

## Skyline Operators for Document Spanners

Antoine Amarilli ${ }^{1}$, Sébastien Labbé², Benny Kimelfeld³${ }^{3}$, Stefan Mengel ${ }^{4}$<br>March 27th, 2024<br>${ }^{1}$ Télécom Paris<br>${ }^{2}$ Technion<br>³École normale supérieure

${ }^{4}$ CNRS CRIL

## Document spanners

We use document spanners, a declarative formalism for information extraction tasks

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Text document

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# "Extract all email addresses in the document" <br> Document spanner 



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"Extract all email addresses in the document"
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\Sigma^{*} \sqcup \vdash_{\text {email }}[\mathrm{a}-\mathrm{z}]^{+} @[\mathrm{a}-\mathrm{z}]^{+} .[\mathrm{a}-\mathrm{z}]^{+} \dashv_{\text {email }} \sqcup \Sigma^{*}
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## "Extract all last names with possibly a phone number" <br> Document spanner



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$$
\Sigma^{*} \vdash_{\text {name }}[\mathrm{A}-\mathrm{Z}][\mathrm{a}-\mathrm{z}]^{+} \dashv_{\text {name }}\left(\epsilon \mid \sqcup \vdash_{\text {phone }}[0-9]^{+} \dashv_{\text {phone }}\right) \sqcup \Sigma^{*}
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Results

- Several formalisms to express document spanners
$\rightarrow$ Focus: regular spanners expressed as Variable-Set Automata (VAs) or regex-formulas


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- Several formalisms to express document spanners
$\rightarrow$ Focus: regular spanners expressed as Variable-Set Automata (VAs) or regex-formulas
- Well-studied task: efficient evaluation, including enumeration algorithms


## Maximal matches

Standard semantics: extract all mappings of the spanner variables. But...

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A3) You
How can I obtain all matches of a regular expression in a string?


ChatGPT
To obtain all matches of a regular expression in a string, you can use the appropriate function provided by your programming language or library.

```
python
pattern \(=\) r'your_pattern_here' \(^{\prime}\)
input_string = "your_input_string_here"
matches \(=\) re.findall(pattern, input_string)
```


## Maximal matches

## Standard semantics: extract all mappings of the spanner variables. But...

Table of Contents
re - Regular expression
operations

- Regular Expression Syntax
- Module Contents
- Flags
- Functions
- Exceptions
- Regular Expression Objects
- Match Objects
- Regular Expression Examoles
re.findall(pattern, string, flags=0)
Return all non-overlapping matches of pattern in string, as a list of strings or tuples. The string is scanned left-to-right, and matches are returned in the order found. Empty matches are included in the result.

The result depends on the number of capturing groups in the pattern. If there are no groups, return a list of strings matching the whole pattern. If there is exactly one group, return a list of strings matching that group. If multiple groups are present, return a list of tuples of strings matching the groups. Non-capturing groups do not affect the form of the result.

```
>>> re.findall(r'\bf[a-z]*', 'which foot or hand fell fastest')
['foot', 'fell', 'fastest']
```

pattern $=$ r'your_pattern_here' $^{\prime}$
input_string = "your_input_string_here"
matches $=$ re.findall(pattern, input_string)

## Maximal matches

Specifically, we may want:

- "Extract all email addresses"

$$
\Sigma^{*} \sqcup \vdash_{\text {email }}[\mathrm{a}-\mathrm{z}]^{+} \text {@ }[\mathrm{a}-\mathrm{z}]^{+} .[\mathrm{[a-z}]^{+} \dashv_{\text {email }} \sqcup \Sigma^{*}
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## Maximal matches

Specifically, we may want:

- "Extract all email addresses", without worrying about delimiters $\Sigma^{*} \sqcup \vdash_{\text {email }}[a-z]^{+}$@ $[a-z]^{+} .[a-z]^{+} \dashv_{\text {email }} \Sigma^{*}$


## Maximal matches

Specifically, we may want:

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## Maximal matches

Specifically, we may want:

- "Extract all maximal email addresses", without worrying about delimiters $\Sigma^{*} \vdash_{\text {email }}[a-z]^{+}$© $[a-z]^{+} .[a-z]^{+} \dashv_{\text {email }} \Sigma^{*}$


## Maximal matches

Specifically, we may want:

- "Extract all maximal email addresses", without worrying about delimiters $\Sigma^{*} \vdash_{\text {email }}[a-z]^{+}$@ $[a-z]^{+} .[a-z]^{+} \dashv_{\text {email }} \Sigma^{*}$
- "Extract all maximal matches of last names with possibly a phone number" $\rightarrow$ If the number is given, do not extract a match without it


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Skyline under a domination relation: the results which are maximal, i.e., not dominated

## Naive skyline computation

$$
\Sigma^{*} \vdash_{\text {email }}[a-z]^{+} @[a-z]^{+} .[a-z]^{+} \dashv_{\text {email }} \Sigma^{*}
$$

Document spanner


## Naive skyline computation

$\Sigma^{*} \vdash_{\text {email }}[\mathrm{a}-\mathrm{z}]^{+} @[\mathrm{a}-\mathrm{z}]^{+} .[\mathrm{a}-\mathrm{z}]^{+} \dashv_{\text {email }} \Sigma^{*}$
Document spanner


Text document
"Maximal substrings" Domination relation

|  |
| :---: |
| email |
| $[42,47\rangle$ |
| $[41,47\rangle$ |
| $[40,47\rangle$ |
| $42,48\rangle$ |
| $[41,48\rangle$ |
| $[40,48\rangle$ |
| Raw result |

## Naive skyline computation

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(maximal results)

## Naive skyline computation

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"Maximal substrings" Domination relation

(maximal results)

- Can we be more efficient, i.e., avoiding materializing the raw result?
- Can we merge both steps, i.e., compile the domination relation in the spanner?


## Paper contributions and talk outline

- Introduce and formalize the skyline problem for regular spanners
$\rightarrow$ Propose a general framework to express domination relations


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$\rightarrow$ State complexity: does it blow up the spanner representation?


## Paper contributions and talk outline

- Introduce and formalize the skyline problem for regular spanners
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- Study if we can compile the skyline operator into the spanner
$\rightarrow$ Expressiveness: is it possible?
$\rightarrow$ State complexity: does it blow up the spanner representation?
- Study the problem of efficiently evaluating the skyline operator
$\rightarrow$ In data complexity and combined complexity


## Table of contents

Defining skylines via domination rules

## Compilation: Building a VA for the skyline

## Evaluation: Computing the skyline

Conclusion and further work

## Basics of spanners

- Document: string over an alphabet


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$$
d=\begin{array}{cccccccccccc}
J & \circ & h & \mathrm{n} & \sqcup & 4 & 5 & 6 & 1 & 2 & 3 & \\
0 & 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 & 9 & 10 & 11
\end{array}
$$

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$\rightarrow$ ex: for $X=\{x, y, z\}$, map $x$ to $[0,4\rangle$ and leave $y$ and $z$ unassigned


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$\rightarrow$ ex: for $X=\{x, y, z\}$, map $x$ to $[0,4\rangle$ and leave $y$ and $z$ unassigned
- Spanner: function that maps each document to a set of mappings

Defining spanners
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In practice, often more convenient to write in the subclass of regex-formulas:

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Other more general classes:

- Core spanners: featuring string equality selection
- Generalized core spanners: featuring difference


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Can we have a unified framework covering those?

## Examples of domination relations

## Extracted

mappings:

$$
\begin{array}{cc}
x & y \\
\hline[1,2\rangle & {[2,3\rangle} \\
- & {[2,3\rangle} \\
{[0,2\rangle} & {[2,3\rangle} \\
{[4,6\rangle} & {[4,10\rangle}
\end{array}
$$

## Examples of domination relations

| Extracted | Skyline under |
| :--- | :--- |
| mappings: | variable inclusion: |


| $x$ | $y$ |
| :---: | :---: |
| $[1,2\rangle$ | $[2,3\rangle$ |
| - | $[2,3\rangle$ |
| $[0,2\rangle$ | $[2,3\rangle$ |
| $[4,6\rangle$ | $[4,10\rangle$ |


| $x$ | $y$ |
| :---: | :---: |
| $[1,2\rangle$ | $[2,3\rangle$ |

$[0,2\rangle \quad[2,3\rangle$
$[4,6\rangle \quad[4,10\rangle$
$[4,6\rangle \quad[4,10\rangle$

## Examples of domination relations

| Extracted mappings: |  | Skyline under variable inclusion: |  | Skyline under spa inclusion: |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
| $x$ | y | $x$ | $y$ | $x$ | $y$ |
| [1, 2> | $[2,3\rangle$ | [1, 2> | [2,3> | - | [2,3> |
| - | [2,3> | [0, 2) | [2,3> | [0, 2) | $[2,3\rangle$ |
| [0, 2> | $[2,3\rangle$ | [4, 6> | $[4,10\rangle$ | [4, 6> | $[4,10\rangle$ |
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| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| $x$ | y | $x$ | $y$ | $x$ | $y$ | $x$ | $y$ |
| $[1,2\rangle$ | $[2,3\rangle$ | [1, 2> | [2,3> | - | [2,3) | - | $[2,3\rangle$ |
| - | $[2,3\rangle$ | [0, 2> | [2,3) | [0, 2> | $[2,3\rangle$ | $[4,6\rangle$ | $[4,10\rangle$ |
| [0, 2) | $[2,3\rangle$ | [4, 6> | $[4,10\rangle$ | [4, 6> | $[4,10\rangle$ |  |  |
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## Formalizing domination relations

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- Idea: domination pair $\left(m, m^{\prime}\right)$ can be seen as a mapping $\mu$, if we rename variables!
- Variables are $X \cup X^{\dagger}$, i.e., $\left\{x, y, x^{\dagger}, y^{\dagger}\right\}$
- Variables of $X$ are mapped by $\mu$ like in $m$
- For each variable $\boldsymbol{z} \in X$, variable $\boldsymbol{z}^{\dagger}$ is mapped by $\mu$ like $m^{\prime}(z)$


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- Example:
- Mapping $m$ maps $x$ to $[42,51\rangle$ and does not map $y$
- Mapping $m^{\prime}$ maps $x$ to $[42,51\rangle$ and maps $y$ to $[52,58\rangle$
- Then $\mu$ maps $x$ and $x^{\dagger}$ to $[42,51\rangle$, does not map $y$, and maps $y^{\dagger}$ to $[52,58\rangle$


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- Then $\mu$ maps $x$ and $x^{\dagger}$ to $[42,51\rangle$, does not map $y$, and maps $y^{\dagger}$ to $[52,58\rangle$
$\rightarrow$ We can define the domination relation as a spanner $D$, called a domination rule:
$\rightarrow$ Definition of spanner $D$ : given $d$, extract all mappings $\mu$ that code a domination pair


## Expressing domination relations via domination rules

Consider the example domination relations on a single variable $x$

- Trivial domination relation: no mapping dominates another


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Expressing domination relations via domination rules
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$$

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## Expressing domination relations via domination rules

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$\rightarrow$ Not expressible as a regular spanner


## Variable-wise rules

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$\rightarrow$ A rule is variable-wise if it is a product of copies of one single-variable rule $\rightarrow$ Covers all examples so far

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- Evaluation: from $A$ and $D$ and $d$, compute the skyline directly


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## Compilation: Expressiveness results

Remember the task:

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## Theorem

Given a VA A and a domination rule expressed as a VA D, we can compute a VA $A^{\prime}$ extracting the skyline $\eta_{D}(A)$

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$\rightarrow$ Already in the case where $D$ is the span inclusion or variable inclusion rule

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Given a VA A with n states, a VA $A^{\prime}$ computing the skyline $\eta(A)$ under variable inclusion needs $2^{\Omega(n)}$ states in general

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- Combined complexity for fixed rule: fix $D$, the input is the VA $A$ and the document $d$


## Evaluation in data complexity

The skyline is always tractable to compute in data complexity:

## Theorem

For any fixed VA A and domination rule D, given a document d, we can compute the skyline of $A$ on $d$ under D in PTIME in d

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- Then, simply run $A^{\prime}$ on $d$ to compute the maximal mappings


## Evaluation in combined complexity

Computing the skyline is intractable even under the variable inclusion rule

## Theorem

The following problem is NP-hard: given $n \in \mathbb{N}$, a VA $A$, and document $d$, decide if $A$ has more than $n$ mappings on $d$ that are maximal for variable inclusion

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In the paper:
- Sufficient condition for hardness: whenever a rule captures unboundedly many comparable pairs that are "disjoint", then skyline computation is hard
- Dichotomy on a subset of domination rules based on a variant of this condition
- Troubling asymmetry: there is a domination rule $\leq$ such that:
- Computing the skyline under $\leq$ is easy
- Computing the skyline under $\geq$ is hard!


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Main open questions:

- Are there other applications of the nFBDD correspondence?
$\rightarrow$ In the paper: exponential blowup for the join of schemaless regex formulas
- Can we get a dichotomy on all single-variable variable-wise rules?
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Thanks for your attention!

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- Given an nFBDD representing a Boolean function $\phi$ on variables $X$, we can easily compute a VA $A_{\phi}$ and document $d$ such that the mappings extracted by $A_{\phi}$ correspond to the satisfying assignments of $\phi$
$\rightarrow$ For any Boolean valuation $\nu: X \rightarrow\{\mathbf{0}, \mathbf{1}\}$, then $\nu$ satisfies $\phi$ if and only if $A_{\phi}$ extracts a mapping on $d$ that assigns $\{x \in X \mid \nu(x)=1\}$
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- But nFBDDs are exponentially less concise than other representations (read-3 monotone 2-CNF formulas)
- For such a formula $\psi$, we can build a VA $A_{\psi}$ whose skyline under variable inclusion corresponds to the satisfying assignments of $\psi$
$\rightarrow$ not expressible as a small nFBDD, hence not expressible by a small VA


## Hardness sketch

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- $r^{\prime}$ captures one mapping per clause $j$ : all literals $p_{i, j^{\prime}}$ and $n_{i, j^{\prime}}$ with $j^{\prime} \neq j$

Example: for $\Phi=\left(x_{1} \vee x_{2}\right) \wedge\left(\neg x_{2} \vee \neg x_{3}\right) \wedge\left(\neg x_{1} \wedge x_{3}\right)$ :

- $r$ assigns: $p_{1,1}$ or $n_{1,3}$; and $p_{2,1}$ or $n_{2,2}$; and $p_{3,3}$ or $n_{3,2}$
- $r^{\prime}$ assigns: all but $p_{1,1}$ and $p_{2,1}$; or all but $n_{2,2}$ and $n_{3,2}$; or all but $n_{1,3}$ and $p_{3,3}$


## Hardness sketch (cont'd)

- $r$ captures one mapping per valuation of $X$, i.e., for each variable $x_{i} \in X$ :
- either assign all the variables $p_{i, j}$ corresponding to positive literals of $x_{i}$
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- The $m$ mappings of $m^{\prime}$ are maximal (incomparable and not covered by $m$ )
- Other than that:
- If there is a maximal mapping from $m$, then it is not covered by $r^{\prime}$ so contains one literal per clause: $\Phi$ is satisfiable
- Otherwise, all assignments violate some clause: $\Phi$ is unsatisfiable

