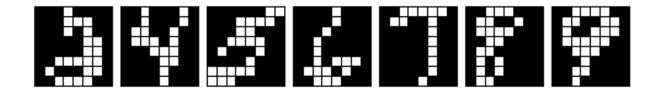


Unsupervised learning



Supervised Learning

Example: digit recognition (postal code)

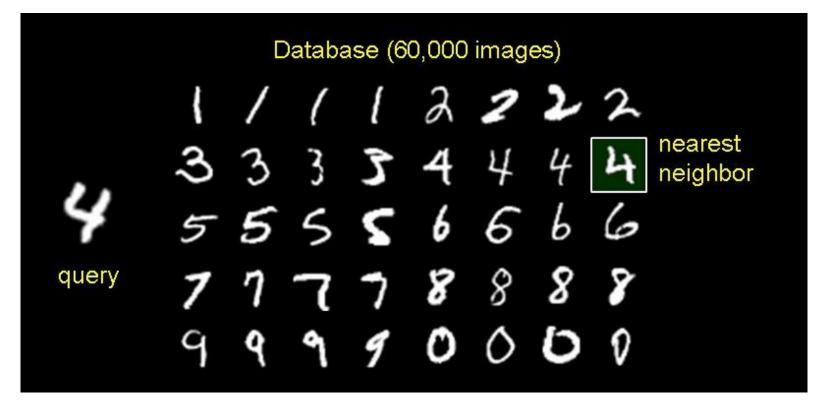


 Simplest approach: memorization



Supervised Learning

Nearest neighbour:





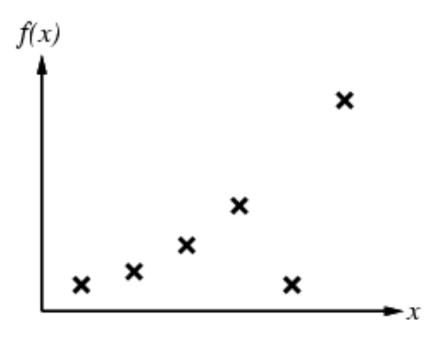
More Formally

Inductive learning (for supervised learning):

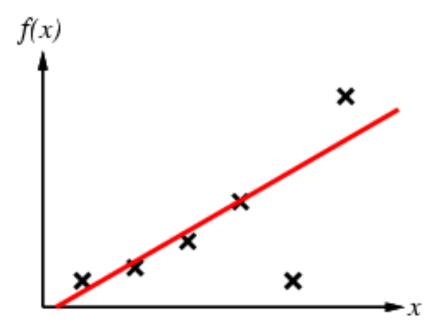
- Given a training set of examples of the form (x, f(x))
 - x is the input, f(x) is the output

- Return a function h that approximates f
 - *h* is called the hypothesis

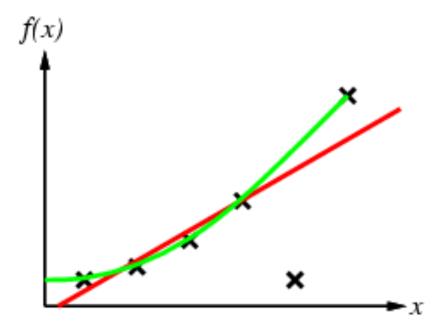




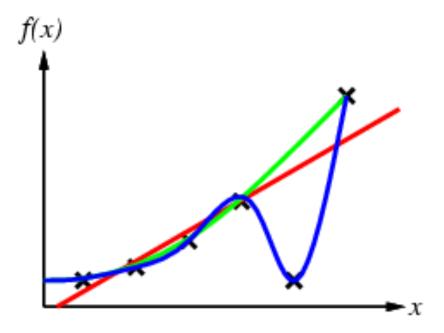




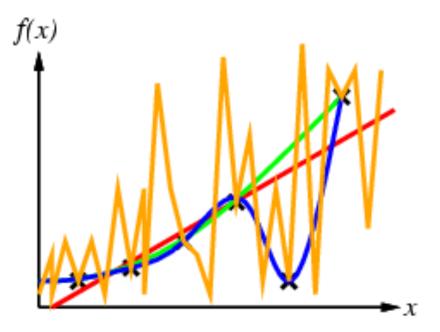














Generalization

• Key: a good hypothesis will generalize well (i.e., predict unseen examples correctly)

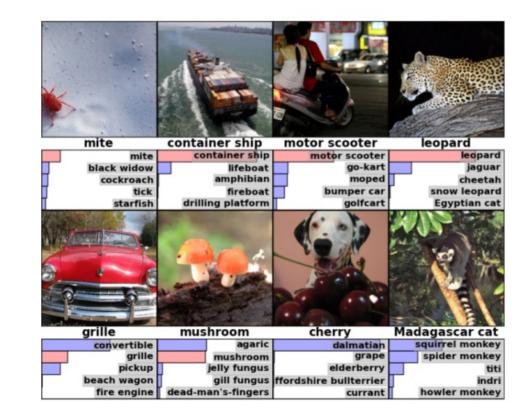
• Ockham's razor: prefer the simplest hypothesis consistent with data



ImageNet Classification

- 1000 classes
- 1 million images

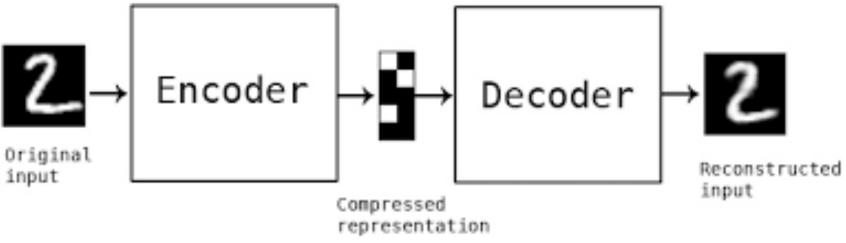
 Deep neural networks (supervised learning)





Unsupervised Learning

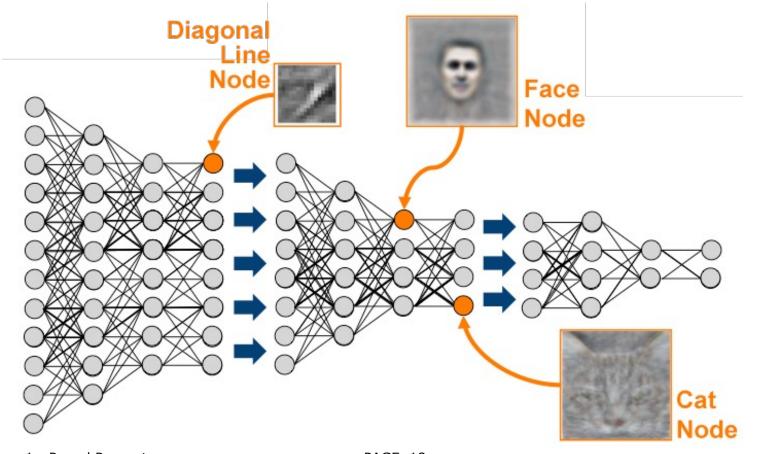
- Output is not given as part of training set
- Find model that explains the data
 - E.g. clustering, compressed representation, features, generative models





Unsupervised Feature Generation

Encoder trained on large number of images





Unsupervised Image Generation

• Which images are real? And which ones are fake?



CelebA (Liu et al., 2015)



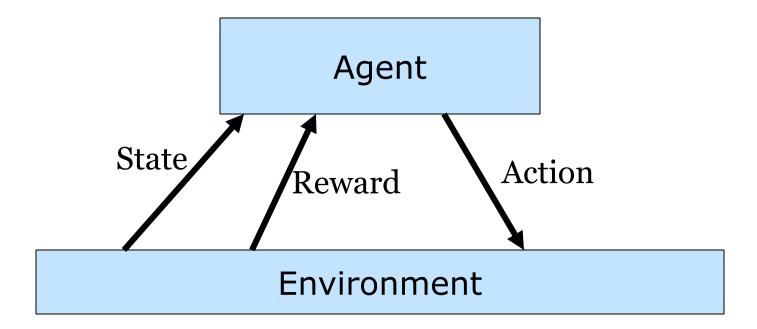
StyleGAN2 (Karras et al., 2020)

 Image generation: variational autoencoders, generative adversarial networks, diffusion models



Fake!

Reinforcement Learning Problem



Goal: Learn to choose actions that maximize rewards



Animal Psychology

- Negative reinforcements:
 - Pain and hunger
- Positive reinforcements:
 - Pleasure and food
- Reinforcements used to train animals



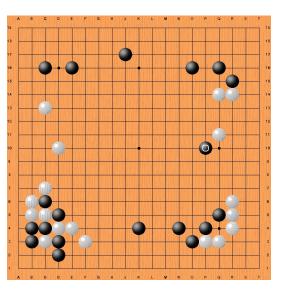
Let's do the same with computers!



Game Playing

- Example: Go (one of the oldest and hardest board games)
- Agent: player
- Environment: opponent
- State: board configuration
- Action: next stone location
- **Reward:** +1 win / -1 loose





2016: AlphaGo defeats top player Lee Sedol (4-1) Game 2 move 37: AlphaGo plays unexpected move (odds 1/10,000)



Combining Unsupervised, Supervised and Reinforcement Learning

- Modern systems:
 - Phase 1: unsupervised feature extraction (no labels)
 - Phase 2: supervised training (exploit labels)
 - Phase 3: fine tune by reinforcement learning (exploit reinforcements)
- Alpha Go: supervised + reinforcement learning
- Sentiment analysis with BERT: unsupervised + supervised learning
- ChatGPT: supervised + reinforcement learning



Applications of Machine Learning

- Speech recognition: Siri, Cortana
- Natural Language Processing: Machine translation, dialog systems
- Computer vision: Image and video analysis
- Robotic Control: Autonomous vehicles
- Intelligent assistants: Activity recognition, recommender systems
- Computational finance: Stock trading, portfolio optimization



This course

Supervised and unsupervised machine learning

- But not reinforcement learning
- See CS 486/686/885 for Reinforcement Learning
 - https://cs.uwaterloo.ca/~ppoupart/teaching/cs885-fall22/

