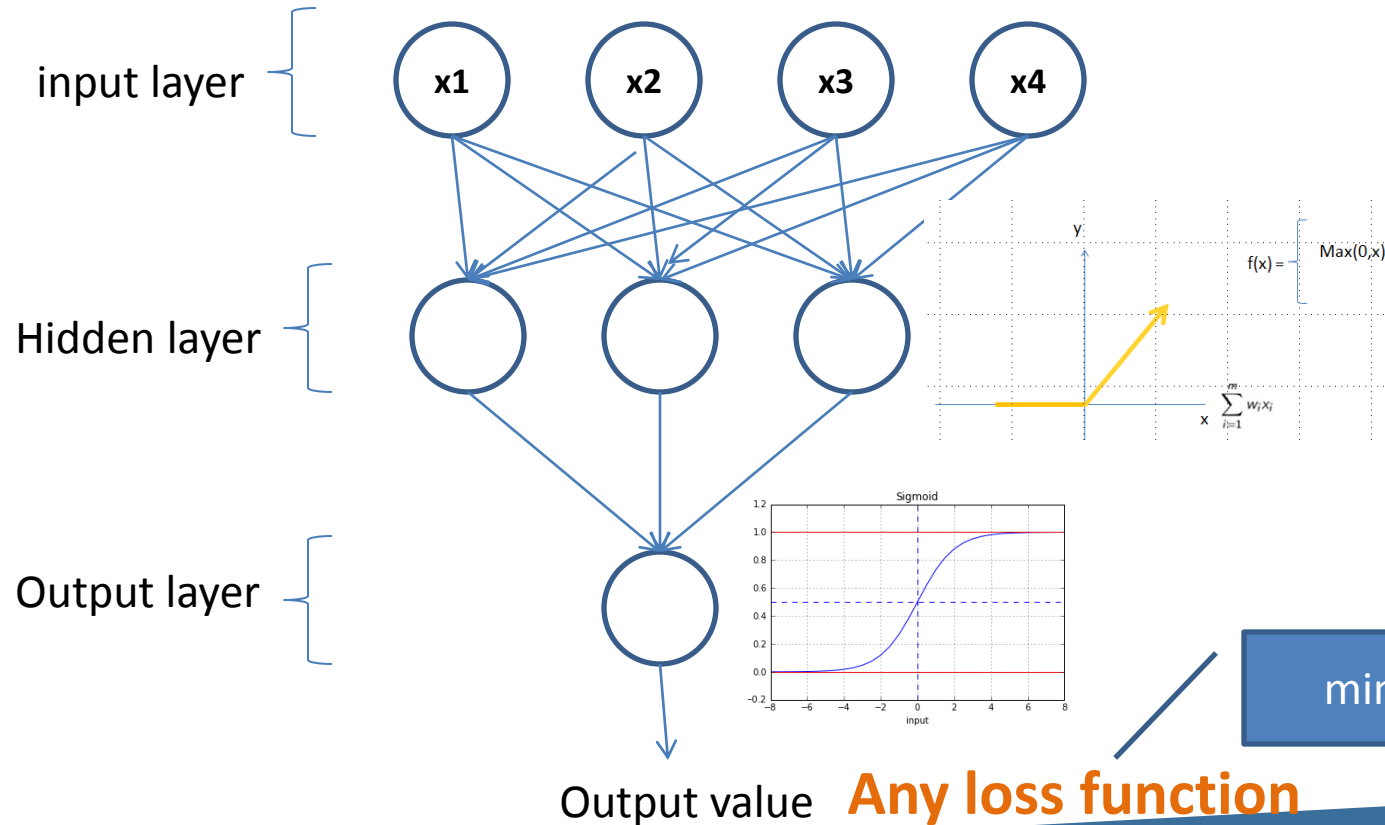
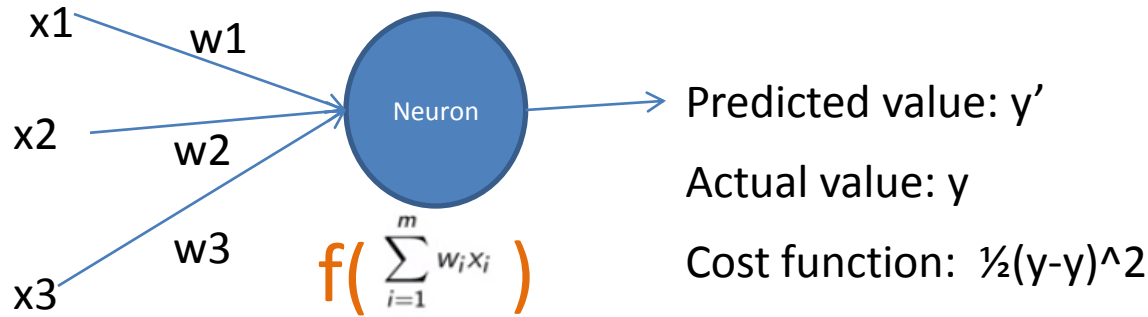


# ❖ Loss function



# ❖ Where is the learning?



## ❖ Steps

1. Random weights
2. Input observation to input layer. Each feature to one node
3. Forward propagation from left to right. Each neuron is activated based on the weights. Till you get an output
4. Compare the predicted result to the actual result. Measure the error (loss)
5. Back-propagate the error from right to left. Update the weights according to how much they are responsible
6. Repeat 1 – 5. Update the weights after each observation or each batch
7. Epoc is when the whole training set is passes through the ANN. Redo the epocs



# Chapter: Reinforcement learning

# Agenda

- 1) What is reinforcement learning
- 2) The bellman equation
- 3) Markov Decision Process
- 4) Policy vs Plan
- 5) Living Penalty
- 6) Q-Learning

# ❖ What is reinforcement learning

## Environment :

Any environment : for eg:

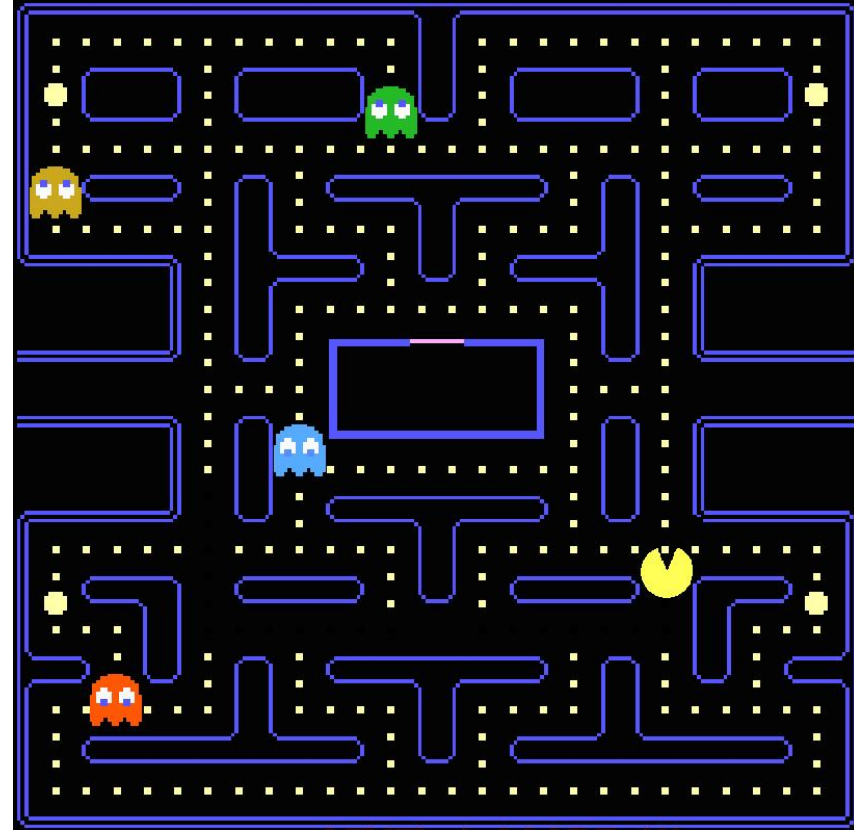
Road, kitchen

## Agent:

Object which works in the environment

Takes actions as performs steps

Gets a reward: Good reward for favorable state, bad reward for unfavorable state



# ❖ What is reinforcement learning

**Agent : Robot**

**Environment: Path**

**Actions that the agent can take:**

**stand,**

**lift right leg,**

**lift left leg,**

**plant right leg,**

**plant left get**

# ❖ Bellman Equation

Created by Richard Ernest Bellman

S – State

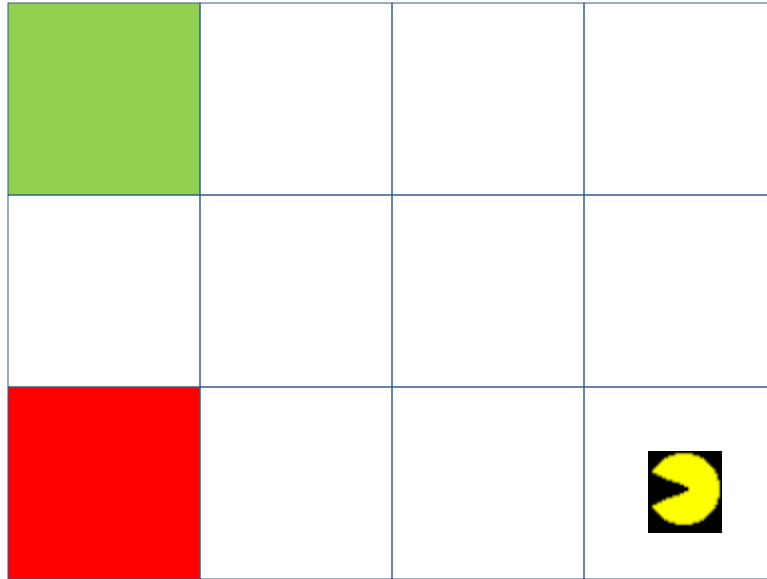
A- Action that the agent can take

R- Reward that the agent gets

Y – Discount



# ❖ Bellman Equation



## ❖ Bellman Equation

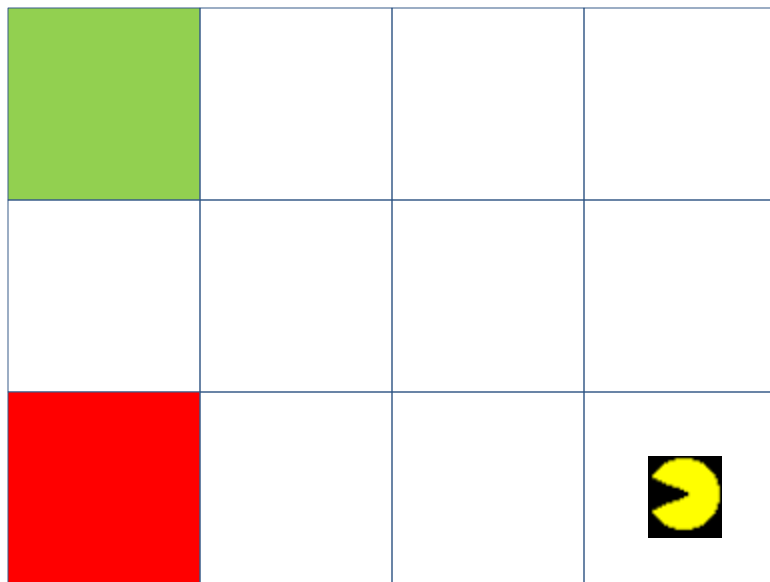
$$V(s) = \max(R(s,a) + \gamma V(s'))$$

$S$  = current state

$S'$  = next state or where you end up

# ❖ Bellman Equation

$$V(s) = \max(R(s,a) + \gamma V(s'))$$



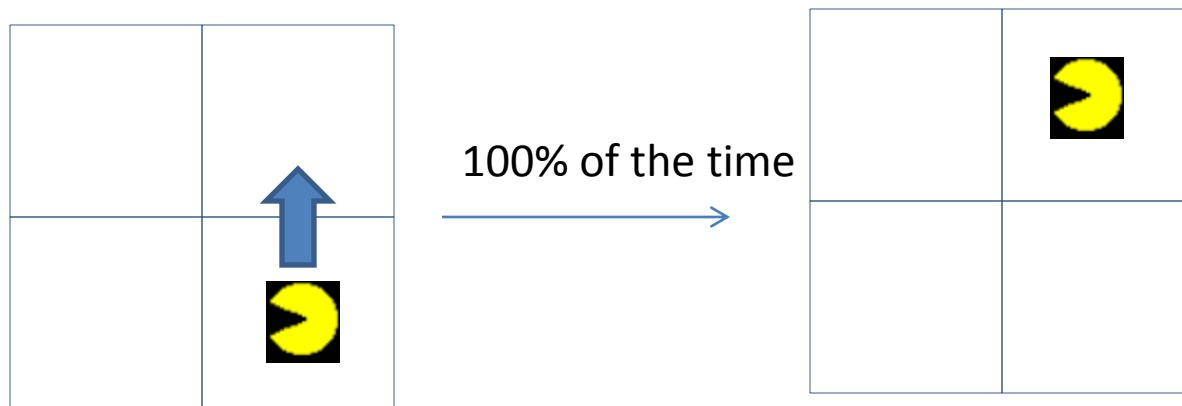
## ❖ Bellman Equation – A Plan

	V=1	V=0.9	V=0.81
V=1	V=0.9	V=0.81	V=0.729
	V=0.81	V=0.729	V=0.66



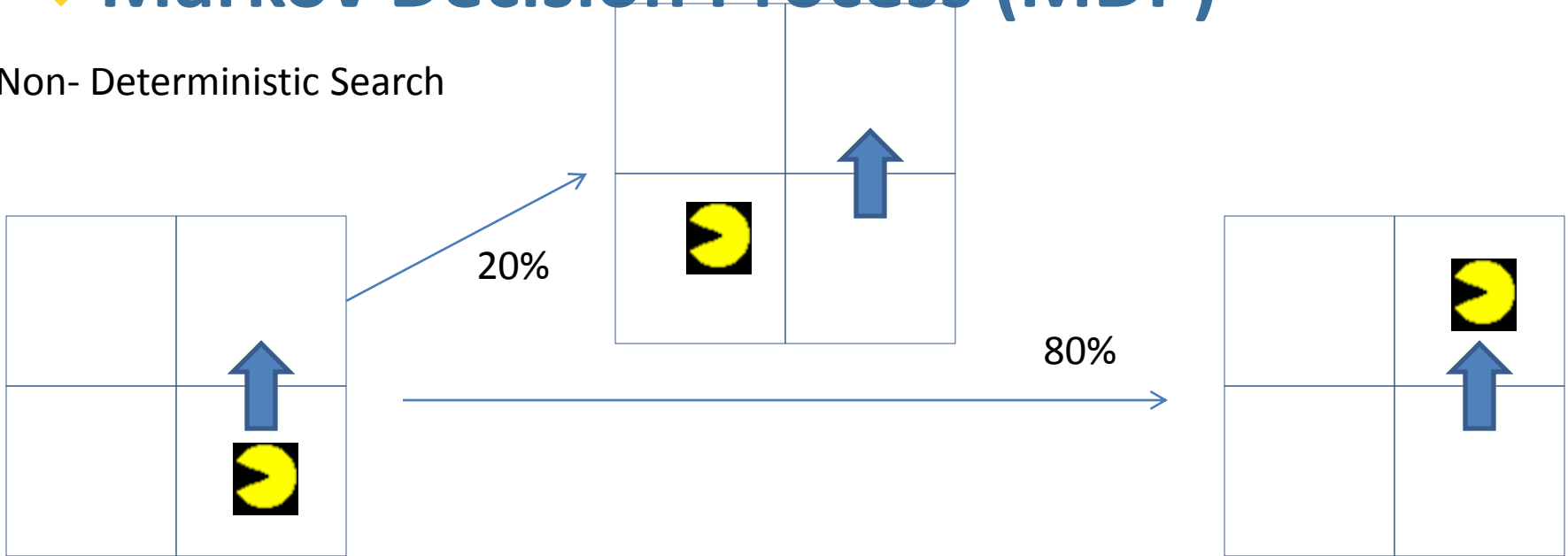
# ❖ Markov Decision Process (MDP)

Deterministic Search



# ❖ Markov Decision Process (MDP)

Non- Deterministic Search



## ❖ Markov Process (MP)

A stochastic process has the Markov Property if the conditional probability distribution of the future states of the process depends only upon the present state, not on the sequence of events that preceded it. A process with this property is called a Markov process

## ❖ Markov Decision Process (MDP)

MDP provide a mathematical framework for modeling decision making in situations where outcomes are partly random and partly under the control of a decision maker



## ❖ Bellman Equation

$$0.8 V(s1') + 0.2V(s2')$$

$$V(s) = \max(R(s,a) + \underbrace{\gamma V(s')})$$

$$V(s) = \max_a \left( R(s, a) + \gamma \sum_{s'} P(s, a, s') V(s') \right)$$

## ❖ Policy Vs Plan

MDP provide a mathematical framework for modeling decision making in situations where outcomes are partly random and partly under the control of a decision maker



## Chapter: Use Cases

## ✦ Use cases

Smarter healthcare

Traffic control

JIT manufacturing

Trading analytics

Search quality

Customer behaviour

Fraud detection

Customer experience management

Predictive maintenance

Churn analysis

Security anomaly detection

## ❖ Use Cases

Reinforcement Learning

- Autonomous systems

Generative

- Image enhancing

Predictive Maintenance

- networks
- Systems
- Human body