

# The Curse of Dimensionality

## Factorized Transitions

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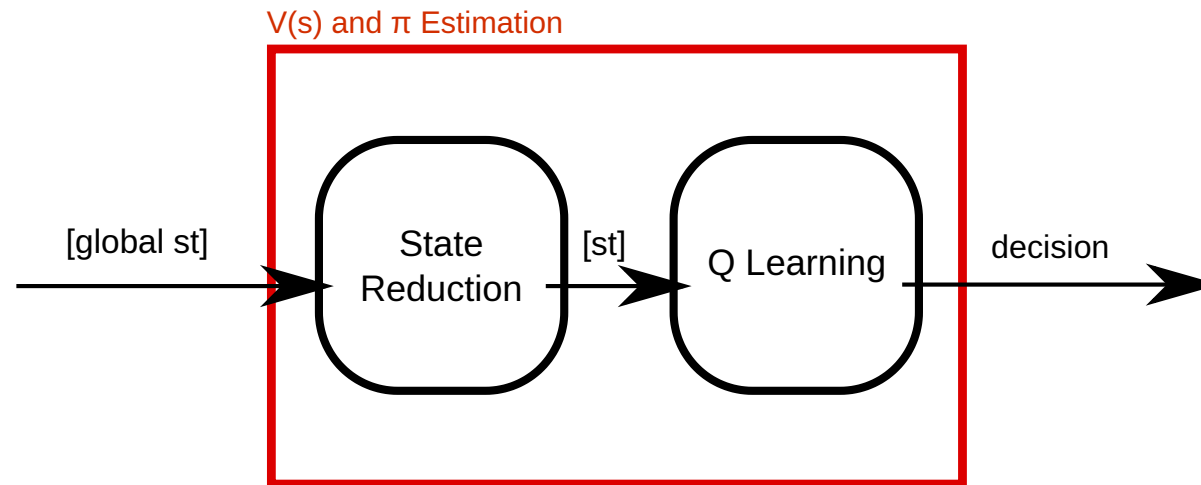


**IMT Nord Europe**  
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1. **Bayesian Network**
2. **Factorized Transition Function**

# Before to go...

Q-Learning over reduced state-space (and actions...)  
necessarily generate averaged decisions...



Need a model to refine decision making from simulations

1. **Bayesian Network**
2. Factorized Transition Function

# Bayesian Network: General idea

Variables are not necessarily correlated over all the others.

## Example:

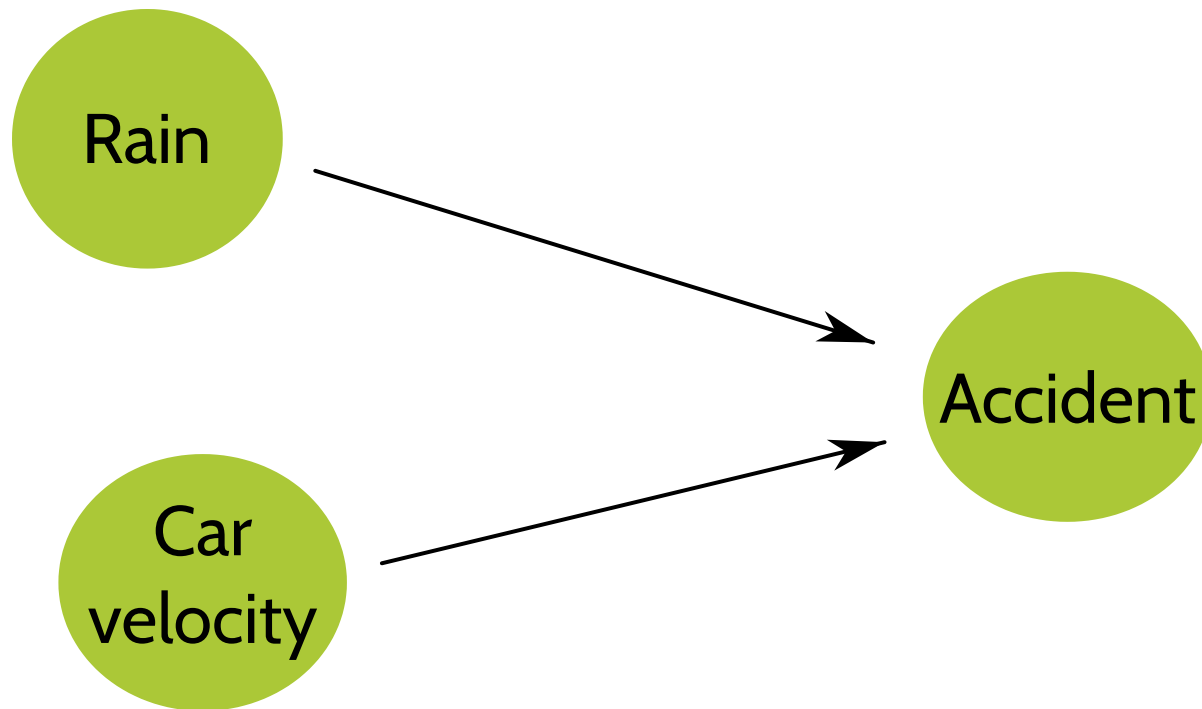
- ▶ The probability of *Rain* does not depend from *Car Velocity*.
- ▶ The *Car Velocity* does not depend from *Rain*.
- ▶ However, the probability of *Accident* depends from both *Rain* and *Car Velocity*.

## Bayesian Network:

A probabilistic graphical model that represents  
a set of variables and their conditional dependencies

# Bayesian Network: Car accident Example

Graphical model: *Node*: variable, *Orriented Edge*, dependancy



**ATTENTION:** *Correlation* is not *Conditional Dependancy*

# Bayesian Network: Node's Probability Table

Propbality dependencies are defined node by node.

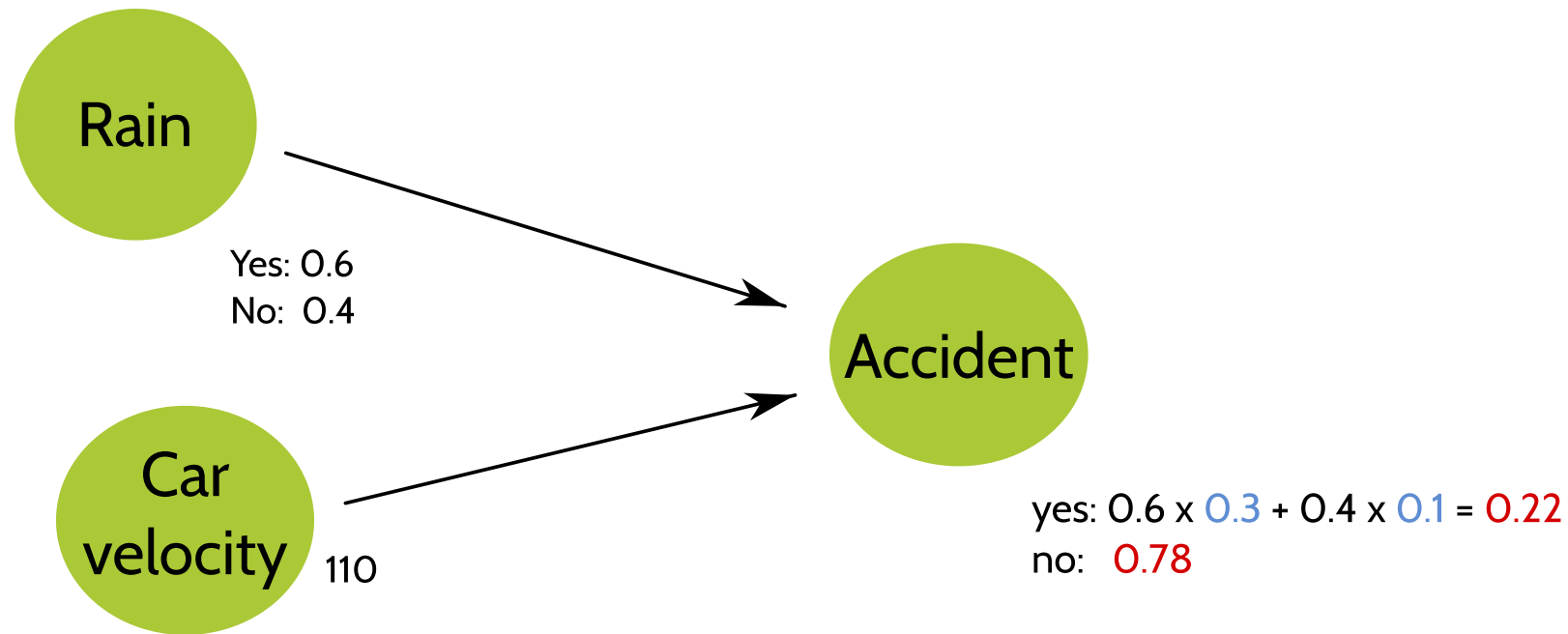


Accident

Rain	C.Vel.	Yes	No
Yes	50	0.1	0.9
	70	0.2	0.8
	90	0.2	0.8
	110	0.3	0.7
	130	0.8	0.2
No	50	0.05	0.95
	70	0.1	0.9
	90	0.1	0.9
	110	0.1	0.9
	130	0.3	0.7

# Bayesian Network: Inference

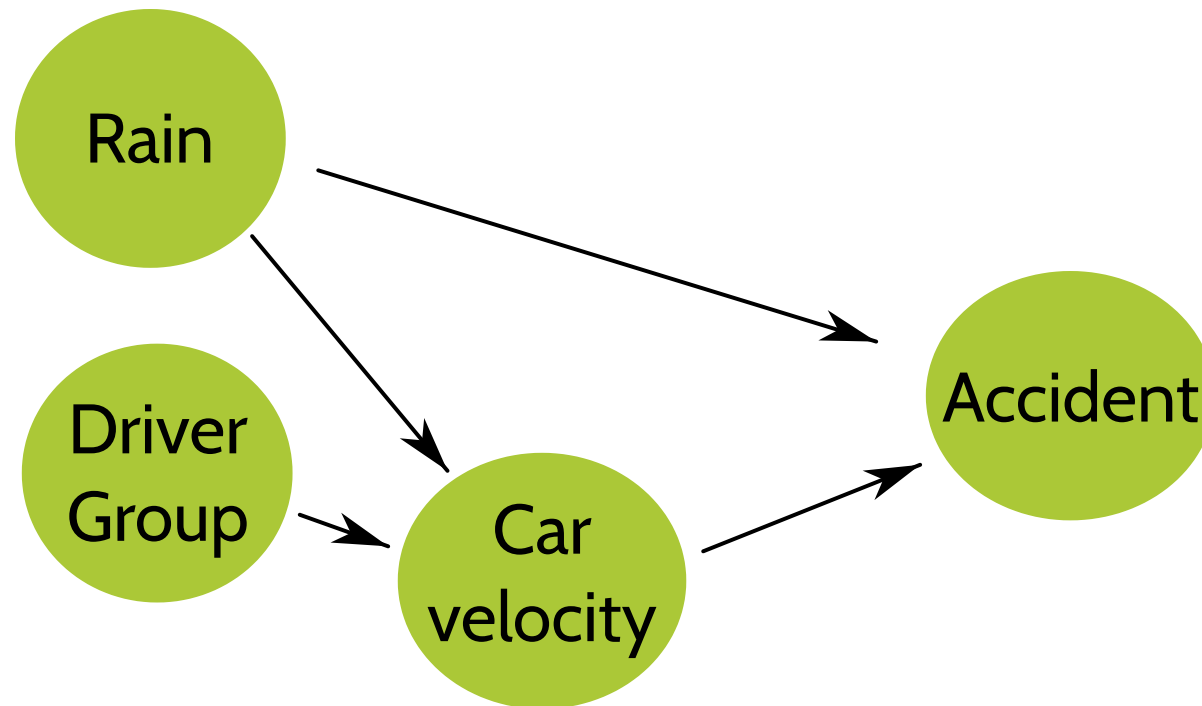
Resulting propbality is computed recursively





# Bayesian Network: Level-up

Growing the graph → no consequences over existing nodes



**ATTENTION:** *Correlation* is not *Conditional Dependancy* → **Directed Acyclic Graph**

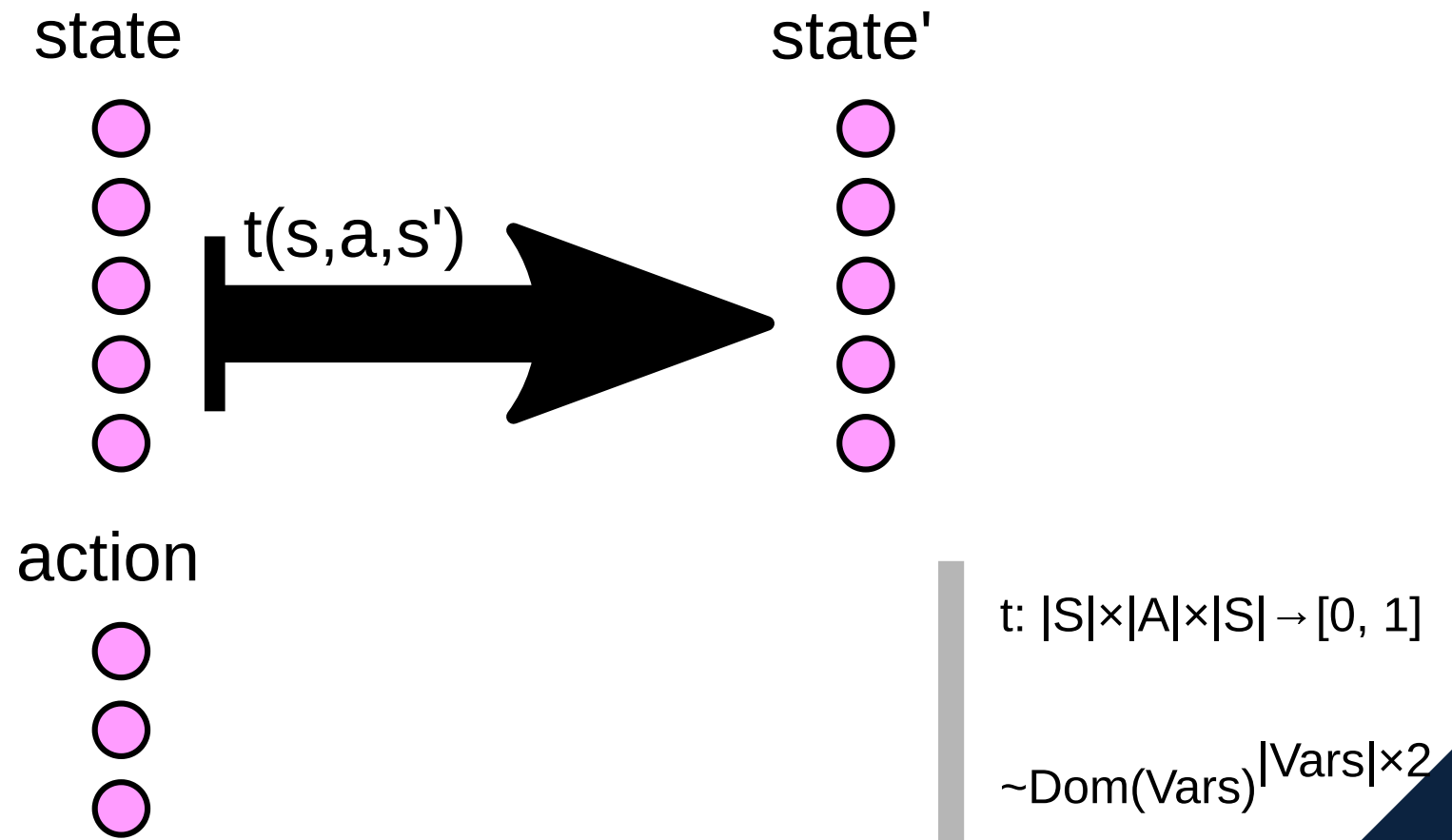
# Bayesian Network: Toolbox

- ▶ Bayesian Network [on Wikipedia](#)
- ▶ In python - numerous implementations
  - [pomegranate](#) - define compute inferences and more.
  - [bnlearn](#) - Learning the bayesian structure (ie. detect the dependencies)

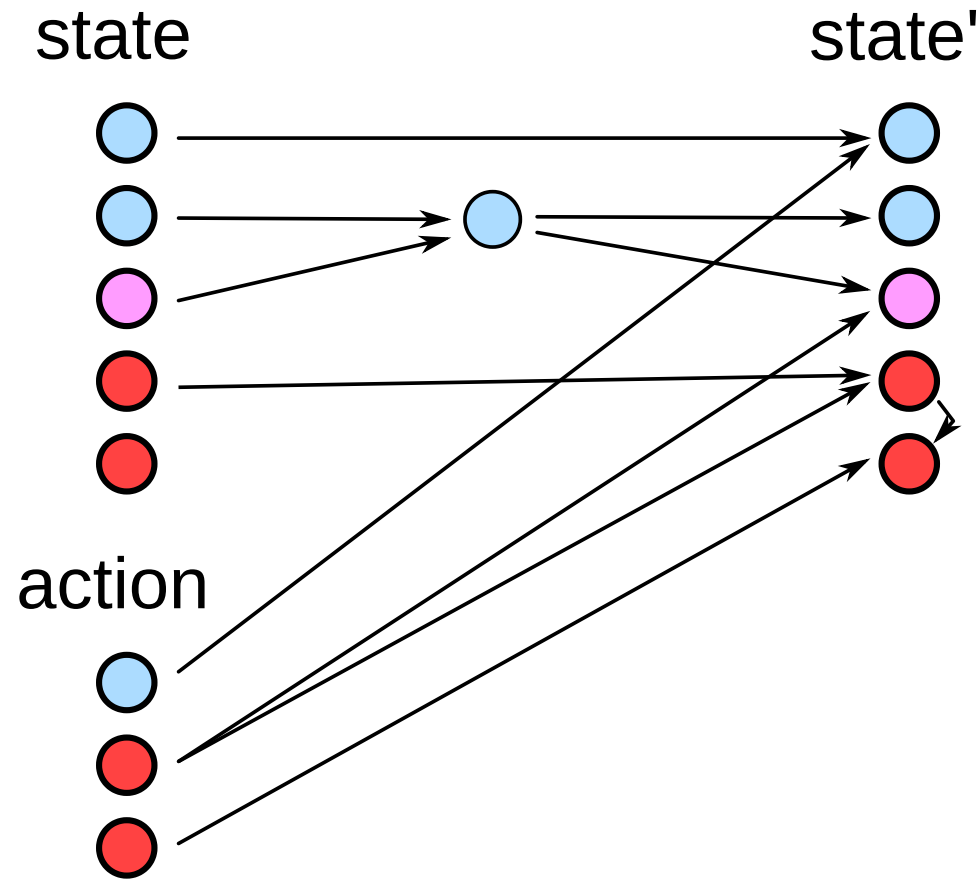
1. Bayesian Network
2. **Factorized Transition Function**

# Factorized Transition Function: Problem

Classically a multi-variables probabilistic evolution problem



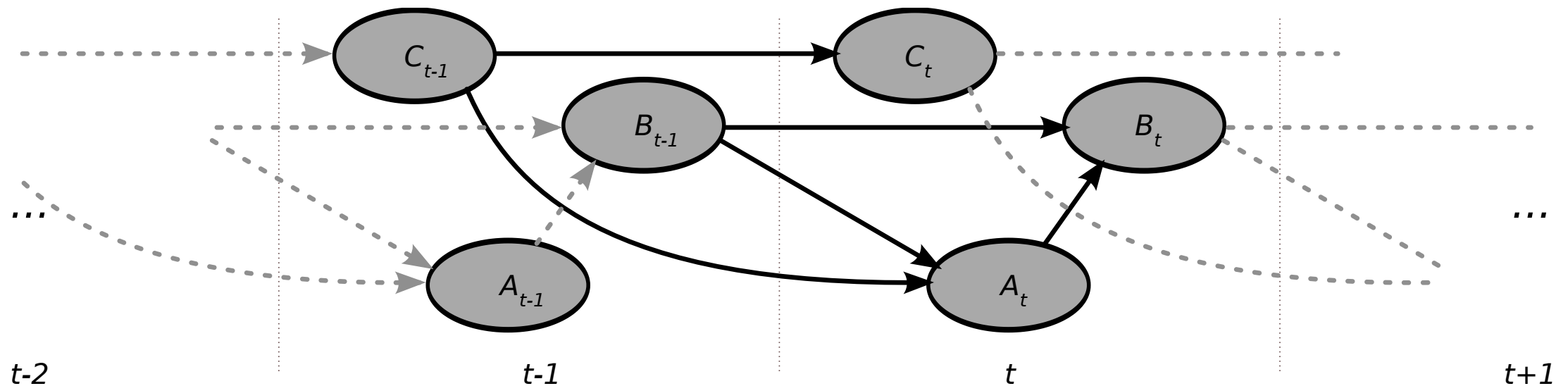
# Factorized Transition Function: As Bayesian Network



Each variable evolution is dependant from few parents  
Potentially with intermediate varaibles.

# Factorized Transition Function: Dynamic Bayesian Network

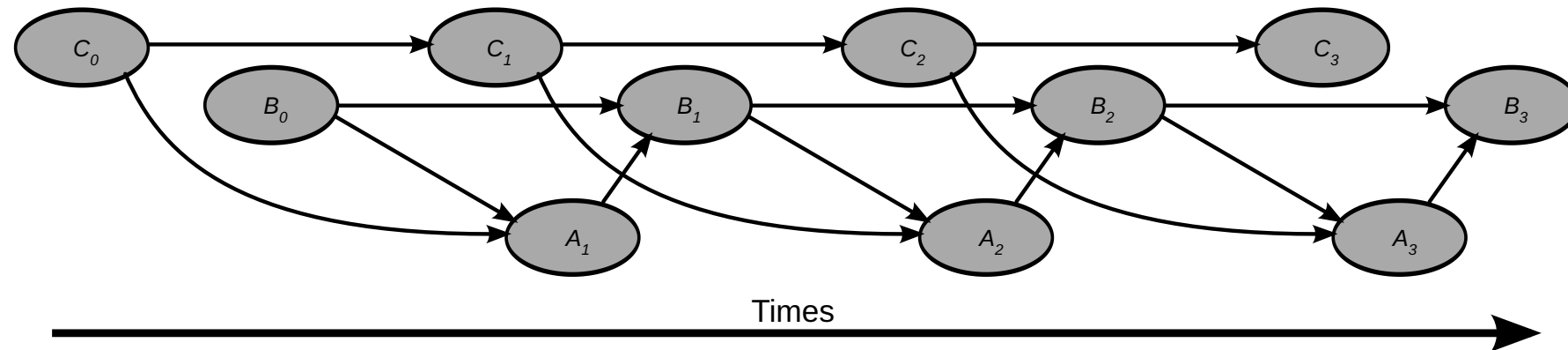
**Bayesian Network** with duplicated variables regarding their evolution in time.



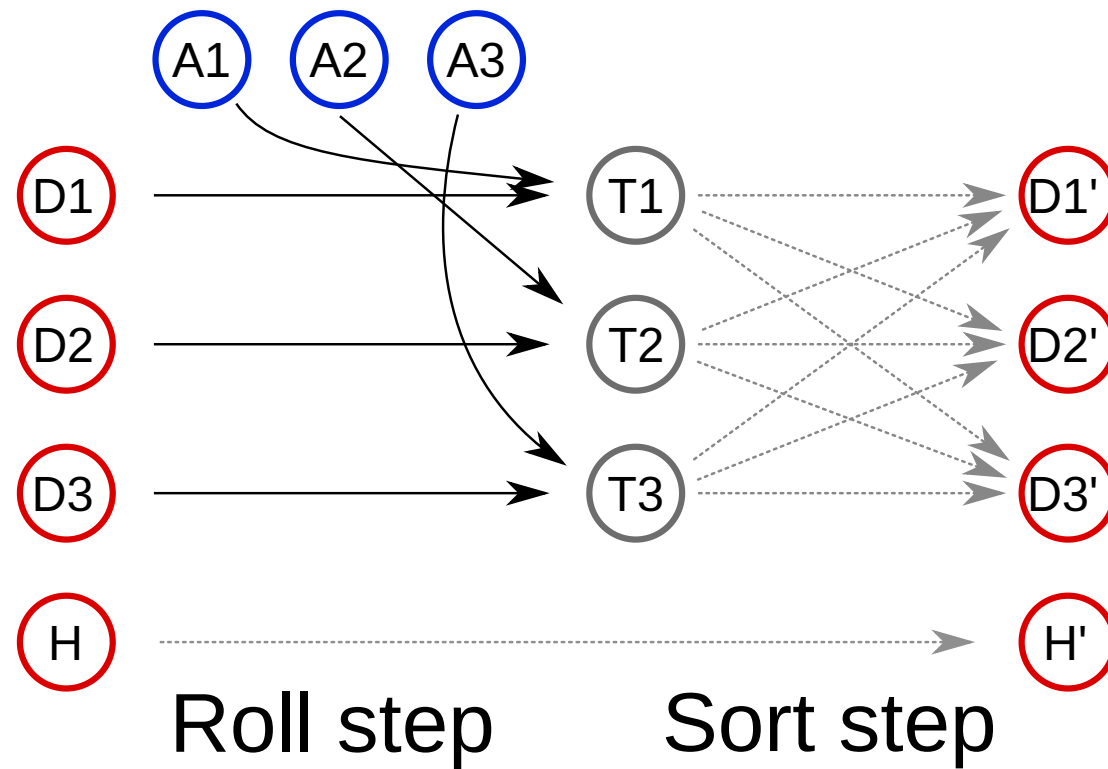
Considering a distribution of probability at time  $t$ ,  
it is possible to infer over the distribution of probabilities at time  $t + N$

# Factorized Transition Function: Dynamic Bayesian Network

Extended definition over 3 times step.



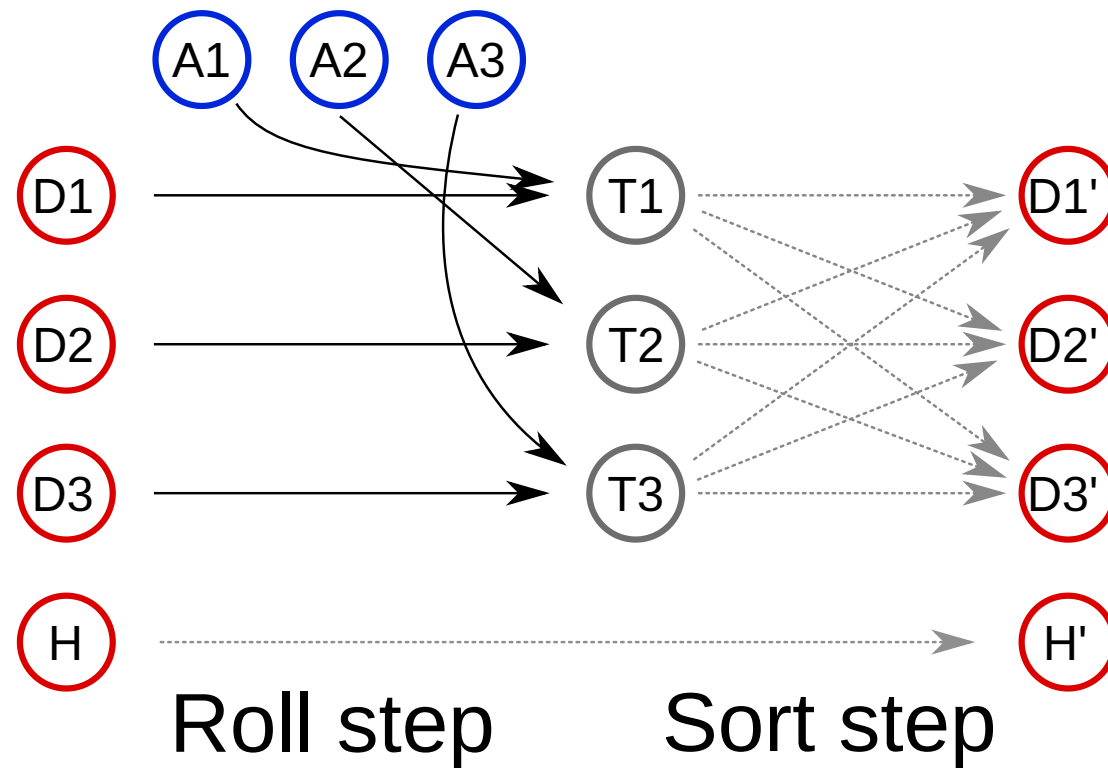
# Factorized Transition Function: 421 example



- Roll step: 2 entrances - Small tables
- Sort step: 3 entrances but deterministic



# Factorized Transition Function: 421 example



- ▶ Let consider cheat dice:  
Learning transitions  $\rightarrow$  computing  $3 \times 12$  probabilities vs  $(168^2 \times 8)$

# Factorized Transition Function: Zombie Dice Example

Naturally the complexity of the networks grow linearly with the complexity of the game



A stop and go game:

2 players or more (2 Zombies)

- The first one to cumulate 13 brains win.
- Player need to pass, before to get 3 damages

# Factorized Transition Function: Zombie Dice Exemple

## A 4 steps' game engine:



3 brains, 2 footprints, 1 shotgun



2 brains, 2 footprints, 2 shotguns



1 brain, 2 footprints, 3 shotguns

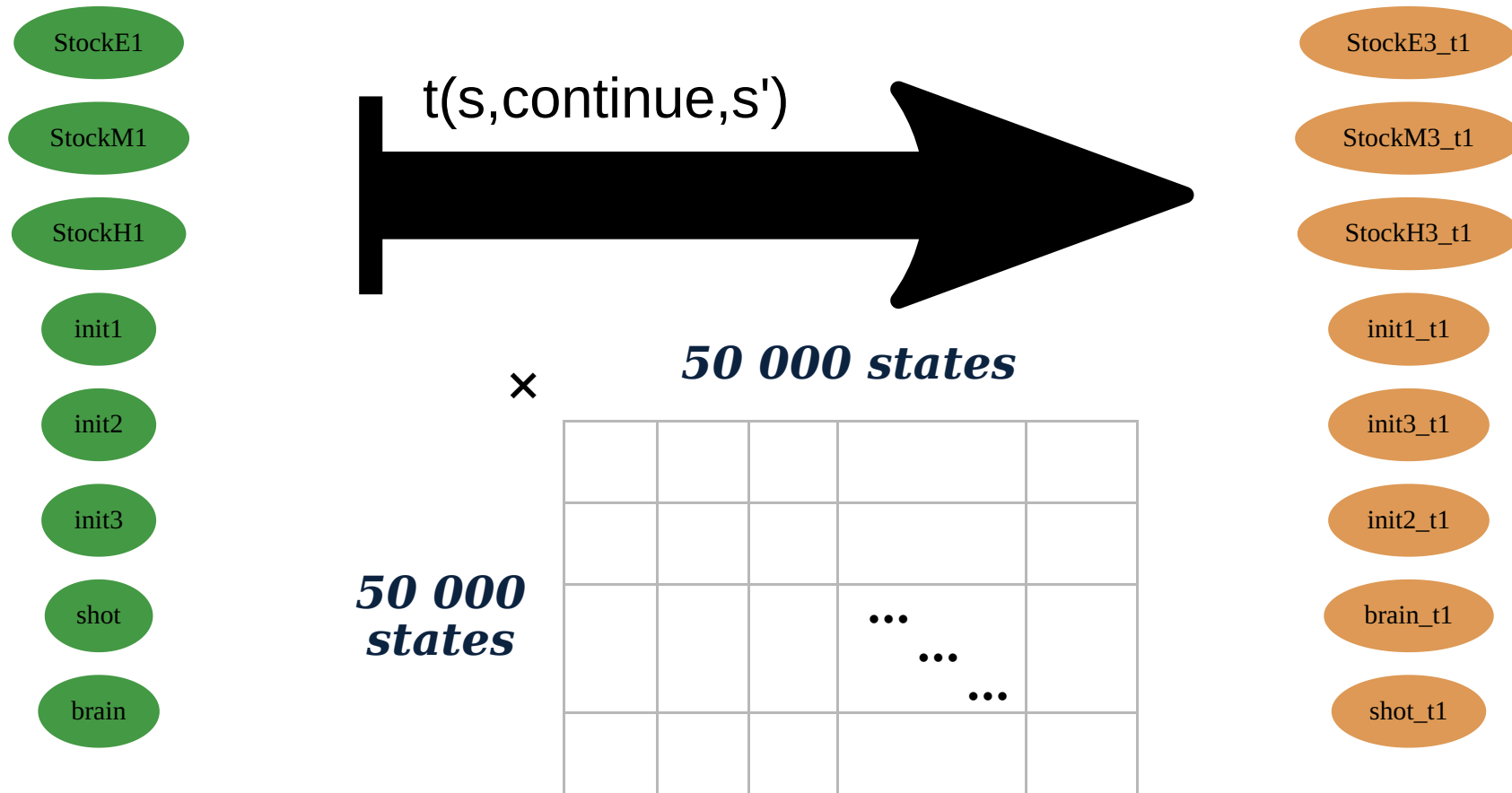
Complete Dice Set:



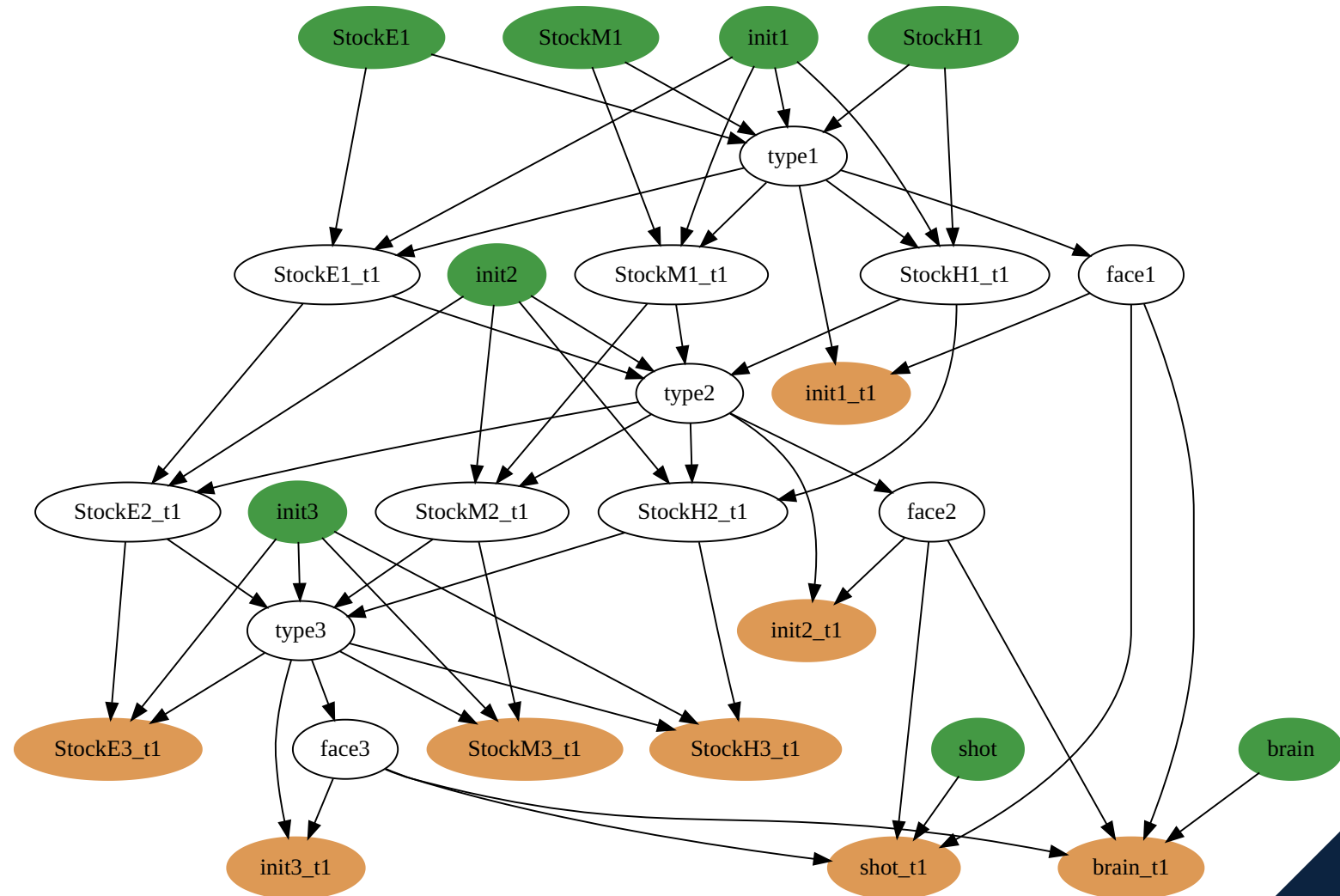
1. player decides to continue or not
2. new dice are picked randomly (to complete the hand)
3. dice are rolled
4. the player cumulates brains and damages

► State Space ? Branching ?

# Zombie Dice : Brut Transition Function (action: go)



# Zombie Dice : Factorized Transition Function (action: go)



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1. Bayesian Network
  2. Factorized Transition Function