



Learning Automata

Chapter 1~1.2

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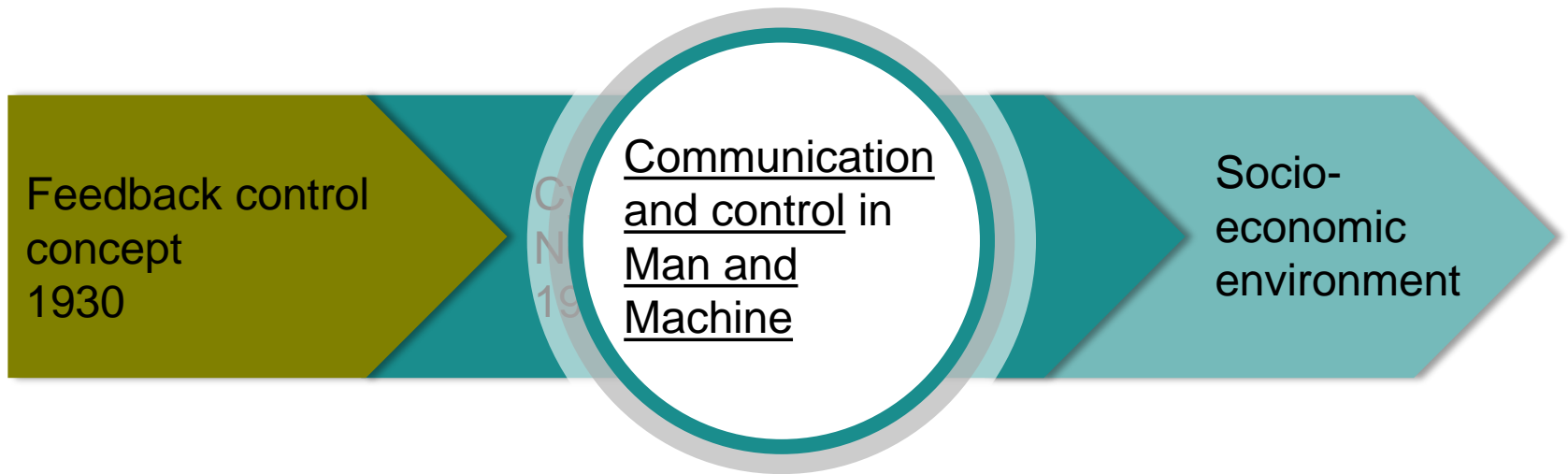
Laboratory for System Health & Risk management

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 - Identification and Control
 - Pattern Recognition
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Introduction

- System theory



Introduction

■ Systems Theory

- The identification and control of well defined deterministic and stochastic systems
- Interest gradually shifted to substantial amount of uncertainty
 - Adaptation
 - Learning
 - Pattern recognition
 - Self-organization
- Deterministic and Stochastic approach

Introduction

■ In more Complex Systems

- Highly Uncertainty
- Lack of parameter information
- Hard to find dynamic relation

Pattern recognition
Adaptive control
Self-organization

■ Distributed processing

- Due to distributed databases
 - Collecting, processing, accessing data

■ Learning Automata

- Input output searching under the influence of reinforcement feedback
- Collective behavior of a number of automata operating in a distributed fashion can be described

Learning

■ Learning in Psychology

■ Empiricism

- Experience is the only source of knowledge
- All complex ideas are made up of simpler ideas
- Complex ideas are connected together through the association of experience

■ Rationalism

- The interrelations among elementary ideas are just as fundamental as the ideas themselves

Learning

■ Learning in Psychology

- Stimulus-response theories
 - Importance of motivations, reward, punishment
 - Behaviorist
 - A detailed understanding of the internal workings of an organism is not necessary for developing a theory of behavior
- Cognitive theories
 - Collection, transmission, storage, retrieval of information

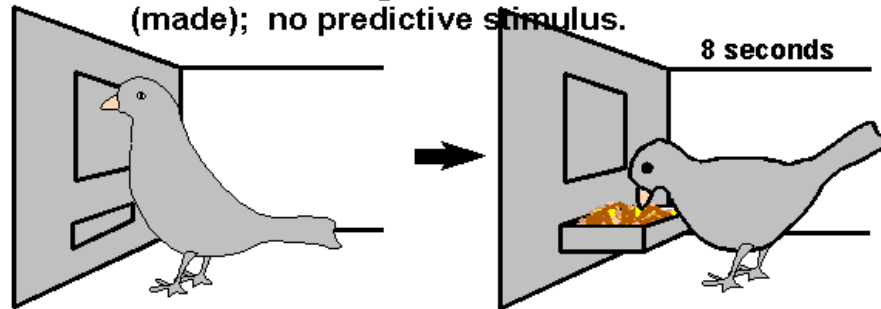
Learning in Psychology

- Expectancy learning
- Classical learning
- Instrumental conditioning
- Operant conditioning

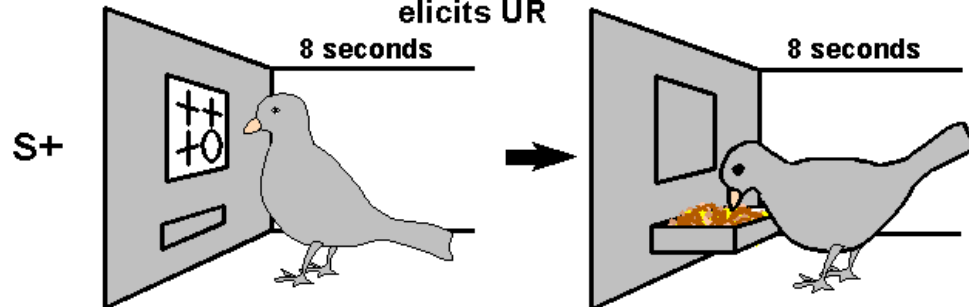
Learning in Psychology

- Expectancy conditioning → inductive inference
- Instrumental conditioning
- Classical conditioning

Instrumental Conditioning: response must be emitted (made); no predictive stimulus.

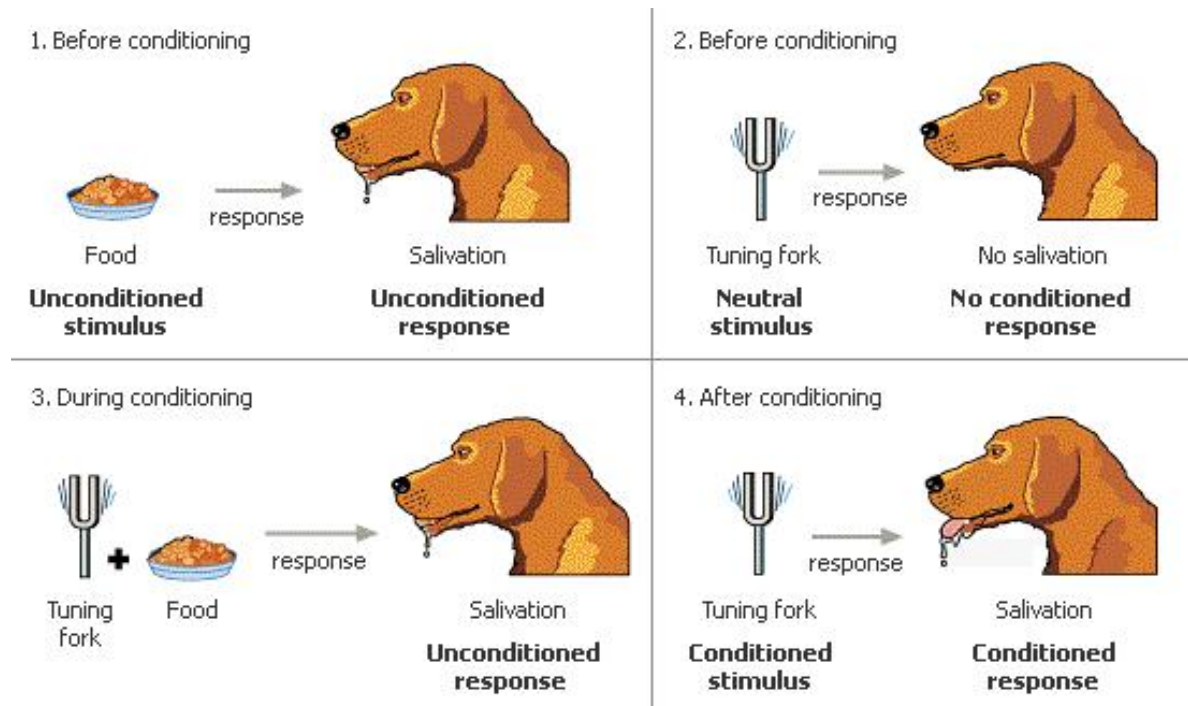


Pavlovian (Classical) Conditioning: predictive CS; US elicits UR



Learning in Psychology

- Expectancy conditioning → inductive inference
- Instrumental conditioning
- Classical conditioning



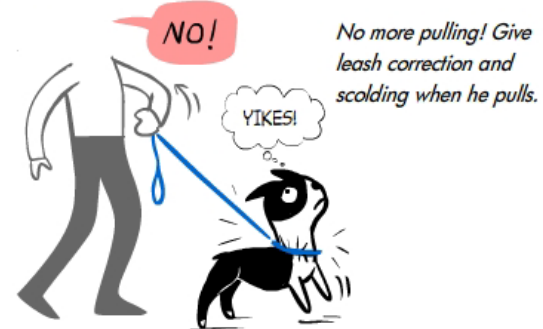
Learning in Psychology

■ Operant conditioning

+R **POSITIVE REINFORCEMENT**
ADDING GOOD STUFF TO
INCREASE A BEHAVIOR



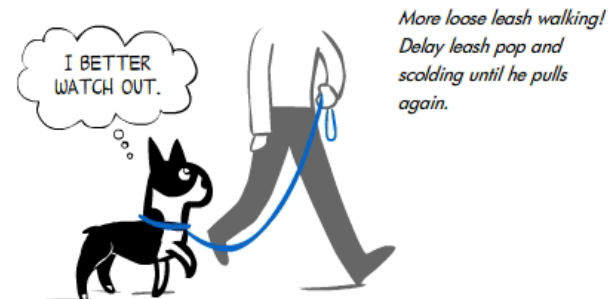
+P **POSITIVE PUNISHMENT**
ADDING BAD STUFF TO
DECREASE A BEHAVIOR



-P **NEGATIVE PUNISHMENT**
DELAYING GOOD STUFF TO
DECREASE A BEHAVIOR



-R **NEGATIVE REINFORCEMENT**
DELAYING BAD STUFF TO
INCREASE A BEHAVIOR



Mathematical Learning Theory

■ Hull (1943)

- Argued for the development of quantitative theories in learning

■ Estes (1959)

- The development of learning theory for individual organisms is an elaboration of association theory
- Association theory
 - Complex stimulus and response patterns at higher levels of learning

Deterministic, Stochastic, and Adaptive Process

■ Adaptive Control Processes - A Guided Tour

(Bellman , 1961)

■ Deterministic

Prior information

■ Stochastic

■ Adaptive Processes

Less Prior information

Deterministic, Stochastic, and Adaptive Process

■ Deterministic Processes

- The controller is designed
 - To satisfy either a set of performance criteria
 - To optimize a given index of performance
- The design of practical estimators and controllers may be difficult due to analytical and computational reasons

Deterministic, Stochastic, and Adaptive Process

■ Stochastic Control Processes

- Probability characteristic
- Relevant distributions
- Optimization of linear system with quadratic performance
- The optimization problems as well as the numerical procedures used to solve them are same
(deterministic process)

Deterministic, Stochastic, and Adaptive Process

■ Adaptive Processes

- Prior information is considerably less
 - Ex. economics, biology, engineering, psychology, operations research, and AI
- So, information needed for their estimation by off-line experiment

E

- Evolutionary algorithm
- Evolutionary programming
- Expectation–maximization algorithm
- Extremal optimization

F

- Fernandez’s method
- Fourier–Motzkin elimination
- Frank–Wolfe algorithm

G

- Gauss–Newton algorithm
- General Subpopulation Framework
- Generalized iterative scaling
- Genetic algorithm
- Genetic algorithms in economics
- Genetic improvement (computer science)
- Golden section search
- Gradient descent
- Gradient method
- Graduated optimization
- Great Deluge algorithm
- Greedy algorithm
 - CMA-ES

- Quasi-Newton inverse least squares method
- Quasi-Newton Least Squares Method
- Quasi-Newton method

R

- Random optimization
- Random search
- Revised simplex method
- Rosenbrock methods

S

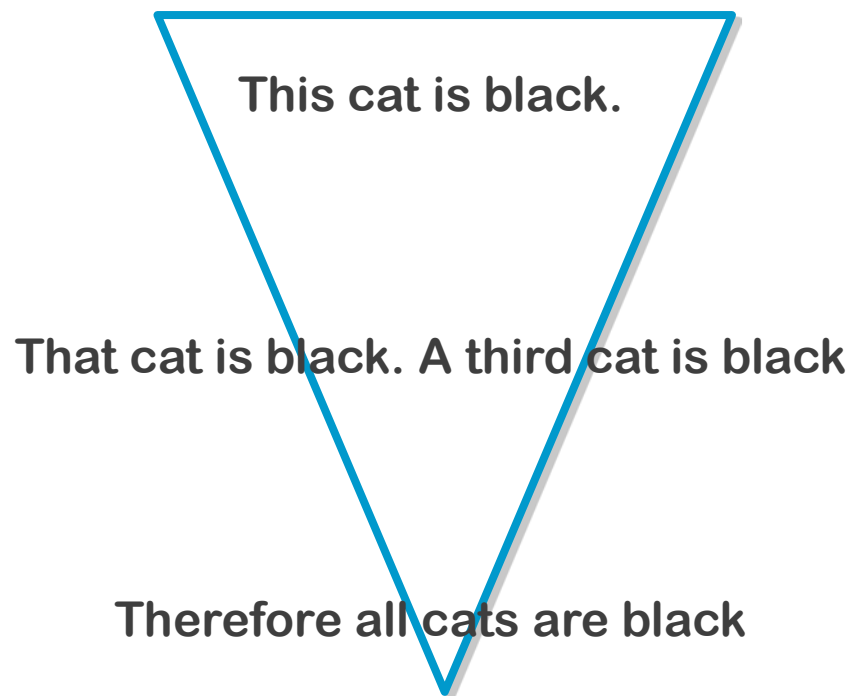
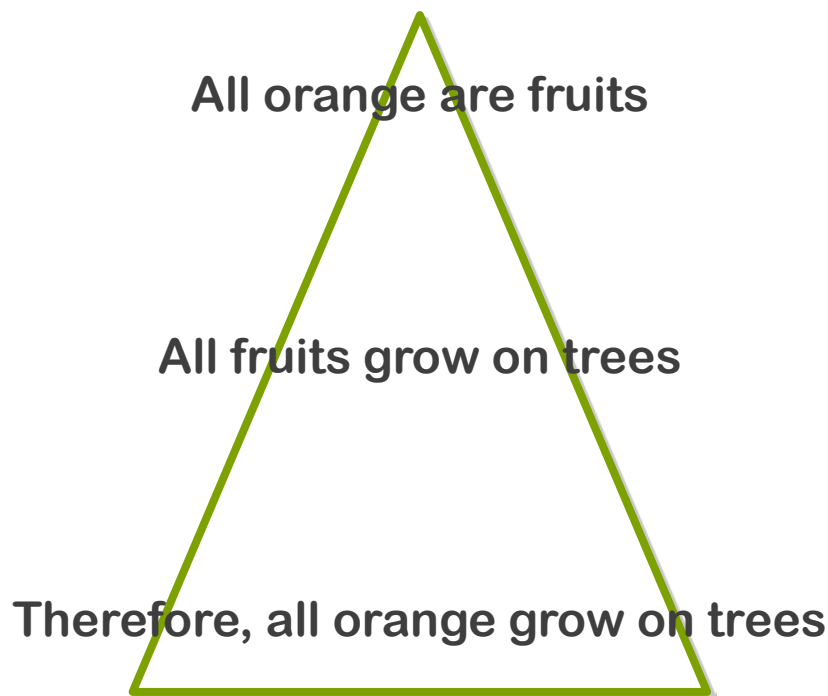
- Search-based software engineering
- Sequential minimal optimization
- Sequential quadratic programming
- Simplex algorithm
- Simulated annealing
- Simultaneous perturbation stochastic appro
- Social cognitive optimization
- Space allocation problem
- Space mapping
- Special ordered set
- Stochastic hill climbing
- Successive parabolic interpolation
- Symmetric rank-one

T

- Network simplex algorithm

Deductive and Inductive Inference

■ Inference reasoning



Inductive and deductive inference do not contradict but merely complement each other and both are found to be essential for the learning.

Deductive and Inductive Inference

■ Credibility and Induction

- Inductive probability : “credibility”

Measure of confidence we place in a hypothesis on the basis of observed data

$$q(H_i) \geq 0, \sum_i q(H_i) = 1$$

Identification and Control

■ Identification vs Control

- Decide better action in the future, or taking the best action on the basis of past experience

- Feldbaum (1965) – Dual control problem

Contains uncertainty

- New inputs it has to improve its knowledge of the characteristics of the system
- New knowledge, it has to determine what actions are necessary for successful control

Pattern Recognition

■ Definition

- Deals with the issue of building a machine or a program that will display some capability of living organisms for classifying or discriminating sensory signals

■ Object

- To sort patterns into different classes so that those patterns which belong to a class share some common properties
 - Medical diagnosis, speech recognition, and scene analysis

Pattern Recognition

■ Pattern recognition process

- Collect raw data
- Raw data of the patterns is converted into n-dimensional feature vectors
 - Feature : contain the essential attributes of the given patterns
 - Preprocessing (Data Transformation, Smoothing, Normalize, Balancing etc)
 - Filtering (FFT, DWT, CWT, etc.),
 - Feature extraction(PCA)
 - Feature selection (ICA)
- Classification

$$p(x|w^1) > p(x|w^2) \Rightarrow x \in w^1 \quad \text{X is the pattern, w is class}$$

Bayesian Learning

- Frequentist vs Bayesian (Degree of Belief)
 - Frequentist
 - Parameters – quantities whose values are fixed but unknown
 - The best estimate of their values – the one that maximizes the probability of obtaining the observed samples
 - Bayesian
 - Parameters – random variables having some known prior distribution
 - Observation of the samples converts to a posterior density; revising our opinion about the true values of the parameters

Bayesian Learning

■ Bayesian approach

- Extract information regarding the unknown parameter θ from observation $x_1, x_2, x_3, \dots, x_N$ on the system

Posterior	Likelihood	Prior
	The probability of getting this evidence if this hypothesis were true	The probability of θ being true, before gathering evidence
$p(\theta x_i)$	$= \frac{p(x_i \theta)p(\theta)}{p(x_i)}$	
The probability that the hypothesis (θ , parameter) is true given the evidence	Evidence The marginal probability of the evidence	

Further discussion

Learning automata

■ Example



Hill Climbing

- Optimization problems

