



Evolution of genes neighborhood within reconciled phylogenies: an ensemble approach

Cedric Chauve, Yann Ponty, João Paulo Pereira Zanetti

► To cite this version:

Cedric Chauve, Yann Ponty, João Paulo Pereira Zanetti. Evolution of genes neighborhood within reconciled phylogenies: an ensemble approach. RECOMB-CG'14, Oct 2014, Cold Spring HArbour, United States. , 16. hal-01216782

HAL Id: hal-01216782

<https://inria.hal.science/hal-01216782>

Submitted on 17 Oct 2015

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers.

L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.

Evolution of genes neighborhood within reconciled phylogenies: an ensemble approach

Cedric Chauve (SFU), Yann Ponty (CNRS/LIX/PIMS), João Paulo Pereira Zanetti (SFU/U. Campinas)

Background: DeCo

Reconstructing ancestral genome features is a classical comparative genomics problem, often addressed with dynamic programming (DP) algorithms. A DP algorithm, called **DeCo**, was recently introduced for computing **parsimonious** evolutionary scenarios for **gene adjacencies** in a **duplication-aware** framework, motivated by the reconstruction of **ancestral gene orders** [Bérard *et al.*, 2012].

DeCo

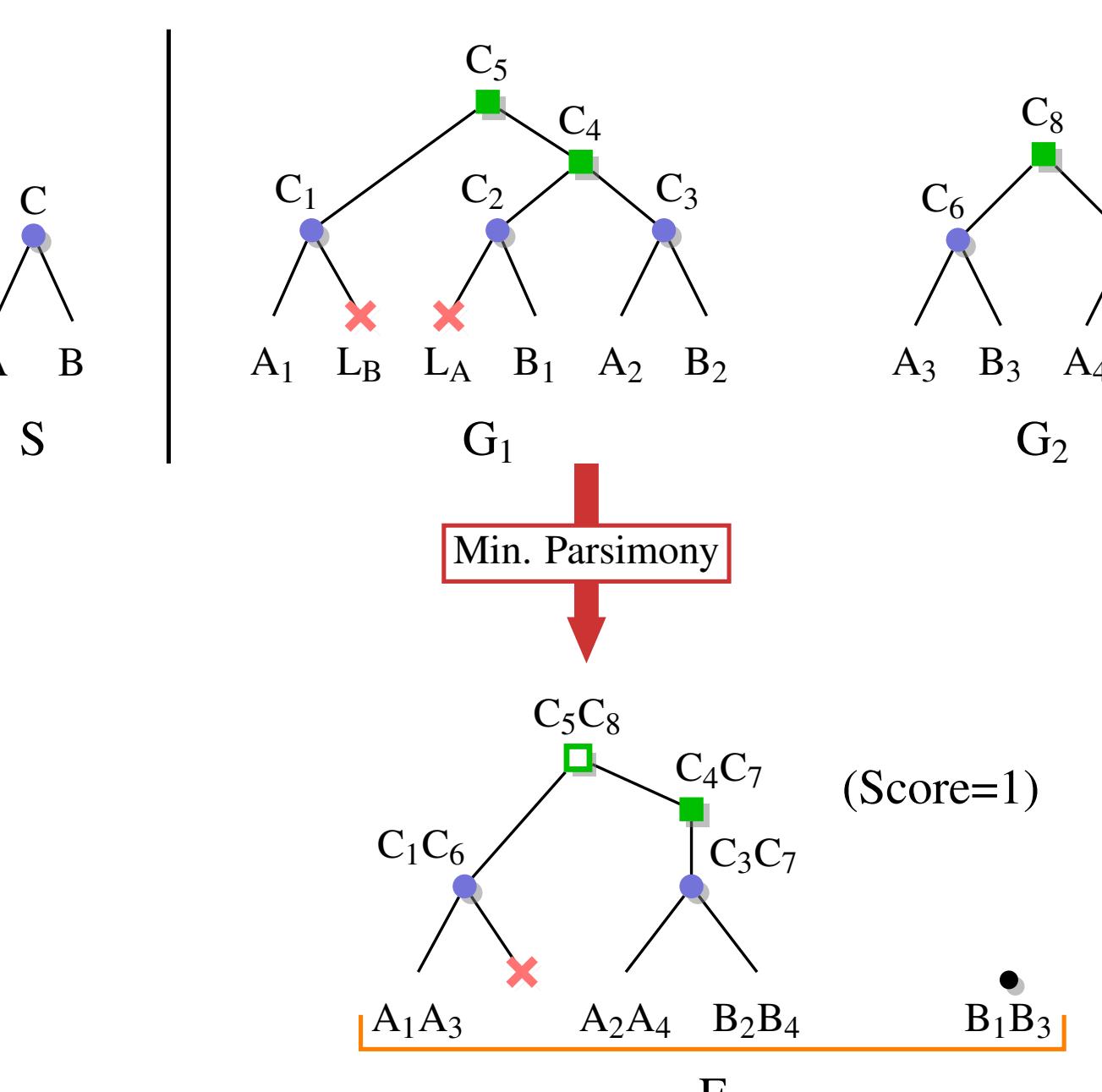
Input:

- Reconciled gene trees G_1 and G_2 ;
- Set of extant adjacencies.

Output:

- Max-parsimony adjacency forest

Parsimony: #adjacencies gains/breaks

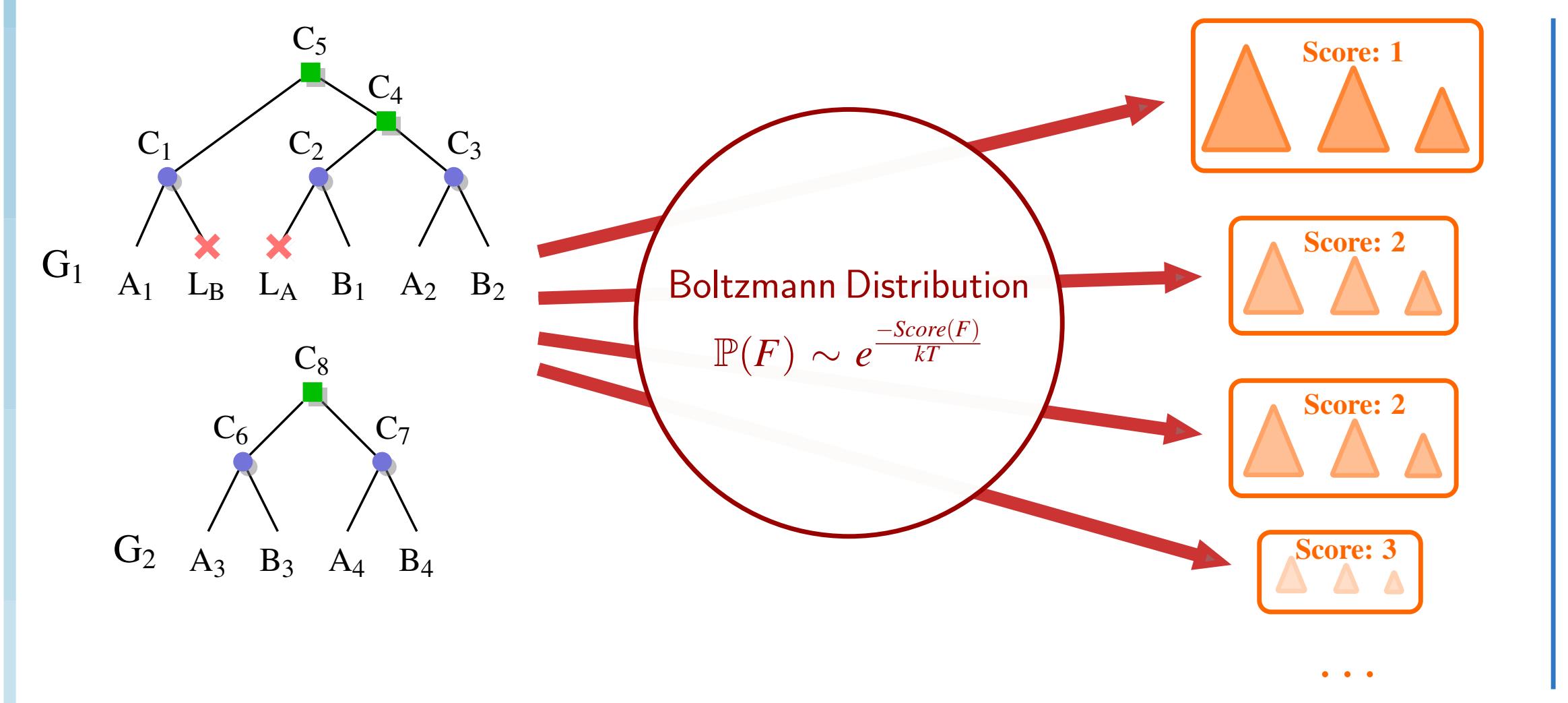


Dynamic programming algorithm (excerpt):

$$c_1(v^1, v^2) = \min \left\{ \begin{array}{l} c_1(ca(v^1), cb(v^2)) + c_1(cb(v^1), ca(v^2)) \\ c_1(ca(v^1), cb(v^2)) + c_0(cb(v^1), ca(v^2)) + \text{Break} \\ c_0(ca(v^1), cb(v^2)) + c_1(cb(v^1), ca(v^2)) + \text{Break} \\ c_0(ca(v^1), cb(v^2)) + c_0(cb(v^1), ca(v^2)) + 2\text{Break} \\ c_1(ca(v^1), ca(v^2)) + c_1(cb(v^1), cb(v^2)) \\ c_1(ca(v^1), ca(v^2)) + c_0(cb(v^1), cb(v^2)) + \text{Break} \\ c_0(ca(v^1), ca(v^2)) + c_1(cb(v^1), cb(v^2)) + \text{Break} \\ c_0(ca(v^1), ca(v^2)) + c_0(cb(v^1), cb(v^2)) + 2\text{Break} \\ \dots \end{array} \right.$$

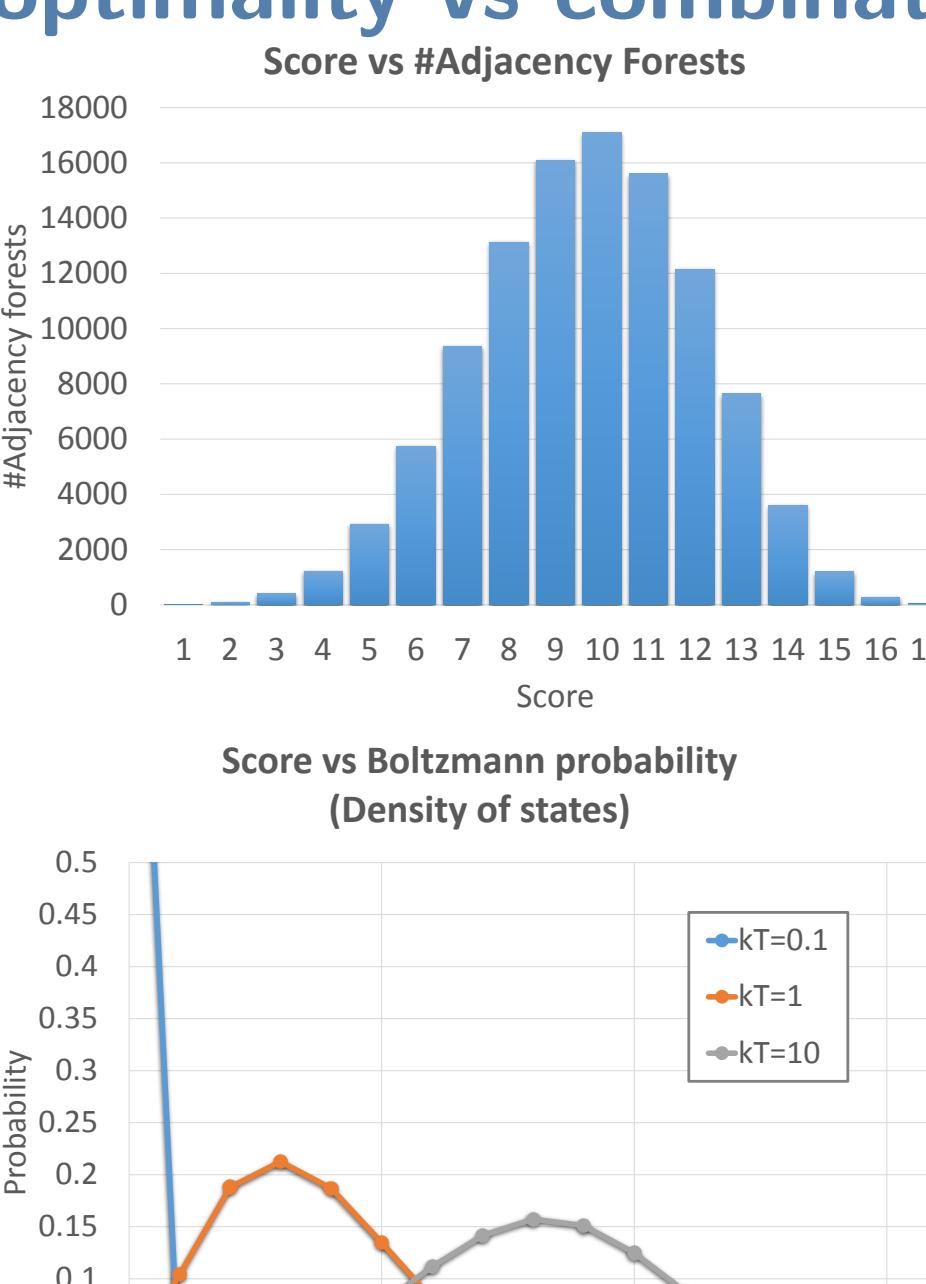
DeCo+Advanced Dynamic Programming=DeClone

The whole (sub)optimal space is modelled as a Boltzmann Ensemble

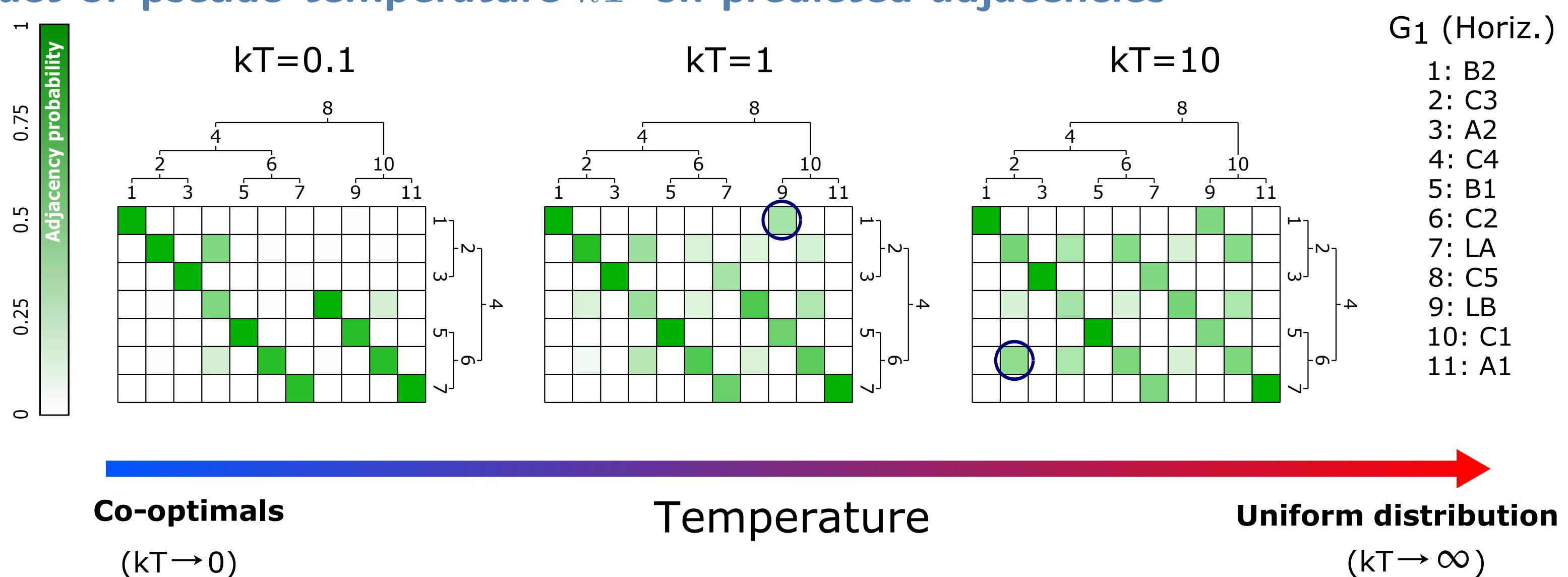


Analysis

- Average properties
- Ancestral adjacencies
- Syntenic inconsistencies
- Recurrent patterns ...
- Max. expected accuracy predictions
- Robustness assessment
- ...



Impact of pseudo-temperature kT on predicted adjacencies



Results: DeClone

We describe **DeClone**, an extension of DeCo, using **Advanced Dynamic Programming** techniques, that allows the **sampling** of adjacency forests under the **Boltzmann distribution**, as well as the computation of **probabilities** of presence/absence of ancestral gene adjacencies, again under the Boltzmann distribution.

Experiments

Data: 6,074 DeCo instances, with genes taken from 36 mammalian genomes from the Ensembl database in 2012. **Syntenic inconsistencies** with parsimonious scenarios concern 5,817 genes over 112,188 ancestral and extant adjacencies.

Methods: for each instance, we sampled 1,000 adjacency forests under a Boltzmann distribution that favours highly co-optimal scenarios.

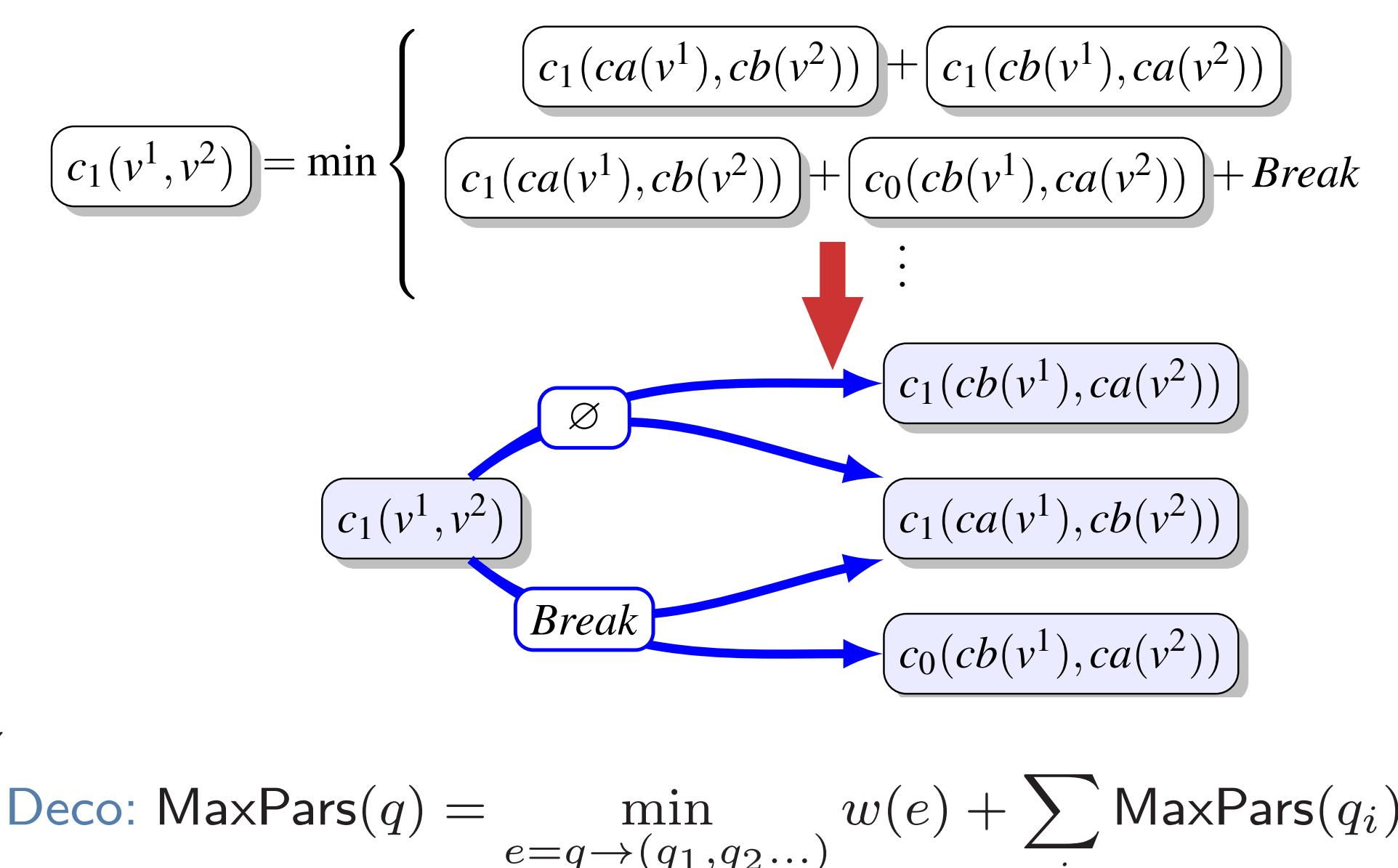
Results:

Adj. freq.	Anc. genes	Anc. adj.	Synt. Inc.
≥ 0.3	118,687	110,180	10,351
≥ 0.4	117,639	106,973	8,330
≥ 0.5	116,231	103,479	6,677
≥ 0.6	114,538	99,720	5,113
≥ 0.7	112,564	96,039	4,092
≥ 0.8	110,086	91,821	3,276
≥ 0.9	107,564	87,790	2,710
= 1	100,443	79,078	1,348

Algorithmic aspects

Algorithmic framework [Ponty/Saule, 2011]

Dyn. Prog. equations → Weighted Hypergraphs



$$\text{Deco: } \text{MaxPars}(q) = \min_{e=q \rightarrow (q_1, q_2, \dots)} w(e) + \sum_i \text{MaxPars}(q_i)$$

Claim: The DeCo DP scheme is **unambiguous** and **complete**

Algorithmic corollaries

- **Computing the partition function \mathcal{Z} in $\mathcal{O}(|G_1||G_2|)$**
Simple change of algebra ($\min, +, C$) → $(\sum, \times, e^{-\frac{C}{kT}})$
- **Boltzmann sampling of adjacency forests**
Each hyperedge chosen with probability proportional to its overall contribution to \mathcal{Z} .
- **Adjacency dot-plot:** $\mathcal{O}(|G_1||G_2|)$ inside/outside algorithm computes the probability of ancestral adjacencies.

Discussion

DeClone allows a controlled exploration of the space of all gene adjacency evolution scenarios: exhaustive enumeration, sampling, probabilities computation.

Experiments show that exploring the whole solution space under a Boltzmann distribution biased toward co-optimals allows to reduce significantly the number of syntenic inconsistencies observed when a single arbitrary optimal scenario is considered.

References

- S. Bérard, *et al.*. 2012. Evolution of gene neighborhoods within reconciled phylogenies. *Bioinformatics*, 28(18):382–388.
Y. Ponty and C. Saule. 2011. A Combinatorial Framework for Designing (Pseudoknotted) RNA Algorithms. *WABI 2011*, pp. 259–269.