

# Efficient Model Transformations For Novices

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An overview of known techniques

# Graph Pattern Matching

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## ◆ Search Plans

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defines traversal order

based on heuristic

## ◆ Constraint Satisfaction Problems (CSP)

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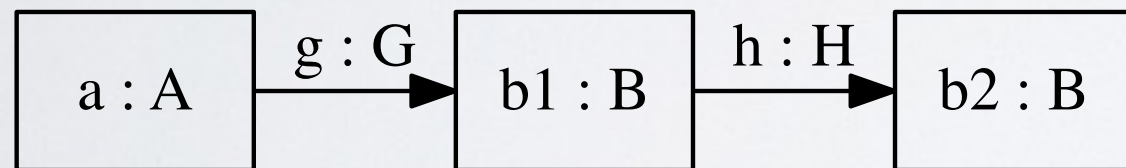
$\langle X, D, C \rangle$

set of objects

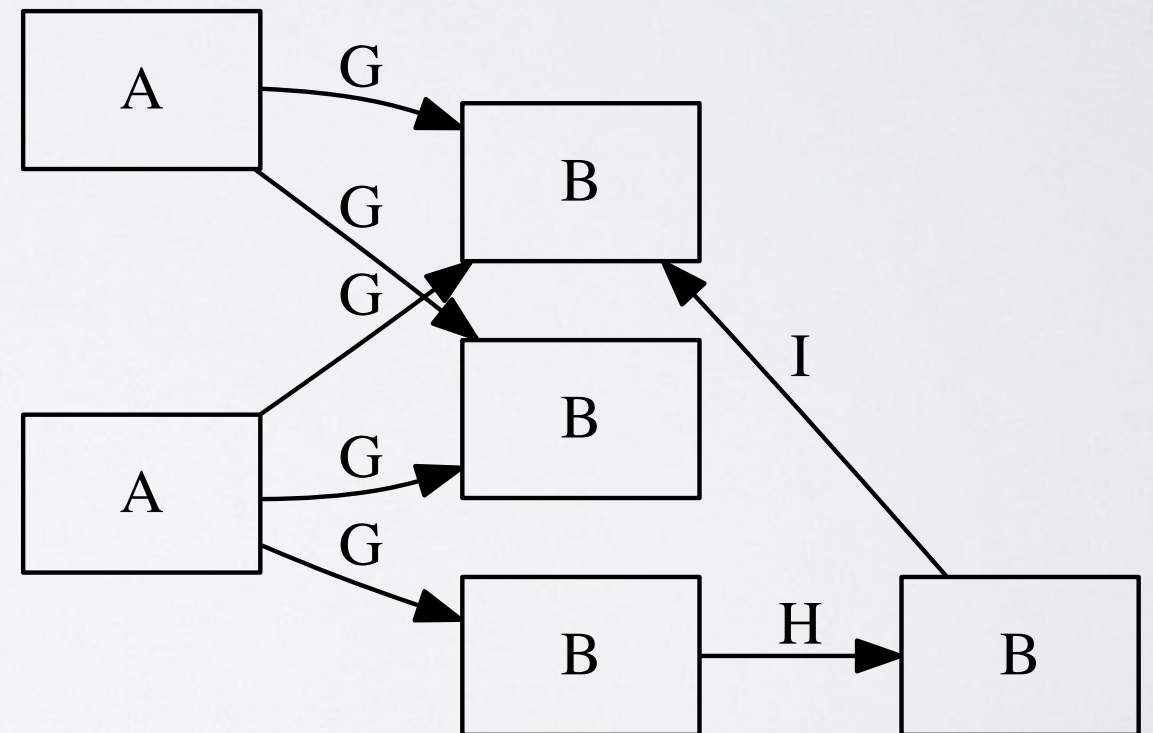
based on constraints

# Search Plan\*

## ◆ Pattern Graph P



## ◆ Host Graph H



\*: G.V. Batz, M. Kroll, R. Geiß, A first experimental evaluation of search plan driven graph pattern matching, in: Applications of Graph Transformations with Industrial Relevance, Springer, 2008, pp. 471–486.

# Primitive Matching Operations

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• <b>lkp</b> ( $x$ )	binds pattern element $x$ to host graph element
• <b>in</b> ( $v, e$ )	binds incoming pattern edge $e$ to incoming edge on already bound $v$
• <b>out</b> ( $v, e$ )	binds outgoing pattern edge $e$ to outgoing edge on already bound $v$
• <b>src</b> ( $e$ )	binds source of already bound edge $e$
• <b>tgt</b> ( $e$ )	binds target of already bound edge $e$



# Valid Search Plan

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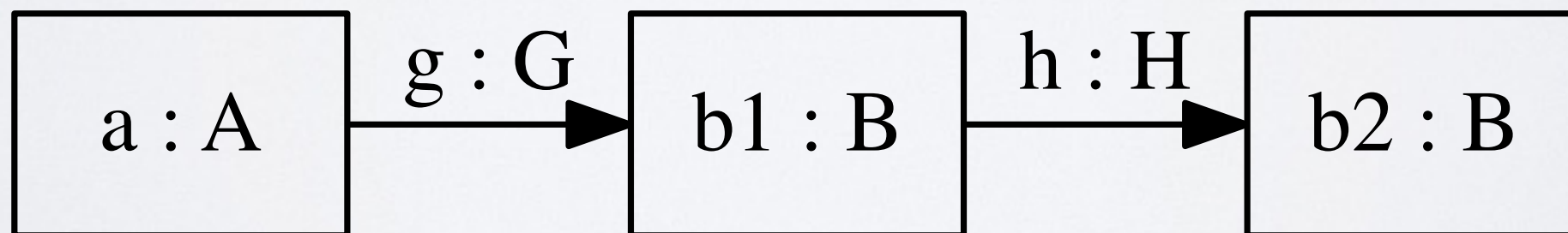
$P = \langle o_1, \dots, o_n \rangle$  is valid:

- If every element in pattern graph is bound **exactly** once
- If  $o_i$  requires a bound element  $x$ ,  $x$  must be bound in one of  $o_1, \dots, o_{i-1}$

# Example Search Plan

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- $P_1 = \langle \mathbf{out}(a, g), \mathbf{lkp}(b_1), \mathbf{tgt}(g), \mathbf{lkp}(h) \rangle$
- $P_2 = \langle \mathbf{lkp}(b_1), \mathbf{out}(b_1, h), \mathbf{in}(b_1, h), \mathbf{tgt}(h), \mathbf{src}(g) \rangle$
- $P_3 = \langle \mathbf{lkp}(h), \mathbf{tgt}(h), \mathbf{src}(h), \mathbf{in}(b_1, g), \mathbf{src}(g) \rangle$

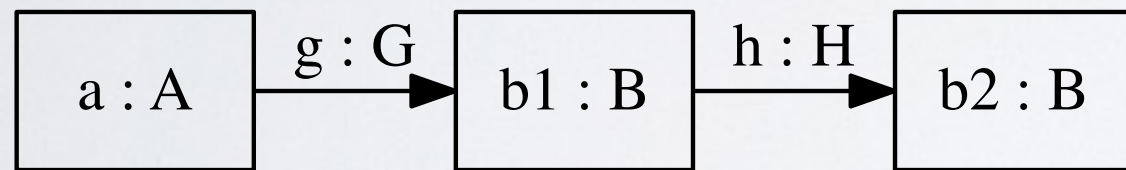


$$P_2 = \langle \mathbf{lkp}(b_1), \mathbf{out}(b_1, h), \mathbf{in}(b_1, h), \mathbf{tgt}(h), \mathbf{src}(g) \rangle$$

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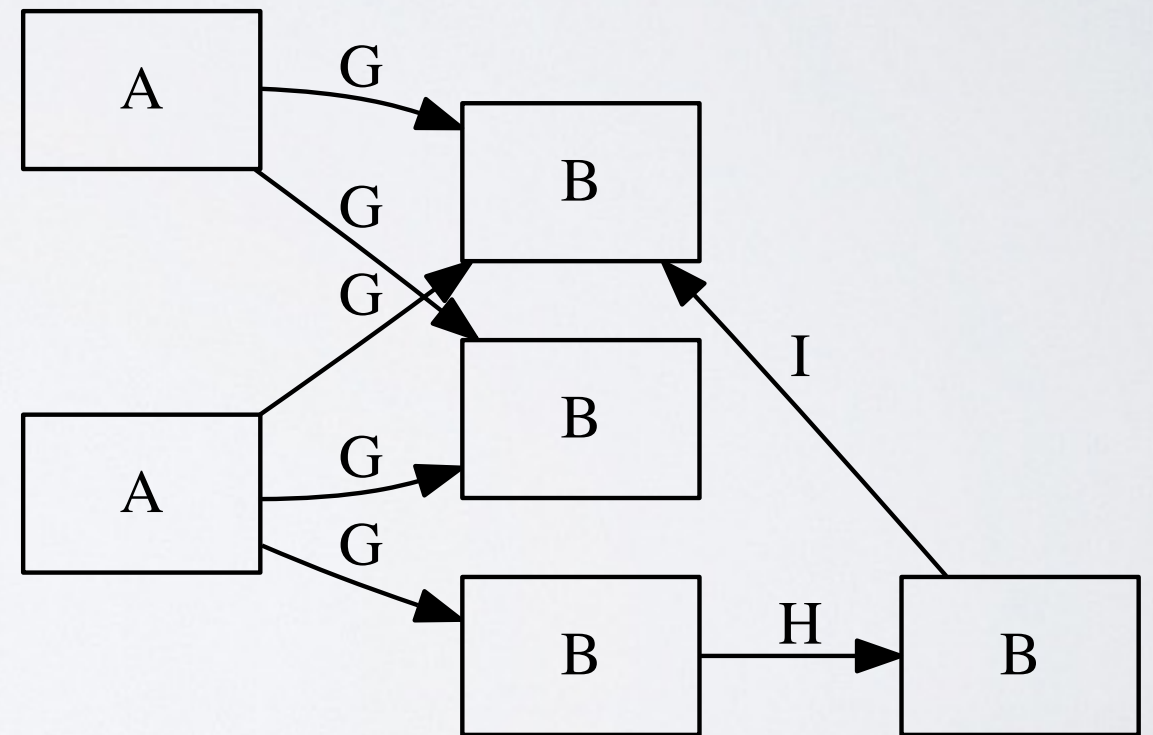
◆ Pattern Graph P

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◆ Host Graph H

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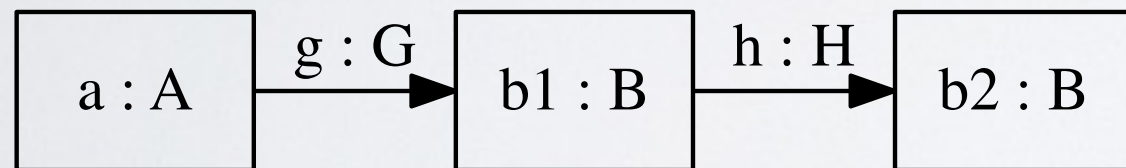


$$P_2 = \langle \mathbf{lkp}(b_1), \mathbf{out}(b_1, h), \mathbf{in}(b_1, h), \mathbf{tgt}(h), \mathbf{src}(g) \rangle$$

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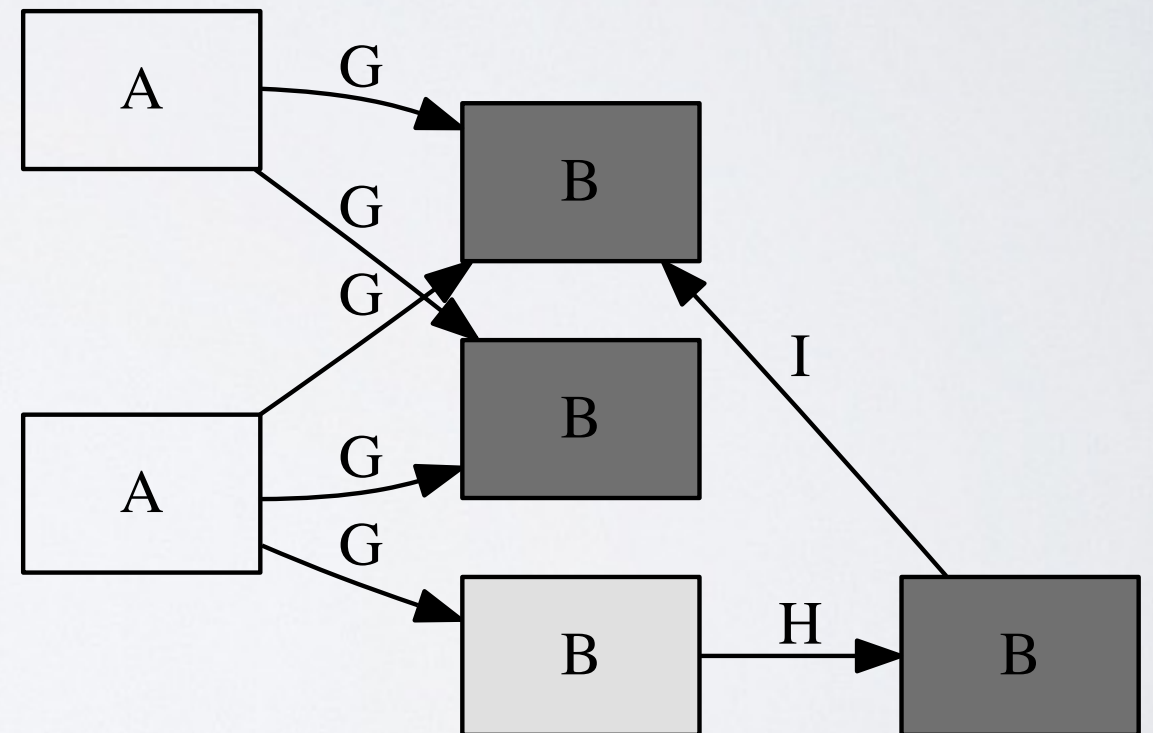
◆ Pattern Graph P

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◆ Host Graph H

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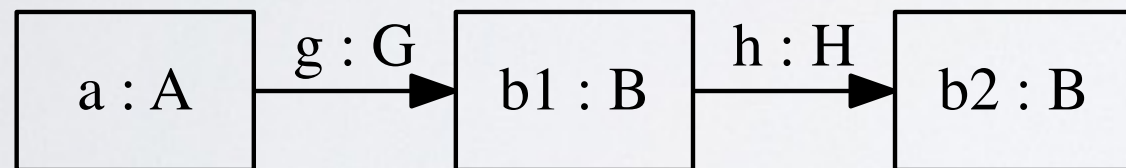


$$P_2 = \langle \mathbf{lkp}(b_1), \mathbf{out}(b_1, h), \mathbf{in}(b_1, h), \mathbf{tgt}(h), \mathbf{src}(g) \rangle$$

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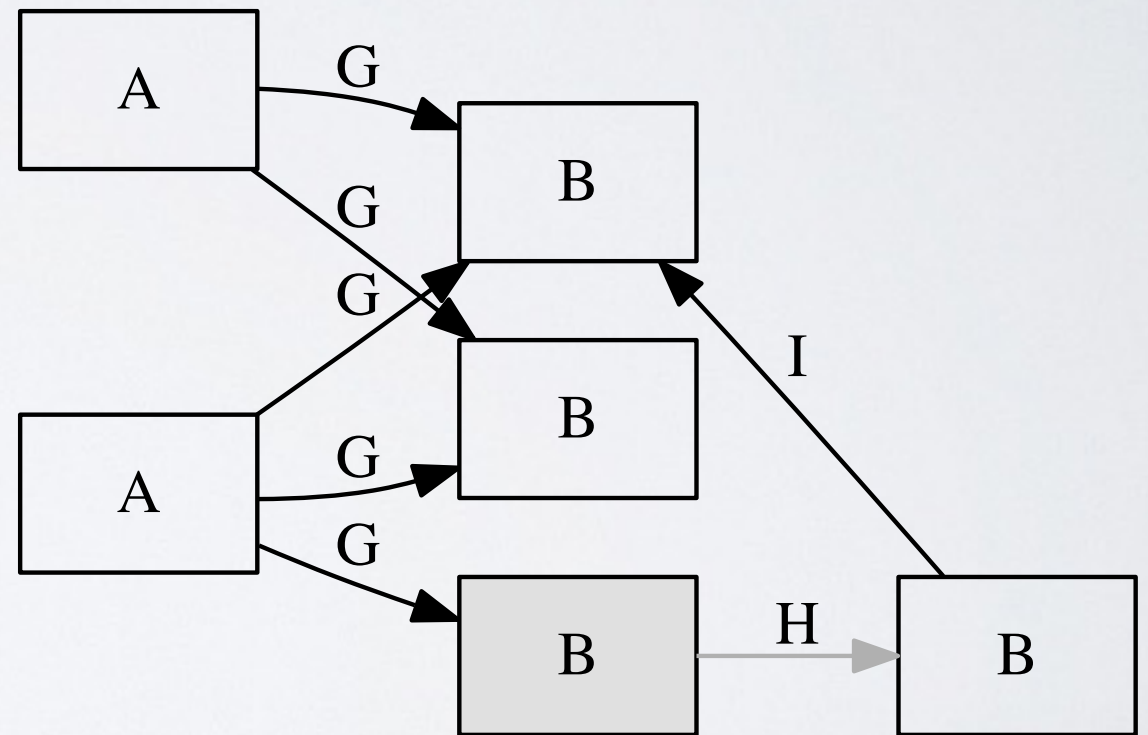
◆ Pattern Graph P

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◆ Host Graph H

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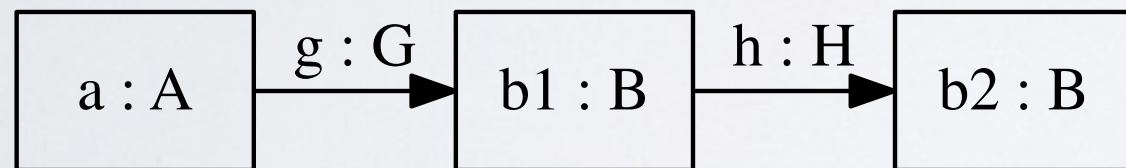


$$P_2 = \langle \mathbf{lkp}(b_1), \mathbf{out}(b_1, h), \mathbf{in}(b_1, h), \mathbf{tgt}(h), \mathbf{src}(g) \rangle$$

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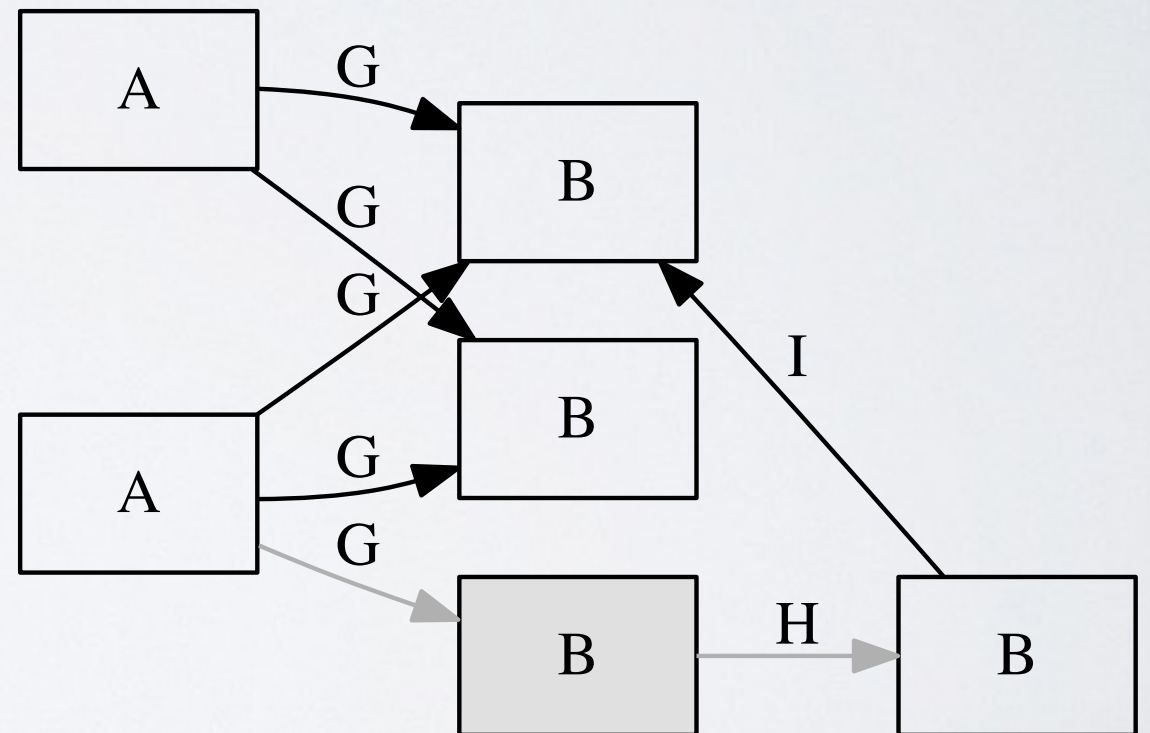
◆ Pattern Graph P

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◆ Host Graph H

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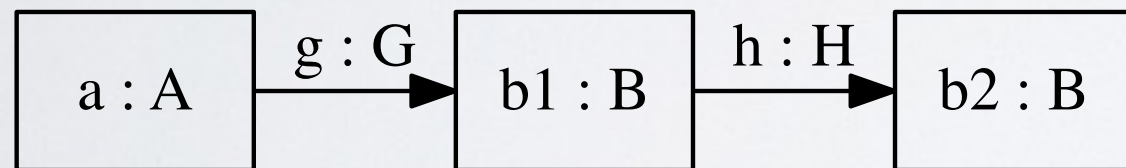


$$P_2 = \langle \mathbf{lkp}(b_1), \mathbf{out}(b_1, h), \mathbf{in}(b_1, h), \mathbf{tgt}(h), \mathbf{src}(g) \rangle$$

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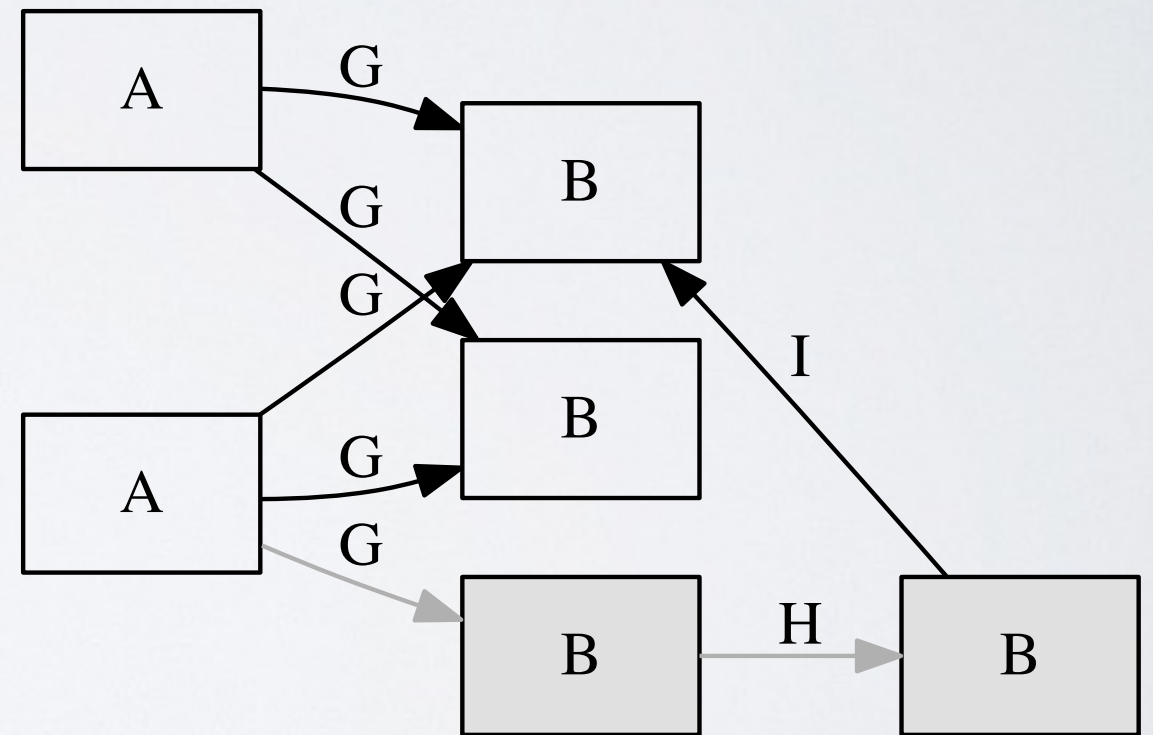
◆ Pattern Graph P

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◆ Host Graph H

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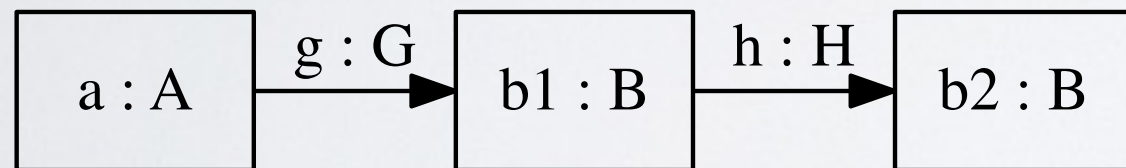


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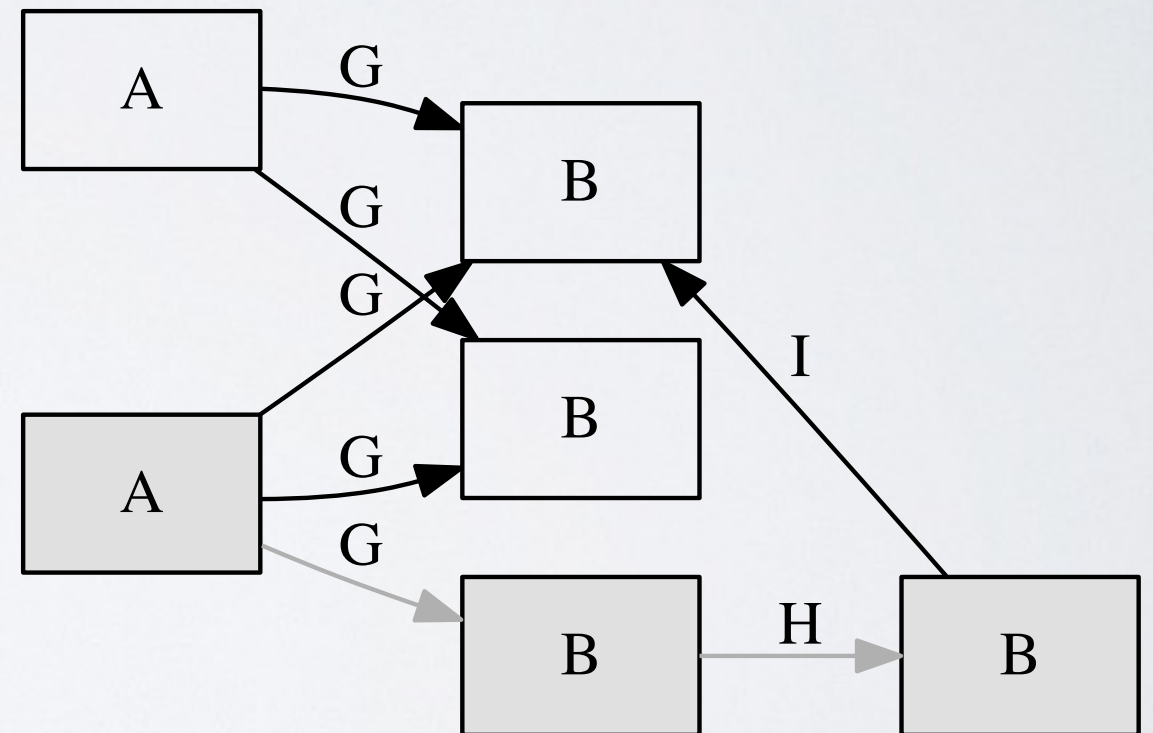
◆ Pattern Graph P

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◆ Host Graph H

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# Cost Model

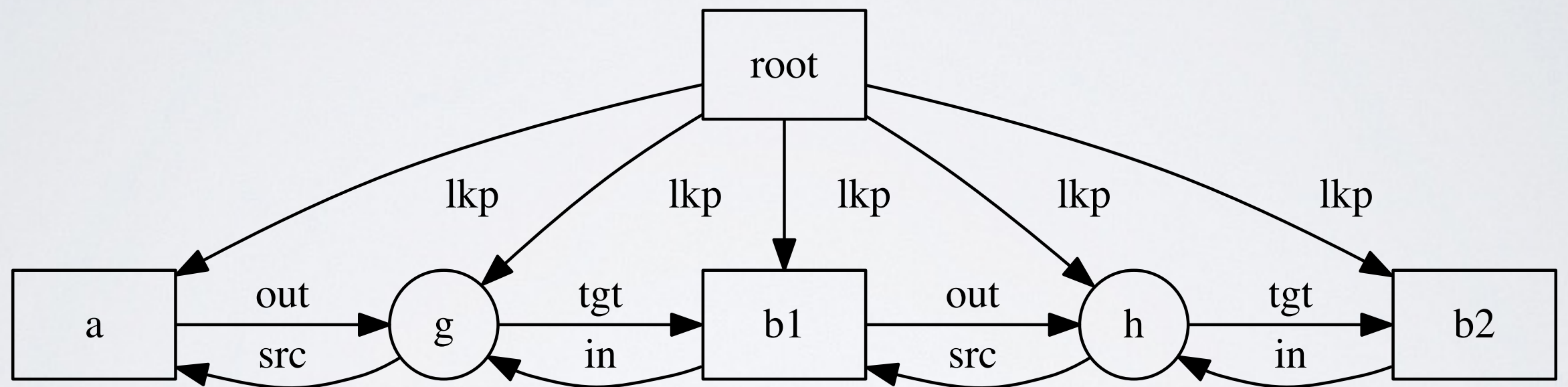
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• <b>lkp</b> ( $x$ )	$ \text{typeof}(x) \text{ in } H $
• <b>in</b> ( $v, e$ ), <b>out</b> ( $v, e$ )	AVG(possibilities)
• <b>src</b> ( $e$ ), <b>tgt</b> ( $e$ )	1
• $P = \langle o_1, \dots, o_2 \rangle$	$\sum_{i=1}^k \prod_{j=1}^i c_j$



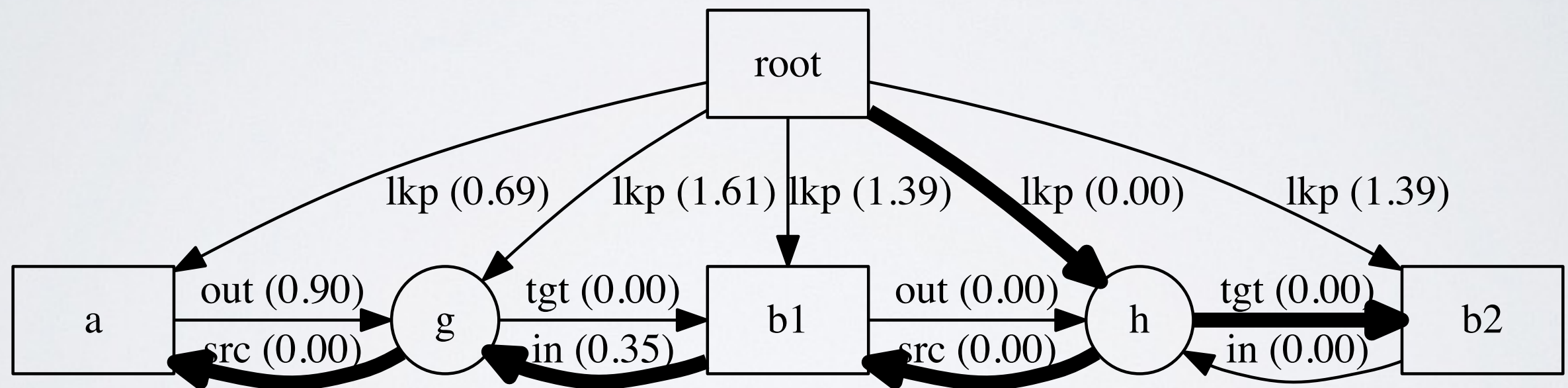
# Plan Graph

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# Plan Graph With Cost

$$c(P) = c_1 + c_1c_2 + \dots + c_1\dots c_k$$



MDST with Edmonds\* algorithm

\*: J. Edmonds, Optimum branchings, Journal of Research of the National Bureau of Standards B 71 (4) (1967) 233–240.

# Alternative Techniques

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- Optimisation problem, different heuristic
- Heuristic with type info, cardinality constraints,...
- Adaptive search plans<sup>1</sup>
- Indexing on type and/or by storing reverse associations, caching, pivoting and overlapped pattern matching<sup>2</sup>

1: G. Varr'ó, K. Friedl, D. Varr'ó, Adaptive graph pattern matching for model transformations using model-sensitive search plans, Electronic Notes in Theoretical Computer Science 152 (2006) 191–205.

2: C. A. G. Gomes, A framework for efficient model transformations.

# Constraint Satisfaction Problems



# Definition

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- $\langle X, D, C \rangle$
- $X$  = elements of pattern graph: variables
- $D$  = set of domains, for each variable
- $C$  = links and attributes

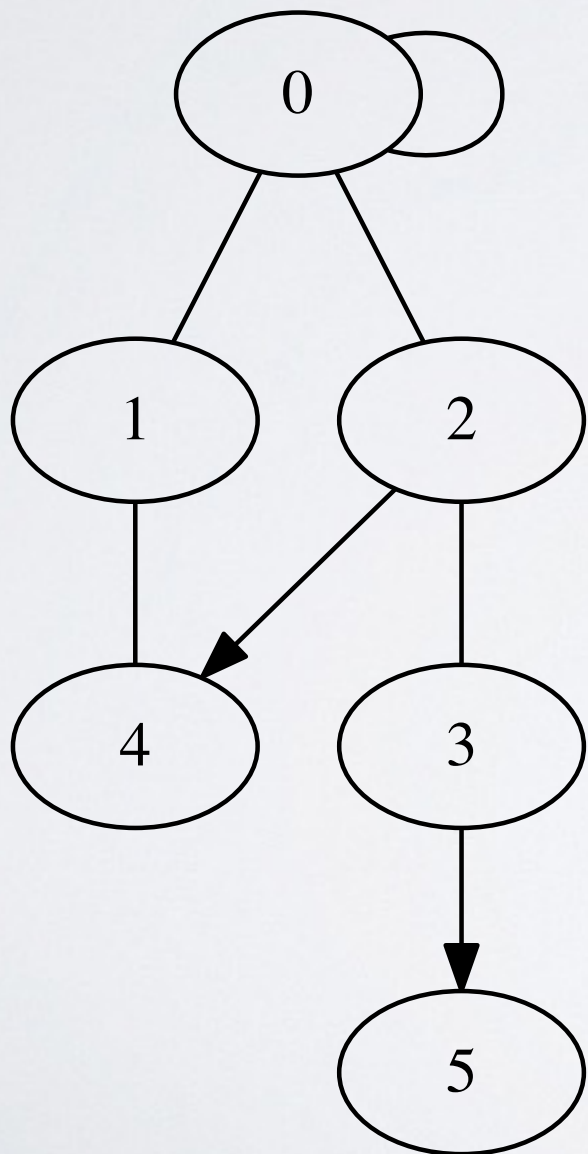


# Adjacency Matrices

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## ◆ Example Graph

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## ◆ Adjacency Matrix I

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x	0	1	2	3	4	5
0	2	1	1	0	0	0
1	1	0	0	0	1	0
2	1	0	0	1	1	0
3	0	0	1	0	0	1
4	0	1	0	0	0	0
5	0	0	0	0	0	0

# Ullmann\*

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$$\mathbf{M}^*[v, w] = \begin{cases} 1, & \text{if } \deg(v) \leq \deg(w) \text{ for } v \in V_P \text{ and } w \in V_H \\ 0, & \text{otherwise.} \end{cases}$$

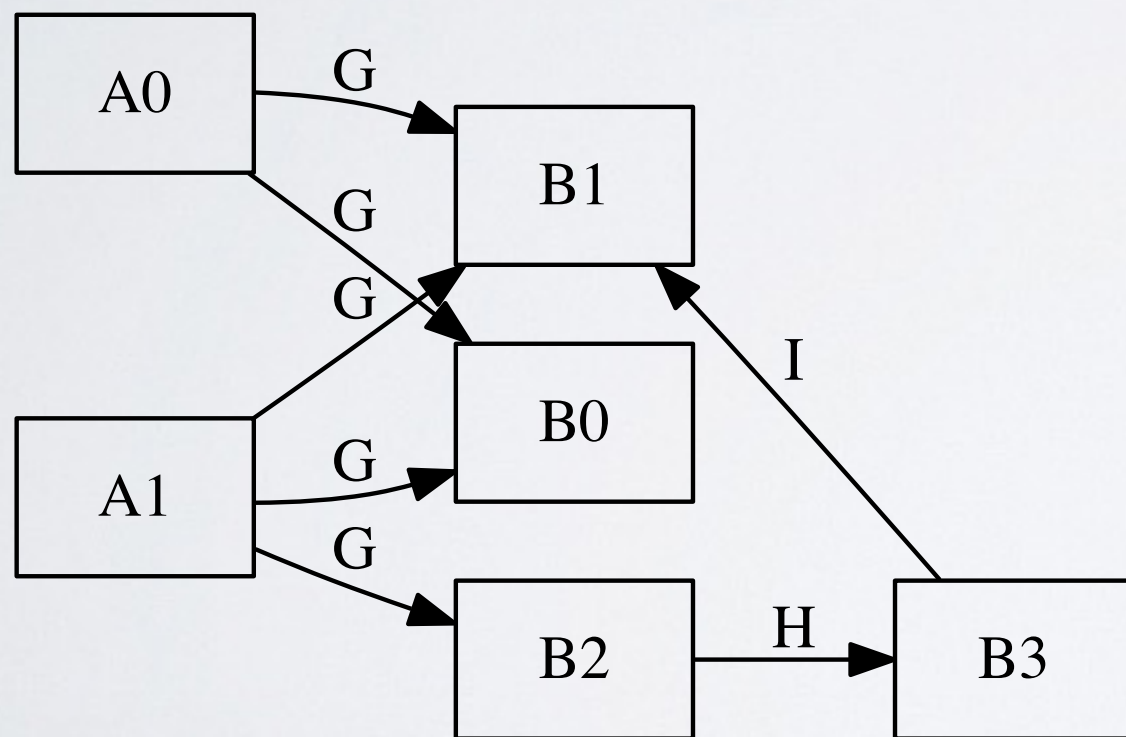
$\deg : V \rightarrow \mathbb{N}^+$  mapping vertex to its degree

isomorph iff  $\mathbf{M}(\mathbf{M}\mathbf{H})^T = \mathbf{P}$

\*: J. R. Ullmann, An algorithm for subgraph isomorphism, Journal of the ACM (JACM) 23 (1) (1976) 31–42.

# Ullmann Example

◆ Relabelled Host Graph

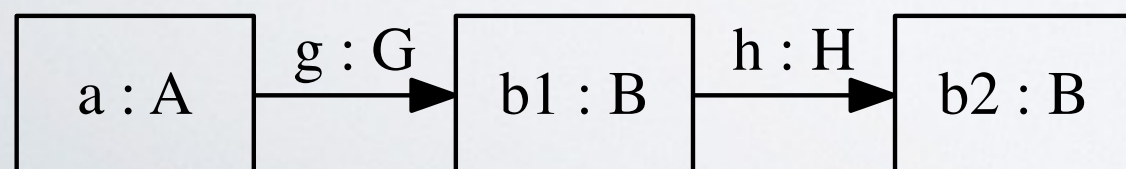
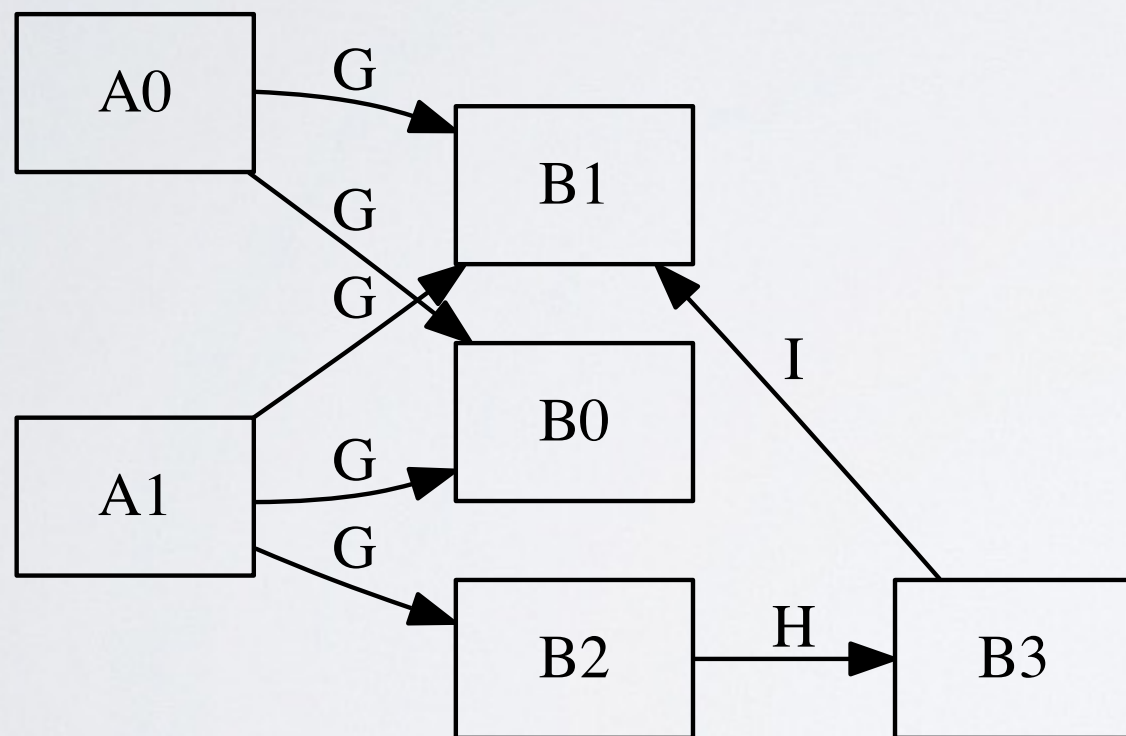


◆ Adjacency Matrix  $H$

$\times$	A0	A1	B0	B1	B2	B3
A0	0	0	1	1	0	0
A1	0	0	1	1	1	0
B0	0	0	0	0	0	0
B1	0	0	0	0	0	0
B2	0	0	0	0	0	1
B3	0	0	0	1	0	0

# Ullmann Example

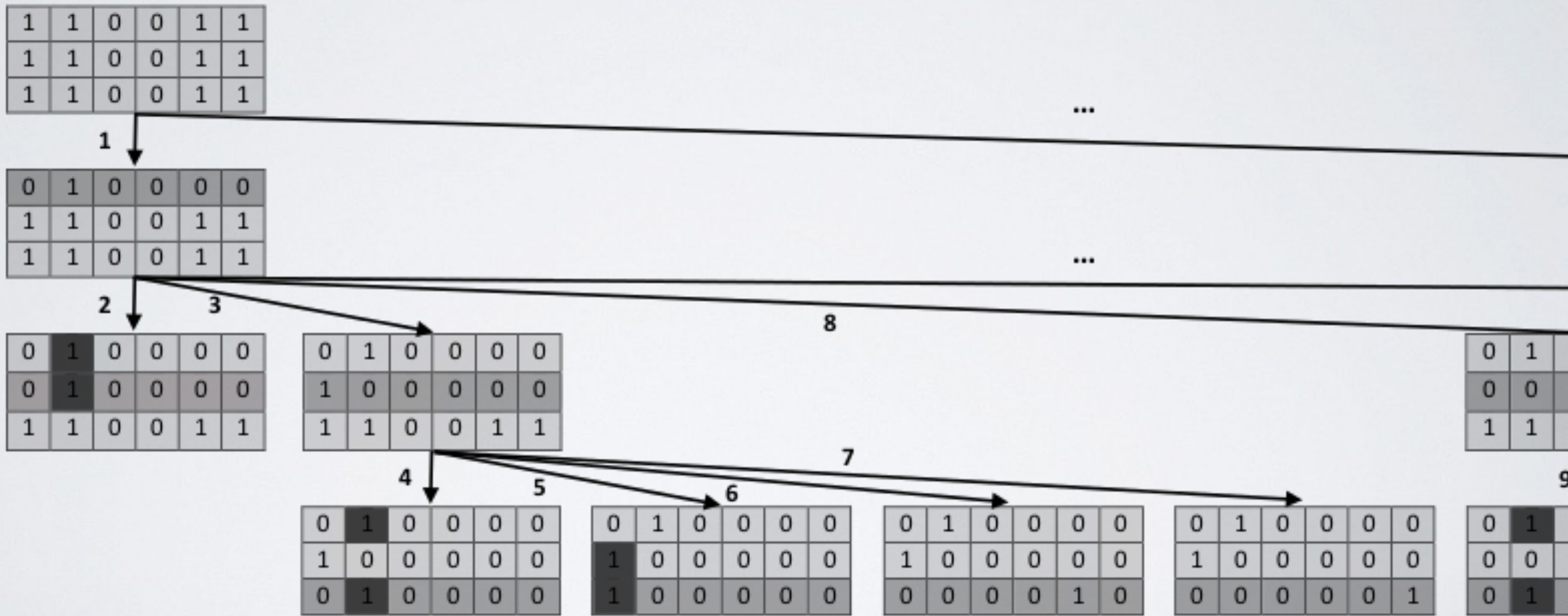
◆ Host and Pattern Graph ◆  $|V_P| \times |V_H|$  matrix  $\mathbf{M}^*$



x	A0	A1	B0	B1	B2	B3
a	1	1	0	0	1	1
b1	1	1	0	0	1	1
b2	1	1	0	0	1	1



# Ullmann Example





# Ullmann Example

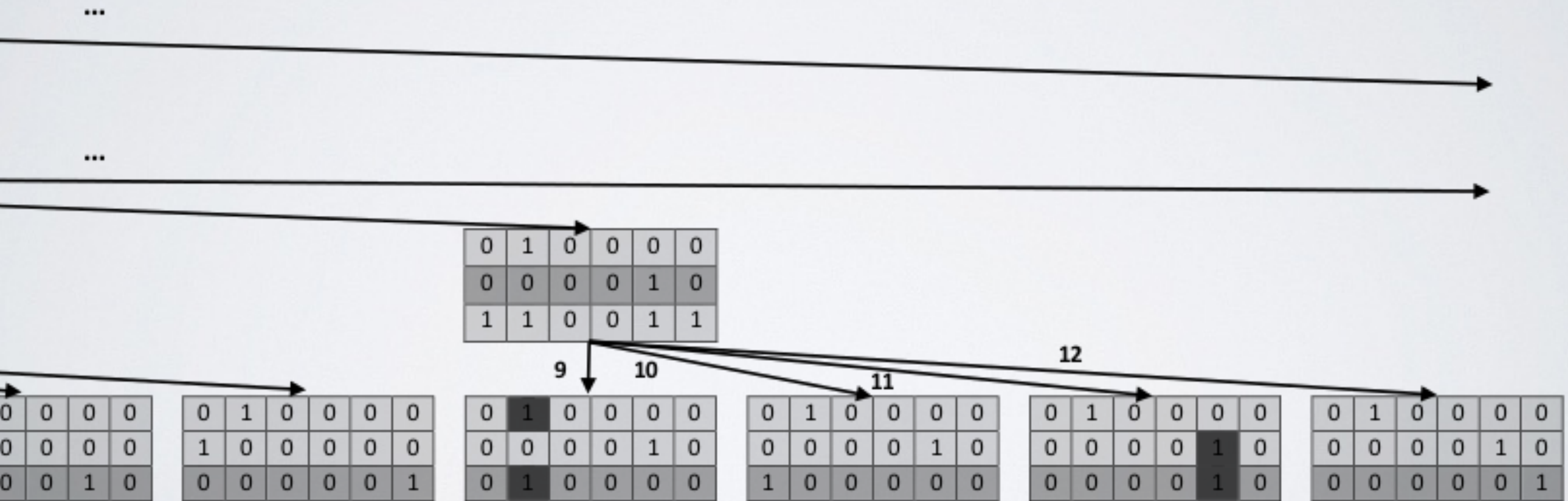
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$\mathbf{P}$  is isomorphic if and only if  $\forall i, j, \mathbf{P}[i, j] = 1 : \mathbf{M}(\mathbf{MH})^T[j, i] = 1$ .

$$\begin{aligned}
 \mathbf{M}(\mathbf{MH})^T &= \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \end{bmatrix} \times \left( \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \end{bmatrix} \times \begin{bmatrix} 0 & 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 & 0 & 0 \end{bmatrix} \right)^T \\
 &= \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \end{bmatrix} \times \left( \begin{bmatrix} 0 & 0 & 1 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \right)^T \\
 &= \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \end{bmatrix} \times \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \\
 &= \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 1 & 0 & 0 \end{bmatrix}
 \end{aligned}$$

# Ullmann Example

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# Ullmann Example

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 &= \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \times \left( \begin{bmatrix} 0 & 0 & 1 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 & 0 & 0 \end{bmatrix} \right)^T \\
 &= \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \times \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 1 & 0 & 0 \\ 1 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \\
 &= \begin{bmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}
 \end{aligned}$$



# VF2\*

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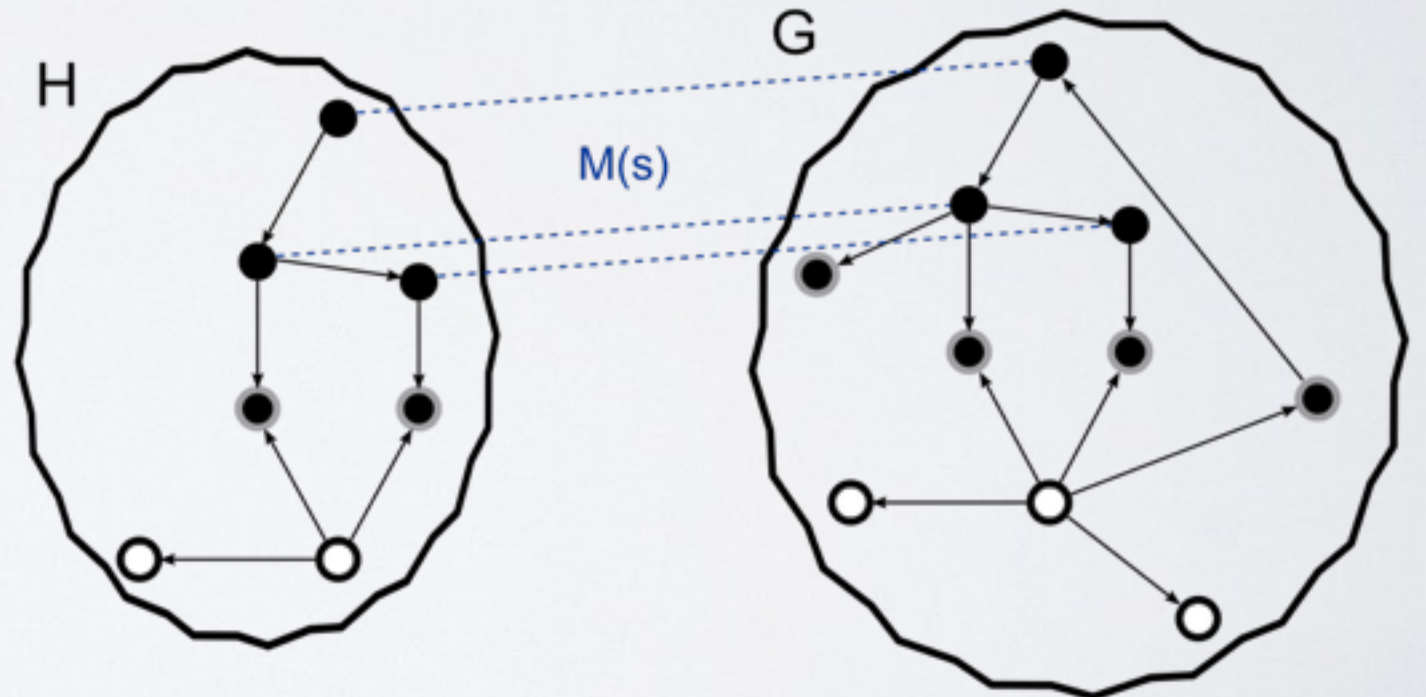
- Similar to Ullmann
- $G = (V_G, E_G), H = (V_H, E_H)$
- $M : V_H \rightarrow V_G$ : isomorphic vertex mapping
- $M(c)$  holds set of current matches
- $M_H, M_G$  represent vertices from  $H, G$  in  $M(s)$

Note:  $G$  represents host graph,  $H$  the pattern

\*: L. P. Cordella, P. Foggia, C. Sansone, M. Vento, A (sub) graph isomorphism algorithm for matching large graphs, Pattern Analysis and Machine Intelligence, IEEE Transactions on 26 (10) (2004) 1367– 1372.

# VF2 Notations

- $N_{H(s)}^{in}$  vertices adjacent to  $M_H(s)$  angling incoming edges
- $N_{H(s)}^{out}$  idem for outgoing edges
- $N_H(s) = N_{H(s)}^{in} + N_{H(s)}^{out}$
- $\tilde{V}_H = V_H - M_H(s) - N_H(s)$
- candidate  $p = (v_p, v_p) \in P(s)$
- priority  $N_{H(s)}^{out}$ ,  $N_{H(s)}^{in}$ ,  $\tilde{V}_H$

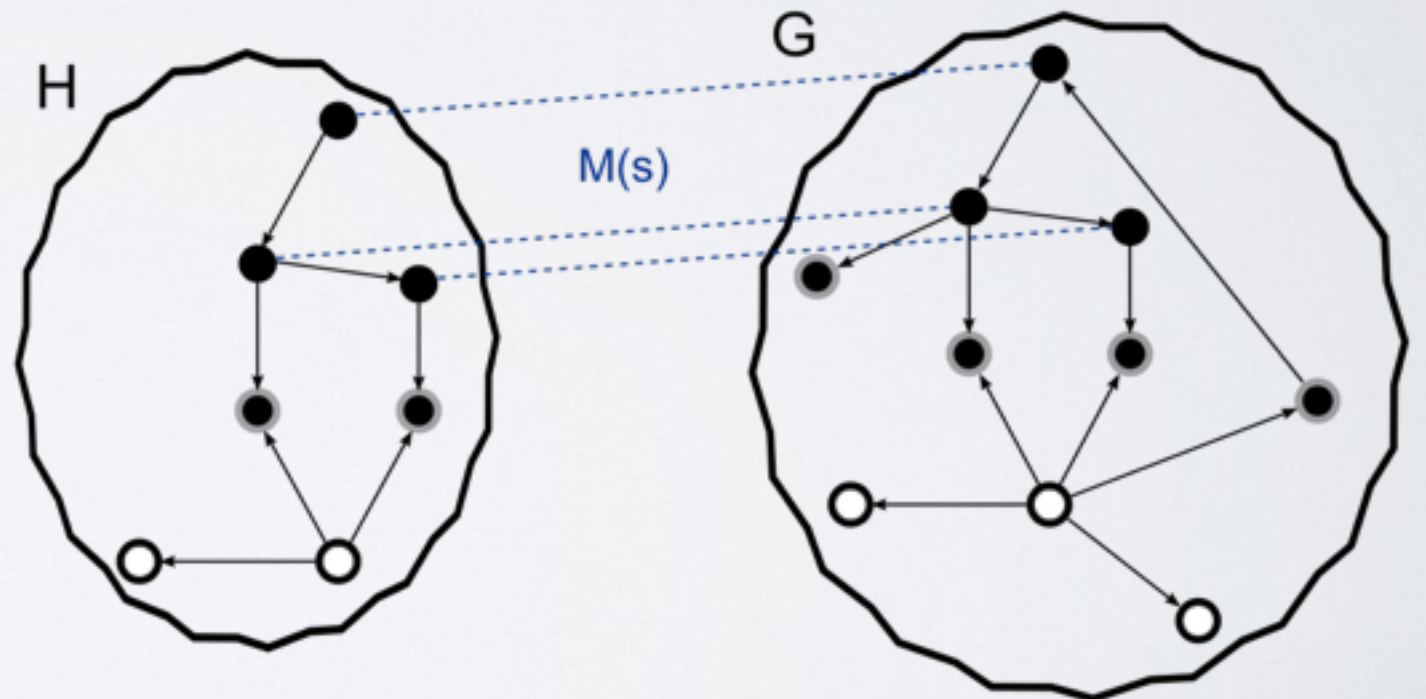




# VF2 Algorithm

- match 1 vertex
- extend match with candidate  $p$  if (else backtrack)
- run test on  $s' = s \cup p$  (in order)

- $M(s')$  valid
- ext. edges between  $M_H(s')$  and  $N_H(s') \leq M_G(s')$  and  $N_G(s')$
- ext. edges between  $N_H(s')$  and  $\tilde{V}_H(s') \leq N_G(s')$  and  $\tilde{V}_{HG}(s')$



# Look-back on Ullmann and VF2

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- Main difference in backtracking step:
  - Ullmann only compares pairs of adjacent vertices
  - VF2 compares verity with neighbourhood
  - Ullmann's  $\mathbf{M}^*$  matrix verifies semantic compatibility between vertices in match
  - VF2 feasibility test ensures correct match

# Ullmann vs VF2\*

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- VF2 best case:  $O(N^2)$ , for  $N = |V_H| + |V_G|$
- VF2 worst case:  $O(N!N)$
- VF2 in both cases order of linear magnitude faster
- VF2 linear spatial complexity vs cubic
- Combination possible for lower time complexity

\*: P. Foggia, C. Sansone, M. Vento, A database of graphs for isomorphism and sub-graph isomorphism benchmarking, in: Proc. of the 3rd IAPR TC-15 International Workshop on Graph-based Representations, 2001, pp. 176–187.