Algorithms for subnetwork mining in heterogeneous networks 11th Symposium on Experimental Algorithms (SEA'12)

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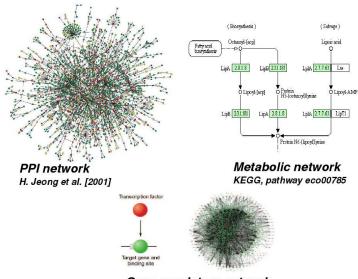
Université de Nantes, LINA - UMR CNRS 6241, France.

Friday 8th June 2012

Outline

- Types of biological networks
- Biological networks comparison
- New approach : Skew SubGraph Mining (SkewGraM)
- Polynomial-time algorithms and harness results
- Conclusion & perspectives

Types of biological networks



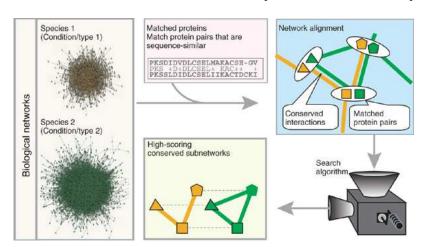
Gene regulatory network M.M. Babu et al. [2004]

Homogeneous and Heterogeneous networks

- ullet Homogeneous networks A and B:
 - same type, same graph representation, different species (interspecies).
- ullet Heterogeneous networks A and B:
 - different type, different graph representation (e.g. a directed graph vs. an undirected graph), same species (intraspecies).

Homogeneous networks comparison

Automatic methods: networks alignment [Kelley et al. 2003, 2006].



Heterogeneous networks comparison

How to compare two heterogeneous networks A and B?

Manual or case-by-case methods :

- Metabolic network vs. linear genome [Zheng et al. 2002], [Rison et al. 2002], [Lee et al. 2004], [Pal 2004.], [Boyer et al. 2005, 2009].
- Metabolic network vs. protein-protein interaction network [Pawel et al. 2008], [Huthmacher et al. 2008].

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.. We need a unified modeling, a unified algorithmic analysis, and unified solutions to this type of problems.

Our model

- D = (V, A) : directed graph;
- G = (V, E): an undirected graph on the same vertex set of G.

Definition. A path P in D is called (D,G)-consistent if G[V(P)] is connected.

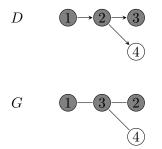
Our model

The problem ONE-TO-ONE SKEWGRAM

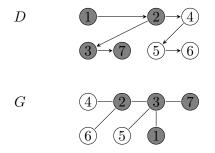
Input. A directed graph D=(V,A), an undirected graph G=(V,A).

Question. Find the longest (D, G)-consistent path.

Longest (\overline{D}, G) -consistent path : example 1



Longest (D, G)-consistent path : example 2



Complexity of ONE-TO-ONE SKEWGRAM

When D is an arbitrarily directed graph, ONE-TO-ONE SKEWGRAM contains the LONGEST PATH PROBLEM as a sub-problem. Thus ONE-TO-ONE SKEWGRAM is NP-complete.

When D is a DAG?

Complexity of ONE-TO-ONE SKEWGRAM when D is a DAG

 D^* is the underlying graph of D.

G	Tree	Outerplanar	General graph
Chordless path or cycle, (bi-)star		Р	
Tree with diameter 4	Р	NPC	NPC
General graph	Р	NPC	NPC
			APX-hard

TABLE: Complexity of the ONE-TO-ONE SKEWGRAM problem.

Theorem. One-to-one SkewGraM (in its decision version) is NP-complete, even when D^* is an outerplanar graph and G is a tree of diameter 4.

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• ONE-TO-ONE SKEWGRAM is in NP.

A reduction from MAX 2SAT:

- $V_n = \{x_1, \dots x_n\}$ is a set of variables;
- ullet $\mathcal{C} = \{C_1, \dots, C_p\}$ is a set of p clauses over \mathcal{V}_n ;
- Each clause is a disjunction of two literals.

Goal: find an assignment that satisfies a **maximum number** of clauses.

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level 0 : a vertex s;

$$\begin{array}{c} p = 2 \text{ and } n = 3 \\ \mathcal{C} = \{C_1, \frac{C_2}{C_2}\} \\ C_1 = (x_1 \vee x_3), \frac{C_2}{C_2} = (\overline{x_1} \vee x_2) \\ \mathcal{X} = \{x_1, x_2, x_3\} \end{array}$$

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D

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 (b_1)





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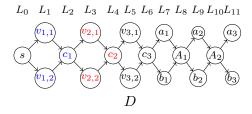
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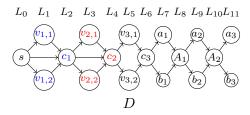
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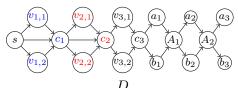
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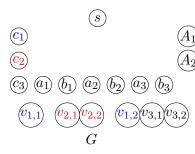
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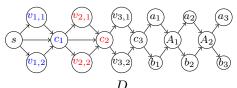
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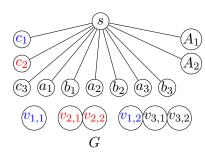




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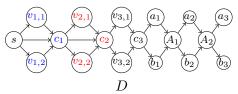


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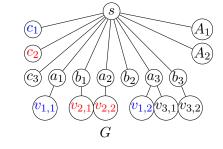
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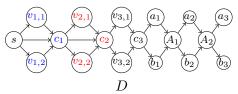


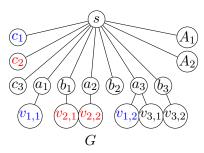
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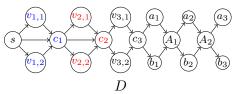


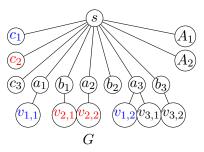
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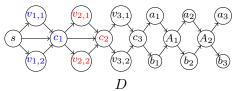
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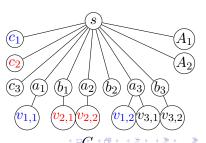
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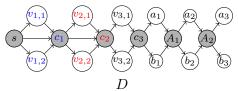
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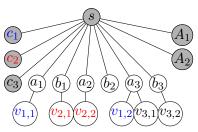
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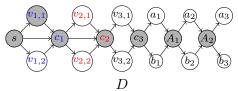
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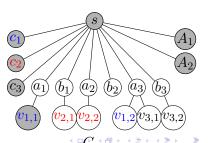
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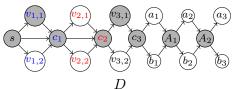
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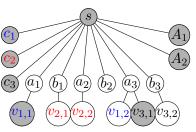
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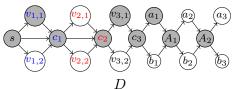
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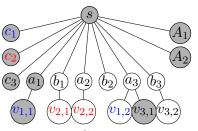
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Complexity of the ONE-TO-ONE SKEWGRAM problem

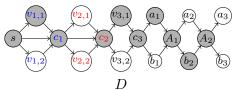
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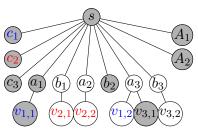
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Property: There is an assignment that satisfies at least k clauses iff there is a (D,G)-consistent path of at least 10+k vertices.



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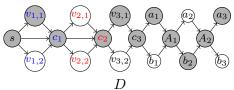
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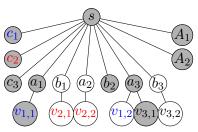
$$C_1 = (x_1 \lor x_3), C_2 = (\overline{x_1} \lor x_2)$$

$$\mathcal{X} = \{x_1, x_2, x_3\}$$

 L_0 L_1 L_2 L_3 L_4 L_5 L_6 L_7 L_8 L_9 L_{10} L_{11}



 $\begin{array}{l} \textbf{Property}: \text{There is an assignment} \\ \text{that satisfies at least } k \text{ clauses iff} \\ \text{there is a } (D,G)\text{-consistent path of} \\ \text{at least } 10+k \text{ vertices.} \\ \end{array}$



Two algorithms for ONE-TO-ONE SKEWGRAM

- Heuristic called AlgoH.
- Exact exponential-time algorithm called ALGOBB using the branch and bound method.

Two algorithms for ONE-TO-ONE SKEWGRAM

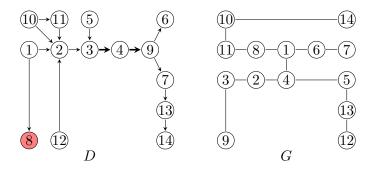
- Heuristic called AlgoH.
- Exact exponential-time algorithm called ALGOBB using the branch and bound method.

Input : D = (V, A), G = (V, E), an arc $xy \in A$.

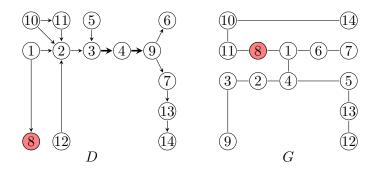
Output : A longest (D,G)-consistent path going through xy.

To solve ONE-TO-ONE SKEWGRAM, an execution is needed for every arc xy of D.

Rule 1 : Connectivity in D (reachability);

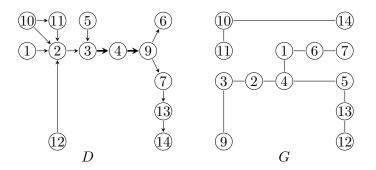


Rule 1 : Connectivity in D (reachability);



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Rule 2 : Connectivity in G (bridges);



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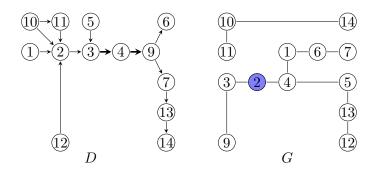


FIGURE: Computation of the cover set of the path $3 \rightarrow 4 \rightarrow 9$.

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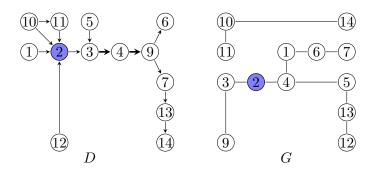


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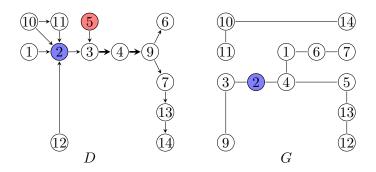


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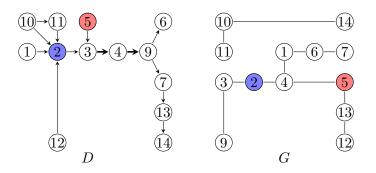


FIGURE: Computation of the cover set of the path $3 \rightarrow 4 \rightarrow 9$.

Rule 1 : Connectivity in D (reachability);

Rule 2 : Connectivity in G (bridges);

Rule 3 : Connectivity in D and G simultaneously (common connected components).

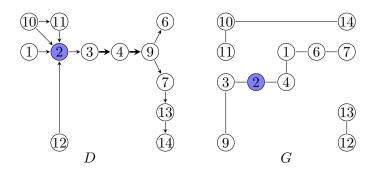


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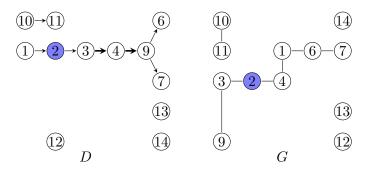
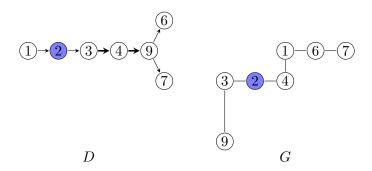


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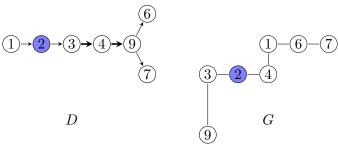
Rule 2 : Connectivity in G (bridges);

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COVERSET
$$(3 \to 4 \to 9) = \{1, 2, 3, 4, 6, 7, 9\}$$

Rule 3 : Connectivity in D and G simultaneously (common connected components).



CoverSet
$$(3 \to 4 \to 9) = \{1, 2, 3, 4, 6, 7, 9\}$$

Property: Let p be a path in D. If D[COVERSET(p)] is Hamiltonian then its Hamiltonian path is the longest (D,G)-consistent path going through p.

Scoring function: for each directed path p, value(p) is the number of vertices of the longest path in D[COVERSET(p)].

- $\bullet \ \operatorname{AlgoH}(D,G,x\to y)$
- ① $cp := x \rightarrow y / * current path * /$
- **3** S := CoverSet(cp); G := G[S]; D := D[S];
- lacktriangle We compute the list L of paths obtained by extending cp by one vertex;
- **5** We compute p_{max} s.t. $value(p_{max}) = max\{value(p) : p \in L\}$.
- **(i)** We compute ph_{max} s.t. $value(ph_{max}) = max\{value(p) : p ∈ L \text{ and } D[COVERSET(p)] \text{ is Hamiltonian}\}.$
- $\bigcirc update(cs, ph_{max});$
- $cp := p_{max} ; return to step 3;$

End of the process : when $L = \emptyset$ or $value(p_{max}) \leq |V(cs)|$.

Performances of ALGOH

- |ALGOH| is the number of vertices of a solution found by the heuristic ALGOH.
- |ALGOBB| is the number of vertices of a solution found by ALGOBB.
- $\bullet \ \, \mathsf{Performance} \, \, \rho = \frac{|\mathtt{ALGOH}|}{|\mathtt{ALGOBB}|}.$

We vary two parameters :

- The number n of vertices of D and G (in the range 20, 30, 40, 50, 60);
- The probability p that an edge between any given two vertices exists (in the range 0.05, 0.1, 0.15, 0.2).

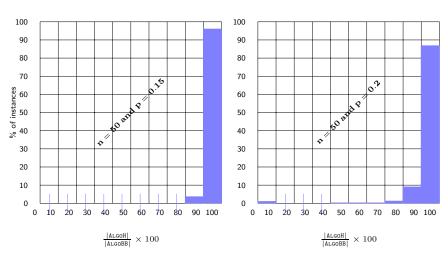
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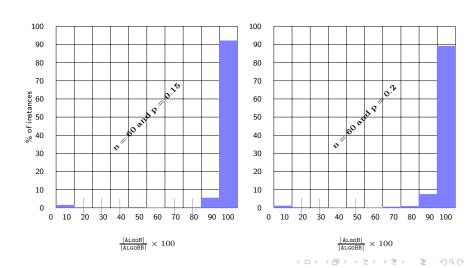
For a fixed n and p we generated 500 instances :

- 100 couples (D,G);
- 5 arcs per couple;

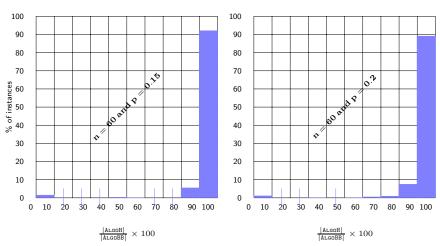
$$n = 50$$



$$n = 60$$

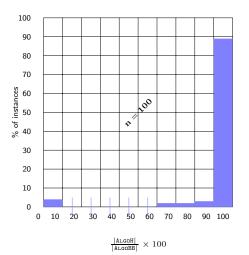


 $n=60 \label{eq:norm}$ Speed-up : AlgoH was 11 times faster than AlgoBB



Performances of ALGOH on scale-free graphs

- \bullet Generation of 100 couples (D,G) of 100 vertices using the public toolkit NGCE (http://ngce.sourceforge.net/);
- ullet 10 arcs per couple.



Conclusion & perspectives

- ONE-TO-ONE SKEWGRAM: New approach to compare heterogeneous biological networks
- ONE-TO-ONE SKEWGRAM is NP-complete.
- Two algorithms for ONE-TO-ONE SKEWGRAM.

Perspectives

- Investigate the complexity of ONE-TO-ONE SKEWGRAM in terms of :
 - Approximability (on specific graph classes)
 - Fixed parameter algorithms
- Consider inexact variants of the problem.

