

# Solving larger instances by reducing to a computationally harder problem

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# Motivation



## Problem

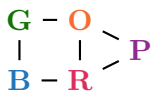
- ▶ There will always be an instance whose encoding is too big.
- ▶ Smaller encoding does not always lead to faster solving.



## Goal

Make encoding **exponentially** more succinct without sacrificing speed.

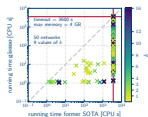
# Main contributions



A **case study** that reduces the NP-hard **generalised identifying code set (GICS)** problem to the computationally harder **independent support** problem.



A new solver: **gismo**.

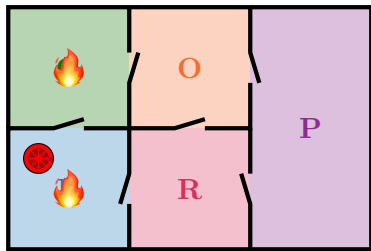


Experiments that **demonstrate the effectiveness** of our reduction and **gismo**.

# Problem (1/4)

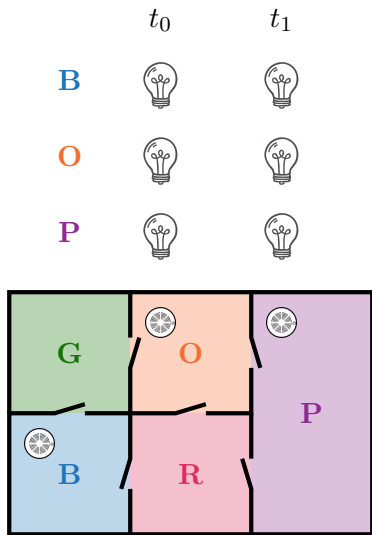



# Problem (2/4)



	signature	
	$t_0$	$t_1$
$\{B\}$	$\{B\}$	$\{B\}$
$\{G\}$	$\emptyset$	$\{B\}$
$\{O\}$	$\emptyset$	$\emptyset$
$\{R\}$	$\emptyset$	$\{B\}$
$\{P\}$	$\emptyset$	$\emptyset$
$\{B, G\}$	$\{B\}$	$\{B\}$
$\{B, O\}$	$\{B\}$	$\{B\}$
$\vdots$	$\vdots$	$\vdots$
$\{O, R, P\}$	$\emptyset$	$\{B\}$
$\vdots$	$\vdots$	$\vdots$
$\emptyset$	$\emptyset$	$\emptyset$

# Problem (3/4)

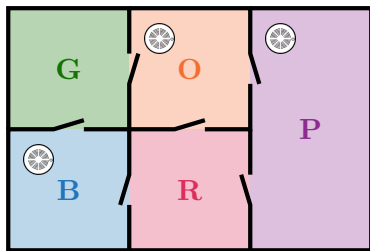



	signature	
		
	$t_0$	$t_1$
<b>B</b>	$\{B\}$	$\{B\}$
<b>O</b>	$\emptyset$	$\{B, O\}$
<b>P</b>	$\{O\}$	$\{O, P\}$
<b>R</b>	$\emptyset$	$\{B, O, P\}$
<b>P</b>	$\{P\}$	$\{O, P\}$
$\{B, G\}$	$\{B\}$	$\{B, O\}$
$\{B, O\}$	$\{B, O\}$	$\{B, O\}$
$\vdots$	$\vdots$	$\vdots$
$\{O, R, P\}$	$\{O, P\}$	$\{B, O, P\}$
$\vdots$	$\vdots$	$\vdots$
$\emptyset$	$\emptyset$	$\emptyset$

## Problem (4/4)

The set of rooms with a detector,  $D$ , is called a **generalised identifying code set (GICS)** (Karpovsky, Chakrabarty, and Levitin 1998) for positive integer  $k$  if each set of at most  $k$  fires has a unique signature.

**Problem:** minimise  $|D|$



	signature	
	$t_0$	$t_1$
{B}	{B}	{B}
{G}	$\emptyset$	{B}
{O}	{O}	{O, P}
{R}	$\emptyset$	{B, O, P}
{P}	{P}	{O, P}
$\emptyset$	$\emptyset$	$\emptyset$

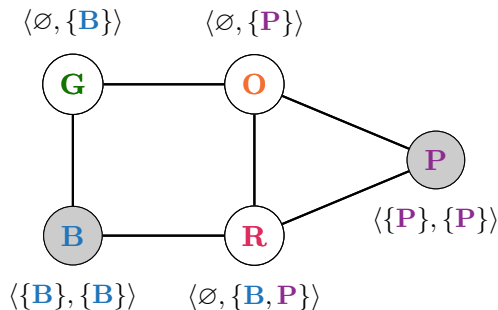
**Example:**


$$k = 1, D = \{B, O, P\}$$

## Problem (4/4)

The set of rooms with a detector,  $D$ , is called a **generalised identifying code set (GICS)** (Karpovsky, Chakrabarty, and Levitin 1998) for positive integer  $k$  if each set of at most  $k$  fires has a unique signature.

**Problem:** minimise  $|D|$



	signature	
	$t_0$	$t_1$
$\{B\}$	$\{B\}$	$\{B\}$
$\{G\}$	$\emptyset$	$\{B\}$
$\{O\}$	$\emptyset$	$\{P\}$
$\{R\}$	$\emptyset$	$\{B, P\}$
$\{P\}$	$\{P\}$	$\{P\}$
$\emptyset$	$\emptyset$	$\emptyset$

**Example:**

$k = 1$ ,  $D = \{B, P\}$



# Applications of Identifying Code Sets



Identifying sources of misinformation (Basu and Sen 2021a).



Identifying criminals in social networks (Basu and Sen 2021b).



Satellite deployment (Sen, Goliber, Basu, Zhou, and Ghosh 2019),  
(Latour, Sen, Basu, Zhou, and Meel 2024).

# Solving the GICS problem

## Former state of the art

(Padhee, Biswas, Pal, Basu, and Sen 2020)

Encode as integer-linear program (ILP).

- ▶ #constraints **exponential** in  $k$ .
- ▶ Checking if candidate is a solution: **polytime**.
- ▶ **Cardinality-minimal** solution  $D$ .

## New approach (contribution)

Reduce GICS problem to finding a minimal **independent support** (IS).

- ▶ #clauses **linear** in  $k$ .
- ▶ Checking if candidate is an IS: **co-NP**.
- ▶ **Set-minimal** solution  $D$ .

## Background: Propositional Logic

Solution  $\sigma : X \mapsto \{0, 1\}$  maps variables to truth values.

Example:  $F(X) := (x_1 \vee x_2) \leftrightarrow x_3$

	$x_1$	$x_2$	$x_3$
$\sigma_1$	1	1	1
$\sigma_2$	1	0	1
$\sigma_3$	0	1	1
$\sigma_4$	0	0	0

Projection set:  $S := \{x_1, x_3\}$

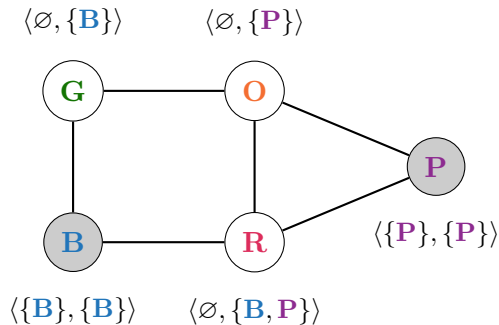
$$|Sol_{\downarrow S}(F)| \leq |Sol(F)|$$

	$x_1$	$x_2$	$x_3$
$\sigma_1$	1	1	1
$\sigma_2$	1	0	1
$\sigma_3$	0	1	1
$\sigma_4$	0	0	0

Projection set:  $I := \{x_1, x_2\}$  is an **independent support** (Chakraborty, Fremont, Meel, Seshia, and Vardi 2014) of  $F(X)$ .

$$|Sol_{\downarrow I}(F)| = |Sol(F)|$$

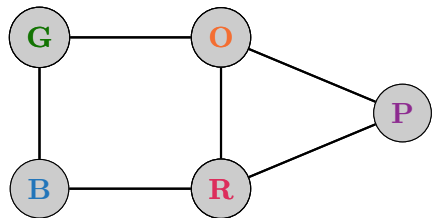
# Contribution: Reduction of GICS to GIS



## Our method

- ▶ Encode GICS in CNF formula
  - ▶ each solution corresponds to the signature  $s_U$  of a  $U \subseteq V$  with  $|U| \leq k$ ;
  - ▶ **linear** size.
- ▶ Two variables per node, *grouped*.
- ▶ Independent support encodes solution  $D$ .


## Example



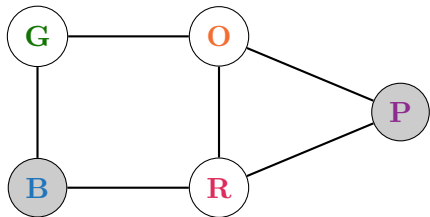
Two variables per node, e.g.,

►  $x_B$  models  at  $t_0$

►  $y_B$  models  at  $t_1$


	$X (S_U^0, \text{at } t_0)$					$Y (S_U^1, \text{at } t_1)$					$S_U^0$	$S_U^1$
	$x_B$	$x_G$	$x_O$	$x_R$	$x_P$	$y_B$	$y_G$	$y_O$	$y_R$	$y_P$		
$\emptyset$	0	0	0	0	0	0	0	0	0	0	$\emptyset$	$\emptyset$
$\{B\}$	1	0	0	0	0	1	1	0	1	0	$\{B\}$	$\{B, G, R\}$
$\{G\}$	0	1	0	0	0	1	1	1	0	0	$\{G\}$	$\{B, G, O\}$
$\{O\}$	0	0	1	0	0	0	1	1	1	1	$\{O\}$	$\{G, O, P, R\}$
$\{R\}$	0	0	0	1	0	1	0	1	1	1	$\{R\}$	$\{B, O, P, R\}$
$\{P\}$	0	0	0	0	1	0	0	1	1	1	$\{P\}$	$\{O, P, R\}$


## Example



Two variables per node, e.g.,

►  $x_B$  models  at  $t_0$

►  $y_B$  models  at  $t_1$

	$X (S_U^0, \text{at } t_0)$					$Y (S_U^1, \text{at } t_1)$					$S_U^0$	$S_U^1$
	$x_B$	$x_G$	$x_O$	$x_R$	$x_P$	$y_B$	$y_G$	$y_O$	$y_R$	$y_P$		
$\emptyset$	0	0	0	0	0	0	0	0	0	0	$\emptyset$	$\emptyset$
$\{B\}$	1	0	0	0	0	1	1	0	1	0	$\{B\}$	$\{B\}$
$\{G\}$	0	1	0	0	0	1	1	1	0	0	$\emptyset$	$\{B\}$
$\{O\}$	0	0	1	0	0	0	1	1	1	1	$\emptyset$	$\{P\}$
$\{R\}$	0	0	0	1	0	1	0	1	1	1	$\emptyset$	$\{B, P\}$
$\{P\}$	0	0	0	0	1	0	0	1	1	1	$\{P\}$	$\{P\}$

# Results

## Size

Largest network ( $|V|$ ):

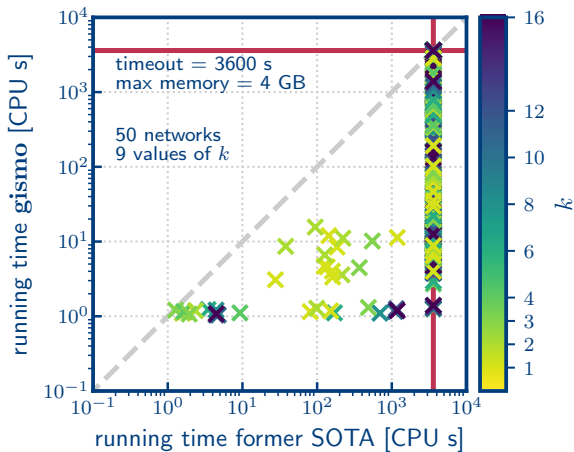
	encoded	solved
SOTA	494	494
<b>gismo</b>	227 320	21 363
improvement	460×	43×

SOTA:  $k = 1$

**gismo**: for all tested  $k$ .

Majority of instances: cardinality of solution close or equal to optimum.

## Time



Reducing to a **computationally harder problem**  
allows us to model and  
**solve much larger problem instances.**

full paper



[www.ijcai.org/proceedings/2023/219](http://www.ijcai.org/proceedings/2023/219)

gismo



[github.com/meelgroup/gismo](https://github.com/meelgroup/gismo)

more info



[www.annalatur.nl/publications](http://www.annalatur.nl/publications)



# References I

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