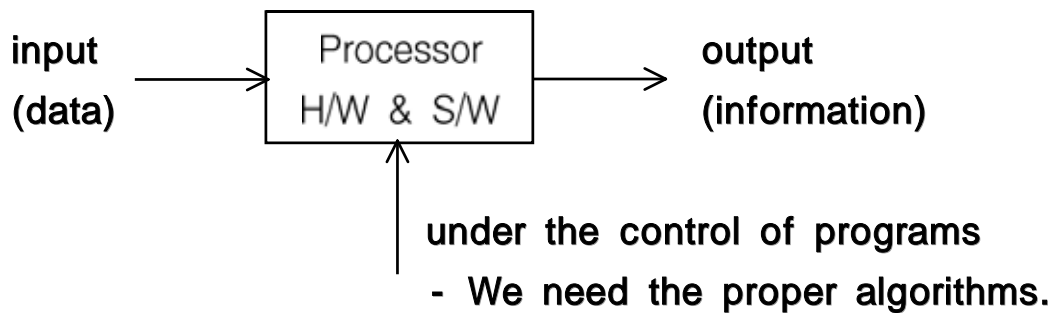


Introduction

- information processing system



1990s software crisis: hard to find out algorithms
(especially related to intelligence)

- machine learning

- . programming computers to *learn* instead of finding out algorithms
- . definition of learning:
a computer program is said to learn from experience E (data) with respect to some class of tasks T and performance P, if its performance at tasks in T, as measured by P, improves with experience E.

example: learning to recognize spoken words

task T: recognizing spoken words

performance measure P: recognition rate

training experience E: speech data

example: learning to play checkers

task T: playing checkers

performance measure P: percentage of games won

training experience E: playing practice games against itself

- **some disciplines related to machine learning:**

artificial intelligence, computational learning theory (COLT), information theory, statistics, psychology and neurobiology, philosophy, etc.,

- **design procedure of machine learning system**

Step 1. Choosing the training experience

Step 2. Choosing the target function

(what type of knowledge will be learned)

Step 3. Choosing a representation for the target function

(functional form of targets)

Step 4. Choosing a learning algorithm

Step 5. Measuring the performance and updating
the learning system

- **some issues in machine learning**

- . What algorithms can approximate functions well?
- . How does training examples influence accuracy?
(sample complexity)
- . What is the upper bounds of general errors for a learning system? (generalization bounds)
- . How does noisy data influence accuracy?
- . How can prior knowledge of learner help?
- . What clues can we get from biological learning system?

Reference: Machine Learning, chapter 1.