Introduction to Reinforcement Learning

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Lecture Plan

- Lecture material
 - Lecture slides will be provided
- Performance evaluation
 - Attendance: 10%
 - Assignment : 10%
 - Exam: 80% (Mid-term: 40, Final: 40)
- Lecture Method
 - offline



What is Machine Learning?

• "Learning is any process by which a system improves performance from experience." -- Herbert Simon

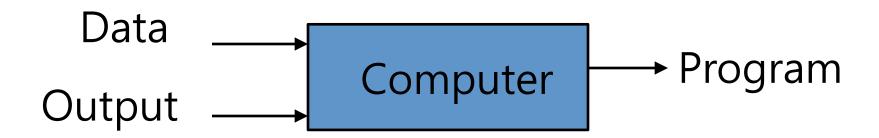
- Definition by Tom Mitchell (1998):
 - Machine Learning is the study of algorithms that
 - *▶* improve their performance P
 - > at some task T
 - *>* with experience *E*.
 - A well-defined learning task is given by $\langle P, T, E \rangle$.



Traditional Programming



Machine Learning





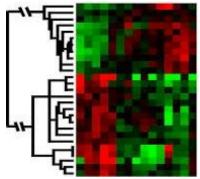
When Do We Use Machine Learning?

- ML is used when:
- Human expertise does not exist (navigating on Mars)
- Humans can't explain their expertise (speech recognition)
- Models must be customized (personalized medicine)
- Models are based on huge amounts of data (genomics)









- Learning isn't always useful:
- There is no need to "learn" to calculate payroll



A classic example of a task that requires machine learning: It is very hard to say what makes a 2





Some more examples of tasks that are best solved by using a learning algorithm

Recognizing patterns:

- Facial identities or facial expressions
- Handwritten or spoken words
- Medical images

Generating patterns:

- Generating images or motion sequences
- Recognizing anomalies:
 - Unusual credit card transactions
 - Unusual paderns of sensor readings in a nuclear power plant
- Prediction:
 - Future stock prices or currency exchange rates



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Sample Applications

- Web search
- Computational biology
- Finance
- E-commerce
- Space exploration
- Robotics
- Information extraction
- Social networks
- Debugging software
- [Your favorite area]



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Samuel's Checkers-Player

- "Machine Learning: Field of study that gives computers the ability to learn without being explicitly programmed."
 - Arthur Samuel (1959)





Improve on task T, with respect to performance metric P, based on experience E

Dfining the Learning Task

- T: Playing checkers
- P: Percentage of games won against an arbitrary opponent
- E: Playing practice games against itself
- T: Recognizing hand-written words
- P: Percentage of words correctly classified
- E: Database of human-labeled images of handwritten words
- T: Driving on four-lane highways using vision sensors
- P: Average distance traveled before a human-judged error
- E: A sequence of images and steering commands recorded while observing a human driver
- T: Categorize email messages as spam or legitimate
- P: Percentage of email messages correctly classified
- E: Database of emails, some with human-given labels



Types of Learning

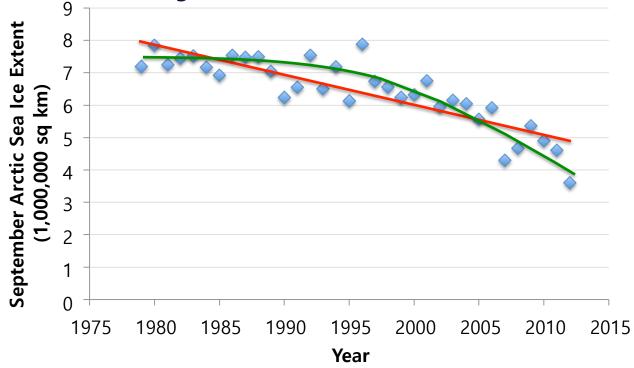
- Supervised (inductive) learning
 - Given: training data + desired outputs (labels)
- Unsupervised learning
 - Given: training data (without desired outputs)
- Semi-supervised learning
 - Given: training data + a few desired outputs
- Reinforcement learning
 - Rewards from sequence of actions



Supervised Learning: Regression

- Given (x1, y1), (x2, y2), ..., (xn, yn)
- Learn a function f(x) to predict y given x

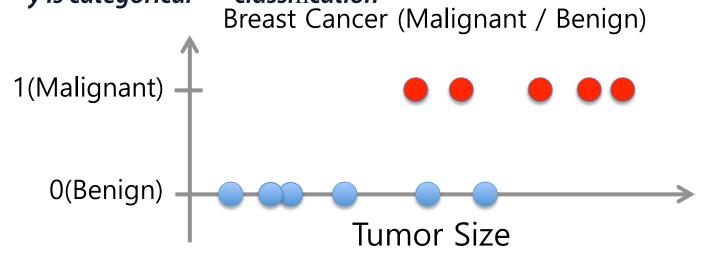
— y is real-valued == regression





Supervised Learning: Classification

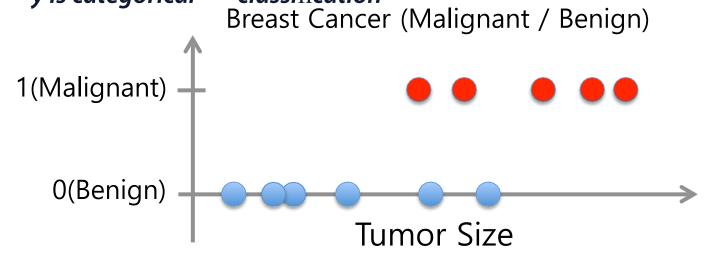
- Given (x1, y1), (x2, y2), ..., (xn, yn)
- Learn a function f(x) to predict y given x
 - y is categorical == classification

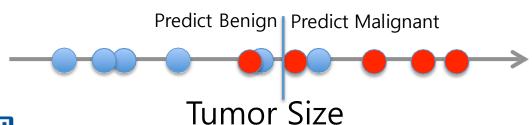




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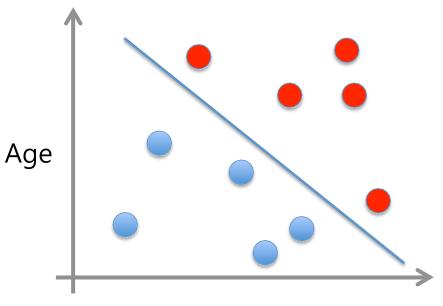




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Supervised Learning

- x can be multi-dimensional
 - Each dimension corresponds to an attribute



Tumor Size

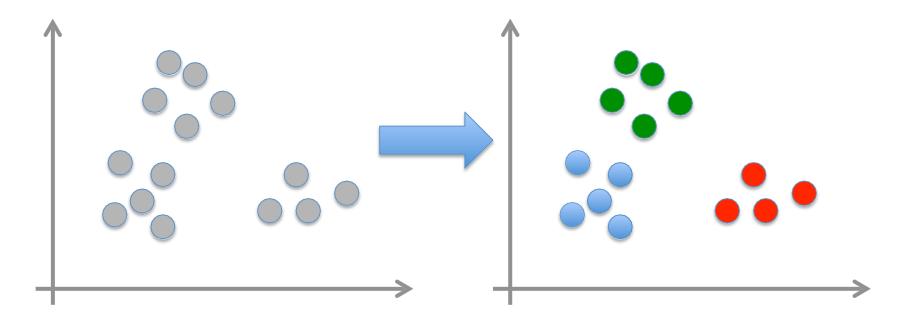
- Clump Thickness
- Uniformity of Cell Size
- Uniformity of Cell Shape

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Unsupervised Learning

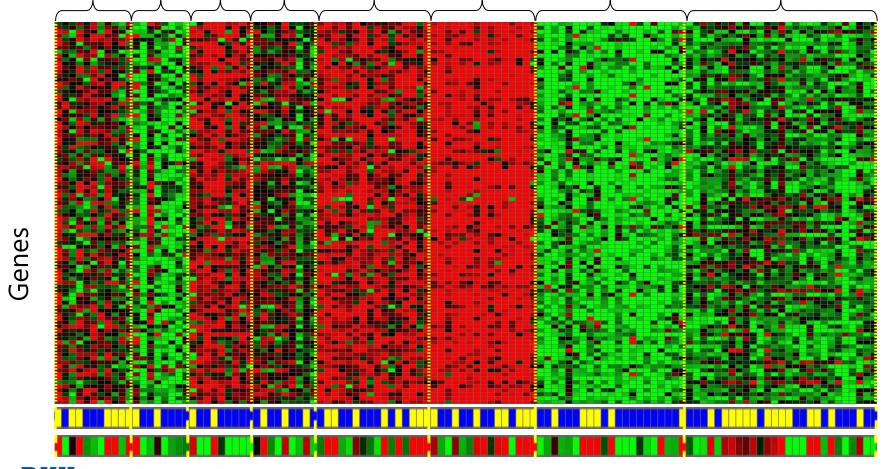
- Given x1, x2, ..., xn (without labels)
- Output hidden structure behind the x's
 - E.g., clustering





Unsupervised Learning

Genomics application: group individuals by genetic similarity





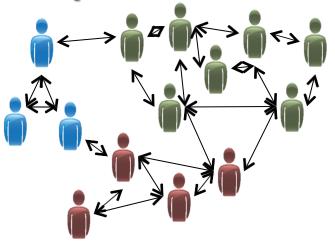
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Organize computing clusters

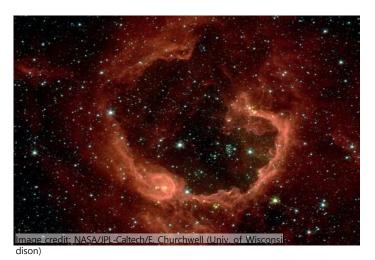


Market segmentation 단국대학교

Unsupervised Learning



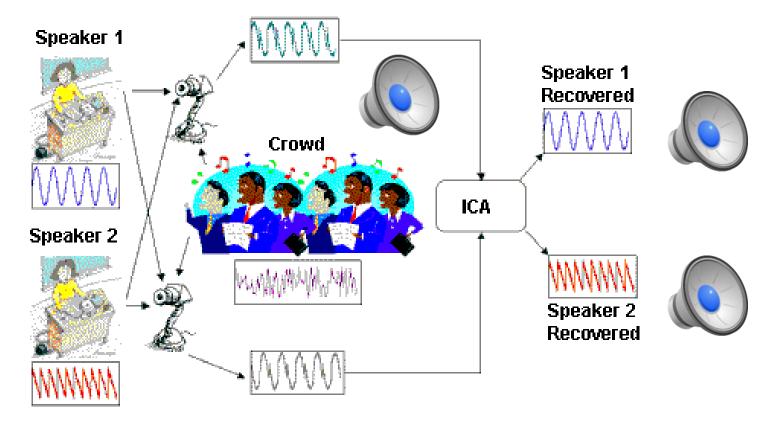
Social network analysis



Astronomical data analysis

Unsupervised Learning

- Independent component analysis
 - separate a combined signal into its original sources





Reinforcement Learning

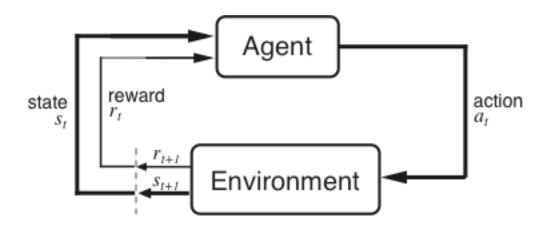
- Given a sequence of states and actions with (delayed) rewards, output a policy
 - Policy is a mapping from states → actions that tells you what to do in a given state

Examples:

- Credit assignment problem
- Game playing
- Robot in a maze
- Balance a pole on your hand



The Agent-Environment Interface



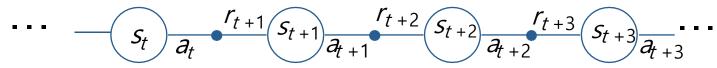
Agent and environment interact at discrete time steps: t = 0, 1, 2, K

Agent observes state at step t: $s_t \in S$

produces action at step t: $a_t \in A(s_t)$

gets resulting reward: $r_{t+1} \in \Re$

and resulting next state : S_{t+1}





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Reinforcement Learning



