

# Grammatical inference: an introductory example

Module X3IT040, Colin de la Higuera, Nantes & Le Mans, 2020



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Statistical and symbolic  
language modeling





# A synopsis

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- There are many runs of algorithms one can take as first examples of what a grammatical inference algorithm is about.
- The following one is « fun » and contains many key ideas.



# Bibliography

D. Carmel and S. Markovitch. Model-based learning of interaction strategies in multi-agent systems. *Journal of Experimental and Theoretical Artificial Intelligence*, 10(3):309-332, 1998

D. Carmel and S. Markovitch. Exploration strategies for model-based learning in multiagent systems. *Autonomous Agents and Multi-agent Systems*, 2(2):141-172, 1999





# The problem:

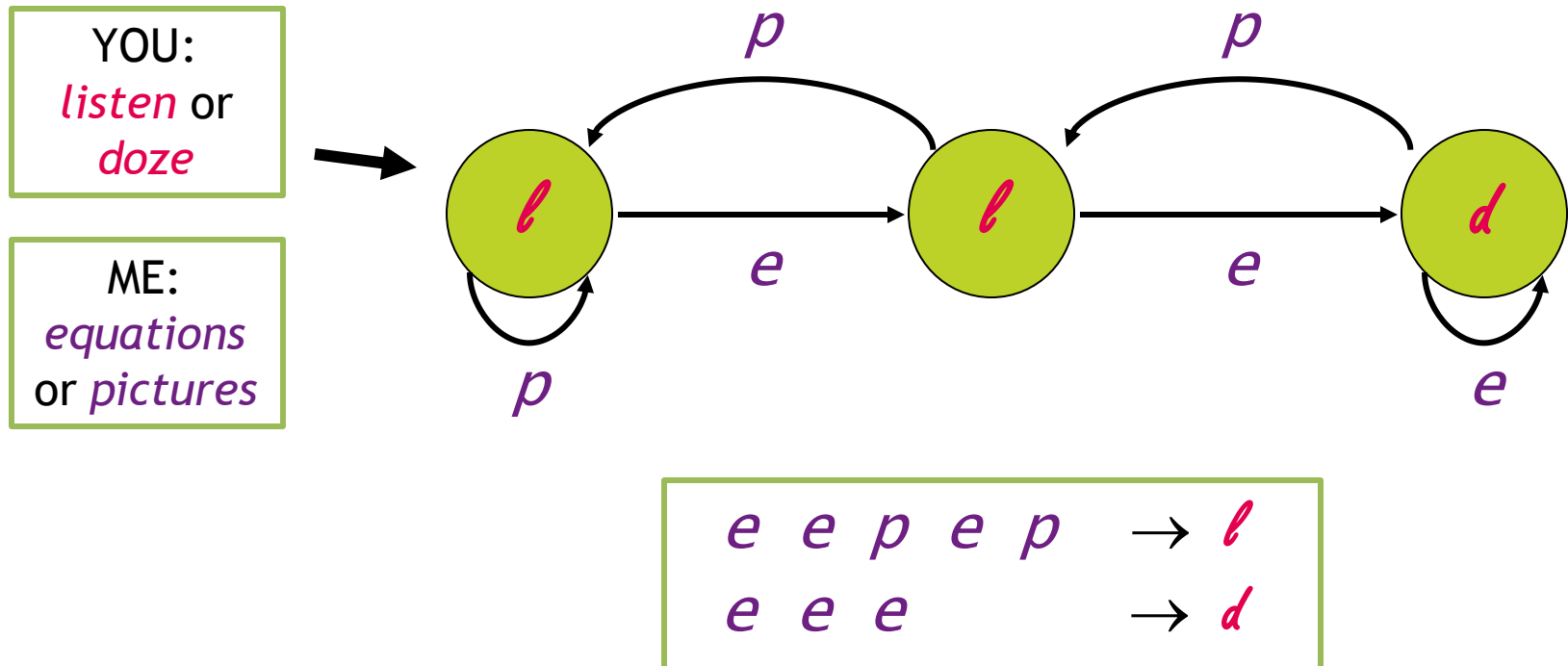
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- An agent must take cooperative decisions in a multi-agent world
- His decisions will depend:
  - on what he hopes to win or lose
  - on the actions of other agents



# Hypothesis:

The opponent follows a rational strategy (given by a *DFA* / Moore machine)





## Example:

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- Each prisoner can admit (*a*) or stay silent (*s*)

### The prisoner's dilemma

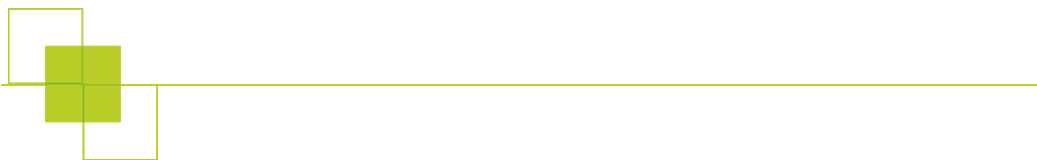
- If both admit: 3 years (prison) each
- If A admits but not B: A=0 years, B=5 years
- If B admits but not A: B=0 years, A=5 years
- If neither admits: 1 year each



# Example:

		B	
A		<i>a</i>	<i>s</i>
	<i>a</i>	<div>-3</div> <div>-3</div>	<div>-5</div> <div>0</div>
	<i>s</i>	<div>0</div> <div>-5</div>	<div>-1</div> <div>-1</div>





- In our version we study an iterated version against an opponent who follows a rational strategy
- Gain Function: limit of means (average over a very long series of moves)
- For example, if we get into a recurrent situation where we both admit, the gain will be -3







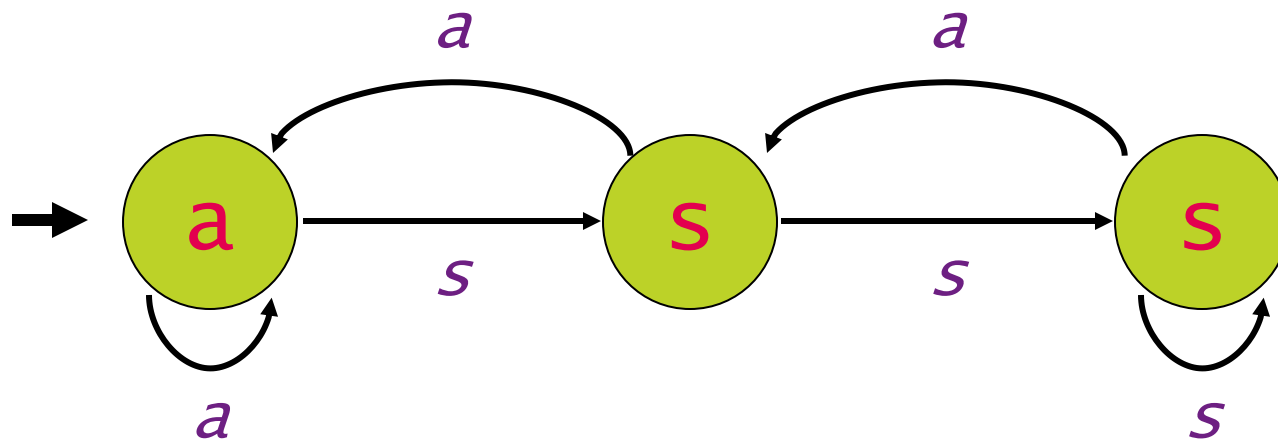
# The general problem

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- We suppose that the strategy of the opponent is given by a deterministic finite automaton (DFA)
- Can we imagine an optimal strategy?



# Running example





# Running example

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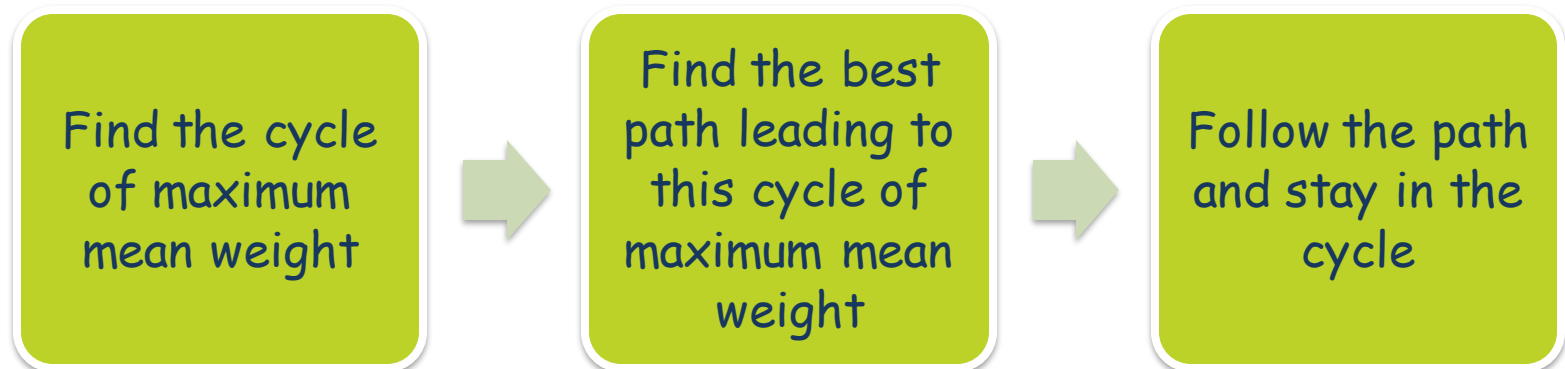
- Then (game theory):

Suppose we know the opponent's strategy

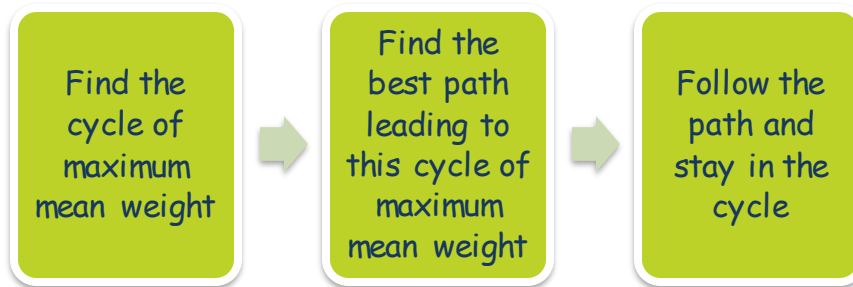
- Consider the opponent's graph in which we value the edges by *our own gain* and find the best (infinite) path in the graph



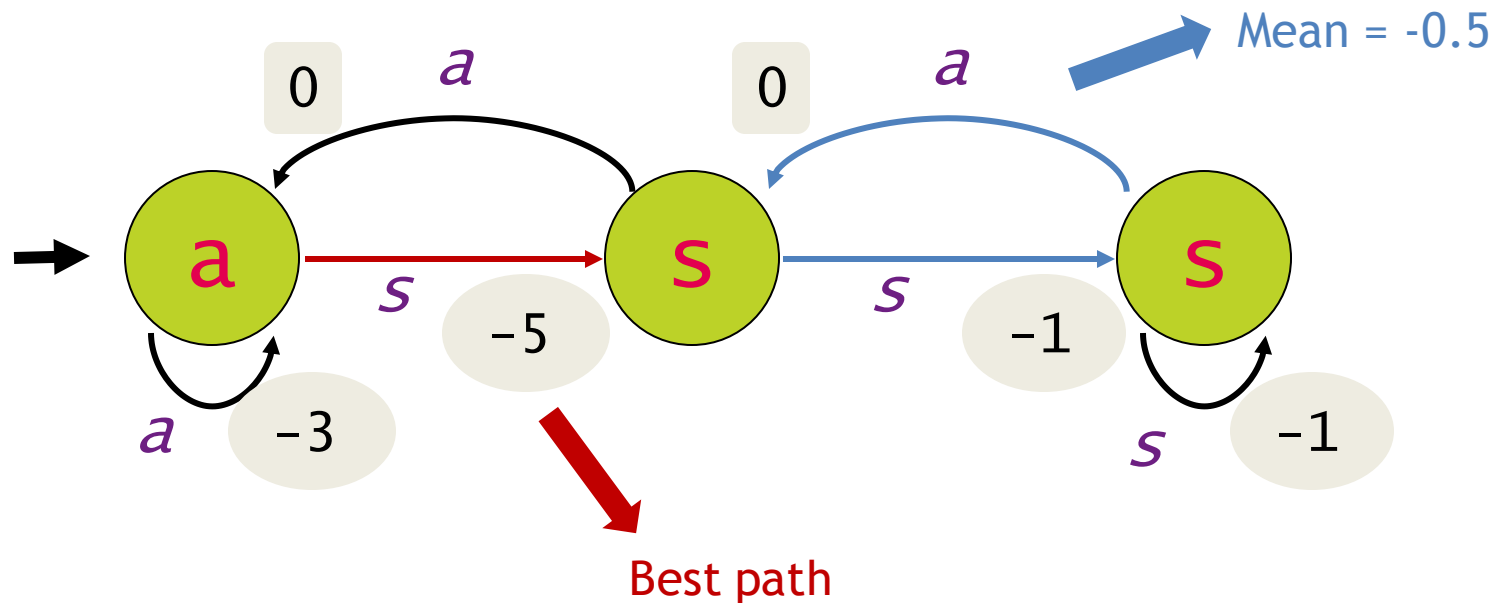
# Running example



# Running example



	<i>a</i>	<i>s</i>
<i>a</i>	-3	0
<i>s</i>	-5	-1





# Question

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Can we play a game against this opponent and...

can we then **reconstruct** his strategy ?



# The data (him, me)

HIM	ME
<i>a</i>	<i>a</i>
<i>a</i>	<i>s</i>
<i>s</i>	<i>a</i>
<i>a</i>	<i>a</i>
<i>a</i>	<i>s</i>
<i>s</i>	<i>s</i>
<i>s</i>	<i>s</i>
<i>s</i>	<i>a</i>
<i>s</i>	<i>a</i>

If I play *asa*, his move is *a*

$\lambda \rightarrow a$

$a \rightarrow a$

$as \rightarrow s$

$asa \rightarrow a$

$asaa \rightarrow a$

$asaas \rightarrow s$

$asaass \rightarrow s$





# The logic of the algorithm

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- The goal is to be able to parse and to have a partial solution consistent with the data
- The algorithm is loosely inspired by a number of grammatical inference algorithms
- It is greedy



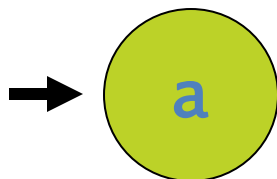


# The algorithm

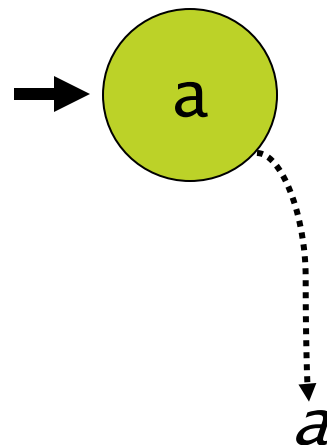
## The first decision

$\lambda \rightarrow a$   
 $a \rightarrow ?$

Sure:

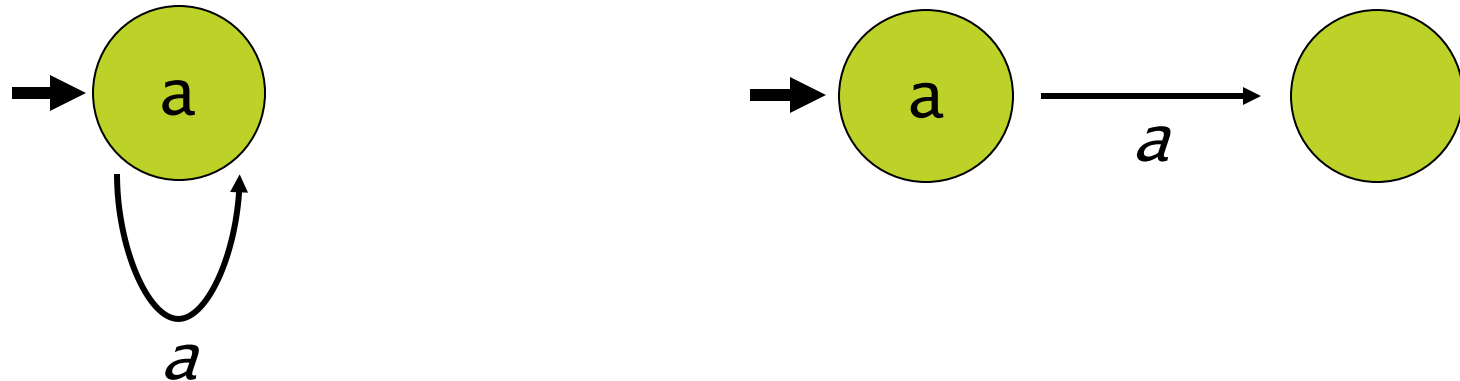


Have to deal with:



# The algorithm

## The candidates



### Occam's razor

*Entia non sunt multiplicanda praeter necessitatem*  
"Entities should not be multiplied unnecessarily"

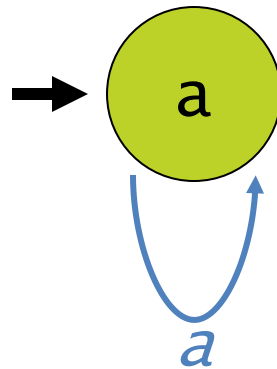


# The algorithm

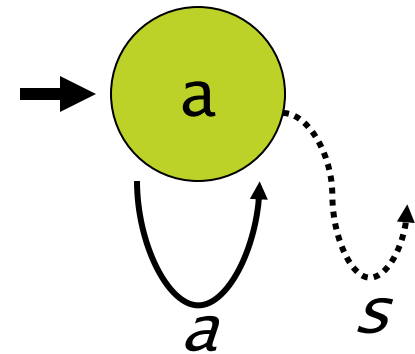
## The second decision

$\lambda \rightarrow a$   
 $a \rightarrow a$   
 $as \rightarrow ?$

Sure:



Have to deal with:

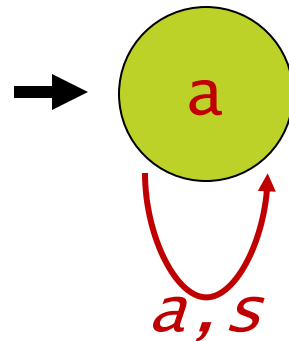


# The algorithm

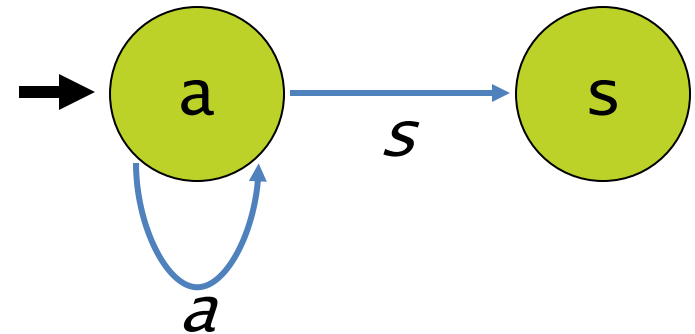
## The third decision

$\lambda \rightarrow a$   
 $a \rightarrow a$   
 $as \rightarrow s$   
 $asa \rightarrow ?$

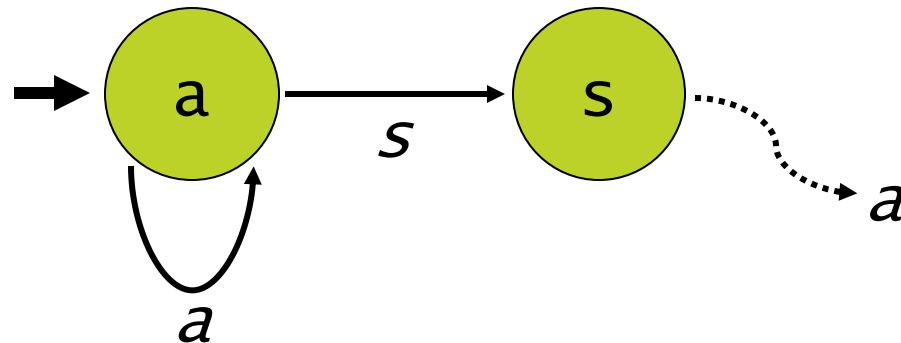
Inconsistent:



Consistent:

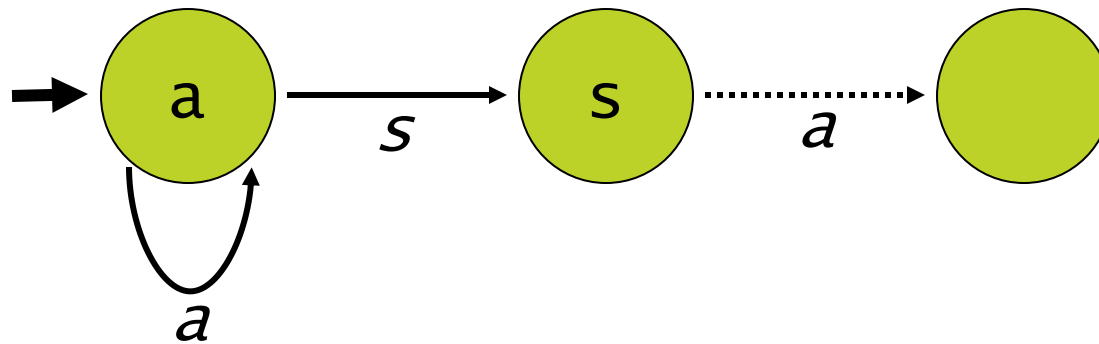
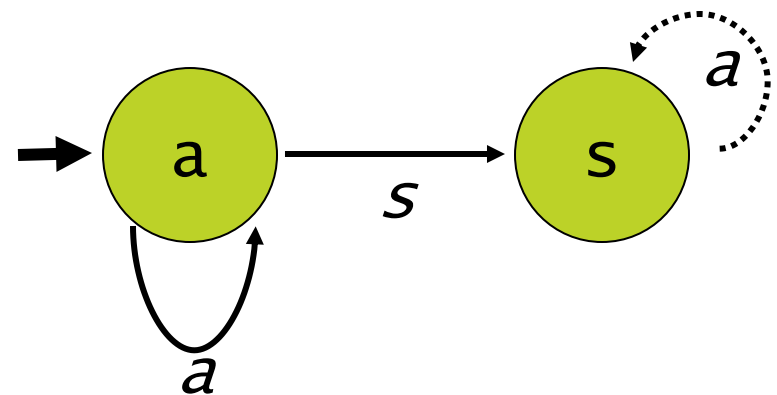
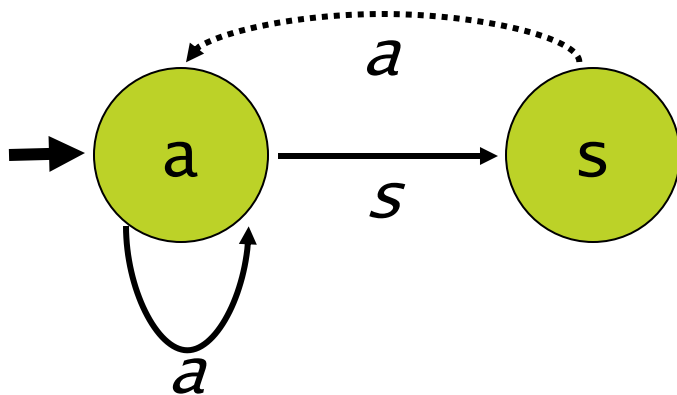


Have to deal with:



# The algorithm

## The three candidates



# The algorithm

## The fourth decision

Consistent:

$\lambda \rightarrow a$

$a \rightarrow a$

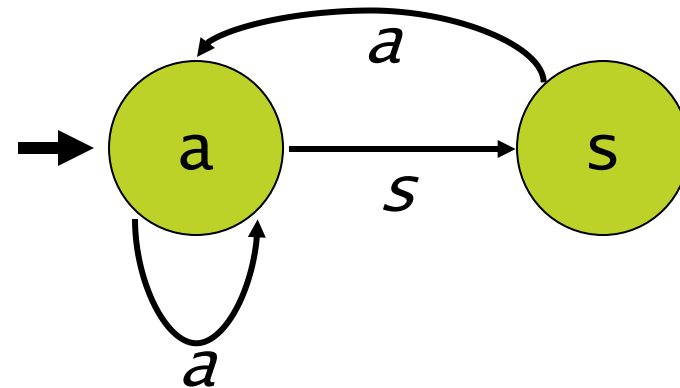
$as \rightarrow s$

$asa \rightarrow a$

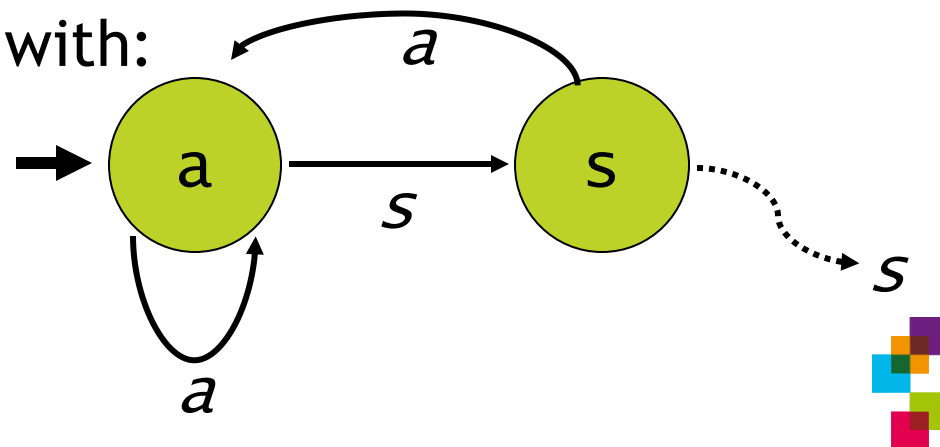
$asaa \rightarrow a$

$asaas \rightarrow s$

$asaass \rightarrow ?$



Have to deal with:



# The algorithm

## The fifth decision

Inconsistent:

$\lambda \rightarrow a$

$a \rightarrow a$

$as \rightarrow s$

$asa \rightarrow a$

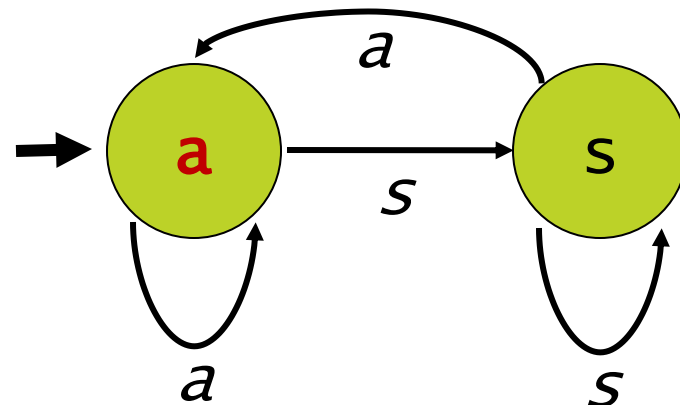
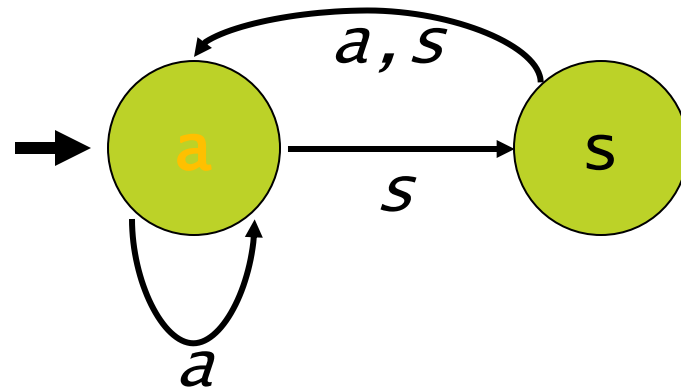
$asaa \rightarrow a$

$asaas \rightarrow s$

$asaass \rightarrow s$

$asaasss \rightarrow s$

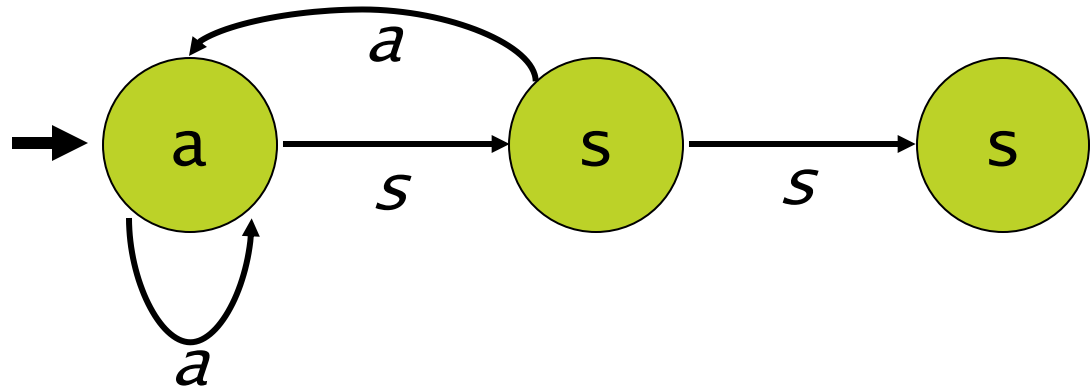
$asaasssa \rightarrow s$



# The algorithm

## The fifth decision

Consistent:



$\lambda \rightarrow a$

$a \rightarrow a$

$as \rightarrow s$

$asa \rightarrow a$

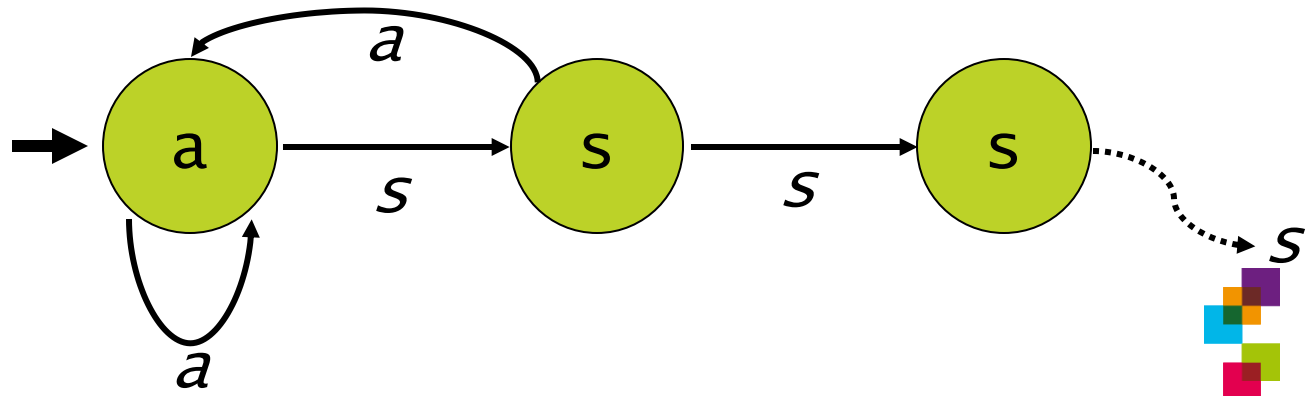
$asaa \rightarrow a$

$asaas \rightarrow s$

$asaass \rightarrow s$

$asaasss \rightarrow ?$

Have to deal with:





# The algorithm

## The sixth decision

Inconsistent:

$\lambda \rightarrow a$

$a \rightarrow a$

$as \rightarrow s$

$asa \rightarrow a$

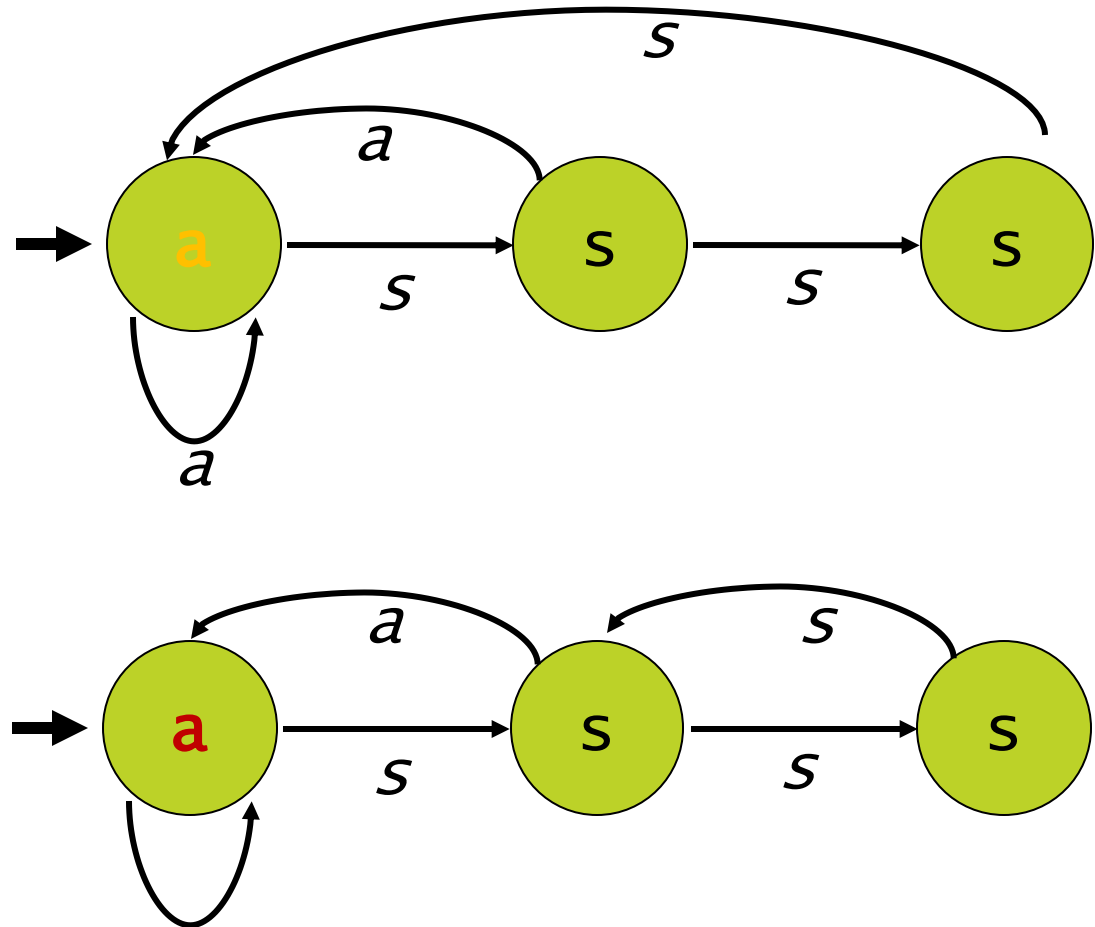
$asaa \rightarrow a$

$asaas \rightarrow s$

$asaass \rightarrow s$

$asaasss \rightarrow s$

$asaasssa \rightarrow s$



# The algorithm

## The sixth decision

Consistent:

$\lambda \rightarrow a$

$a \rightarrow a$

$as \rightarrow s$

$asa \rightarrow a$

$asaa \rightarrow a$

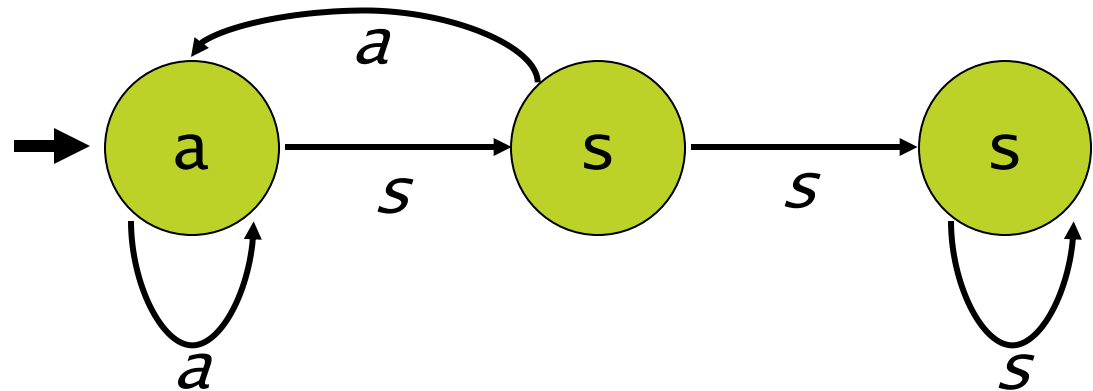
$asaas \rightarrow s$

$asaass \rightarrow s$

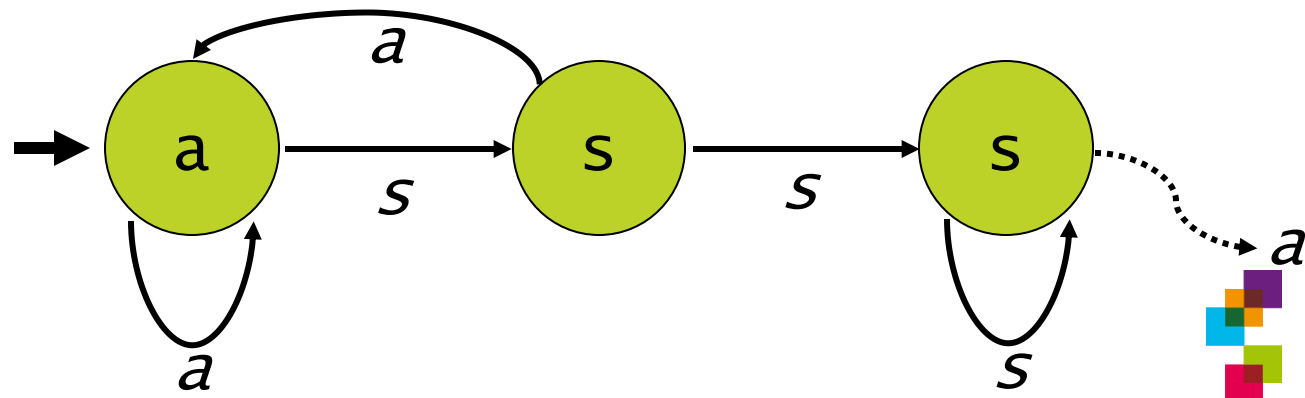
$asaasss \rightarrow s$

$asaasssa \rightarrow s$

$asaasssa \rightarrow ?$



Have to deal with:



# The algorithm

## The seventh decision

$\lambda \rightarrow a$

$a \rightarrow a$

$as \rightarrow s$

$asa \rightarrow a$

$asaa \rightarrow a$

$asaas \rightarrow s$

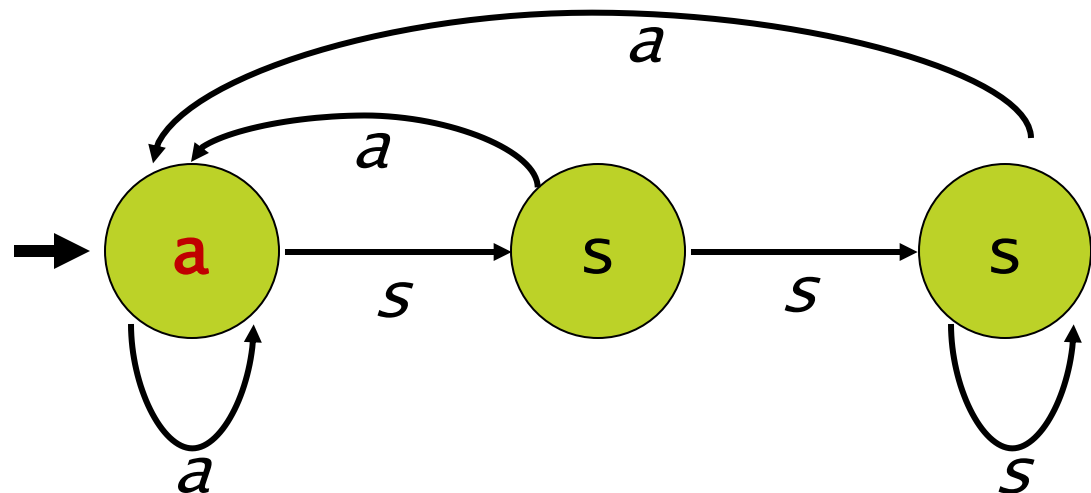
$asaass \rightarrow s$

$asaasss \rightarrow s$

$asaasssa \rightarrow s$

**$asaasssa \rightarrow s$**

Inconsistent:



# The algorithm

## The seventh decision

$\lambda \rightarrow a$

$a \rightarrow a$

$as \rightarrow s$

$asa \rightarrow a$

$asaa \rightarrow a$

$asaas \rightarrow s$

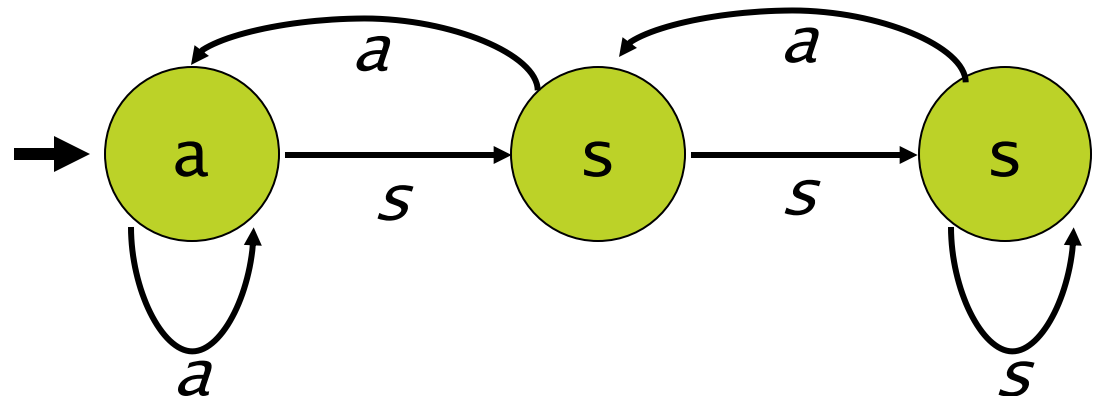
$asaass \rightarrow s$

$asaasss \rightarrow s$

$asaasssa \rightarrow s$

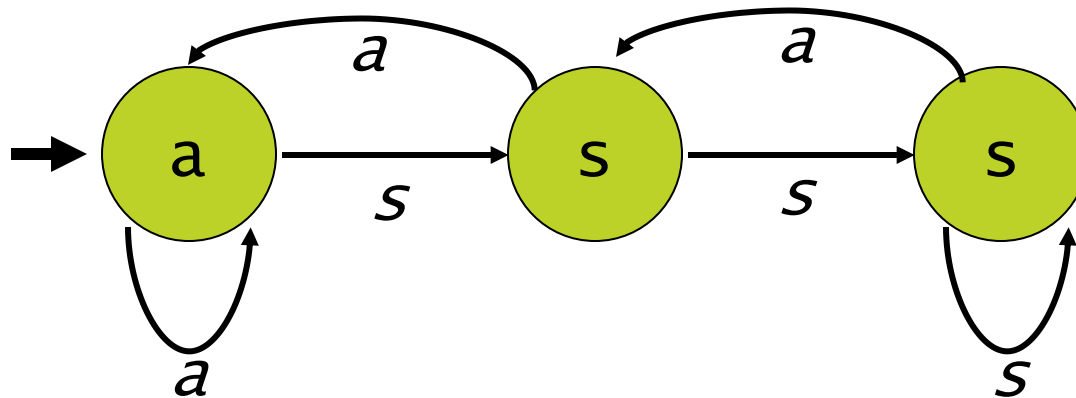
$asaasssa \rightarrow s$

Consistent:



# The algorithm

The result





# How do we get hold of the learning data?

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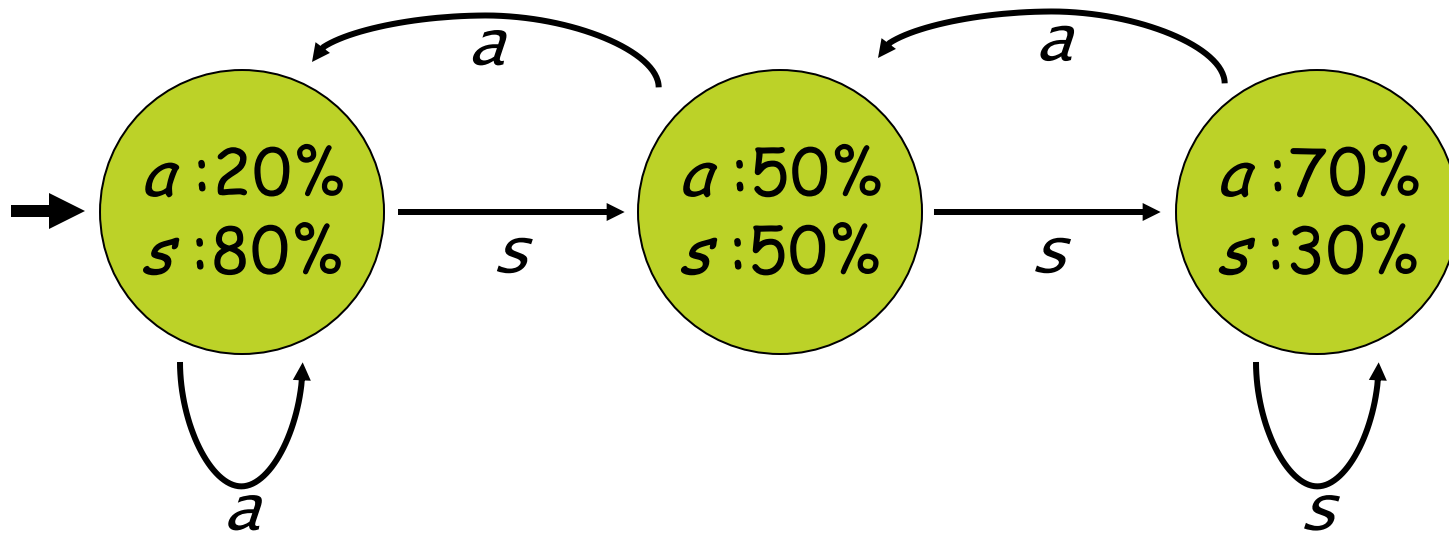
a) through observation

b) through exploration (like here)

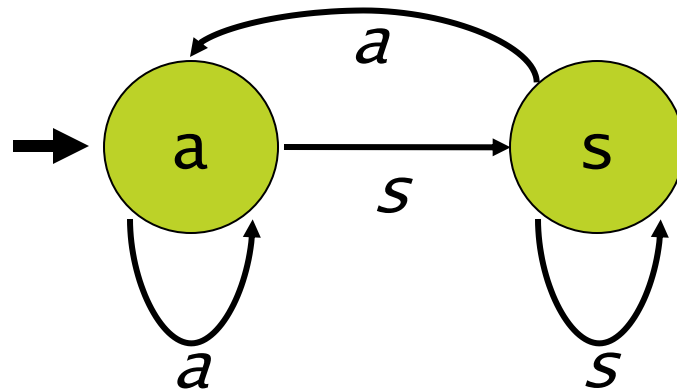


# An open problem

The strategy is probabilistic:



# Tit for tat







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