

Model-based learning

the other RL technic

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Model-based learning

Main Idea:

- ▶ Random trajectories (a lot)
- ▶ Until each transition is visited several times.
- ▶ Compute an optimal policy.

Potentially:

- ▶ Require driving exploration
- ▶ Only incomplete exploration can be performed

Markov Decision Process

MDP: $\langle S, A, T, R \rangle$:

S : set of system's states

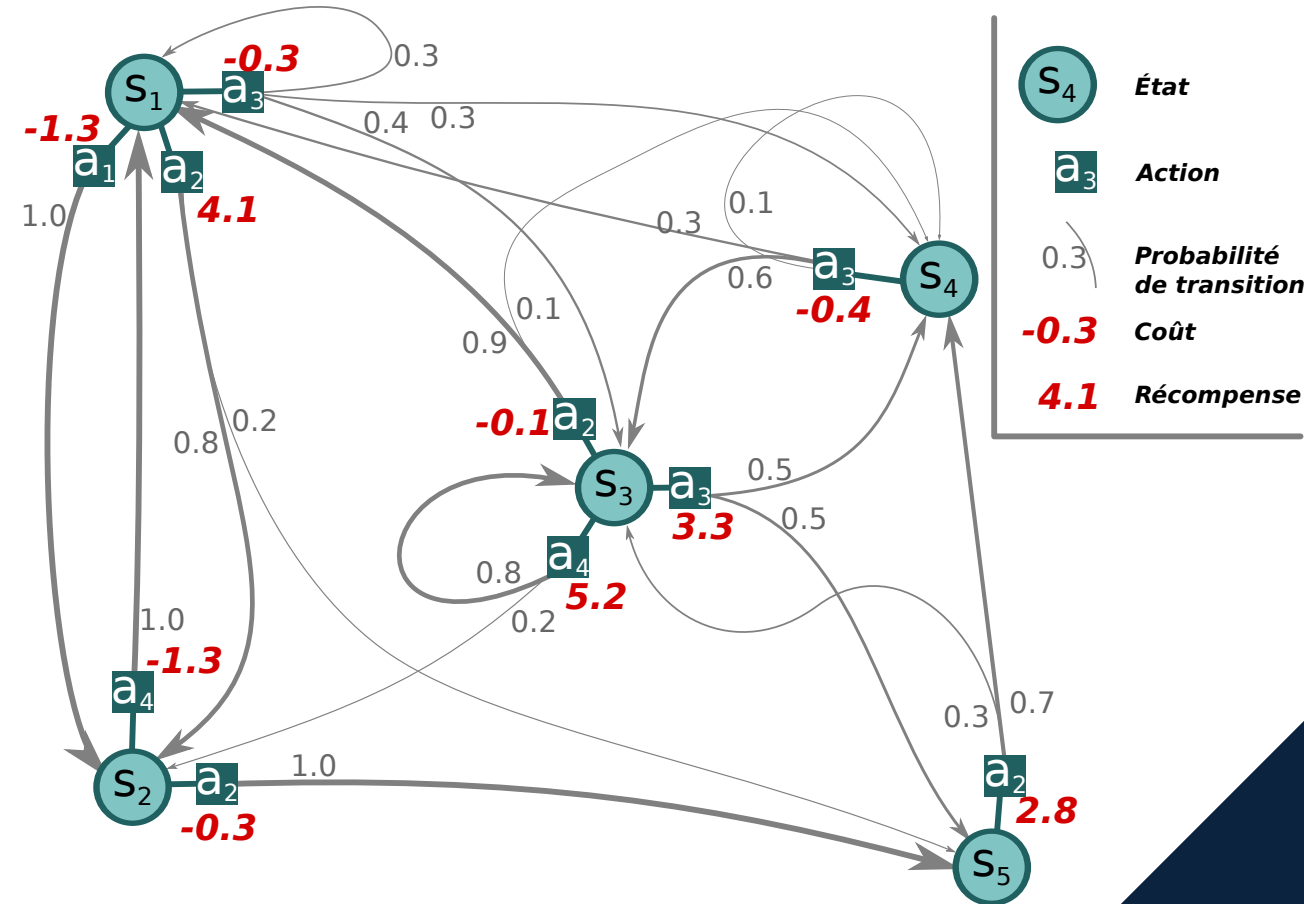
A : set of possible actions

$T: S \times A \times S \rightarrow [0, 1]$: transitions

$R: S \times A \rightarrow \mathbb{R}$: cost/rewards

Optimal policy:

The policy π^* maximizing Bellman



Solving MDP: Value Iteration

Input: an **MDP**: $\langle S, A, T, R \rangle$; precision error: ϵ ; discount factor γ ; initial $V(s)$

1. Repeat until the **maximal delta** $< \epsilon$

For each state $s \in S$

- Search the action a^* maximizing the Bellman Equation
- Update $\pi(s)$ and $V(s)$ by considering action a^*
- Compute the delta value between the previous and the new $V(s)$

Output: an optimal π^* and associated **V-values**.

$$V^\pi(s) = R(s, a) + \gamma \sum_{s' \in S} T(s, a, s') \times V^\pi(s')$$

Application to 421

- ▶ Python implementation - [playerMDP.py](#) :

```
solver= MDP()  
solver.learnModel( Engine() )  
solver.valueIteration()
```

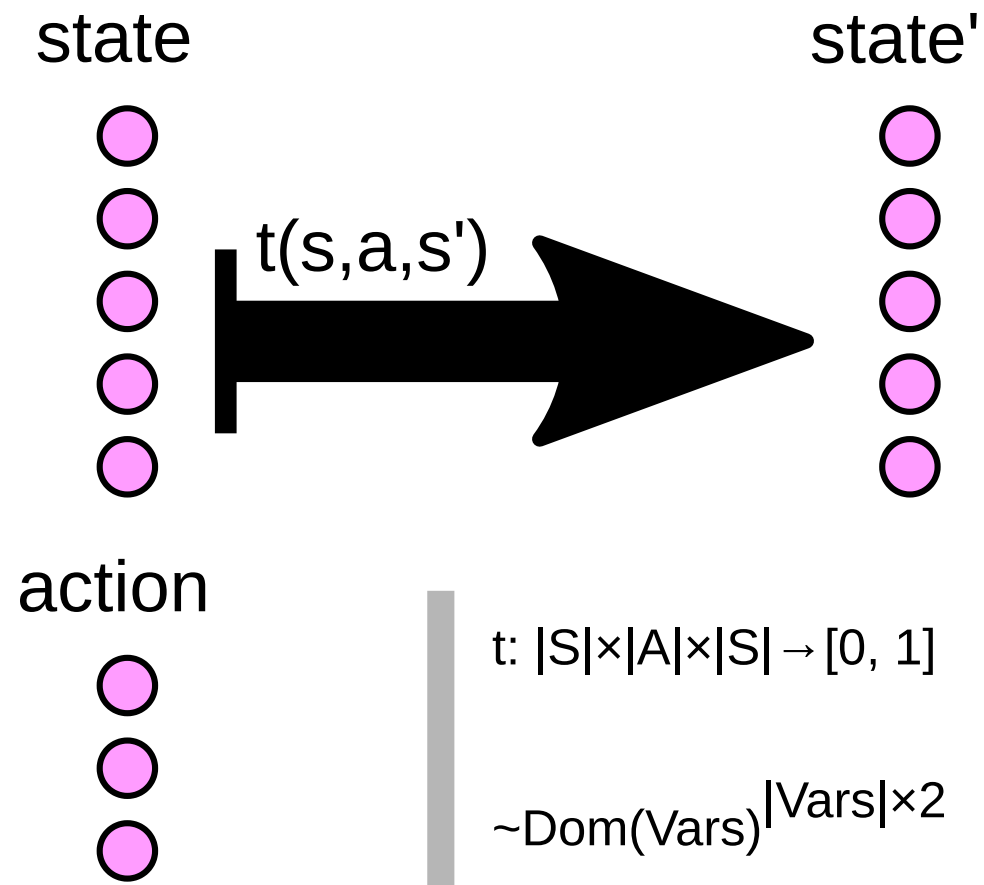
```
player= PiPlayer( solver.policy() )
```

- ▶ Learning phase: Estimate **t** and **r**:
 - **10 000** simulations for each couple (s, a)
- ▶ Value iteration:
 - **3** iterations (directed and finit game)
- ▶ Average score (100 000 games): **~338**

(To notice: decreasing the learning phase impact the average score)

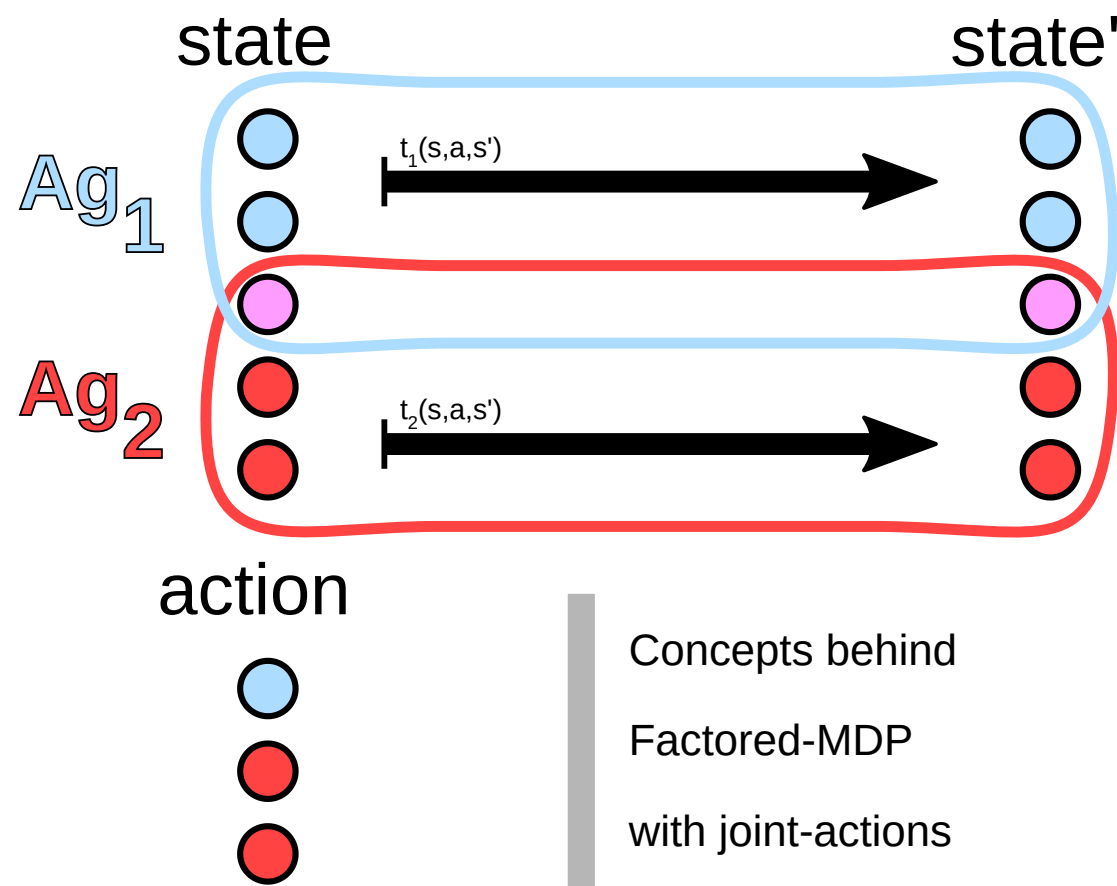
The Curse of Dimensionality

Fonction de Transition:



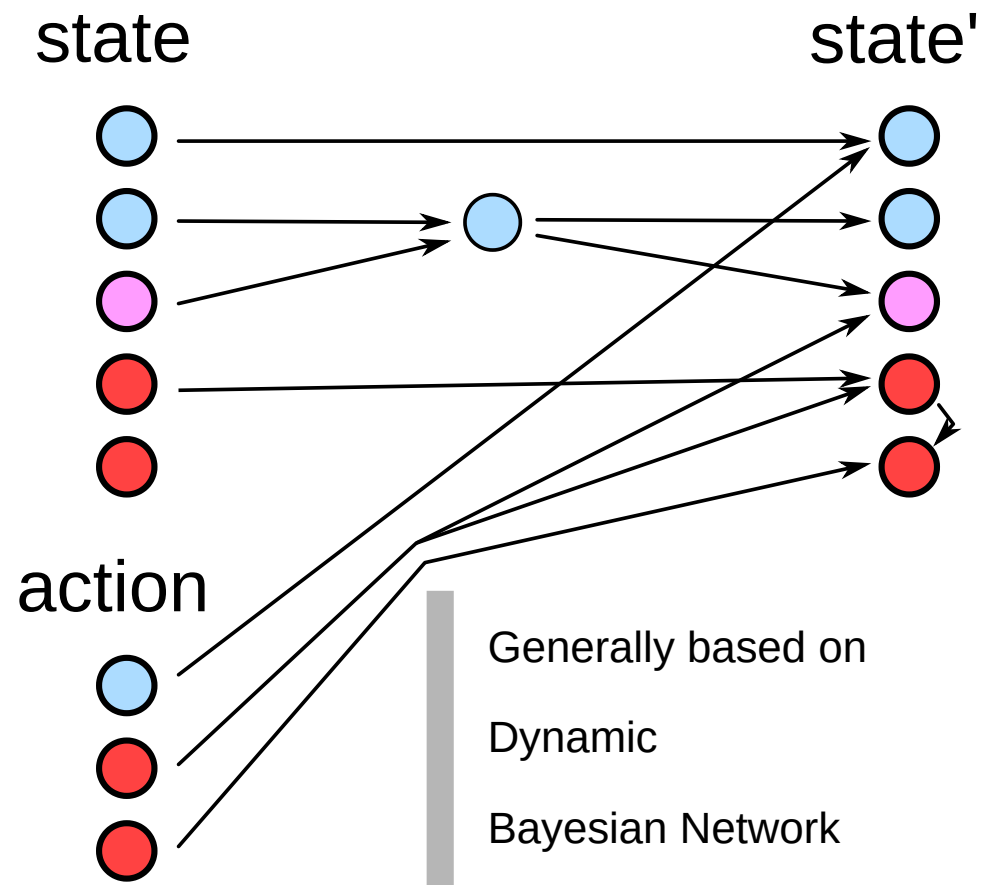
Le fléau de la dimension

Factored Transition function:



Le fléau de la dimension

Factored Transition function:



Example: Zombie Dice



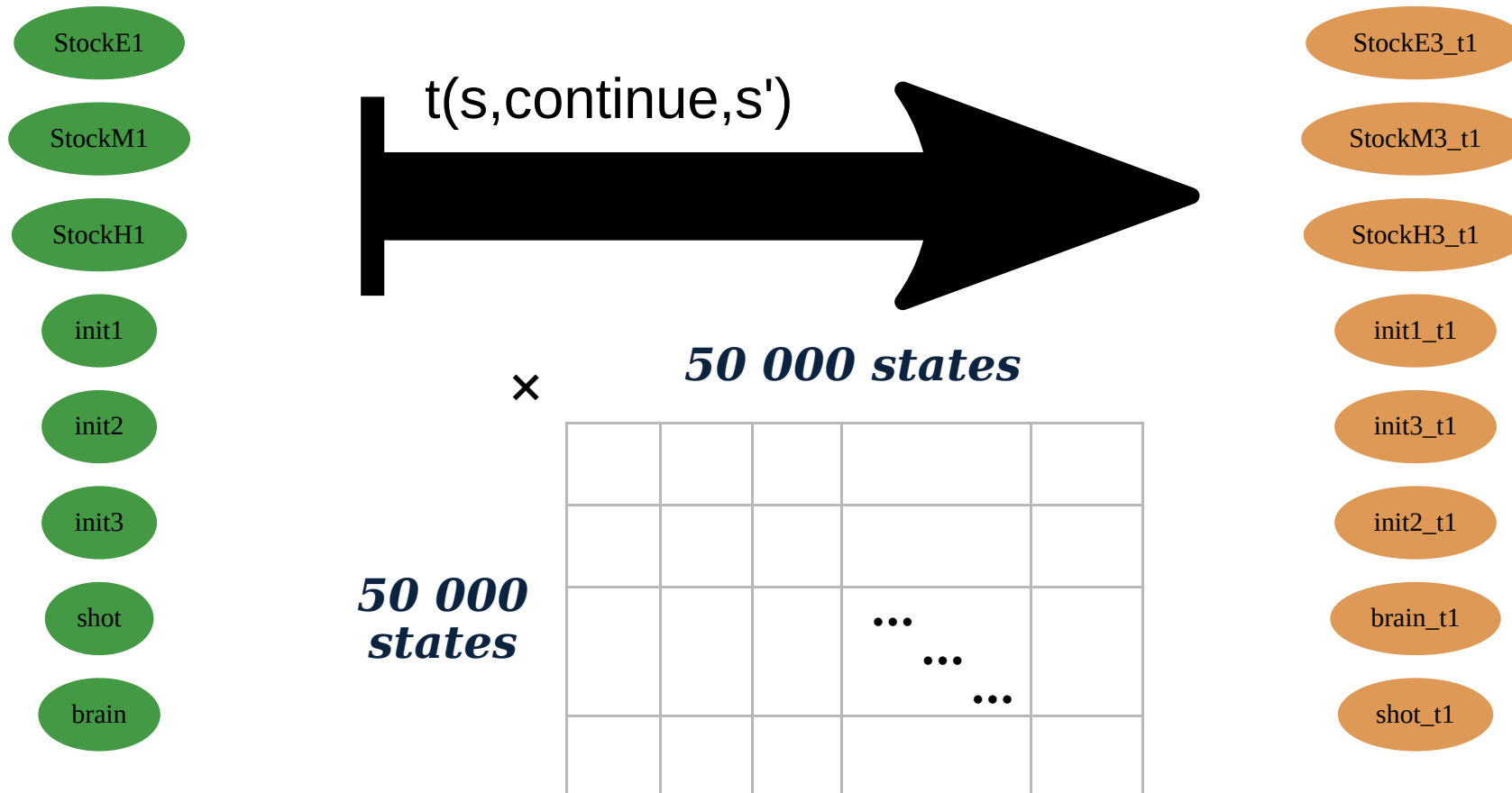
Eat maximum brains

without dying (3 damages)

- ▶ Players are zombies.
- ▶ They try to catch humans three at a time.
- ▶ Humans are dice with probability to fight back.

Example: Zombie Dice

Matrice complète



Example: Zombie Dice

Dynamic Bayesian Network (Continue)

