# More ideas about new orthology inference methods 

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$34^{\text {th }}$ TBI Winterseminar
Bled, $12{ }^{\text {th }}$ February 2019

## Why Orthology Analysis?

Orthology analysis is an important part of data analysis in many areas such as comparative genomics and molecular phylogenetics

## Phylogenetic Tree of Life


source: https://en.wikipedia.org/wiki/Tree_of_life_(biology)

Idea: There is only one true tree of life - we just need good methods to detect it!

## Tree-based vs. graph-based methods

## Tree-based:

- species tree must be known, gene tree via sequence alignments
$\rightarrow$ tree reconciliation gives orthology relation
- accuracy highly depends on quality of trees
- high computational costs


## Graph-based:

- construction of the trees from sequence data
- lower computational costs
- many tools restricted to small number of species (except ProteinOrtho ${ }^{1}$ )
- some tools even include manual correction

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- many tools restricted to small number of species (except ProteinOrtho ${ }^{1}$ )
- some tools even include manual correction
$\rightarrow$ Our overall-goal: improve orthology inference/develop new methods

[^1]
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$\rightarrow$ Goal: Deeper understanding of (reciprocal) Best Match Graphs to make the process more efficient

## Best Match Graphs (BMGs)

here: "closest" = closest last common ancestor (lca)

## Definition

The leaf $y$ is a best match of the leaf $x$ in a tree $T$ if $\sigma(x) \neq \sigma(y)$, and
(i) $\operatorname{Ica}(x, y) \preceq \operatorname{Ica}\left(x, y^{\prime}\right)$ for all leaves $y^{\prime}$ from species $\sigma\left(y^{\prime}\right)=\sigma(y)$.
We write $x \rightarrow y$.
$\sigma=$ colors (= species)


## Reciprocal Best Match Graphs (RBMGs)

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$\rightarrow$ Which (un-)directed graphs are (Reciprocal) Best Match Graphs, i.e., have a tree representation?

## Mathematical Results about (Reciprocal) Best Match Graphs

## BMGs ${ }^{1}$

- Two characterizations for 2-cBMGs via triples and neighborhoods
$\rightarrow$ Recognition in polynomial time
- Characterization for $n$-cBMGs via Aho-Tree from 2-cBMGs
$\rightarrow$ Recognition and tree reconstruction in polynomial time
- Unique least resolved tree

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## RBMGs ${ }^{2}$

- Classification of three distinct groups of 3-cRBMGs
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- Characterization for $n-\mathrm{cRBMGs}$ via supertree from 3-cRBMGs $\rightarrow$ Recognition and tree reconstruction presumably not in polynomial time
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## RBMGs ${ }^{2}$

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- No unique least resolved tree $\rightarrow$ Much information lost by only looking at RBMGs!


## How can we use all this?

## Theorem ${ }^{1}$

In pure DL scenarios (i.e. in the absence of HGT events) the reciprocal best match graph can only produce false positive but not false negative orthology assignments.
$\Rightarrow$ The true orthology relation has to be contained in the RBMG.

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$\Rightarrow$ The true orthology relation has to be contained in the RBMG.
$\rightarrow$ Some false positive edges can be identified using best match graphs


Remove middle edge: $P_{4}$-Editing (P4E)

[^6]
## Simulation results with 0 HGT events



FPR
before
VS.
after
P4E


## Simulation results with 0 HGT events




FPR
before
VS.
after
P4E



## Simulation results with 1 HGT event



FPR
before
vS.
after
P4E




FPR
before
VS.
after
P4E





FPR
before
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## Summary \& Outlook

## Results so far:

- Characterization and tree reconstruction algorithms for BMGs and RBMGs
- RBMG contains no false positive orthologs in the absence of HGT
- $P_{4}$-Editing in the absence of HGT
- RBMG contains false negative orthologs in the presence of HGT
$\rightarrow$ RBMG loses much information that is still contained in the BMG!


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Next steps:

- BMGs might help to detect HGT events
- Improved graph editing based on characterization of BMGs and RBMGs

Special Thanks to:
Peter F. Stadler
Marc Hellmuth
Nicolas Wieseke
Edgar Chávez
Marcos González
Maribel Hernández Rosales Alitzel López


Dulce Valdivia

## Thank you for your attention!



## Some basics: Rooted Trees and Triples

Rooted Tree $T$ :

acyclic, connected graph

Triples:

- $T$ displays a triple $a b \mid c$ if the path from $c$ to the root is not intersected by the path from $a$ to $b$.
- $\mathcal{R}(T)=\{a b|c, a b| d, a b \mid e\}$
- A set of triples $R$ is said to be consistent if there is a tree $T$ with $R \subseteq \mathcal{R}(T)$.
- Consistency-check via BUILD-algorithm in polynomial time. In case of consistency, it returns a tree $T$ (the "Aho Tree") with $R \subseteq \mathcal{R}(T)$.


## How do n-cBMGs look like?

## Theorem

A colored digraph $(G, \sigma)$ is a n-cBMG if and only if all induced subgraphs on two colors are $2-c B M G$ 's and the union of the triples obtained from their least resolved trees forms a consistent set.
least resolved $=$ "lowest possible resolution"
$\rightarrow$ The unique least resolved tree for $(G, \sigma)$ can be reconstructed in cubic time
$\rightarrow$ All information that is needed, is contained in the 2-colored best match graphs!

## The case of two colors: Characterization via triples

Some 2-colored subgraphs on 3 vertices give us constraints on the tree topology:

$X_{1}, X_{2}, X_{3}$, and $X_{4}$ all give the informative triple ab|c.

## Theorem

A connected 2-colored digraph $(G, \sigma)$ is a 2-cBMG if and only if $(G, \sigma)=G(\operatorname{Aho}(\mathcal{R}(G, \sigma)), \sigma)$, where $\mathcal{R}(G, \sigma)$ is the set of all informative triples of $(G, \sigma)$.

# The case of two colors: Characterization via 

 out-neighborhoods
## Augenkrätze-Theorem

A connected 2-colored digraph $(G, \sigma)$ is a 2-cBMG if and only if $(G, \sigma)$ satisfies properties (N0), (N1), (N2), and (N3). Moreover, the tree $T$ defined by the $\mathcal{H}^{\prime}:=\left\{R^{\prime}(\alpha) \mid \alpha \in \mathcal{N}\right\}$ is the unique least resolved tree that explains $(G, \sigma)$.
(N0) $\beta \subseteq N(\alpha)$ or $\beta \cap N(\alpha)=\emptyset$
(N1) $\alpha \cap N(\beta)=\beta \cap N(\alpha)=\emptyset$ implies
$N(\alpha) \cap N(N(\beta))=N(\beta) \cap N(N(\alpha))=\emptyset$.
(N2) $N(N(N(\alpha))) \subseteq N(\alpha)$
(N3) $\alpha \cap N(N(\beta))=\beta \cap N(N(\alpha))=\emptyset$ and $N(\alpha) \cap N(\beta) \neq \emptyset$ implies $N^{-}(\alpha)=N^{-}(\beta)$ and $N(\alpha) \subseteq N(\beta)$ or $N(\beta) \subseteq N(\alpha)$
properties can be nicely checked by an algorithm

The three classes of 3-cRBMGs
There are exactly three classes of $3-\mathrm{cRBMGs}$ :


(A)
(B)
(C)

The three classes of $3-c$ RBMGs
There are exactly three classes of $3-\mathrm{cRBMGs}$ :

(A)
(B)

(C)

## Theorem

A graph $(G, \sigma)$ is a 3-cRBMG if and only if the construction algorithm returns a tree that explains $(G, \sigma)$.

## What can we say so far about $n$-cRBMGs?

Idea: Similarly to the case of BMGs, all information needed is contained in the 3-colored induced subgraphs of $(G, \sigma)$

## Conjecture

An undirected colored graph $(G, \sigma)$ is an n-cRBMG if and only if for any $\left(G_{r s t}, \sigma_{r s t}\right)$ there exists a tree ( $\left.T_{r s t}, \sigma_{r s t}\right)$ that explains $\left(G_{r s t}, \sigma_{r s t}\right)$, such that $\mathcal{P}:=\bigcup_{r, s, t} T_{r s t}$ is compatible.
$\left(G_{r s t}, \sigma_{r s t}\right):=$ induced subgraph on colors $r, s, t$ of $(G, \sigma)$
$\rightarrow$ It looks like there is no polynomial-time construction algorithm for $n$-cRBMGs


[^0]:    ${ }^{1}$ Lechner M, Findeiß S, Steiner L, Marz M, Stadler PF, Prohaska SJ, 2011. Proteinortho: detection of (co-)orthologs in large-scale analysis. BMC Bioinformatics 12:124

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