## Compact Representation of Sets of Binary Constraints

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## Motivation: planning

- Practically all implementations of planning as satisfiability, have used a quadratic size translation from a problem instance to SAT.

Cliques
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## Motivation: general problem

- A binary relation (graph) on a set of $n$ objects may have $n^{2}$ elements (edges).
- If the relation/graph is dense and $n$ is high $\left(10^{4}>\right)$ the number of elements/edges can be very high $\left(10^{8}>\right)$.
- The representation of the elements/edges may become impractical.
- Goal: succinct representation of the relation/graph.


## Cliques in constraint graphs

## Definition

Let $\langle N, E\rangle$ be an undirected graph. Then a clique is
$C \subseteq N$ such that $\left\{n, n^{\prime}\right\} \in E$ for every $n, n^{\prime} \in C$ such that $n \neq n^{\prime}$.


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$O(n)$ Representation
$O(n \log n)$
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## Representation with $O(n)$ Size and $O(n)$ Auxiliary Variables


[Rintanen et al. 2005]

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## Representation with $\mathcal{O}(n \log n)$ size and $\mathcal{O}(\log n)$ auxiliary variables

Let $C=\left\{l_{0}, l_{2}, l_{3}, l_{4}, l_{5}, l_{6}, l_{7}\right\}$ be a clique consisting of 8 literals. Let $x_{0}, x_{1}, x_{2}$ be new Boolean variables.

$$
\begin{aligned}
& l_{0} \rightarrow\left(\neg x_{0} \wedge \neg x_{1} \wedge \neg x_{2}\right) \\
& l_{1} \rightarrow\left(\neg x_{0} \wedge \neg x_{1} \wedge x_{2}\right) \\
& l_{2} \rightarrow\left(\neg x_{0} \wedge x_{1} \wedge \neg x_{2}\right) \\
& l_{3} \rightarrow\left(\neg x_{0} \wedge x_{1} \wedge x_{2}\right) \\
& l_{4} \rightarrow\left(x_{0} \wedge \neg x_{1} \wedge \neg x_{2}\right) \\
& l_{5} \rightarrow\left(x_{0} \wedge \neg x_{1} \wedge x_{2}\right) \\
& l_{6} \rightarrow\left(x_{0} \wedge x_{1} \wedge \neg x_{2}\right) \\
& l_{7} \rightarrow\left(x_{0} \wedge x_{1} \wedge x_{2}\right)
\end{aligned}
$$

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In general, for $n$ literals there are $n\left\lceil\log _{2} n\right\rceil$ 2-literal clauses.

## Complexity of finding cliques

- Finding a maximum cardinality clique is NP-hard.
- Approximation to any constant factor is NP-hard.
- Of course, polynomial-time algorithms for finding cliques exist but they have no approximation guarantees.
- (Bicliques do have polynomial-time 2-approximation algorithms!)


## General compression procedure

(1) Find a big clique in the constraint graph.
(2) If only small cliques were found, go to the last step.
(3) Represent the clique compactly.
(4) Remove the edges of the clique from the constraint graph.

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## Bicliques

## Definition

Let $\langle N, E\rangle$ be an undirected graph. A biclique is a pair of
$C \subseteq N$ and $C^{\prime} \subseteq N$ such that $C \cap C^{\prime}=\emptyset$ and $\left\{\left\{n_{1}, n_{2}\right\} \mid n_{1} \in C, n_{2} \in C^{\prime}\right\} \subseteq E$.

The $n m$ edges of an $n, m$ biclique can be represented with only one auxiliary variable and $n+m$ edges.


## Every clique is also a biclique



## Every clique is also a biclique



## Example: one 8-clique as three 4,4-bicliques



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## Example: one 8-clique as three 4,4-bicliques


$000 \rightarrow x_{0}, x_{0} \rightarrow \neg 100$
$001 \rightarrow x_{0}, x_{0} \rightarrow \neg 101$
$010 \rightarrow x_{0}, x_{0} \rightarrow \neg 110$
$011 \rightarrow x_{0}, x_{0} \rightarrow \neg 111$

$000 \rightarrow x_{1}, x_{1} \rightarrow \neg 010$
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## Example: one 8-clique as three 4,4-bicliques



## Example: one 8-clique as three 4,4-bicliques

$$
\begin{aligned}
& 000 \rightarrow x_{0}, x_{0} \rightarrow \neg 100 \\
& 001 \rightarrow x_{0}, x_{0} \rightarrow \neg 101 \\
& 010 \rightarrow x_{0}, x_{0} \rightarrow \neg 110 \\
& 011 \rightarrow x_{0}, x_{0} \rightarrow \neg 111 \\
& \\
& \\
& 000 \rightarrow x_{1}, x_{1} \rightarrow \neg 010 \\
& \text { Molivation } \\
& \text { Ciques } \\
& \text { Bicliques } \\
& \text { Cliques vs. } \\
& \text { Bicilques } \\
& 100 \rightarrow x_{1}, x_{1} \rightarrow \neg 011 \\
& \text { Application } \\
& 101 \rightarrow x_{1}, x_{1} \rightarrow \neg 110 \\
& \text { Conclusion }
\end{aligned}
$$

## Example: one 8 -clique as three 4,4-bicliques

It's equivalent to the $n \log _{2} n$ encoding of cliques!

$$
\begin{aligned}
& 000 \rightarrow x_{0}, 100 \rightarrow \neg x_{0} \\
& 001 \rightarrow x_{0}, 101 \rightarrow \neg x_{0} \\
& 010 \rightarrow x_{0}, 110 \rightarrow \neg x_{0} \quad \begin{array}{l}
\text { Molivation } \\
011 \rightarrow x_{0}, 111 \rightarrow \neg x_{0} \\
\\
000 \rightarrow x_{1}, 010 \rightarrow \neg x_{1} \\
\text { Ciques } \\
\text { Biciques } \\
\text { Cliques vs. } \\
\text { Biciques } \\
\text { Application }
\end{array} \\
& 100 \rightarrow x_{1}, 011 \rightarrow \neg x_{1}, 110 \rightarrow \neg x_{1} \\
& 101 \rightarrow x_{1}, 111 \rightarrow \neg x_{1} \\
& \text { Conclusion }
\end{aligned}
$$

## Example: IPC Airport Problem

- Problem represents the movement of airplanes at an airport.
- Constraints on the airplane movement
- Halfway the instance series the formula sizes exceed 1 GB. Culprit: binary invariants/mutexes
- All problems this far solvable in seconds: it's the physical size, not the actual difficulty.


## Example: Compression of the Constraint Graph

Constraint graph with 62 nodes and 653 edges


## Example: Compression of the Constraint Graph

Replacing $13 \times 16=208$ by $13+16=29$ edges.


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## Example: Compression of the Constraint Graph

Replacing $11 \times 18=198$ by $11+18=29$ edges.


## Example: Compression of the Constraint Graph

$$
\text { Replacing } 11 \times 7=77 \text { by } 11+7=18 \text { edges } .
$$



## Example: Compression of the Constraint Graph

Replacing $10 \times 7=70$ by $10+7=17$ edges.


## Example: Compression of the Constraint Graph

Replacing $8 \times 8=64$ by $8+8=16$ edges.


## Example: Compression of the Constraint Graph

Replacing $6 \times 6=36$ by $6+6=12$ edges.


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## Example: Compression of the Constraint Graph

## Total reduction is from 653 to 121 edges.



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## Example: IPC Airport Problem

| instance | clauses for invariants |  | size in MB |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  | before | after | before | after | Motiv |
| 21_4halfMUC_P2 | 182094 | 13191 | 2.59 | 0.37 | clioues |
| 22_4halfMUC_P3 | 275927 | 21388 | 4.06 | 0.58 | Bicial |
| 23_4halfMUC_P4 | 381675 | 31776 | 5.60 | 0.84 |  |
| 24_4halfMUC_P4 | 383791 | 30407 | 5.72 | 0.90 | Biciques |
| 25_4halfMUC_P5 | 478455 | 41719 | 7.24 | 1.18 | Application |
| 26_4halfMUC_P6 | 587951 | 50247 | 8.85 | 1.43 |  |
| 27_4halfMUC_P6 | 572292 | 53721 | 9.01 | 1.57 | Conolusion |
| 28_4halfMUC_P7 | 670530 | 66060 | 10.62 | 1.89 |  |
| 36_5MUC_P2 | 325136 | 18872 | 4.68 | 0.52 |  |
| 37_5MUC_P3 | 490971 | 30681 | 7.40 | 0.93 |  |
| 38_5MUC_P3 | 487600 | 29464 | 7.30 | 0.86 |  |
| 39_5MUC_P4 | 655616 | 44647 | 10.08 | 1.34 |  |
| 40_5MUC_P4 | 657309 | 43872 | 10.04 | 1.27 |  |
| 41_5MUC_P4 | 653940 | 42314 | 9.93 | 1.20 |  |

## Other domains and applications

- The size reduction for many other problems is far less dramatic: 10, 30, 50 per cent.


## Other domains and applications

- The size reduction for many other problems is far less dramatic: 10, 30, 50 per cent.
- Action mutexes / interference constraints:
- Trivial $\mathcal{O}\left(n^{2}\right)$ representation (used in BLACKBOX, SatPlan, ...) catastrophic for big problems.
- We have given (Rintanen et al. 2005, 2007) linear encodings: very good scalability in comparison to BLACKBOX/SatPlan.
- Surprisingly, the biclique reduction is often better than the linear encoding, but in few cases far worse.


## Conclusions

- We presented a biclique based technique for representing sets of 2-literal clauses more compactly (sometimes much more).
- The basic idea is very simple and widely applicable.
- Quadratic worst-case cannot be eliminated (there is a simple argument showing this.)
- We have shown how compression with cliques is a special case of compression with bicliques.
- Challenges: more efficient algorithms for finding big cliques and bicliques

