Hierarchical-Alphabet Automata

Guest lecture for the "Current Trends in Artificial Intelligence" course



Introduction

• Who am I?

- PhD student at the VUB AI lab
- Under supervision of Prof. Dr. Johan Loeckx
- Research interests in applied AI and cybersecurity

What is this guest lecture about?

- Finite state machine variants + their applications
- Our hierarchical extension on finite state machines
- o Involvement in projects of the AI lab Applied Research team



Topic of this guest lecture

Preliminary knowledge

- Basics of languages and automata
- Directed acyclic graphs and ordered sets
- Applications, advantages and disadvantages

My PhD research

- Introduction to Hierarchical-Alphabet Automata (HAA)
- Applied research projects within the AI Lab (using HAA)



Before we begin...

- Slides are as self-contained as possible:
 - Lots of information on slides, but we'll skip where necessary
 - o If you miss guest lecture, you can still read the slides by yourself
- Follow along with the slides: patricks.phd/current-trends/slides.pdf



Terminology

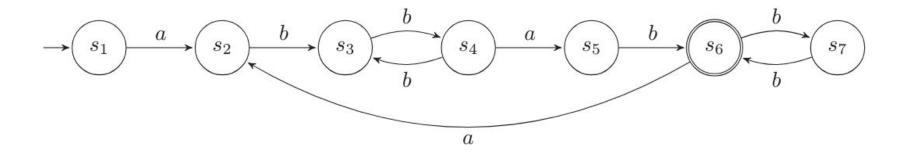
- Alphabet: finite set of symbols
- Words: concatenations of zero or more symbols
- Factors: contiguous subsequences of a word
 - Prefix: a factor at the beginning of a word
 - Suffix: a factor at the end of a word
- Language: subset of the infinite set of possible words we can create using an alphabet of symbols



- Alphabet: { a, b, c, ..., z }
 Words: { "", "a", "aa", "abc", "cars", ... }
- Factors: actor is a factor of factory
 - **Prefix:** <u>fact</u> is a factor and prefix of <u>factory</u>
 - Suffix: tory is a factor and suffix of <u>factory</u>
- Language: a*b*, the set of factors of a word, ...

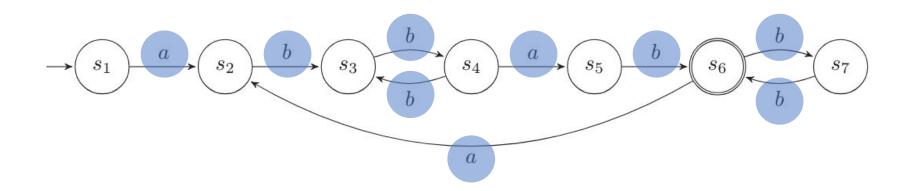


- Σ is a finite set of symbols (the **input alphabet**)
- S is a finite set of states
- $s_0 \in S$ is the **initial state**
- $\delta: S \times \Sigma \to S$ is the state-transition function
- $F \subseteq S$ the set of final or accepting states



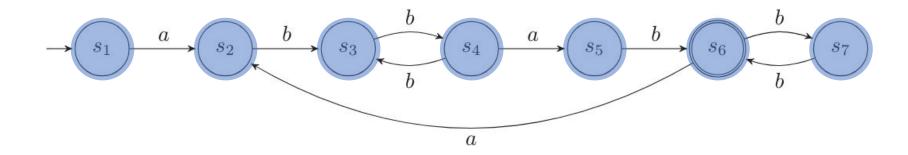


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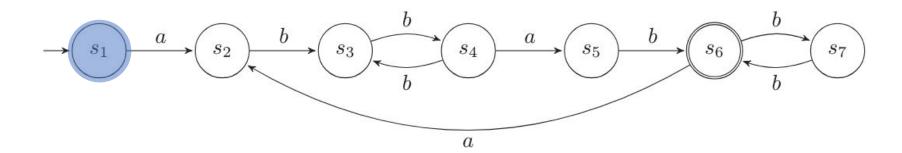


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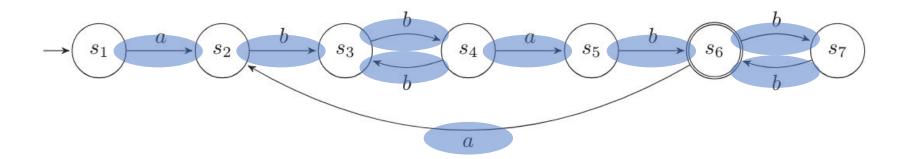


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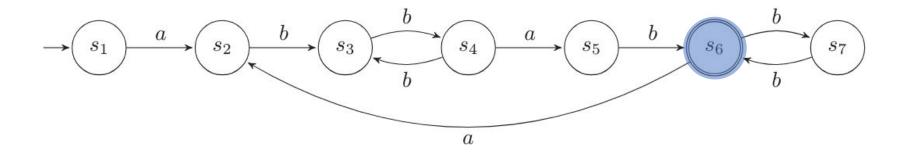


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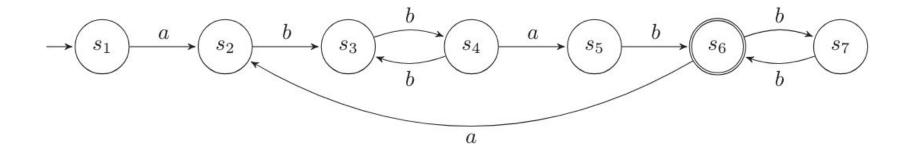
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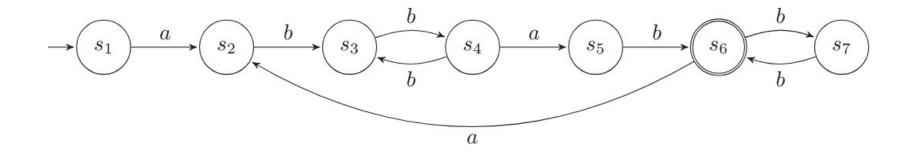
Acceptance and rejection

- An FSM recognises a language:
 - It accepts words that are part of this language
 - It rejects words that are not part of this language
- Start at the initial state + first symbol of the input:
 - Repeatedly follow transition function with current state and symbol
 - Accepts word if you end up in accepting state
 - Rejects word if you end up in normal state, or if there is no transition



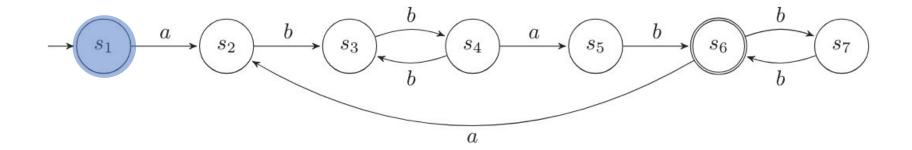


• FSM that recognises the language of alternating even and odd number of occurrences of 'b' separated by a single 'a'



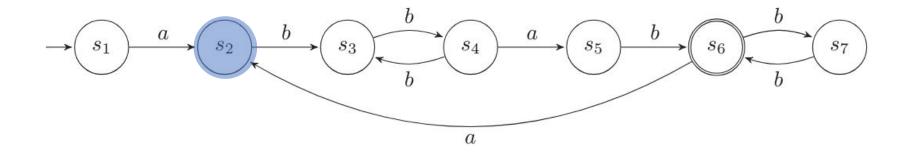


Input the word:



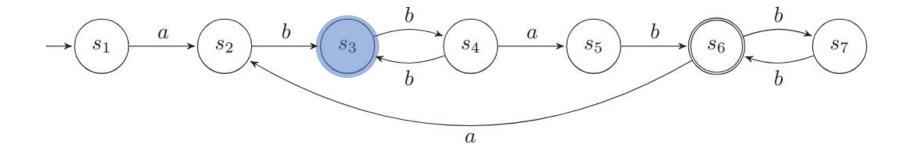


Input the word:



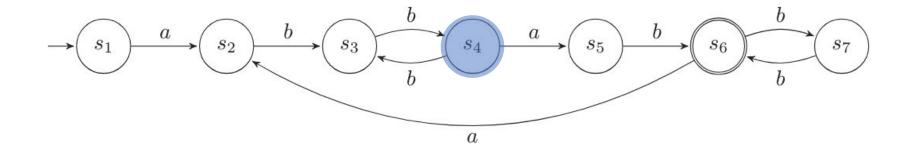


Input the word:



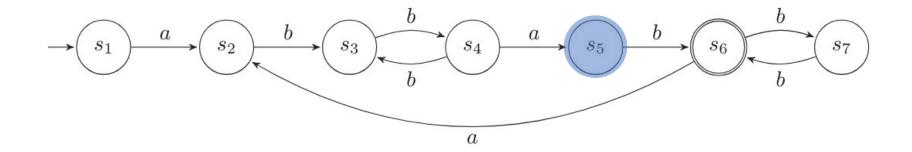


Input the word:



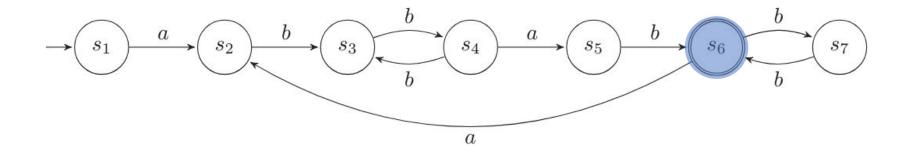


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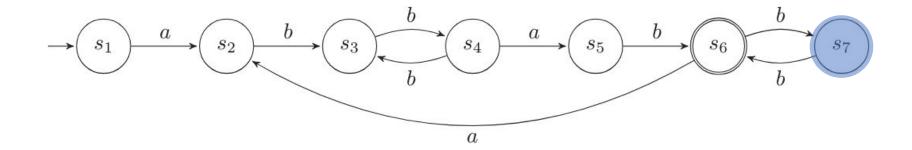


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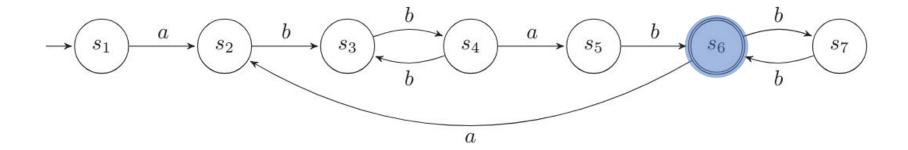


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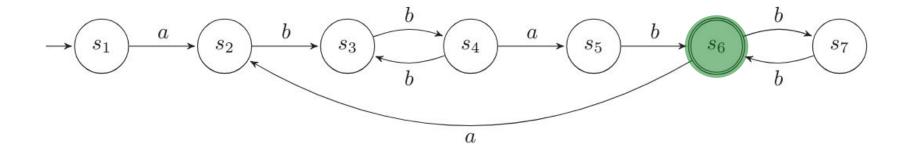


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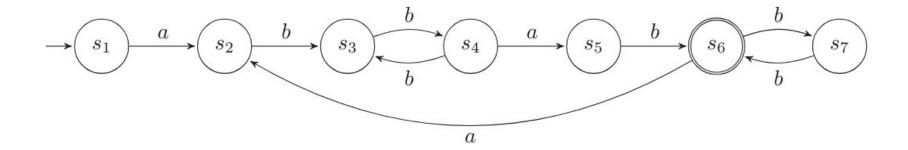






What if we would have:

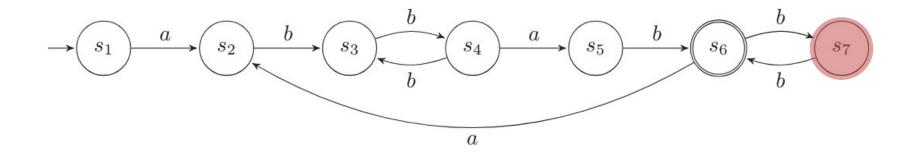
"abbabb"?





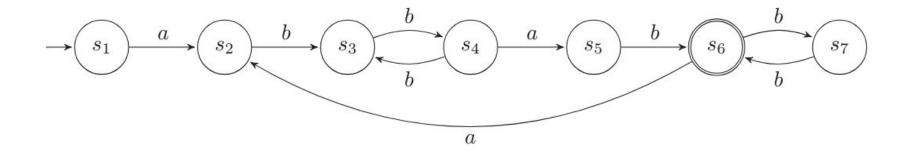
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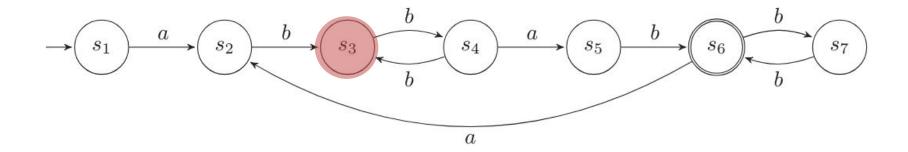
"ababbb"?





What if we would have:







Applications of FSMs

- Anomaly detection:
 - Learning a language from sequences of system calls
- What problems could this specific application have?
 - Scalability issues for programs with non-trivial behaviour
 - "State-explosion problem" (more on this later...)

Figure 1. An example program and associated trigrams. S0,...,S5 denote system calls.

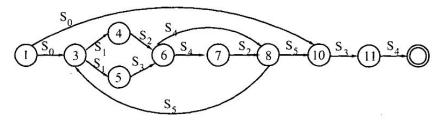
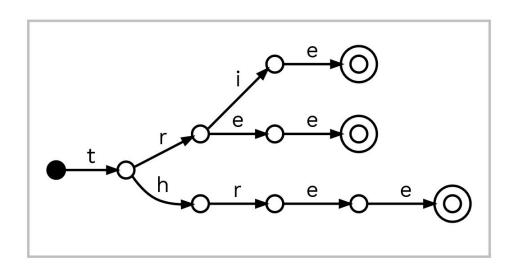


Figure 2. Automaton learnt by our algorithm for Example 1

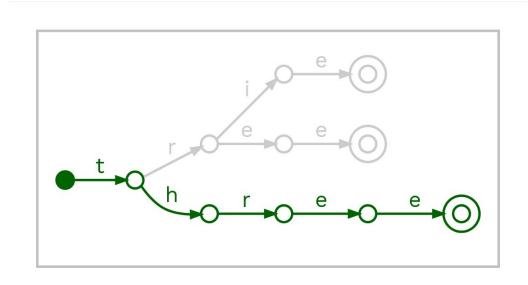


- **Trie:** FSM as a rooted tree associated with a set of words
 - Paths from initial to accepting states represent words from its set
 - Example: trie for the set { three, tree, trie }



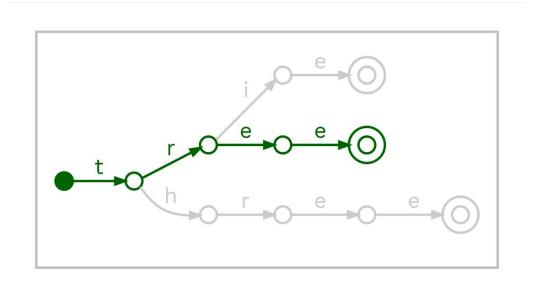


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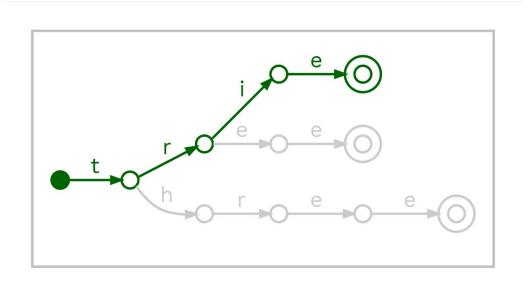


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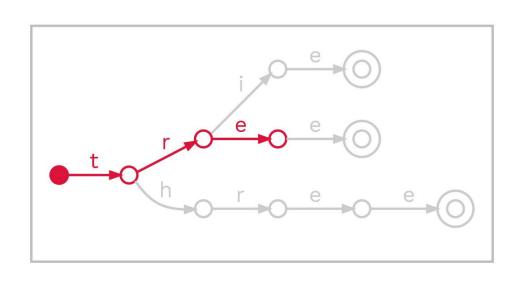
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tre is not part of the set { three, tree, trie }





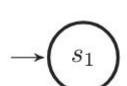
Construction Algorithm - tries

{ trie, tree }



Construction Algorithm - tries

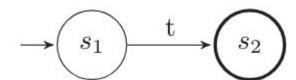
```
def trie(sequences):
    """Constructs a trie from a set of sequences."""
    fsm = Automaton()
    for sequence in sequences:
        current = fsm.initial
        for symbol in sequence:
            if symbol not in fsm.alphabet:
                fsm.add_symbol(symbol)
            if (next_state := fsm.follow(current, symbol)) is None:
                      next_state = fsm.add_state()
                      fsm.set_transition(current, next_state, symbol)
                      current = next_state
                     fsm.accepting.add(current)
                      return fsm
```





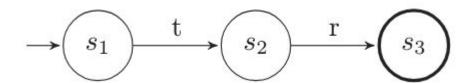


Construction Algorithm - tries

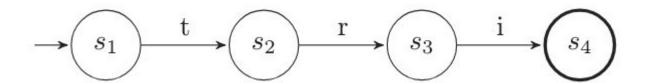


{ **t**rie, tree }

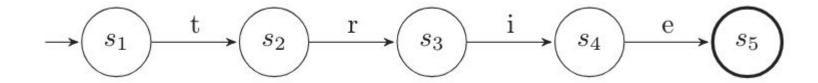




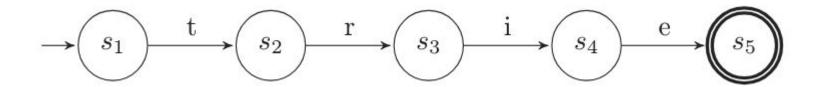






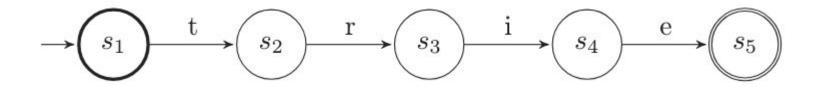








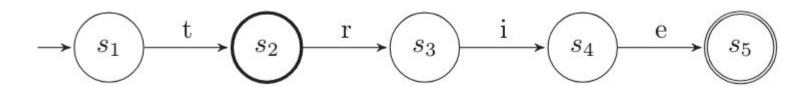
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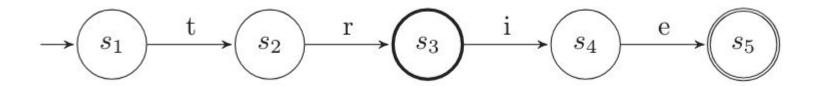
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    return fsm
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                                                                                          s_3
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                                                                                                             s_6
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                                                                                              { trie, tree } 🔽
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            current = next_state
                                                                                                            s_6
        fsm.accepting.add(current)
   return fsm
                                                                                                            s_4
```



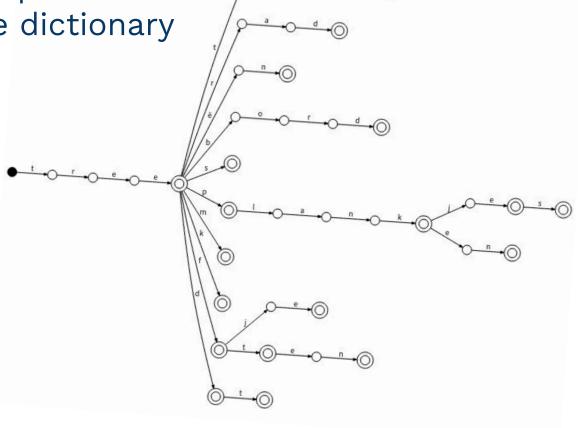
Applications - tries

Efficient data structure:

Storing words based on common prefixes

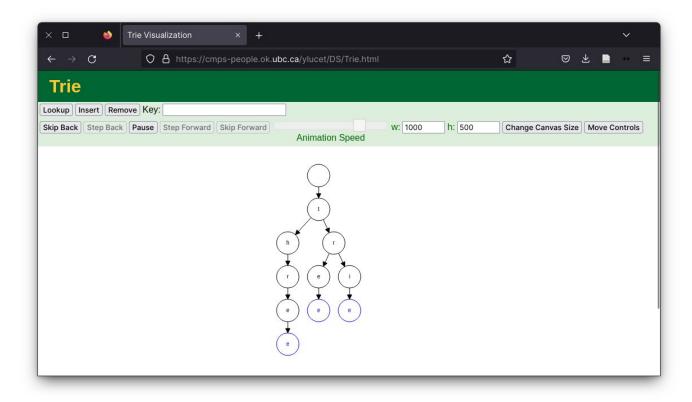
May require less storage for large dictionary

Allows for efficient search





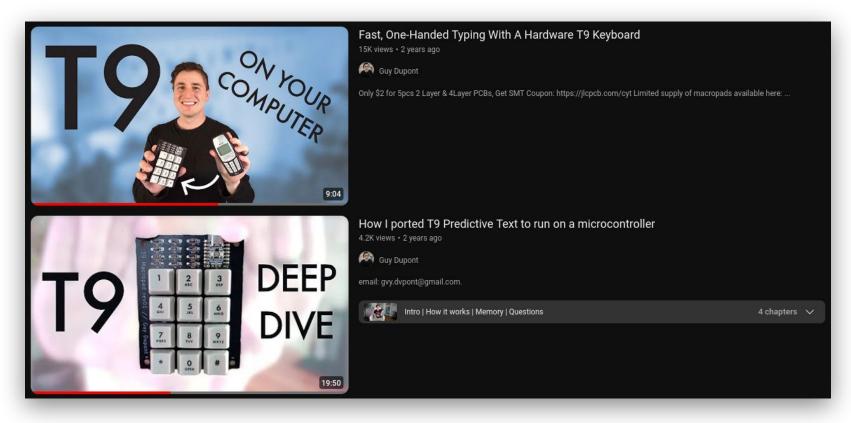
Demonstration



https://cmps-people.ok.ubc.ca/ylucet/DS/Trie.html



What is it used for?



https://youtu.be/6cbBSEbwLUI



Building FSMs - factor automaton

Factor automaton of a word x:

 The minimal deterministic automaton that recognises the factors of x

Remember:

Factors: <u>contiguous</u>
 subsequences of a word

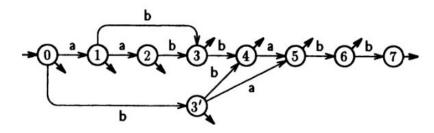
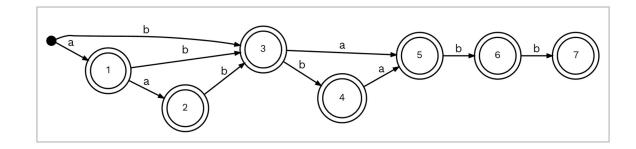


Fig. 7.11. Minimal deterministic automaton recognizing the factors of aabbabb.



Factor oracle:

 Acyclic automaton built on a set of words, recognises at least all factors of every word of the set



Applications:

 Intended for multi-pattern matching



Factor Automata vs. Factor Oracles

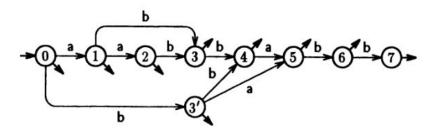
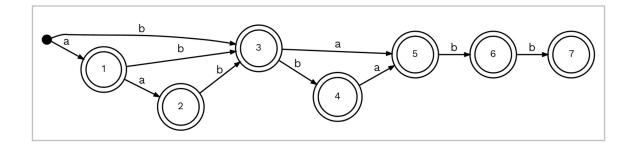


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Factor automaton of a word x:

 The minimal deterministic automaton that recognises the factors of x



• Factor oracle:

 Acyclic automaton built on a set of words, recognises at least all factors of every word of the set

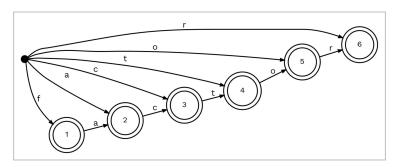
("abab" is not a factor!)

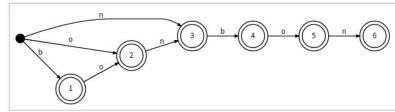


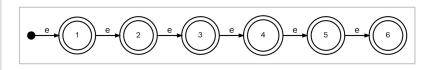
States and transitions:

- Has exactly m + 1 states
- Between m and 2m 1 transitions

• Examples:







"factor"

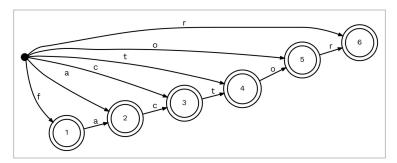
"bonbon"

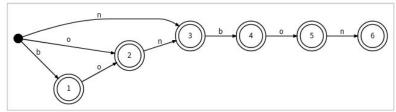
"eeeeee"

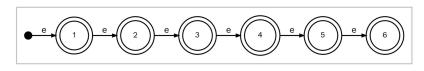


- States and transitions:
 - Has exactly m + 1 states
 - Between m and 2m 1 transitions

• Examples:







"factor"
Worst case

"bonbon" **Average case**

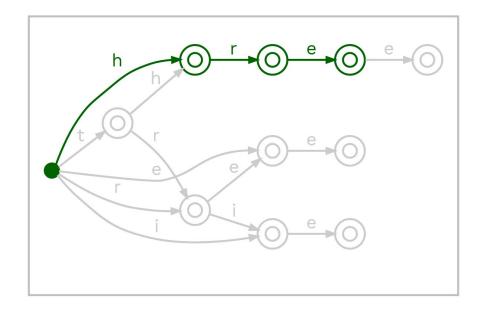
"eeeeee"

Best case



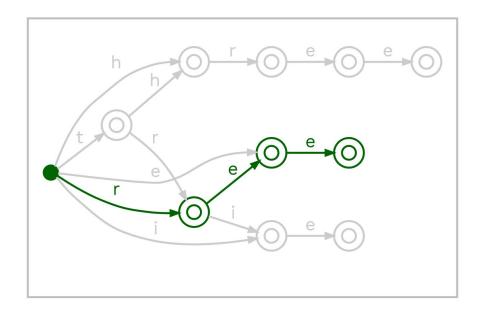


- Accepts at least all factors:
 - Factor oracle for { three, trie, tree }
- Examples:
 - Accepts "hre", factor of three





- Accepts at least all factors:
 - Factor oracle for { three, trie, tree }
- Examples:
 - Accepts "ree", factor of both three and tree

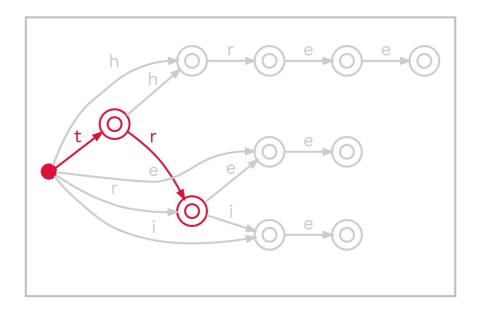




- Accepts at least all factors:
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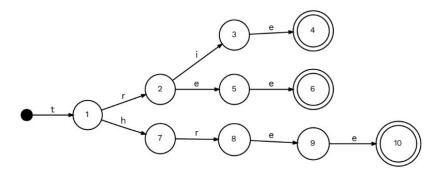
Examples:

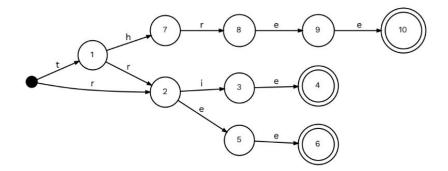
Does not accept "trh", not a factor
 (and thus no transition labelled 'h')





Construction Algorithm - factor oracle

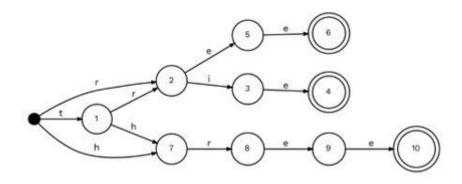


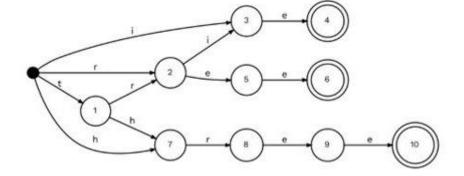


- (a) The trie for $P = \{trie, tree, three\}$, taken from Fig. 2.10. In the first step of the algorithm, we generate the trie which we use as a starting point to generate the factor oracle.
- (b) In this step of the construction algorithm, a new transition from the initial state to state 2 is made, labelled with r. There are now 11 transitions in total.



Construction Algorithm - factor oracle



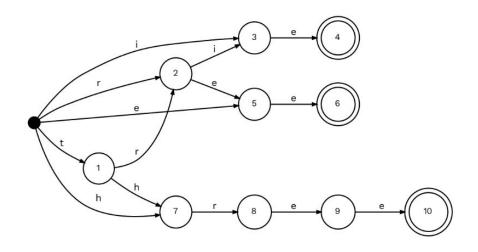


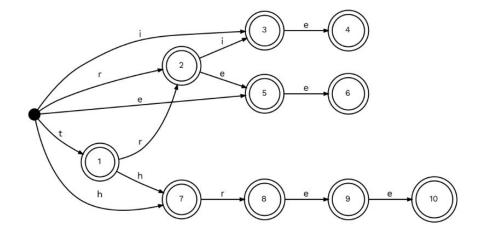
(c) In this step of the construction algorithm, a new transition from the initial state to state 7 is made, labelled with h. There are now 12 transitions in total.

(d) In this step of the construction algorithm, a new transition from the initial state to state 3 is made, labelled with i. There are now 13 transitions in total.



Construction Algorithm - factor oracle

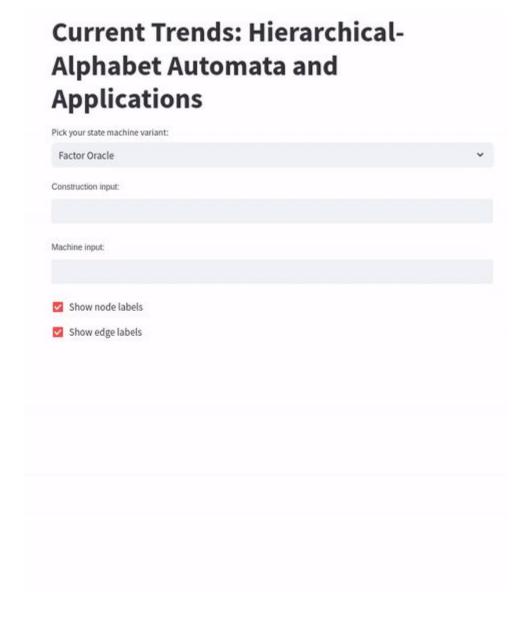




- (e) In this step of the construction algorithm, a new transition **from the initial state to state 5** is made, labelled with *e*. There are now 14 transitions in total.
- (f) In the final step of the construction algorithm, all states are marked as final states. The algorithm transformed the original trie to the factor oracle of $P = \{trie, tree, three\}$.



Demonstration





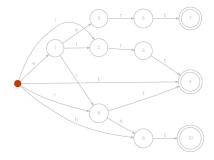
What is it used for?

Pattern matching:

- Backwards Oracle Matching (BOM)
- Set Backwards Oracle Matching (SBOM)

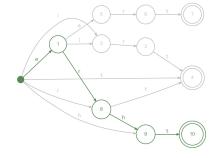
two_or_three_trees

(a) We attempted to input _ in the factor oracle, but the oracle rejects directly. We shift the window after the _ and continue our search.



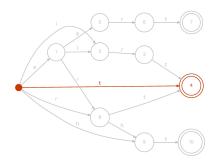
two_or_ thre e_trees

(c) We successfully input e, τ , h, and t in the oracle. We reach accepting state 10, which is associated with the word three. We check for an occurrence of three, and shift the window by 1.



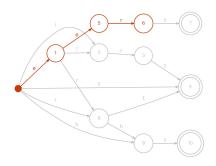
two_or_t hree_trees

(b) We successfully input t in the factor oracle, but the oracle fails on \square . We shift the window after the \square and continue our search.



two_or_t hree_trees

(d) We successfully input e, e, and r in the oracle, but the oracle fail on the character h. We shift the window after the h and continue our search.





Applications of FSMs

Anomaly detection:

 Learning a language from sequences of system calls

FOs for anomaly detection?

 Accept factors of allowed system call sequences

Problems:

- Evolving behaviour
- Difficulties with broad behaviour
 - State-explosion problem?

Figure 1. An example program and associated trigrams. S0,...,S5 denote system calls.

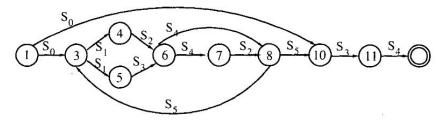


Figure 2. Automaton learnt by our algorithm for Example 1



Example - Even and Odd

!≡ Even-odd binary words

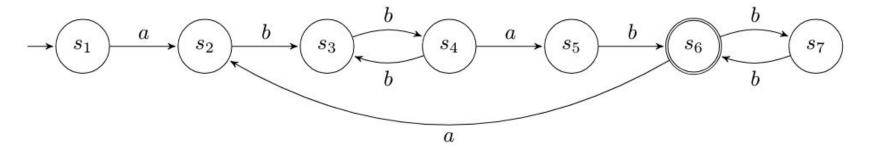
Create a machine that accepts words starting with a, followed by alternating even and odd number of repetitions of b separated by one a, where the word always ends with an odd number of b:

$$(a(bb)^+ab(bb)*)^+$$



• •

Regular FSM

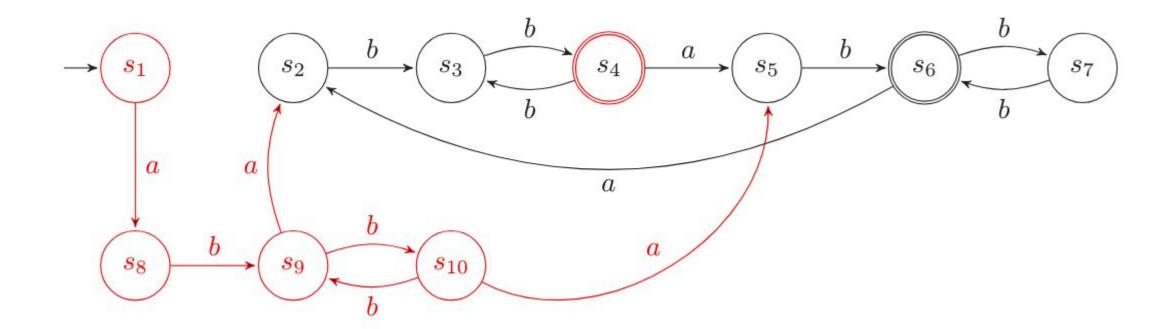




Example - Even and Odd

Evolving behaviour:

 What if we can now also start with an odd number of occurrences of b?





Example - Even and Odd

Evolving behaviour:

 Now, what if we alternate between alternating from even to odd to even, and repeating even to even and odd to odd?

Problem:

- Building on previous machine requires us to make too many changes
- There is no modularity to exploit, and machine gets too complex



Research Goals

Solving previously-mentioned issues:

- Capturing broad behaviour with compact FSMs
- Easily change model when behaviour changes or evolves

Our hypothesis:

 We can exploit hierarchical relationships of symbols to create compact, less complex, modular machines

• How do we do this?

 Hierarchical-alphabet automata (HAA): introducing hierarchy in machines and alphabet



The hierarchical-alphabet automaton is a finite <u>poset</u> $P = (M, \preceq)$ of machines $m_i \in M$, with one machine m_r as the greatest element of the set. We can interpret this ordering as a <u>directed acyclic graph</u>, with machine m_r as the only source of the DAG.

Partial Order

An **order** (or partial order) on P is a binary relation \leq on P such that, for all $x,y,z\in P$,

- 1. $x \leq x$ (reflexivity)
- 2. $x \leq y$ and $y \leq x$ imply x = y (antisymmetry)
- 3. $x \leq y$ and $y \leq z$ imply $x \leq z$ (transitivity)

Partially Ordered Set (poset)

A poset $P = (X, \preceq)$ is a pair consisting of a set X and the <u>partial order</u> \preceq on X



The hierarchical-alphabet automaton is a finite <u>poset</u> $P = (M, \preceq)$ of machines $m_i \in M$, with one machine m_r as the greatest element of the set. We can interpret this ordering as a <u>directed acyclic graph</u>, with machine m_r as the only source of the DAG.



A directed acyclic graph (DAG) is a directed graph with no directed cycles.



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∃ Example

A poset $P=(M,\preceq)$ consisting of a set $M=\{m_1,m_2,m_3\}$ with m_1 as the greatest element. We can represent this poset P as a DAG with $m_i\in M$ as the vertices, m_1 as the source node, and the <u>covering relation</u> $(m_i>m_j)$ as directed edges.

O Covering Relation

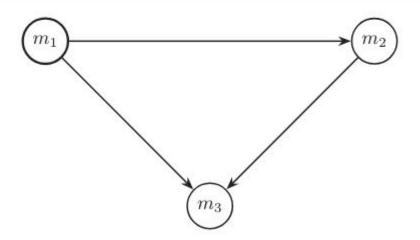
Let X be a set with a partial order \preceq . Let \prec be the relation on X such that $x \prec y \iff x \preceq y \land x \neq y$ with $x,y \in X$. Then y covers x if x < y and there is no element $z \in X$ such that x < z < y.



The hierarchical-alphabet automaton is a finite <u>poset</u> $P = (M, \preceq)$ of machines $m_i \in M$, with one machine m_r as the greatest element of the set. We can interpret this ordering as a <u>directed acyclic graph</u>, with machine m_r as the only source of the DAG.

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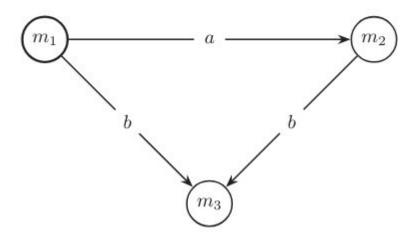




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We also introduce an **edge labelling** to the edges of the DAG:

For every $m_i \in M$, there is at most one outgoing edge labelled with $\sigma \in \Sigma_{m_i}$.





Input of the HAA:

- Sequence of ordered trees
 - **Leafs** are our observations
 - Internal nodes represent groupings of these observations

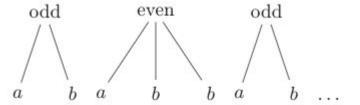
Ordered trees:

 Rooted tree where the order of the subtrees of a node are significant

Rooted tree

A **rooted tree** is a finite set S of one or more nodes such that (a) there is one specially designated node, called the root of the tree, and (b) the remaining nodes are partitioned into $m \geq 0$ disjoint sets S_1, \ldots, S_m , and each of the sets in turn is a rooted tree. The trees S_1, \ldots, S_m are called the subtrees of the root.

- Number of subtrees of a node = the degree of the node
 - Node of degree zero is called a leaf
 - Node of positive degree is called an internal node
 - Node of degree at least 2 is called a branching node
- The level of a node is the number of separations from root





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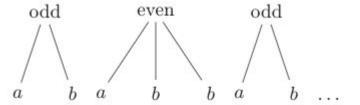
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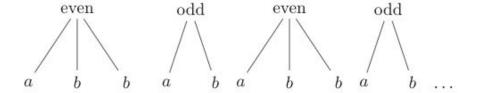


Example - Even and Odd

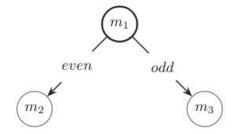
¡ ≡ Even-odd binary words

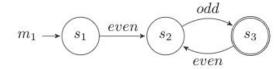
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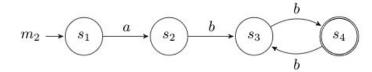
$$(a(bb)^+ab(bb)*)^+$$

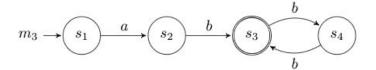


The hierarchical-alphabet automaton is a finite poset $P=(M,\preceq)$ of machines $m_i\in M$, with one machine m_r as the greatest element of the set. We can interpret this ordering as a directed acyclic graph, with machine m_r as the only source of the DAG.



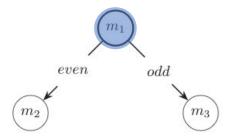


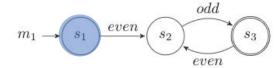


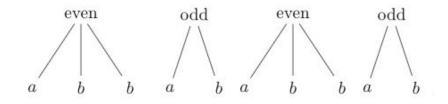


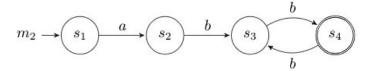


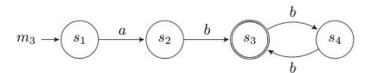
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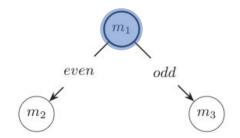




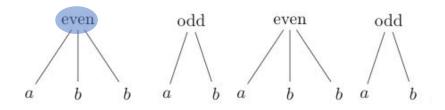


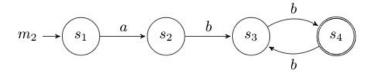


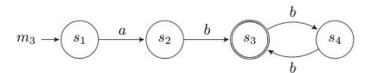
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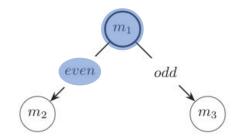




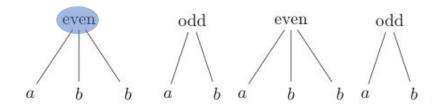
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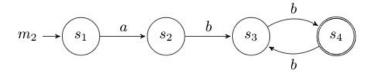
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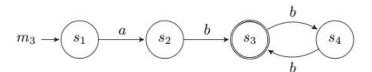
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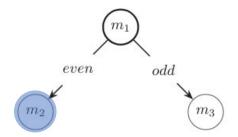




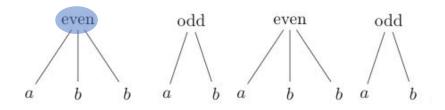


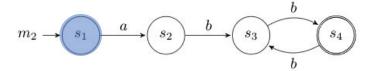


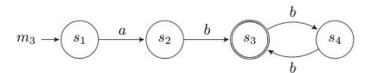
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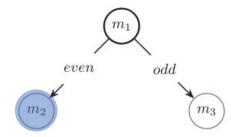




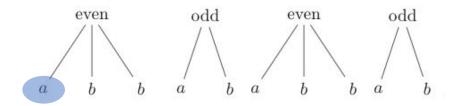


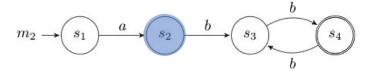
The hierarchical-alphabet automaton is a finite poset

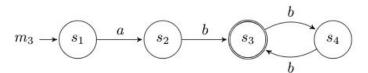
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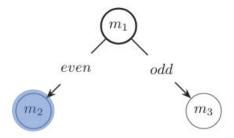




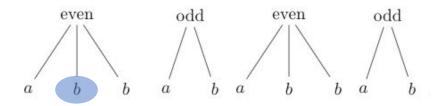


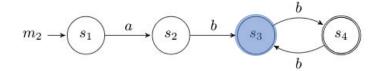


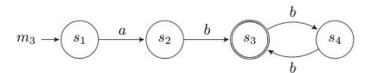
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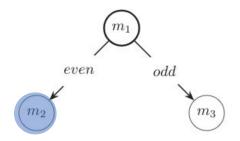




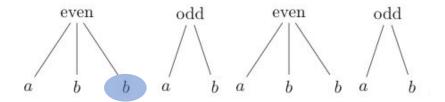


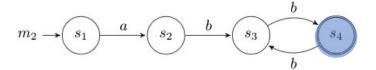
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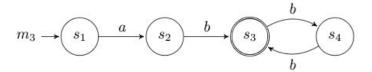
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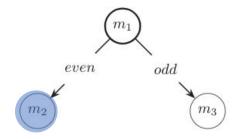




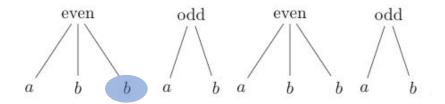


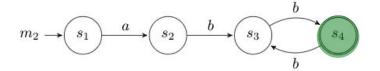
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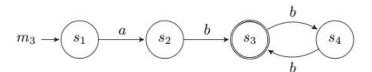
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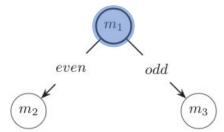


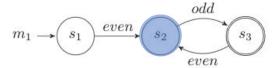


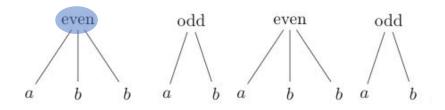


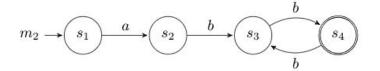


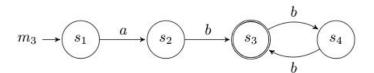
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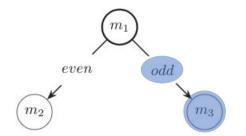


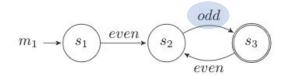


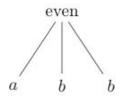


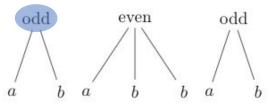


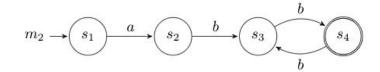
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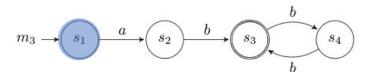








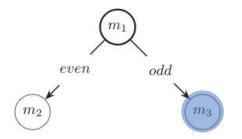


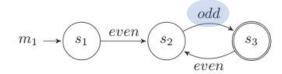


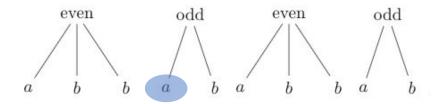


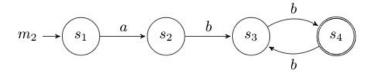
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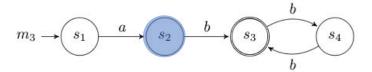
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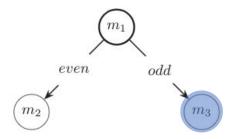


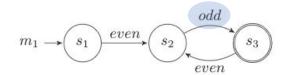


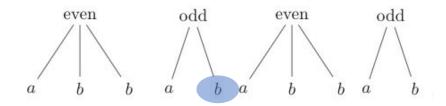
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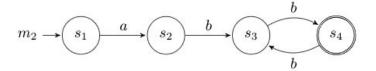
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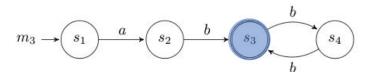
of the DAG.







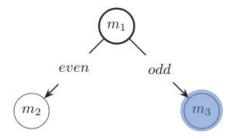


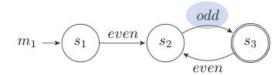


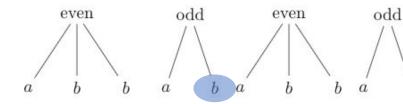


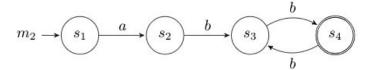
The hierarchical-alphabet

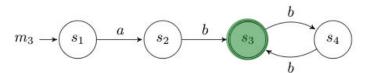
The hierarchical-alphabet automaton is a finite <u>poset</u> $P=(M,\preceq)$ of machines $m_i\in M$, with one machine m_r as the greatest element of the set. We can interpret this ordering as a <u>directed acyclic graph</u>, with machine m_r as the only source of the DAG.





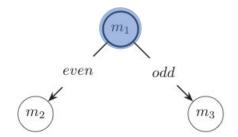




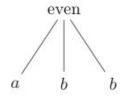


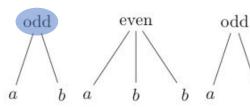


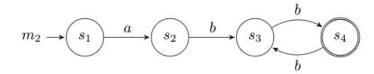
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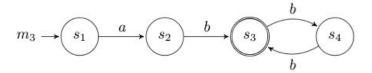






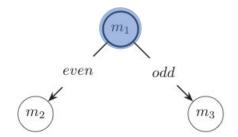


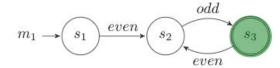
and so on...



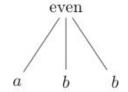


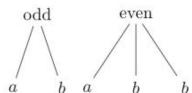
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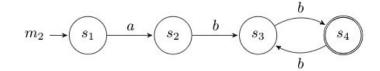
In the end:

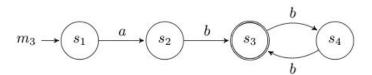










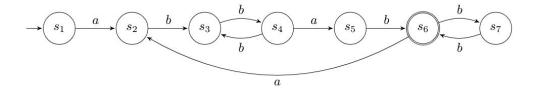


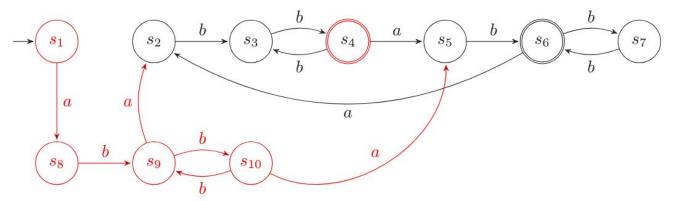


Advantages

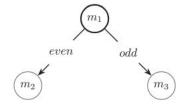
- What about evolving behaviour?
 - Remember:

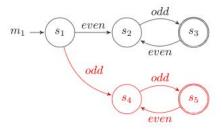
Regular FSM

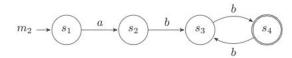


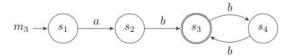














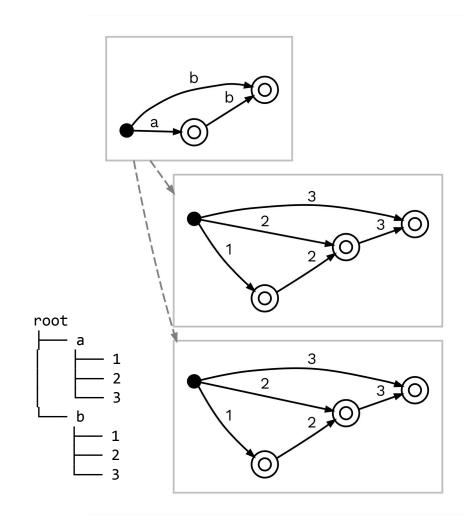
Advantages and Disadvantages

- Modularity, interpretability, ...
 - When behaviour changes, we can easily build on existing HAA without having to change (all) submachines + introduce new submachines on top of them (= modularity)
- Hierarchical inference? X
 - Here, we assume we have the hierarchy already
 - Our How do we infer the hierarchy from data alone?



Building HAA - hierarchical factor oracle

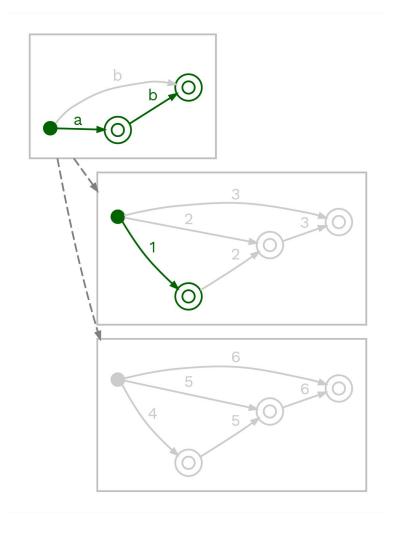
- Intuitively, the HFO is the hierarchical variant of the factor oracle
- Accepts language of hierarchical factors:
 - Accepts at least the set of ordered subtrees which we used to create the HFO





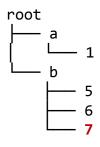
Building HAA - hierarchical factor oracle

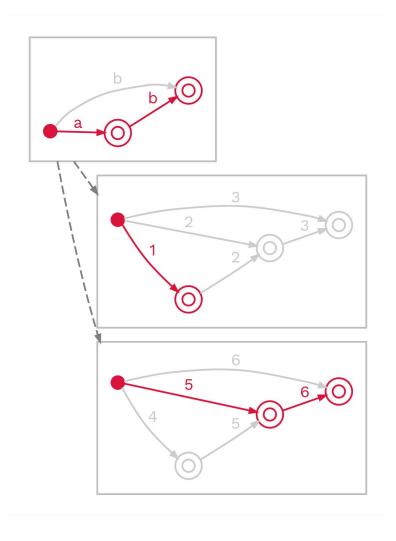






Building HAA - hierarchical factor oracle







Anomaly detection:

 \circ Finding patterns in data that do not conform to expected behaviour

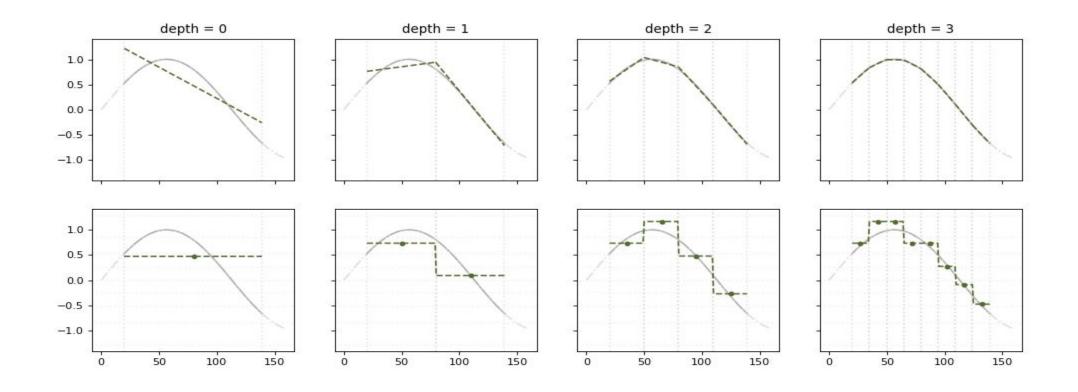
Main focus:

- Needs to be lightweight due to constraints in resources
- We should never forget rarely-occurring events
- We should be able to adapt to new behavior
- o It should be general enough to work on different time series
- It should recognize non-strictly periodic events
- Should have a minimum amount of false positives



- How do we use HAA for time series?
 - Extract overlapping subsequences using a sliding window
 - o Discretise segments, and use these as symbols of alphabet
- How do we have hierarchy in time series?
 - Discretise in increasing granularity







- Some projects of the AI lab's Applied Research team:
 - "For cybersecurity, our primary application involves anomaly detection using both discrete and continuous data. For instance, endpoints can generate discrete data like process creation and termination events, or continuous data such as CPU usage or domain requests over time. Using this data, we model the processes or systems that generate this data, and apply these models for anomaly detection."



- Some projects of the AI lab's Applied Research team:
 - "For predictive building maintenance, the goal is to move away from employing corrective or periodic maintenance approaches. We aim to predict potential system failures by monitoring and modelling the system's functionality over time, facilitating corrective action before an actual failure that impacts building user comfort or energy-efficiency occurs. This is particularly challenging due to complex interactions among various components in large building installations."



• Some projects of the AI lab's Applied Research team:

"For neuroscience, collected data can help gain insights into the spatio-temporal dynamics between brain regions. High-resolution techniques such as magnetoencephalography (MEG) have allowed for the study of brain networks. Therefore, we aim to develop new techniques for context-aware modelling and long-term dependency modelling of these networks, taking into account factors such as subject fatigue and age regression. We apply these techniques to improve our understanding of the impact of Multiple Sclerosis on brain functioning."



