

Efficient Analysis of Unambiguous Automata Using Matrix Semigroup Techniques

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DEPARTMENT OF
**COMPUTER
SCIENCE**



1 Introduction

- What is model checking
- Finite word model checking
- Infinite word model checking

2 Cuts

3 Pseudo-cuts

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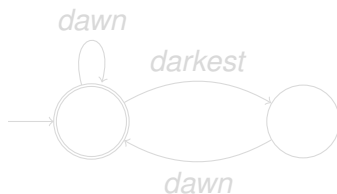
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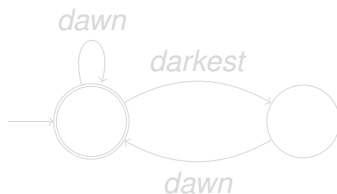
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- Calculating possibility of a scenario happening
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- Example: $\mathbf{G}(\text{darkest}(x) \rightarrow \mathbf{X}\text{dawn}(x))$
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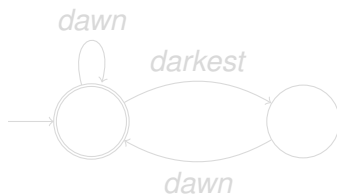
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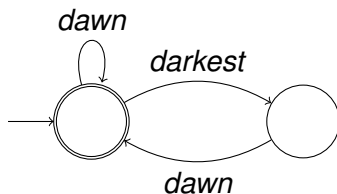
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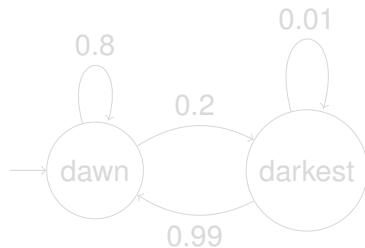
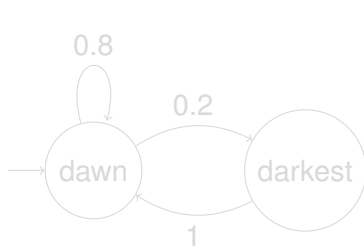
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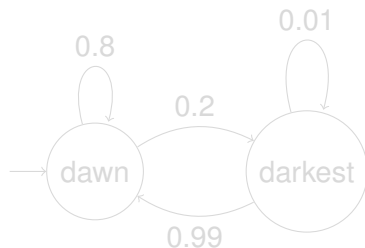
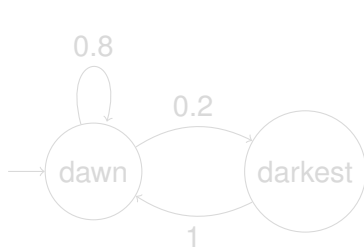
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- Generally: Markov chains
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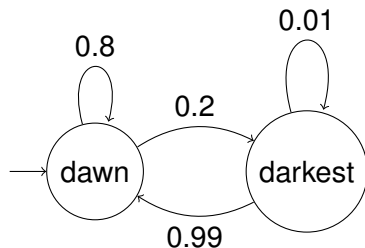
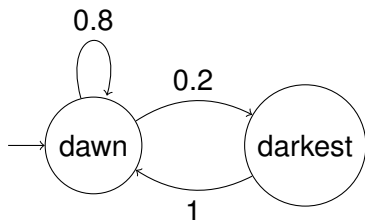
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a Markov chain (P, M) , with $P_{m,m'}$ denoting the probability of going to $m' \in M$ from $m \in M$,
an initial vector $i \in M$ stating the initial distribution of the Markov chain,
and a vector $\omega \in [0, 1]^M$ denoting the probability of stopping the word after reading a character in M .
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- For UFAs:

$$\zeta_{q,m} = |\delta(q,m) \cap F| \omega(m) + (1 - \omega(m)) \sum_{m' \in M} \sum_{q' \in \delta(q,m)} P_{m,m'} \zeta_{q',m'}$$

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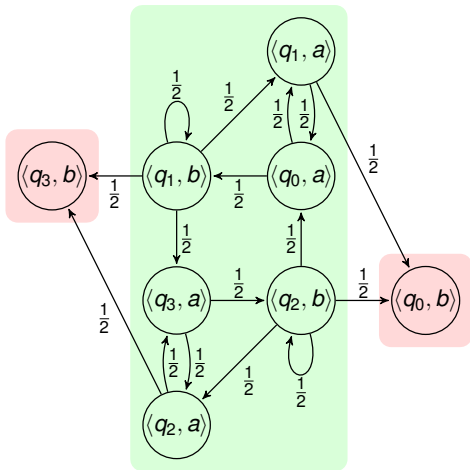
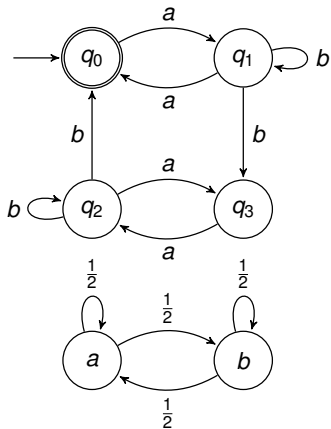
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Running example



- Notice that if the probability of generating an accepted word is greater than zero, then there exist prefixes such that almost every word beginning with that prefix is accepted
- Given such a prefix p for a UBA (S, M, δ, Q_0, F) , we see that for any $m \in M$, $\sum_{q_0 \in Q_0} \sum_{s \in \delta(q_0, p)} \zeta_{s, m} \geq 1$. $\delta(q_0, p)$ is called a *cut*.
- But since no two co-reachable states can accept the same words, we see that $\sum_{q_0 \in Q_0} \sum_{s \in \delta(q_0, p)} \zeta_{s, m} = 1$. We call the characteristic vector of a cut a *normaliser*.
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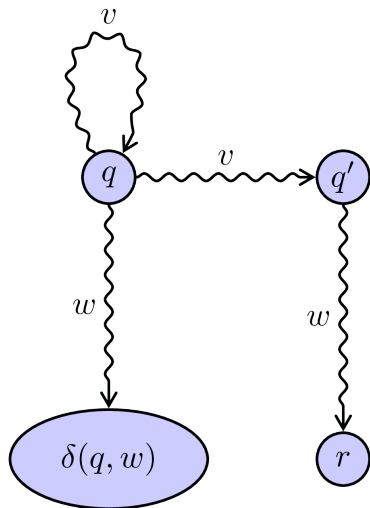
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- Given a state q and a word w such that $\delta(q, w)$ is not a cut, an extender is a word v such that $\delta(q, v) \supseteq \{q, q'\}$, $q \neq q'$, and $\delta(q', w) \neq \emptyset$. Then $\delta(q, vw) \supseteq \delta(q, w)$.

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Extenders increase set size



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- By repeatedly calculating extenders we get a word p such that almost any infinite word prefixed by p is accepted

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- Sums of cut vectors also invariant under transition matrices
- Matrix semigroups with constant spectral radius have invariant affine plane (Protasov, Voynov, 2017)

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Calculating $Co(d)$ -pseudo-cuts

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- Find the affine plane with BFS on M^*
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- (Almost) complementation
- ...