



# INF421, Lecture 9

## Drawing graphs

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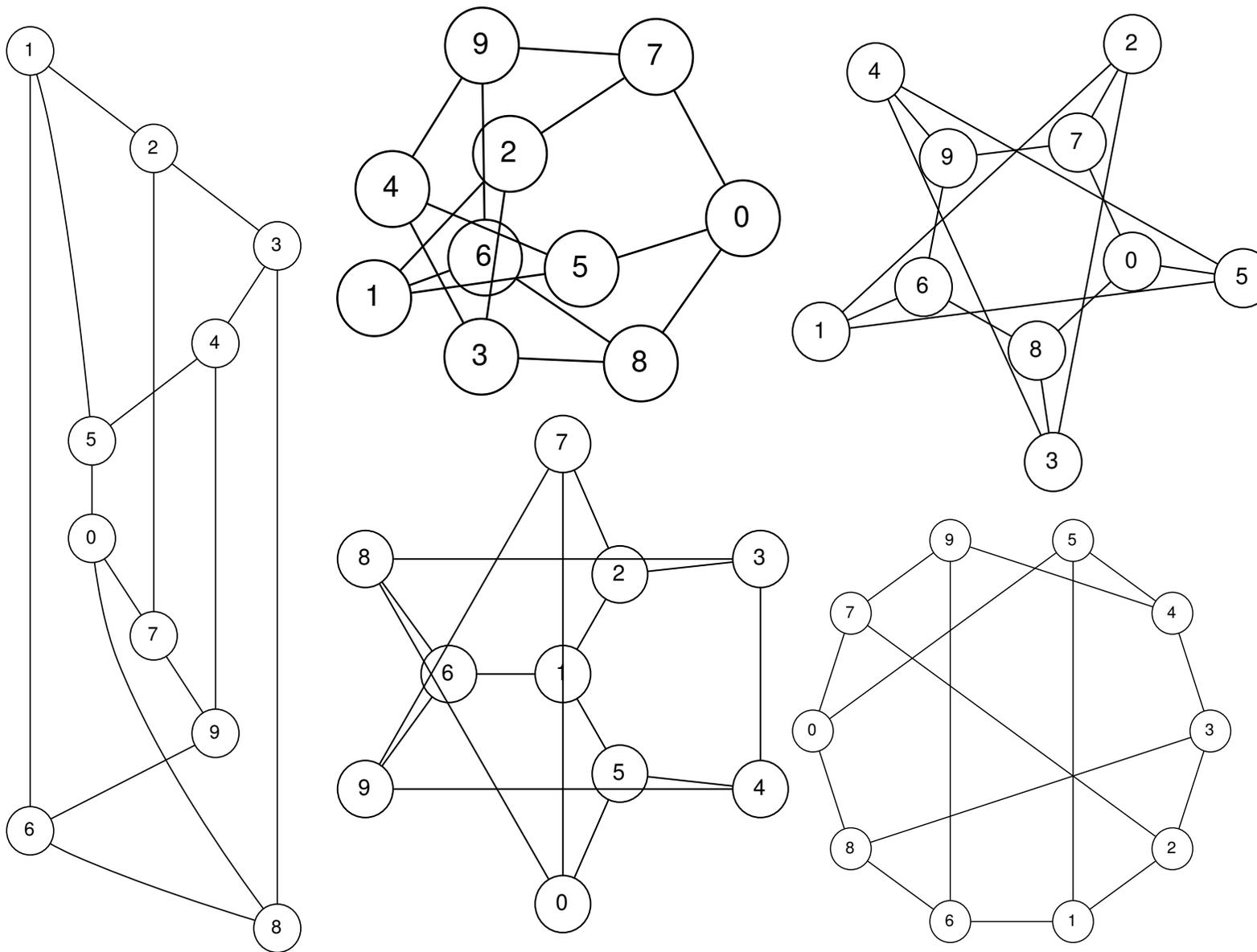
# Course

- **Objective:** teach notions AND develop intelligence
- **Evaluation:** TP noté en salle info, Contrôle à la fin. Note:  
 $\max(CC, \frac{3}{4}CC + \frac{1}{4}TP)$
- **Organization:** fri 31/8, 7/9, 14/9, 21/9, 28/9, 5/10, 12/10, 19/10, 26/10,  
amphi 1030-12 (Arago), TD 1330-1530, 1545-1745 (SI:30-34)
- **Books:**
  1. K. Mehlhorn & P. Sanders, *Algorithms and Data Structures*, Springer, 2008
  2. D. Knuth, *The Art of Computer Programming*, Addison-Wesley, 1997
  3. G. Dowek, *Les principes des langages de programmation*, Editions de l'X, 2008
  4. Ph. Baptiste & L. Maranget, *Programmation et Algorithmique*, Ecole Polytechnique (Polycopié), 2006
- **Website:** [www.enseignement.polytechnique.fr/informatique/INF421](http://www.enseignement.polytechnique.fr/informatique/INF421)
- **Blog:** [inf421.wordpress.com](http://inf421.wordpress.com)
- **Contact:** [liberti@lix.polytechnique.fr](mailto:liberti@lix.polytechnique.fr) (e-mail subject: INF421)



**Today, a “research seminar”!**

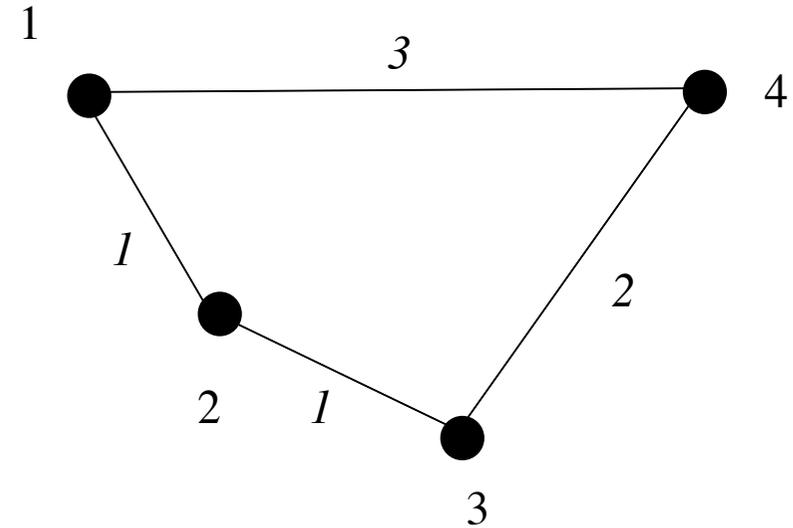
# At a glance



**Which graph has most symmetries?**



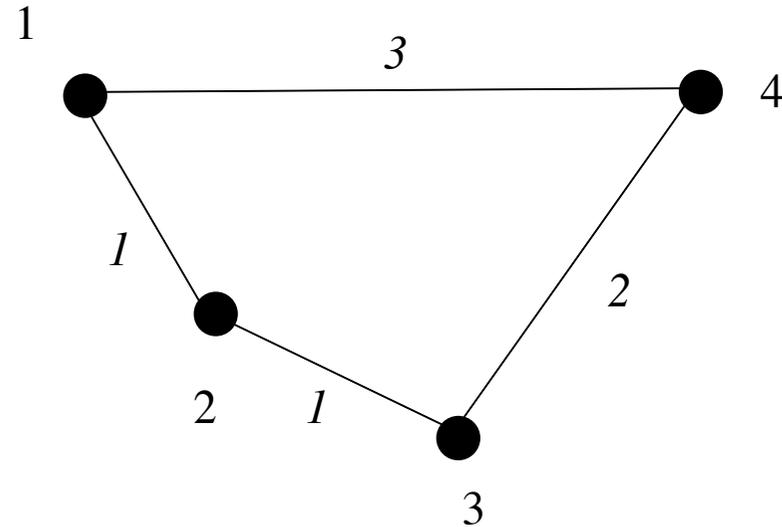
# How does a weighted graph look?



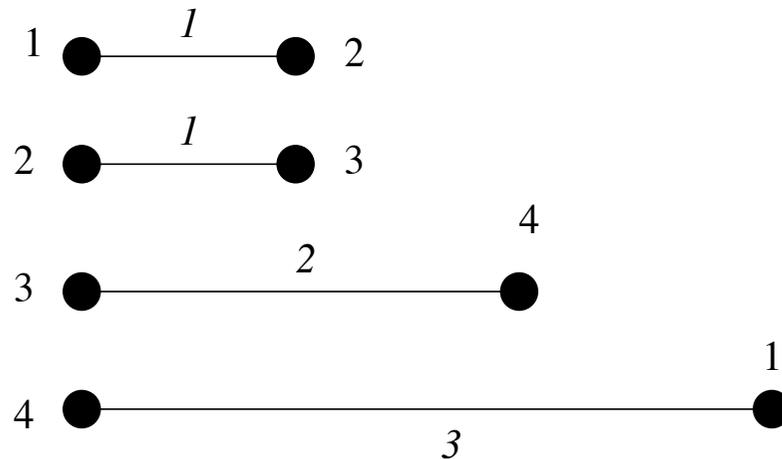
● Like this?



# How does a weighted graph look?



● Like this?



● Perhaps like this?



**Don't confuse a graph with its  
drawing**



# Clean energy

- Use hydrogen to produce chemical energy
- How to produce “pure hydrogen”?
- **Photosystem II**: *complex molecular conglomerate*
- Molecular function  $\leftrightarrow$  3D shape
- Molecule = graph
  - Atoms = vertices
  - Known inter-atomic distances = edges

Draw a weighted graph in 3D



# Other applications

## Applications:

- Clock synchronization, phase retrieval (A. D'Aspremont, CMAP) — 1D
- Wireless sensor network localization — 2D
- Molecule conformation (me, LIX) / submarine localization — 3D
- Multidimensional scaling — (whatever)D



# Drawing a graph

- Given a simple weighted undirected graph  $G = (V, E)$  with a distance function  $d : E \rightarrow \mathbb{R}_+$ , solve the constraint system:

$$\forall \{u, v\} \in E \quad \|x_u - x_v\| = d_{uv} \quad (1)$$

- Obtain an embedding  $x : V \rightarrow \mathbb{R}^2$



# Global optimization

- Reformulate (1) to

$$\min_x \sum_{\{u,v\} \in E} (\|x_u - x_v\|^2 - d_{uv}^2)^2 \quad (2)$$

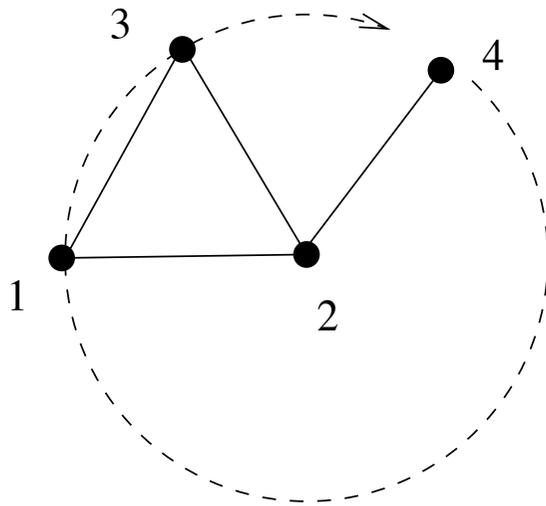
- $G$  has an embedding  $\Leftrightarrow$  optimum  $x^*$  of (2) has value 0.
- Eq (2) is nonconvex in  $x$ , many local optima

Try it on Matlab/Octave/Maple/whatever for simple data, you won't get very far (< 10 vertices)



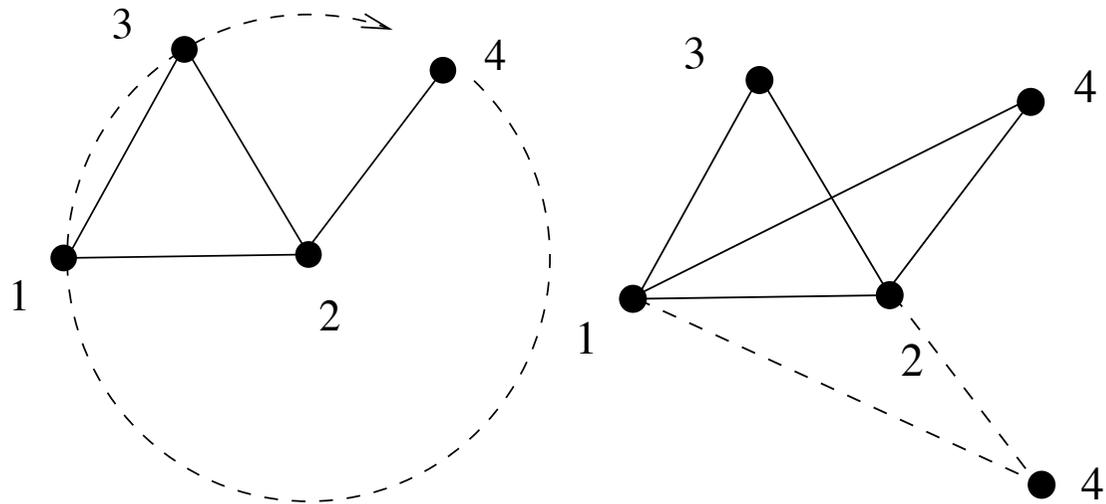
# The number of embeddings

- Uncountably many (incongruent) embeddings



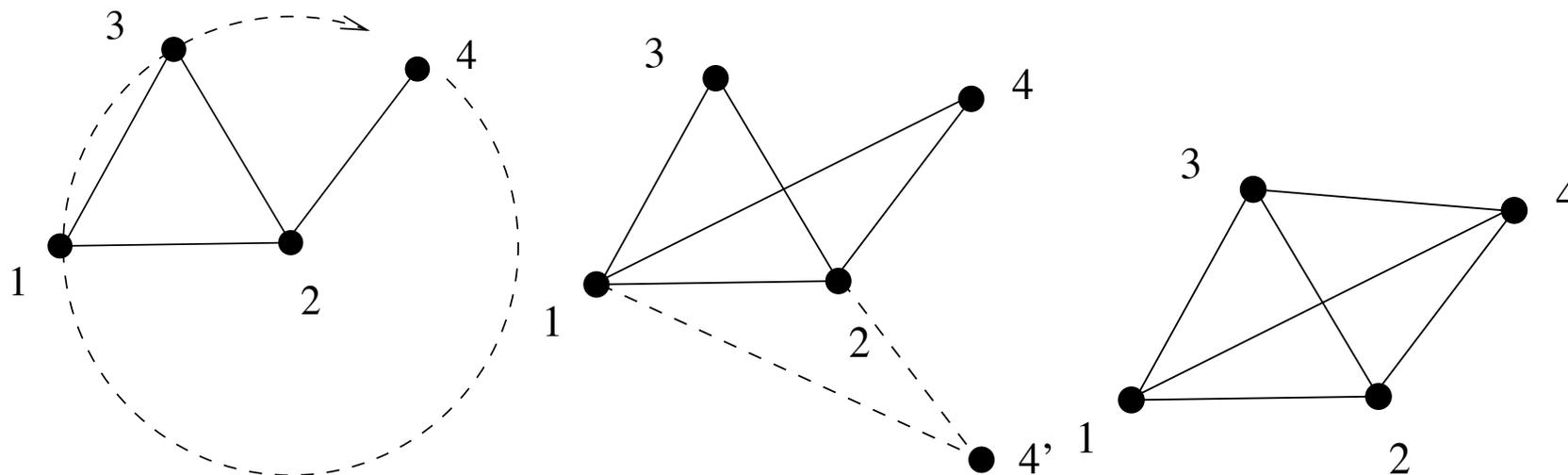
# The number of embeddings

- Uncountably many (incongruent) embeddings
- Finitely many

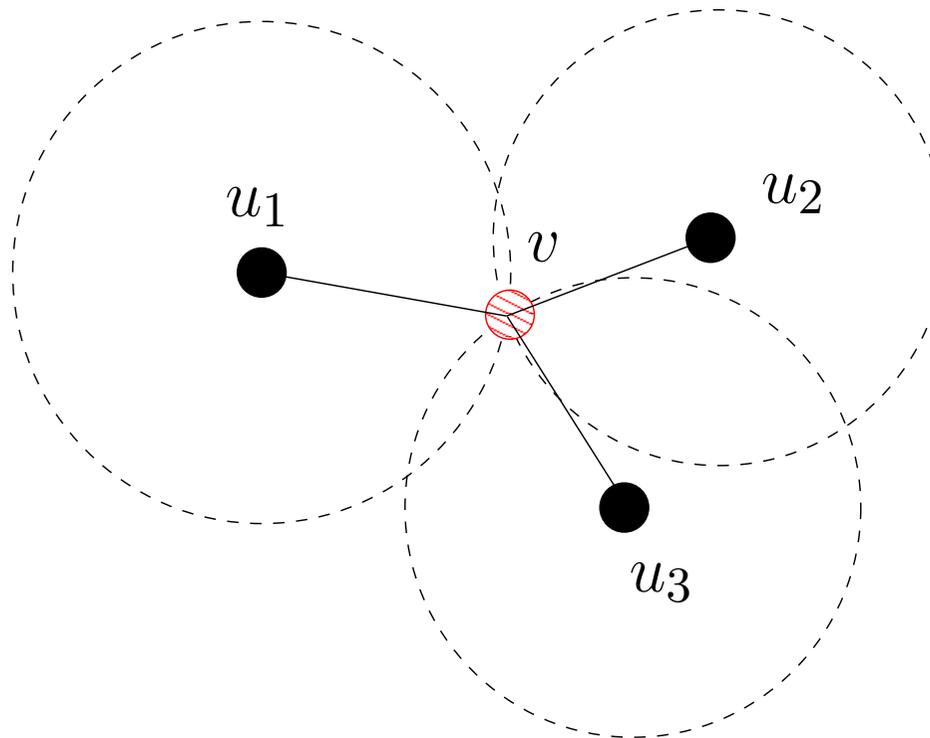


# The number of embeddings

- Uncountably many (incongruent) embeddings
- Finitely many
- At most one



# $K$ -lateration



$v$  has  $\geq K + 1$  adjacencies with known general positions  $\Rightarrow$

**Find unique position for  $x_v$  in  $\mathbb{R}^K$  in polytime**



# Example with $K = 3$

Given  $U = \{1, 2, 3, 4\} \subseteq V$  and a partial embedding  $x_1, x_2, x_3, x_4 \in \mathbb{R}^3$

1. Consider  $v$  adjacent to all  $u \in U$

2. Extend  $x$  to  $v$  by solving a linear system:

$$\begin{array}{l} \|x_v - x_1\|^2 = d_{1v}^2 \\ \|x_v - x_2\|^2 = d_{2v}^2 \\ \|x_v - x_3\|^2 = d_{3v}^2 \\ \|x_v - x_4\|^2 = d_{4v}^2 \end{array} \Rightarrow \begin{array}{l} \|x_v\|^2 - 2x_v \cdot x_1 + \|x_1\|^2 = d_{1v}^2 \quad (3) \\ \|x_v\|^2 - 2x_v \cdot x_2 + \|x_2\|^2 = d_{2v}^2 \quad (4) \\ \|x_v\|^2 - 2x_v \cdot x_3 + \|x_3\|^2 = d_{3v}^2 \quad (5) \\ \|x_v\|^2 - 2x_v \cdot x_4 + \|x_4\|^2 = d_{4v}^2 \quad (6) \end{array}$$

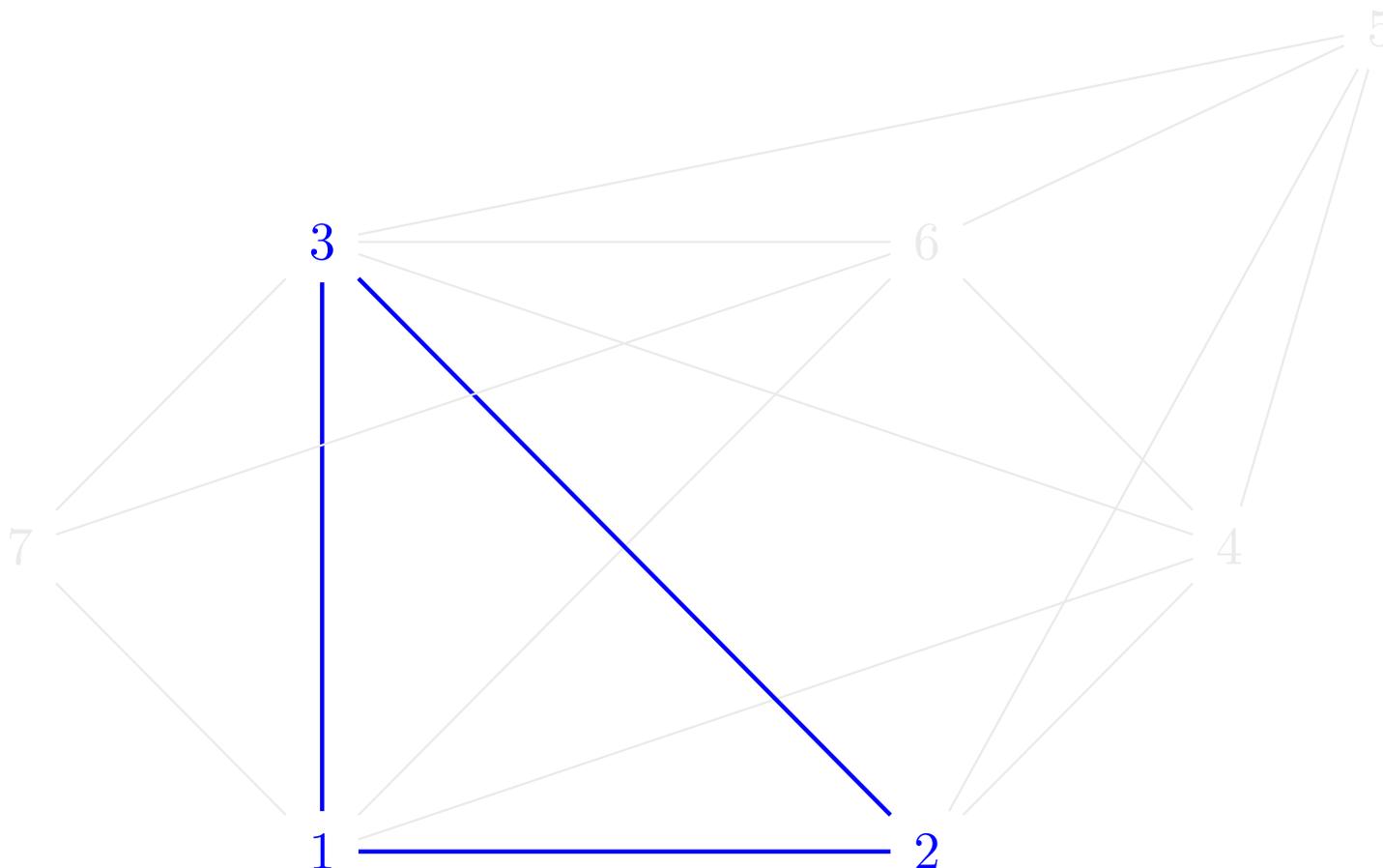
$$\begin{array}{l} (6)-(7) \\ (6)-(8) \\ (6)-(9) \end{array} \Rightarrow \begin{pmatrix} 2(x_1 - x_4)^\top \\ 2(x_2 - x_4)^\top \\ 2(x_3 - x_4)^\top \end{pmatrix} x_v = \begin{pmatrix} (\|x_1\|^2 - \|x_4\|^2) - (d_{1v}^2 - d_{4v}^2) \\ (\|x_2\|^2 - \|x_4\|^2) - (d_{2v}^2 - d_{4v}^2) \\ (\|x_3\|^2 - \|x_4\|^2) - (d_{3v}^2 - d_{4v}^2) \end{pmatrix}$$

Can do this in  $O(K^3)$ , if  $K$  is fixed, this is  $O(1)$



# Combinatorial iterative approach

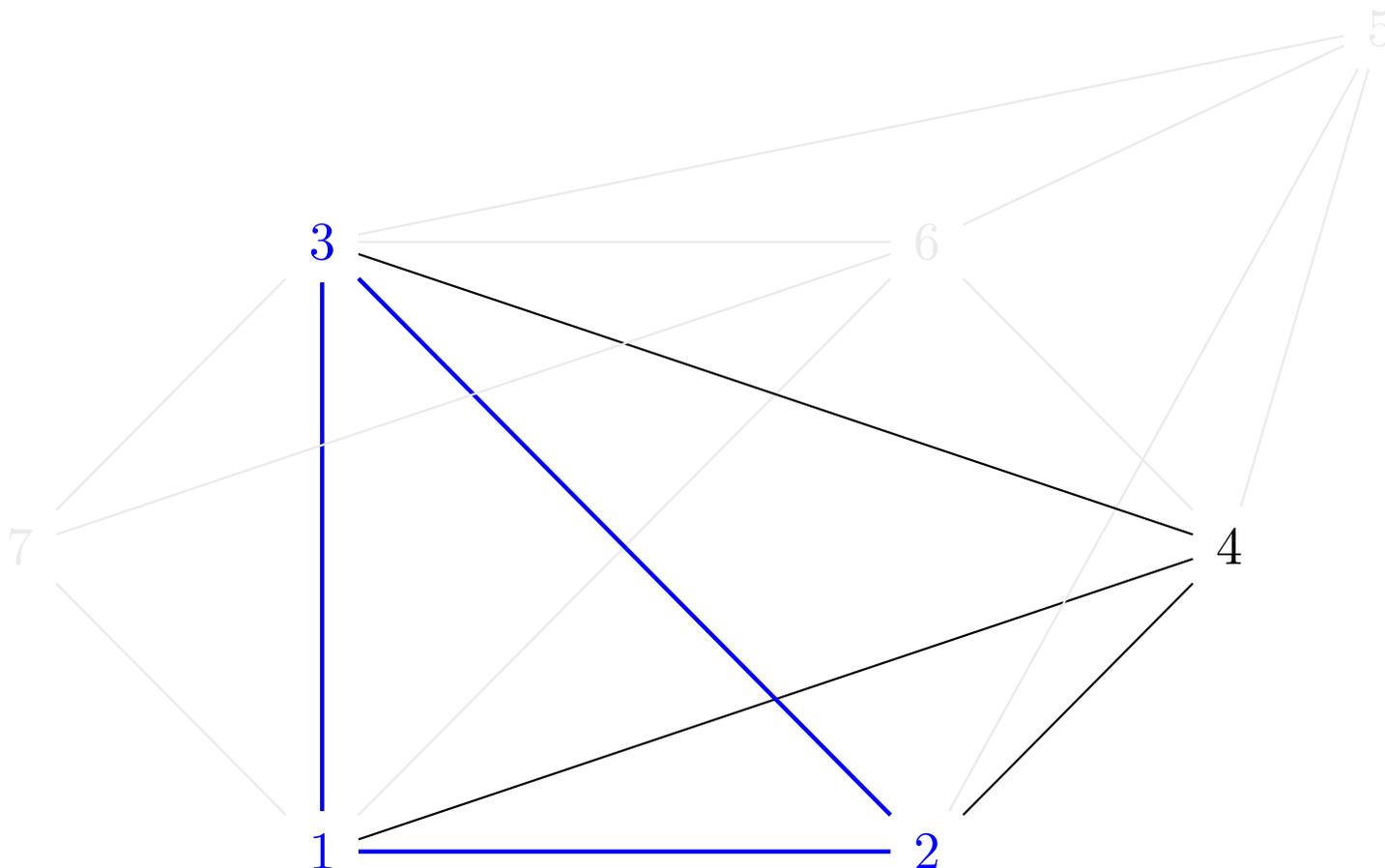
$K = 2$ ; if  $\exists$  vertex order s.t. **next vertex** has enough **adjacent predecessors** :





# Combinatorial iterative approach

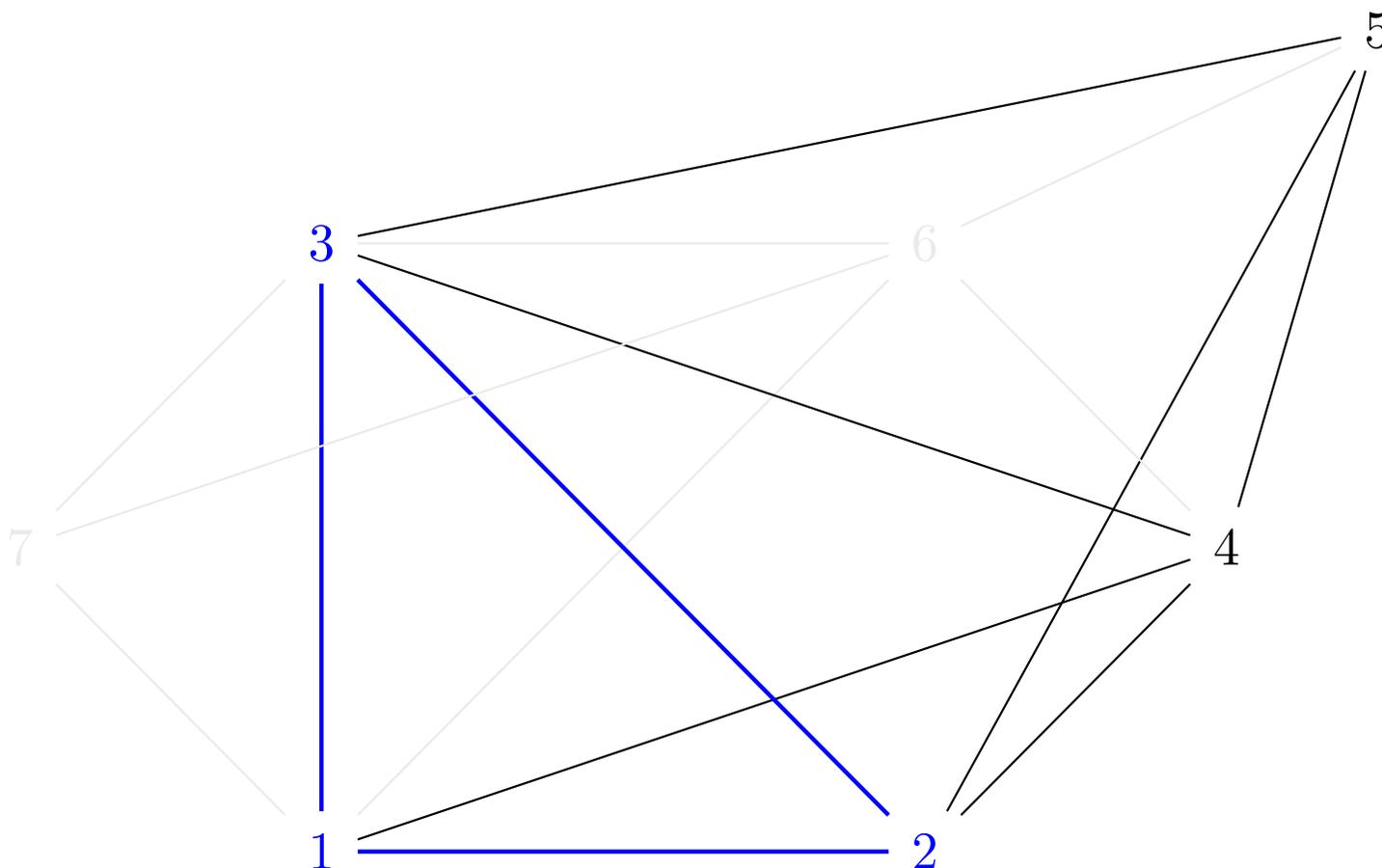
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# Combinatorial iterative approach

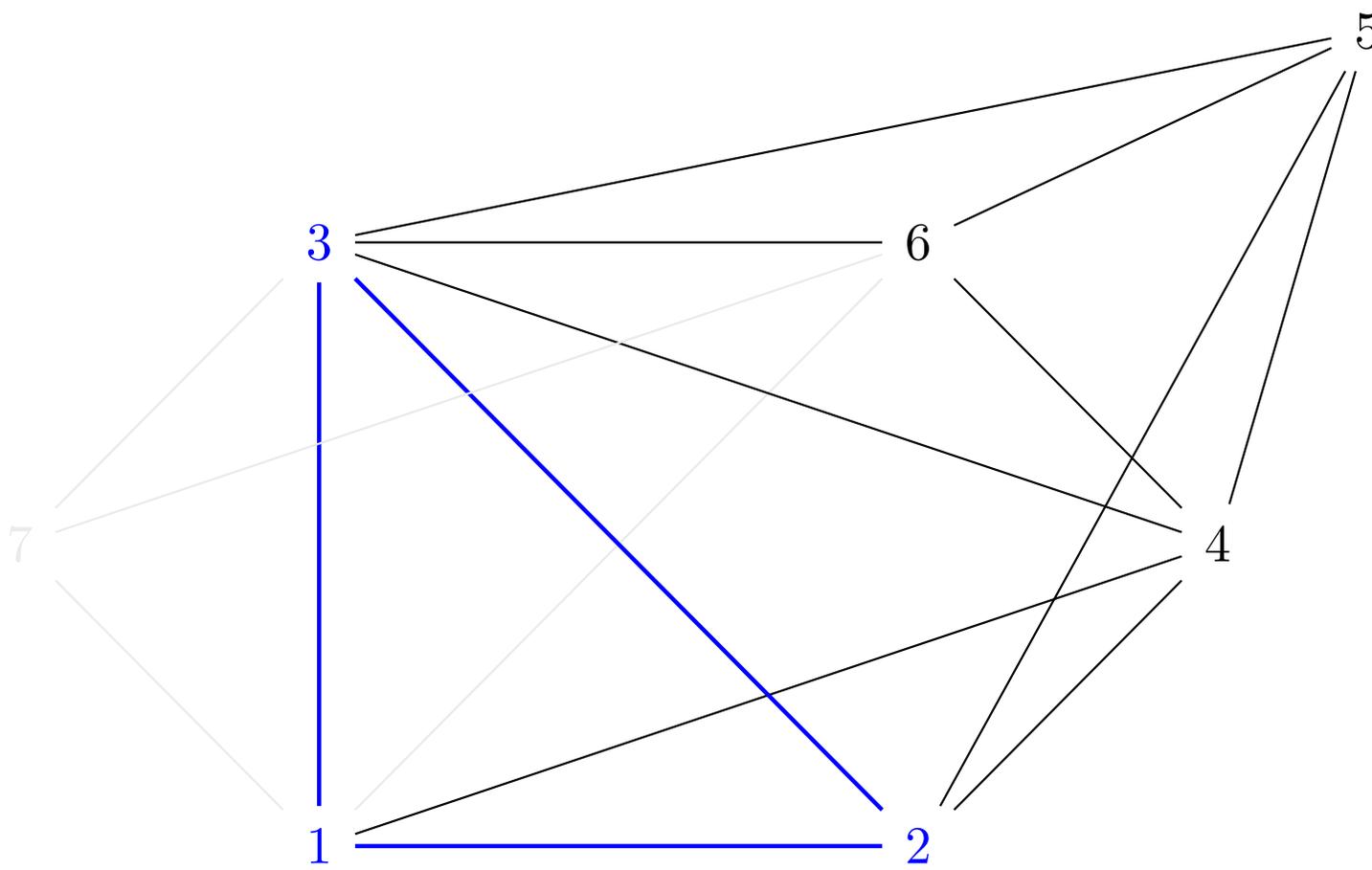
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# Combinatorial iterative approach

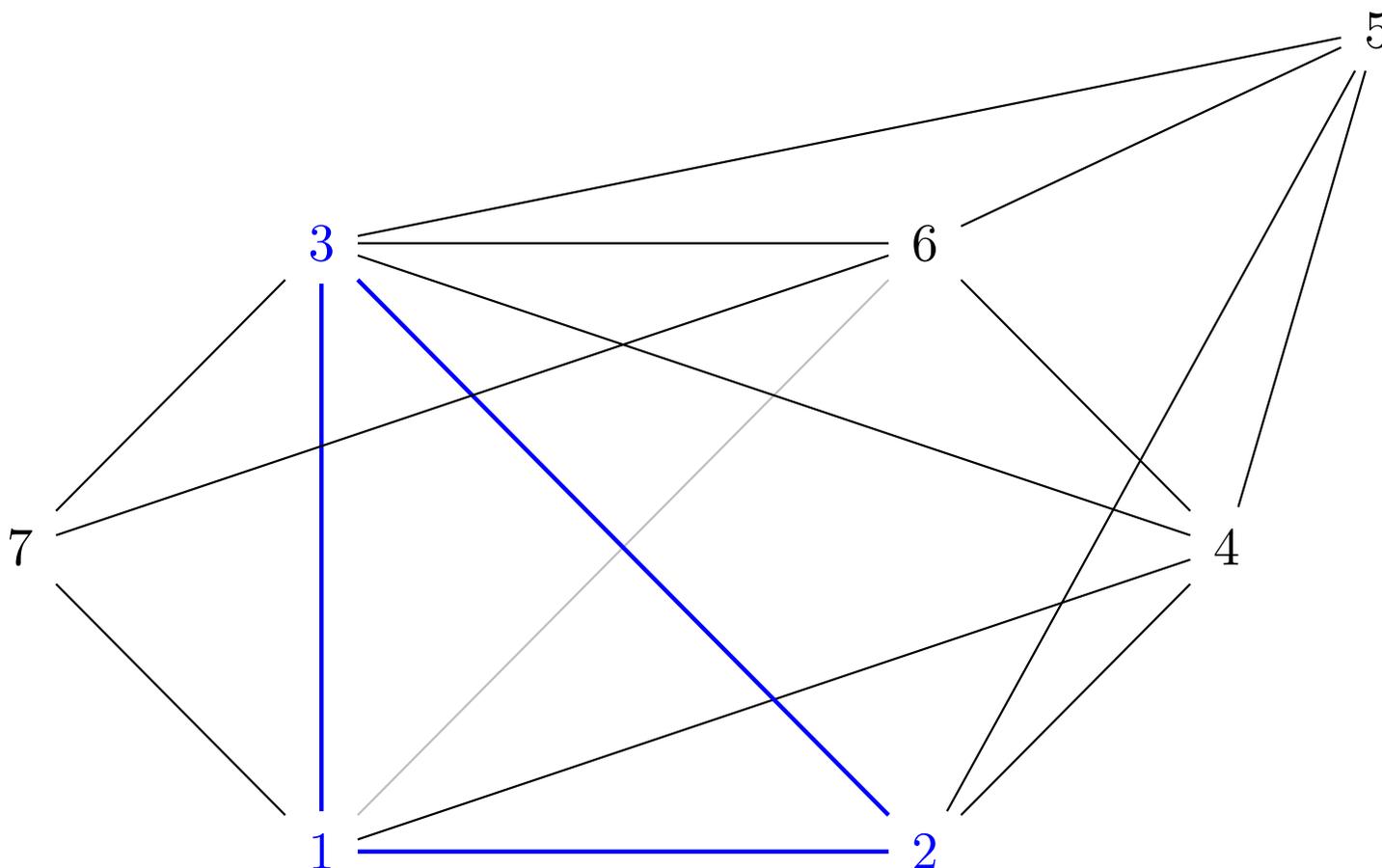
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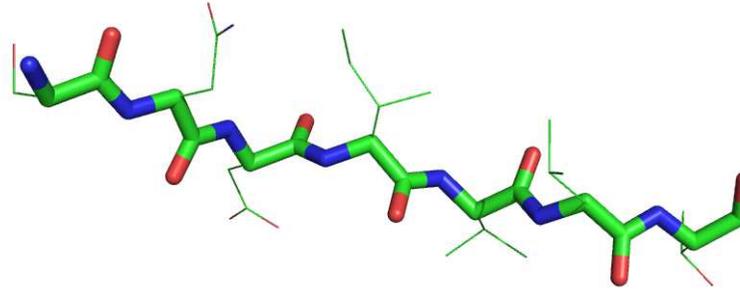
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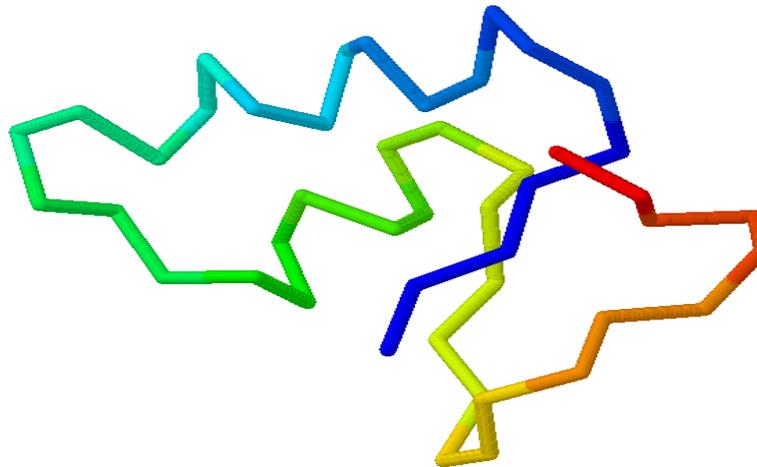


# Proteins

- Proteins: *backbone* and *side chains*



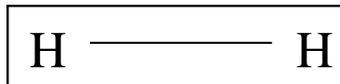
- Backbone: total order  $<$  on a set  $V$  of atoms



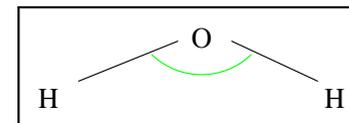
- Decompose the problem: embed the backbone, then plug the side chains in

# Protein distances

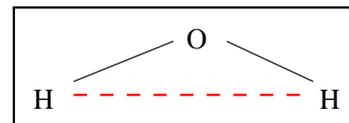
- Covalent bond distances  $d_{v-1,v}$  are known



- Angles between covalent bonds are known



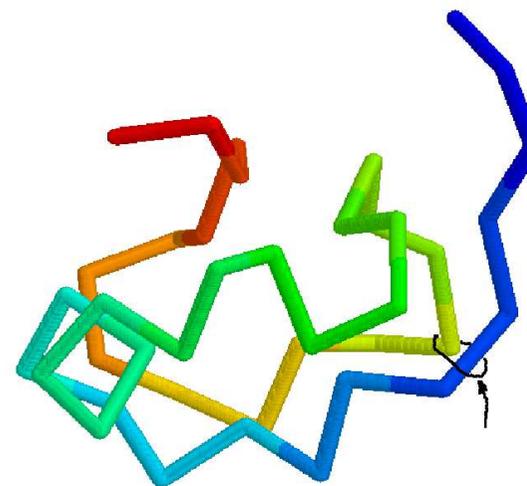
- $\Rightarrow d_{v-2,v}$  is known for all  $v > 3$



- Distances  $d_{v-3,v}$  are always  $< 6\text{\AA}$ , so they can be measured using NMR techniques

- NMR might give other distances too

**Atoms may be distant order-wise but  
closer than  $6\text{\AA}$  in space**





# Discretizable MDGP

- Protein backbones: 3 consecutive predecessors in 3D
- Weaken the condition  $\geq K + 1$  adjacent predecessors in  $\mathbb{R}^K$  to:

**$\geq K$  consecutive adjacent predecessors in  $\mathbb{R}^K$**

- DMDGP: given  $x_1, \dots, x_K \in \mathbb{R}^3$ , and a vertex order as above, find  $x_{K+1}, \dots, x_n$  satisfying

$$\forall \{u, v\} \in E \quad \|x_u - x_v\| = d_{uv}$$

- An **NP-hard** problem



# Can we adapt the iterative method?



# Sphere intersection

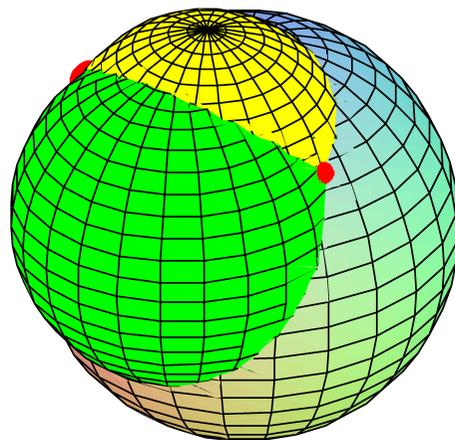
For given  $v > 3$ ,

- $x_{v-3}, x_{v-2}, x_{v-1}$  are known
- $d_{v,v-1}, d_{v,v-2}, d_{v,v-3}$  are known

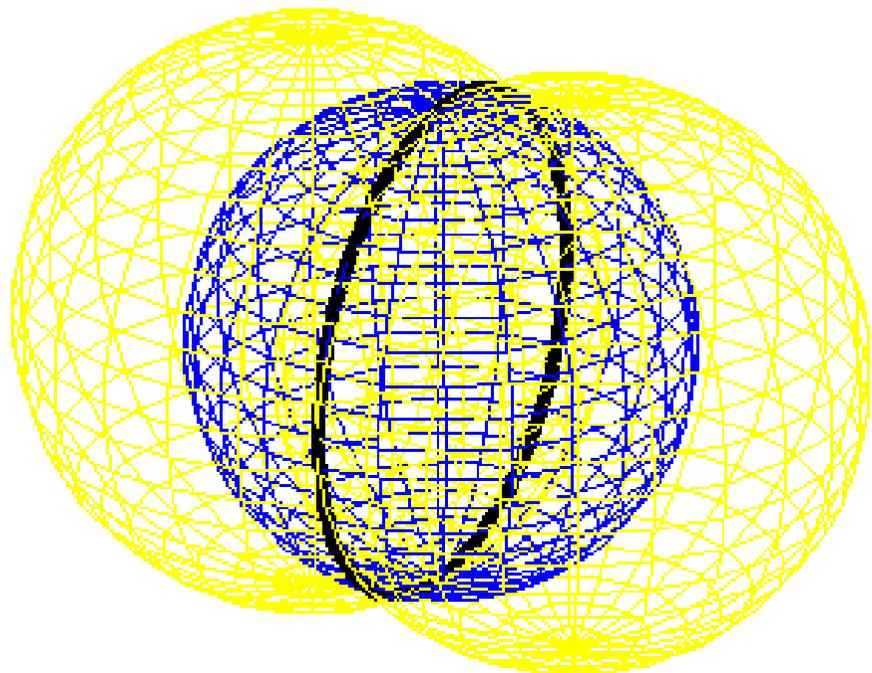
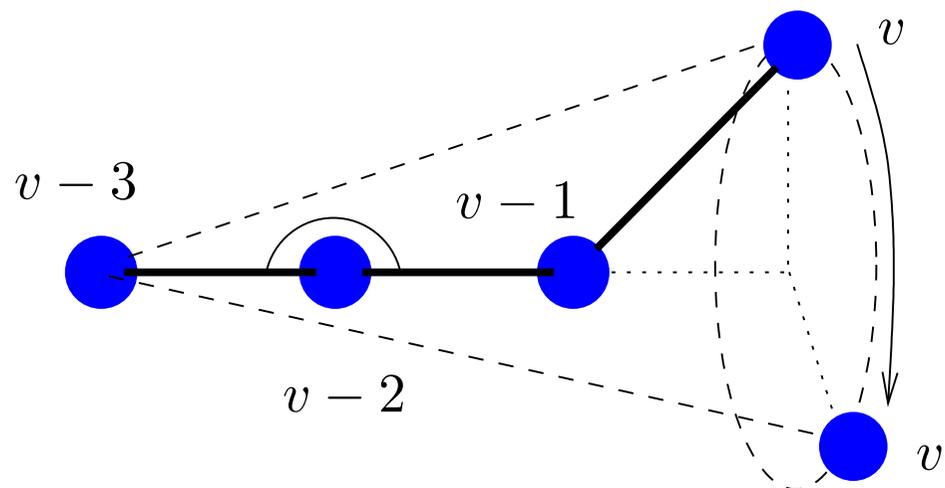
find  $x_v$

---

Non-empty intersection of  $K$  spheres in  $\mathbb{R}^K$  contains 2 points in general



# Failure: collinearity



# Probability 1



- We can develop a theory “modulo collinearity”
- Set of (configurations of  $n$  points in  $\mathbb{R}^K$ ): all  $\mathbb{R}^K$
- Collinearity in general: all points obey an equation  $g(x) = 0$
- $\{x \mid g(x) = 0\}$ : lower-dimensional manifold in  $\mathbb{R}^K$ , volume in  $\mathbb{R}^K$  is 0
- Probability of sampling collinear embedding  $x$ : 0
- Results holding “with probability 1”  $\equiv$  apart from a set of cases having volume 0 in the set of all possible cases



# Finding the 2 points ( $K = 3$ )

Given  $U = \{1, 2, 3\} \subseteq V$  and a partial embedding  $x_1, x_2, x_3 \in \mathbb{R}^3$

1. Consider  $v$  adjacent to all  $u \in U$
2. Extend  $x$  to  $v$  by solving a linear system:

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3. Diagonalize the  $2 \times 3$  linear system (one pivot)
4. Express  $x_{v1}, x_{v2}$  in function of  $x_{v3}$  linearly (\*)
5. Replace  $x_{v3}$  in Eq. (9), solve quadratic in  $x_{v3}$
6. Obtain two values for  $x_{v3}$ , use (\*) to find two points for  $x_v$



# Branch-and-Prune

$v$ : rank of current atom     $x_{<v}$ : partial embedding to rank  $v - 1$

$G$ : instance     $X$ : current pool of embeddings

$S(y, r)$ :  $\mathbb{R}^K$  sphere centered at  $y$  with radius  $r$

BRANCHANDPRUNE( $v, x_{<v}, G, X$ ):

Let  $\mathcal{S} \leftarrow \bigcap_{i \in \{1, \dots, K\}} S(x_{v-i}, d_{v-i, v}) = (\{s_1, s_2\} \text{ or } \emptyset)$

**for**  $s \in \mathcal{S}$  **do**

Extend current embedding to  $x = (x_{<v}, s)$

**if**  $\forall u \in \text{AdjPred}(v) \|x_u - x_v\| = d_{uv}$  **then**

**if** ( $v = n$ ) **then**

Let  $X \leftarrow X \cup \{x\}$

**else**

BRANCHANDPRUNE( $v + 1, x, G, X$ )

**end if**

**end if**

**end for**



# BP properties

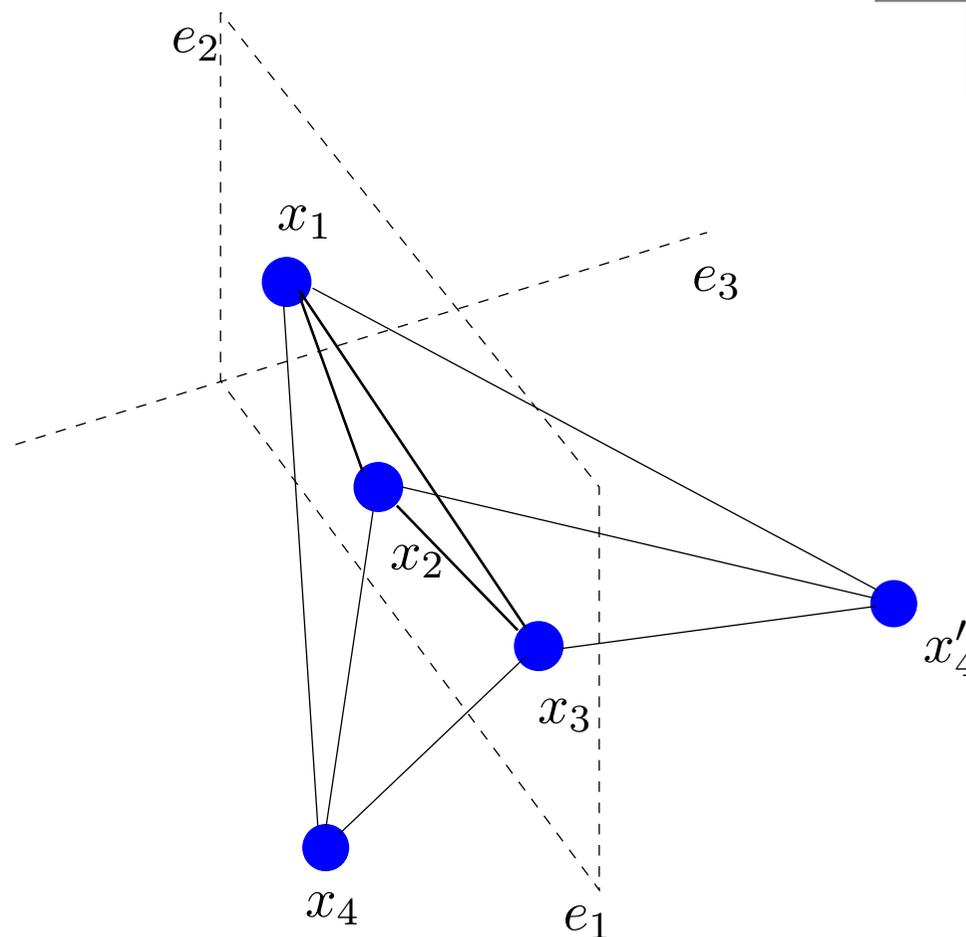
- BP: worst-case exponential time
- With probability 1, find **all incongruent embeddings** of  $G$  extending initial partial embedding
- Performs very efficiently (speed and accuracy)  
**Embed 10,000 vertices in a 13 seconds of CPU time**
- Two empirical observations:
  1. **the number of solutions it finds is always a power of two**
  2.  **$|V|$  versus CPU time plots are always linear-like for PDB**



# Symmetry

# BP root node symmetry

- $x'_4$  is a reflection of  $x_4$  w.r.t. the plane defined by  $x_1, x_2, x_3$
- $\Rightarrow$  BP tree symmetric below level 3
- Start branching from level 4, not 3





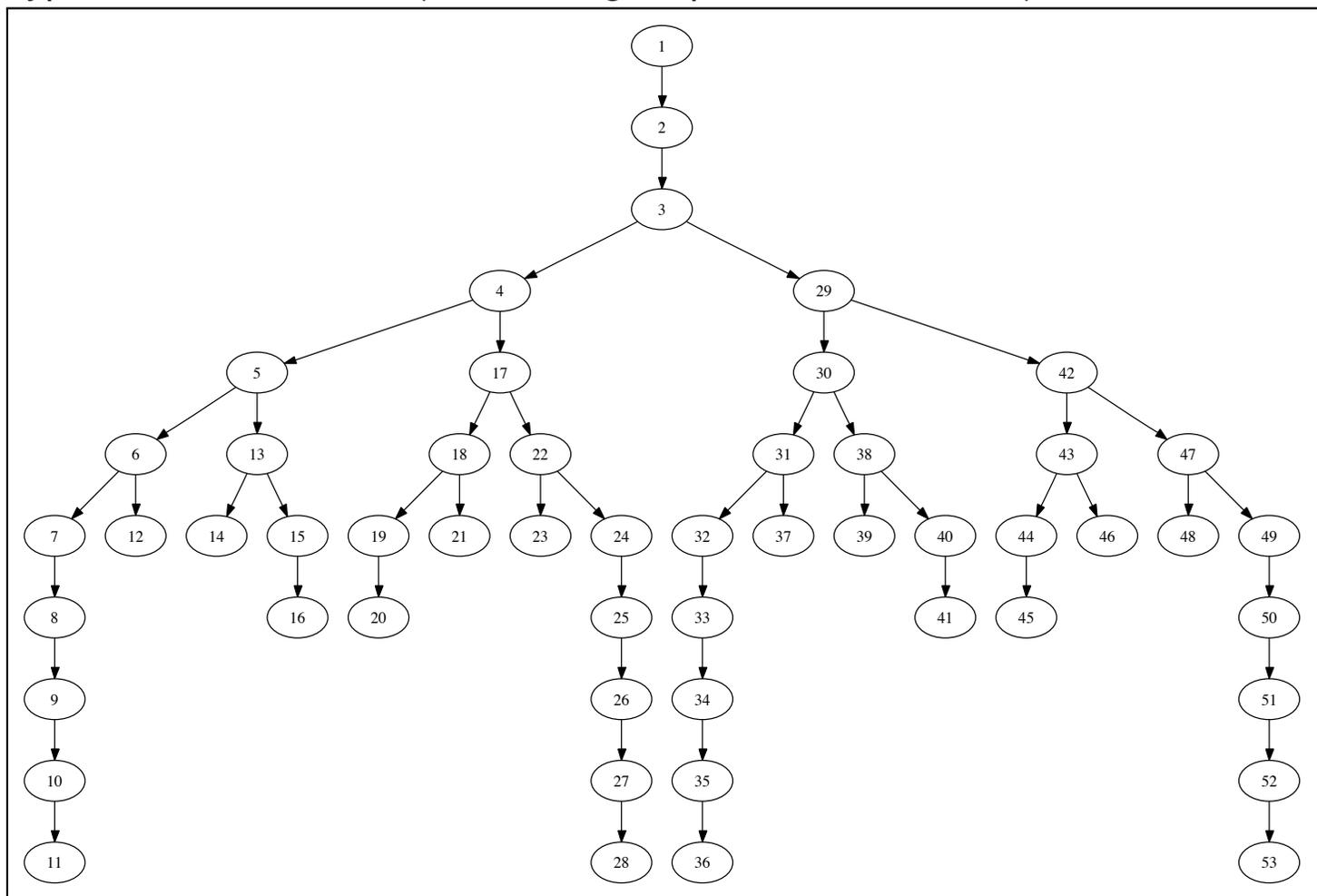
# Number of solutions

<i>Instance</i>	$ X $	<i>Instance</i>	$ X $
		1brv	2
		1aqr	4
		2erl	2
		1crn	2
mmorewu-2	2	1ahl	16
mmorewu-3	2	1ptq	2
mmorewu-4	4	1brz	4
mmorewu-5	4	1hoe	2
mmorewu-6	4	1lfb	2
		1pht	2
lavor10_0	4	1jk2	2
lavor15_0	16	1f39a	2
lavor20_0	8	1acz	8
lavor25_0	8	1poa	2
lavor30_0	2	1fs3	2
lavor35_0	64	1mbn	2
lavor40_0	2	1rgs	2
lavor45_0	2	1m40	2
lavor50_0	4096	1bpm	2
lavor55_0	64	1n4w	2
lavor60_0	64	1mqq	2
		1rwh	2
		3b34	2
		2e7z	2
		1epw	2

For all tested DMDGP instances,  $\exists \ell \in \mathbb{N}$  such that  $|X| = 2^\ell$

# A BP search tree example

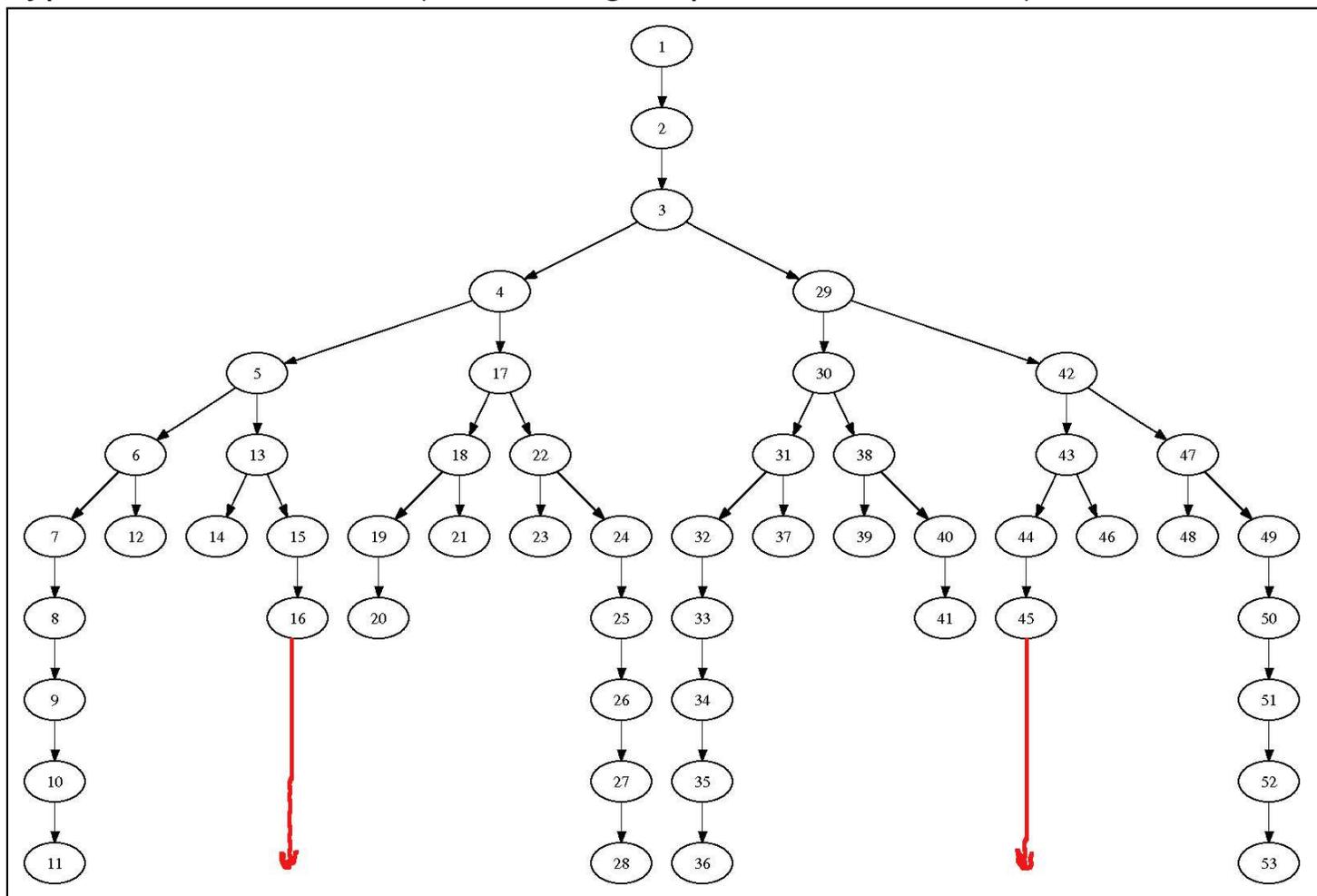
Typical BP search tree (embeddings = paths root→leaves)



- Root node symmetry:  $|X|$  is even
- No evident reason why  $|X|$  should be a power of two

# A BP search tree example

Typical BP search tree (embeddings = paths root→leaves)



- Root node symmetry:  $|X|$  is even
- No evident reason why  $|X|$  should be a power of two  
(why not symmetric paths to level  $|V|$  from nodes 16 and 45?)

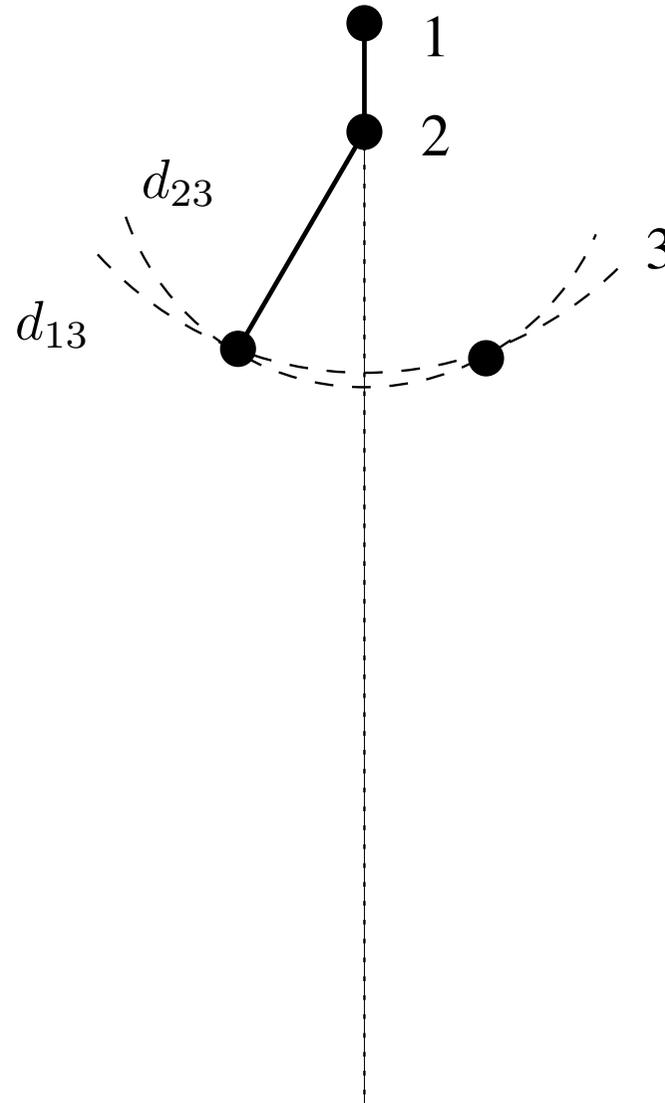


# Discretization/pruning distances

- Let  $E_D = \{\{u, v\} \mid |u - v| \leq K\}$  and  $E_P = E \setminus E_D$
- $E_D$ : **discretization distances**
  - they guarantee that the instance is a DMDGP
  - they allow the construction of the complete BP tree
  - this tree has  $2^{|V|-3}$  leaves,  $2^{|V|-4}$  if we consider root node symmetry
- $E_P$ : **pruning distances**
  - they allow pruning of the BP tree
  - not clear why they should prune branches symmetrically

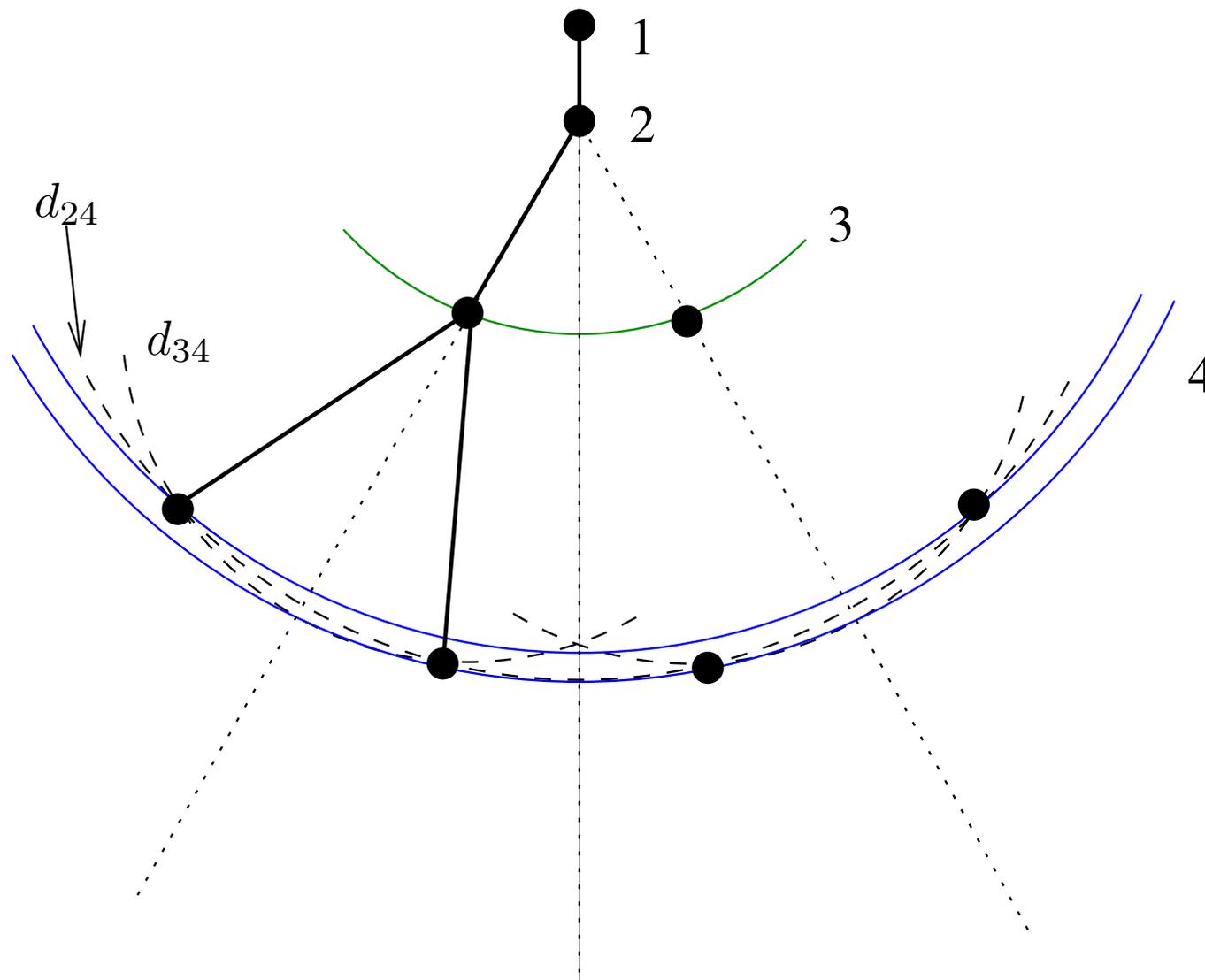


# Structure of the BP tree ( $\mathbb{R}^2$ )

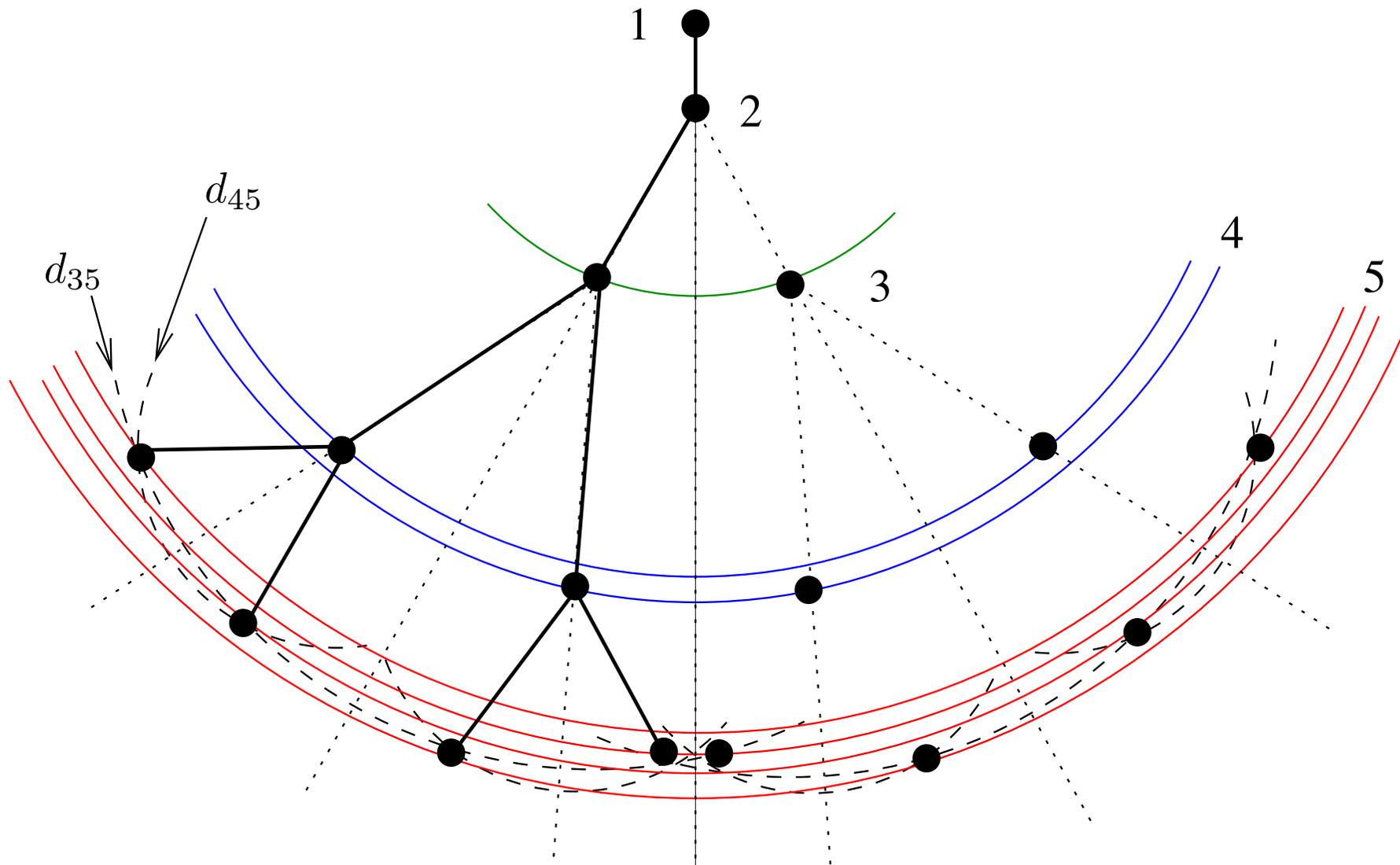




# Structure of the BP tree ( $\mathbb{R}^2$ )



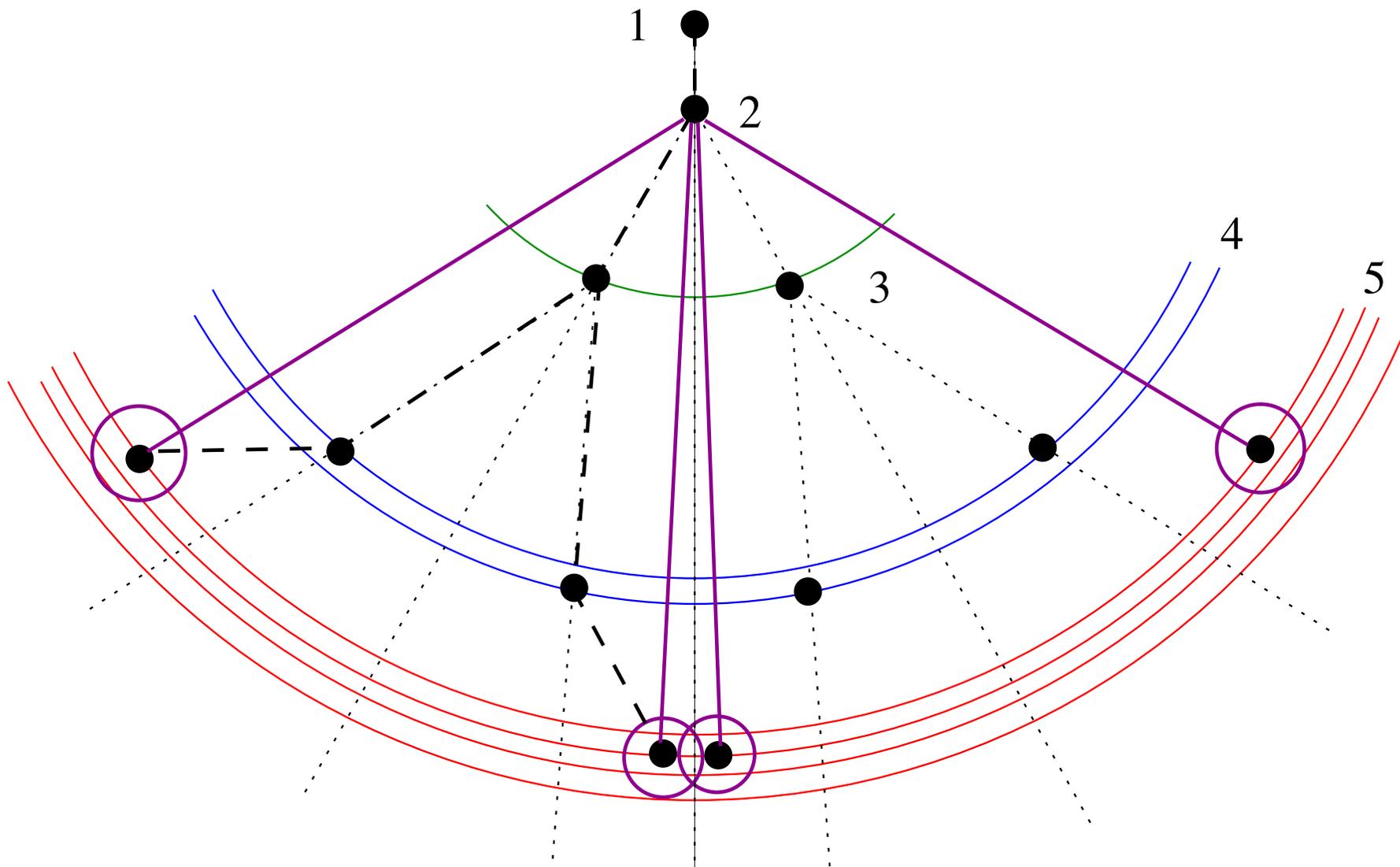
# Structure of the BP tree ( $\mathbb{R}^2$ )





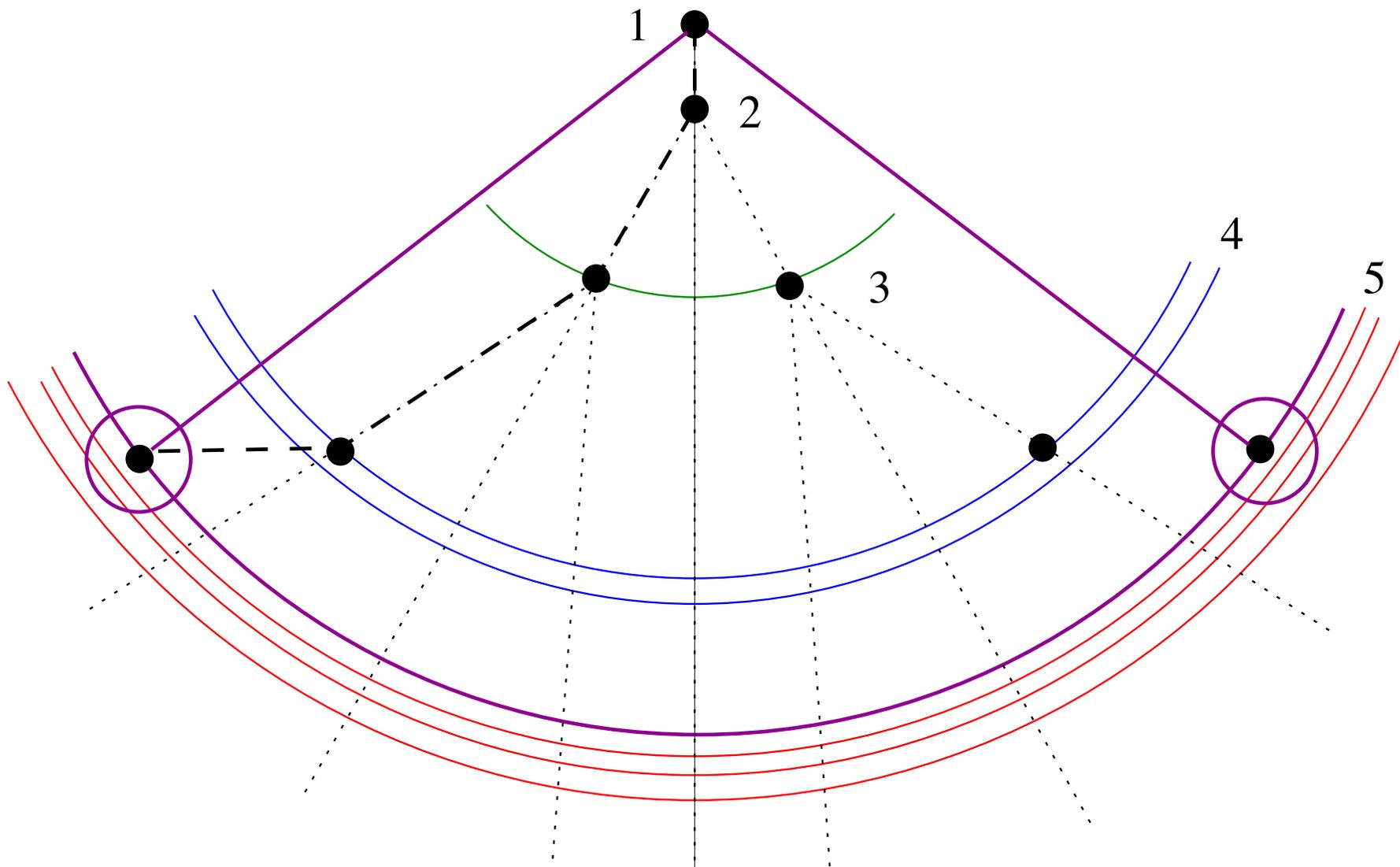


# Effect of pruning distance $d_{25}$



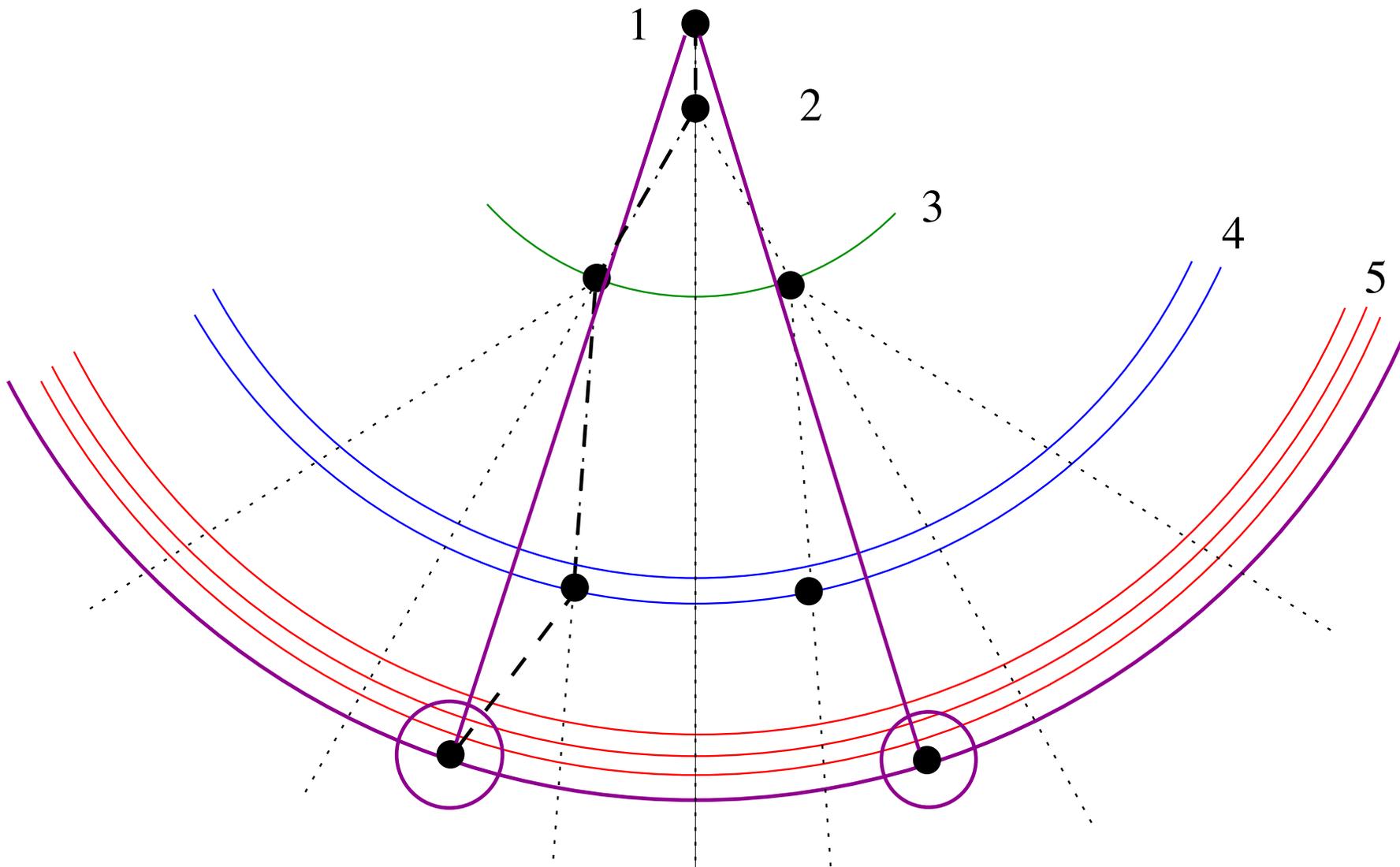


# Effect of pruning distance $d_{15}$

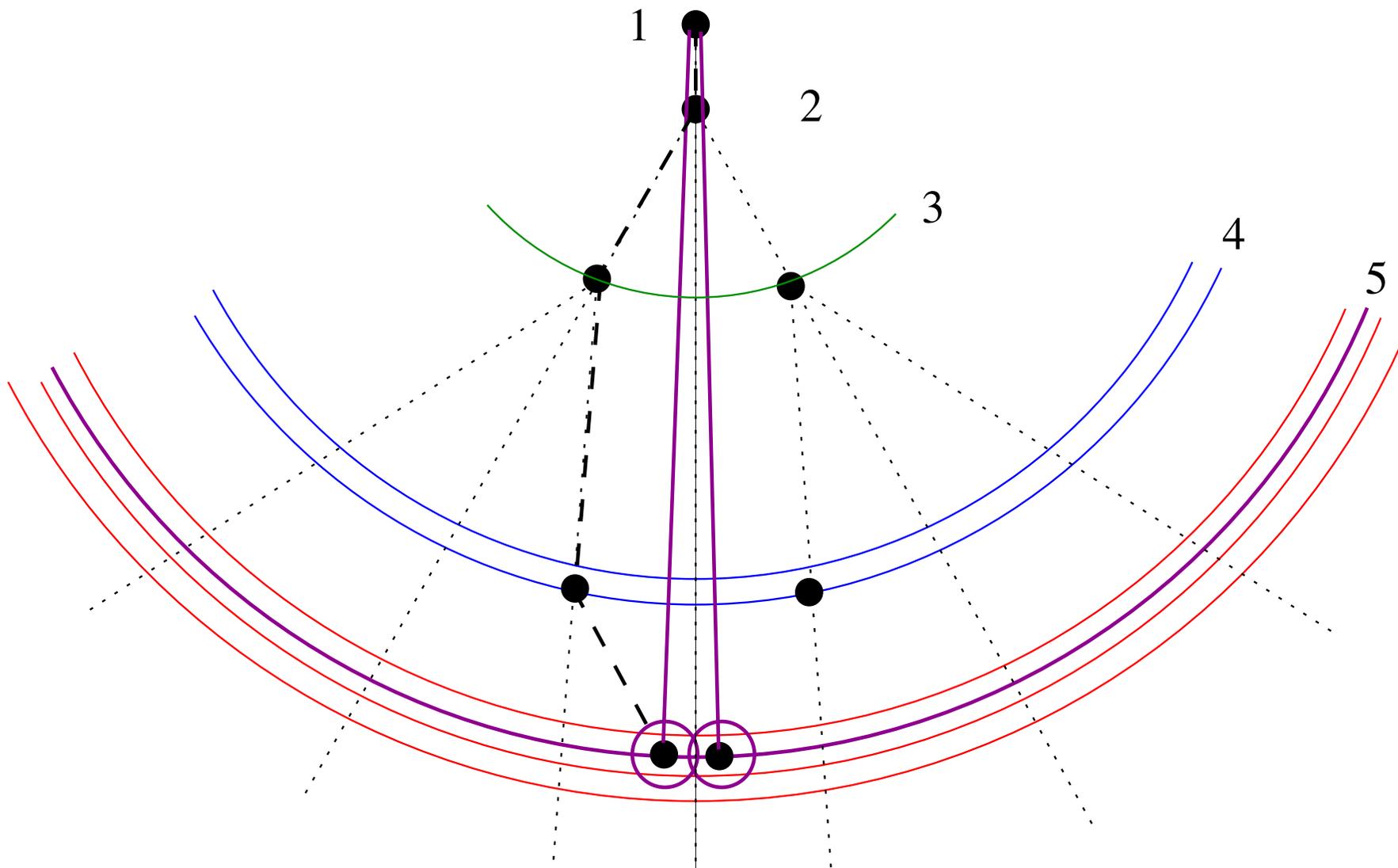




# Effect of pruning distance $d_{15}$



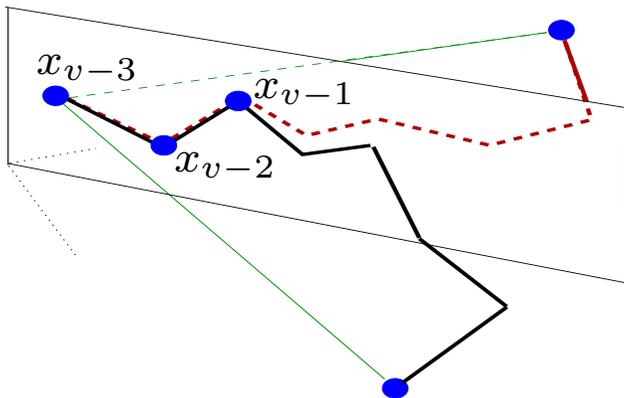
# Effect of pruning distance $d_{15}$





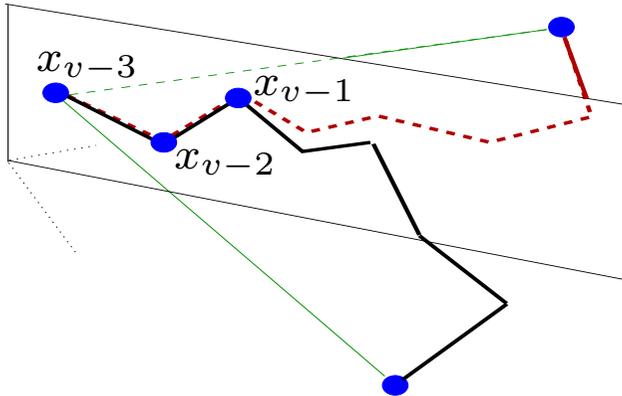
# Symmetry by pruning distances

Given embedding  $x$ ,  $R_x^v =$  reflection w.r.t. hyperplane  $x_{v-K}, \dots, x_{v-1}$



# Symmetry by pruning distances

Given embedding  $x$ ,  $R_x^v =$  reflection w.r.t. hyperplane  $x_{v-K}, \dots, x_{v-1}$



*Thm.*

With prob. 1, for each  $u, v \in V$  with  $v > K, u < v - K$ ,

$$\forall x \neq x' \in X \quad \|x_u - x_v\| = \|x'_u - x'_v\| \Leftrightarrow x'_v = R_x^{u+K}(x_v)$$

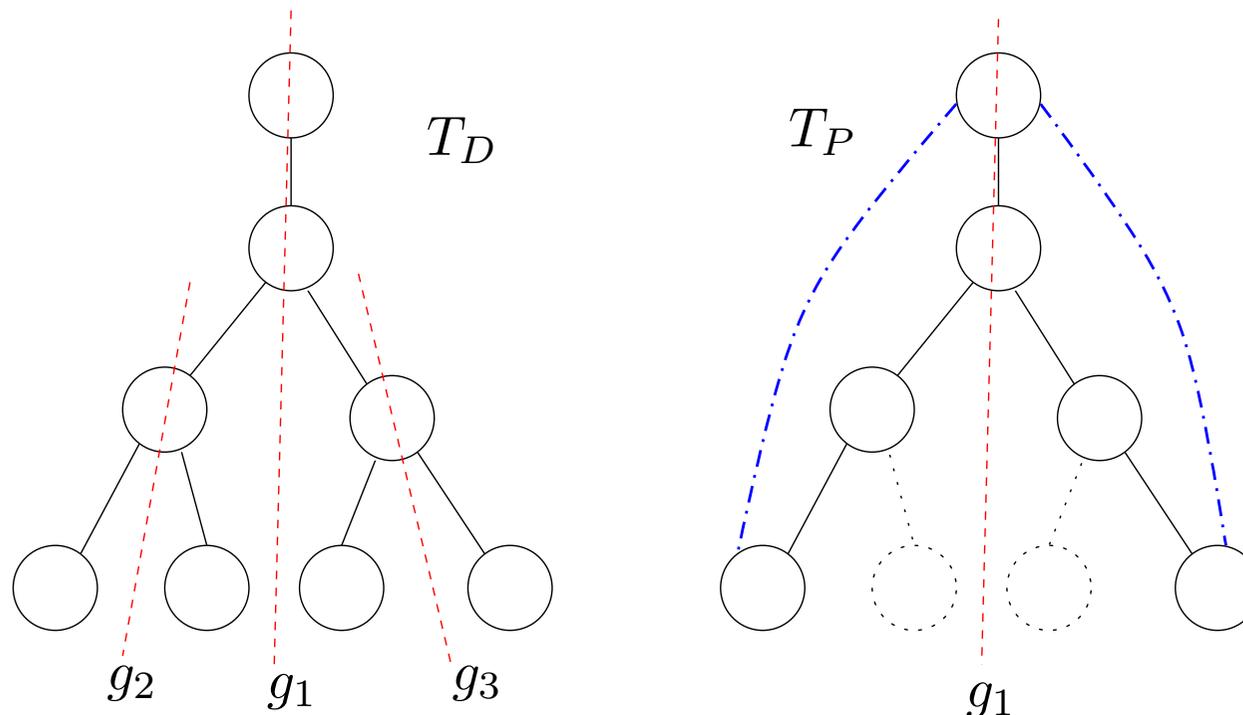
Moreover,  $\exists$  a finite set  $H^{uv} \subseteq \mathbb{R}_+$  with  $|H^{uv}| = 2^{v-u-K}$  s.t.

$$\forall x \in X \left( \underbrace{\|x_u - x_v\|}_{\text{plays the role of pruning dist.}} \in H^{uv} \right)$$

plays the role of pruning dist.

# Groups fixing the trees

- Let  $T_D$  be a full BP binary search tree
- Let  $T_P$  be the subtree of  $T_D$  representing only feasible branches
- Draw them so  $T_P \subseteq T_P$
- Invariant group for  $T_D$ : all partial reflections ( $g_1, g_2, g_3$ )
- Invariant group for  $T_P$ : only some partial reflections ( $g_1$ )





# Partial reflections

$$g_v(x) = (x_1, \dots, x_{v-1}, R_x^v(x_v), \dots, R_x^v(x_n))$$

**Only reflect starting from vertex  $v$**

# Discretization group

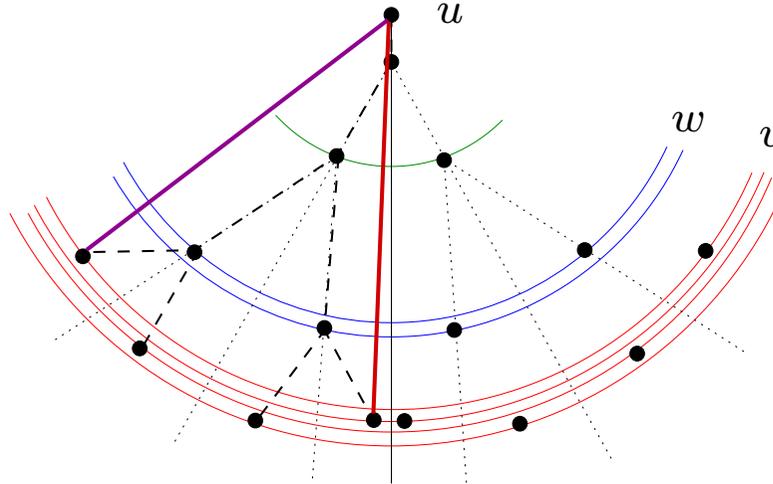
## Group of partial reflections fixing the complete BP tree (no pruning distances)

- The following hold with probability 1  $\forall v > K$ :
  1.  $g_v$  is injective with probability 1 (by reflection)
  2.  $g_v$  is idempotent (by reflection)
  3.  $\forall u > K, u \neq v, g_u$  and  $g_v$  commute (nontrivial)
- Thus,  $\mathcal{G}_D = \langle g_v \mid v > K \rangle$  is an Abelian group under composition  
 $\Rightarrow$  isomorphic to  $C_2^{n-K}$
- By previous thm, discretization distances are invariant under  $\mathcal{G}_D$
- The action of  $\mathcal{G}_D$  on  $X$  is transitive,  
i.e.  $\forall x, x' \in X \exists g \in \mathcal{G}_D (x' = g(x))$
- This action has only one orbit, i.e.  $X = \mathcal{G}_D x$

# Pruning group

## Group of partial reflections fixing the actual BP tree (with pruning distances)

- Assume DMDGP instance is YES, consider  $\{u, v\} \in E_P$
- With probability 1,  $d_{uv} \in H^{uv}$  (otherwise the instance would be NO)
- Notice  $d_{uv} = \|x_v - x_u\| \neq \|g_w(x)_v - g_w(x)_u\|$  for all  $w \in \{u + K + 1, \dots, v\}$



- In order to keep invariance we remove such  $g_w$ 's from the group
- Pruning group:**  $\mathcal{G}_P = \langle g_w \mid w > K \wedge \forall \{u, v\} \in E_P (w \notin \{u + K + 1, \dots, v\}) \rangle$
- $\mathcal{G}_P \leq \mathcal{G}_D$  and all distances are invariant w.r.t. the pruning group
- Again, **action of  $\mathcal{G}_P$  on  $X$  is transitive** (nontrivial proof)

# Power of two

*Thm.*

$$\exists \ell \in \mathbb{N} (|X| = 2^\ell)$$

*Proof*

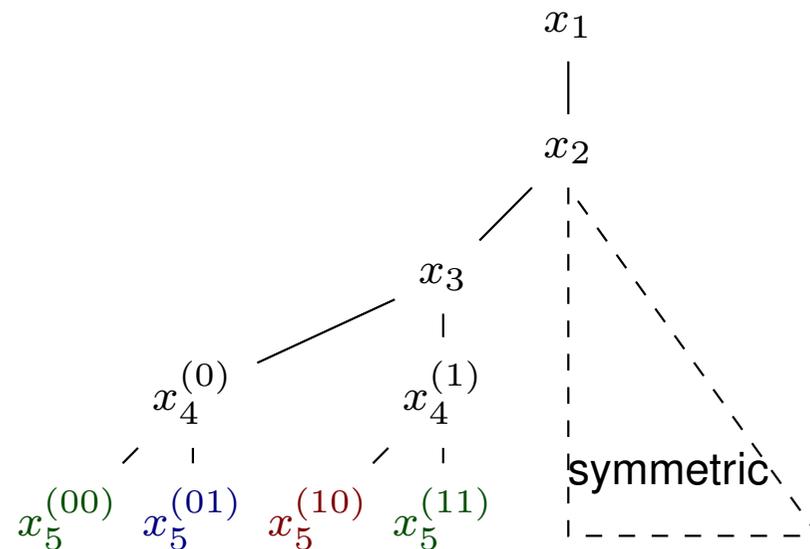
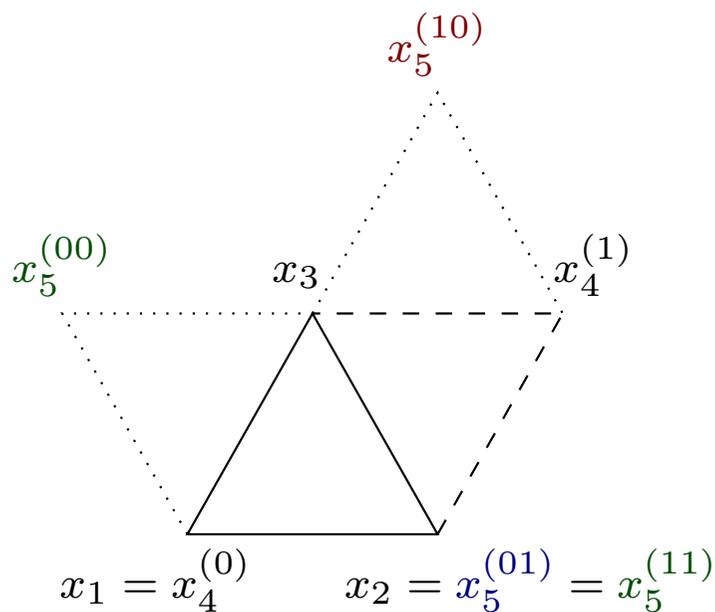
With probability 1:

- $\mathcal{G}_D \cong C_2^{n-K} \Rightarrow |\mathcal{G}_D| = 2^{n-K}$
- $\mathcal{G}_P \leq \mathcal{G}_D \Rightarrow |\mathcal{G}_P| \mid |\mathcal{G}_D| \Rightarrow \exists \ell \in \mathbb{N} |\mathcal{G}_P| = 2^\ell$
- Action of  $\mathcal{G}_P$  on  $X$  is transitive  $\Rightarrow \mathcal{G}_P x = X$
- Idempotency  $\Rightarrow$  for  $g, g' \in \mathcal{G}_P$ , if  $gx = g'x$  then  $g = g' \Rightarrow |\mathcal{G}_P x| = |\mathcal{G}_P|$
- Thus,  $|X| = |\mathcal{G}_P x| = |\mathcal{G}_P| = 2^\ell$



# Why the “probability 1”?

- Not all “YES” DMDGP instances have  $|X| = 2^\ell$
- But the set of such instances (with real data) has Lebesgue measure zero in the set of all DMDGP instances



*Happens when  $> 1$  vertices are embedded in the same position*

$x_5^{(01)}$  should be infeasible, but  $x_5^{(01)} = x_5^{(11)}$  (event with prob. 0)



# FPT behaviour

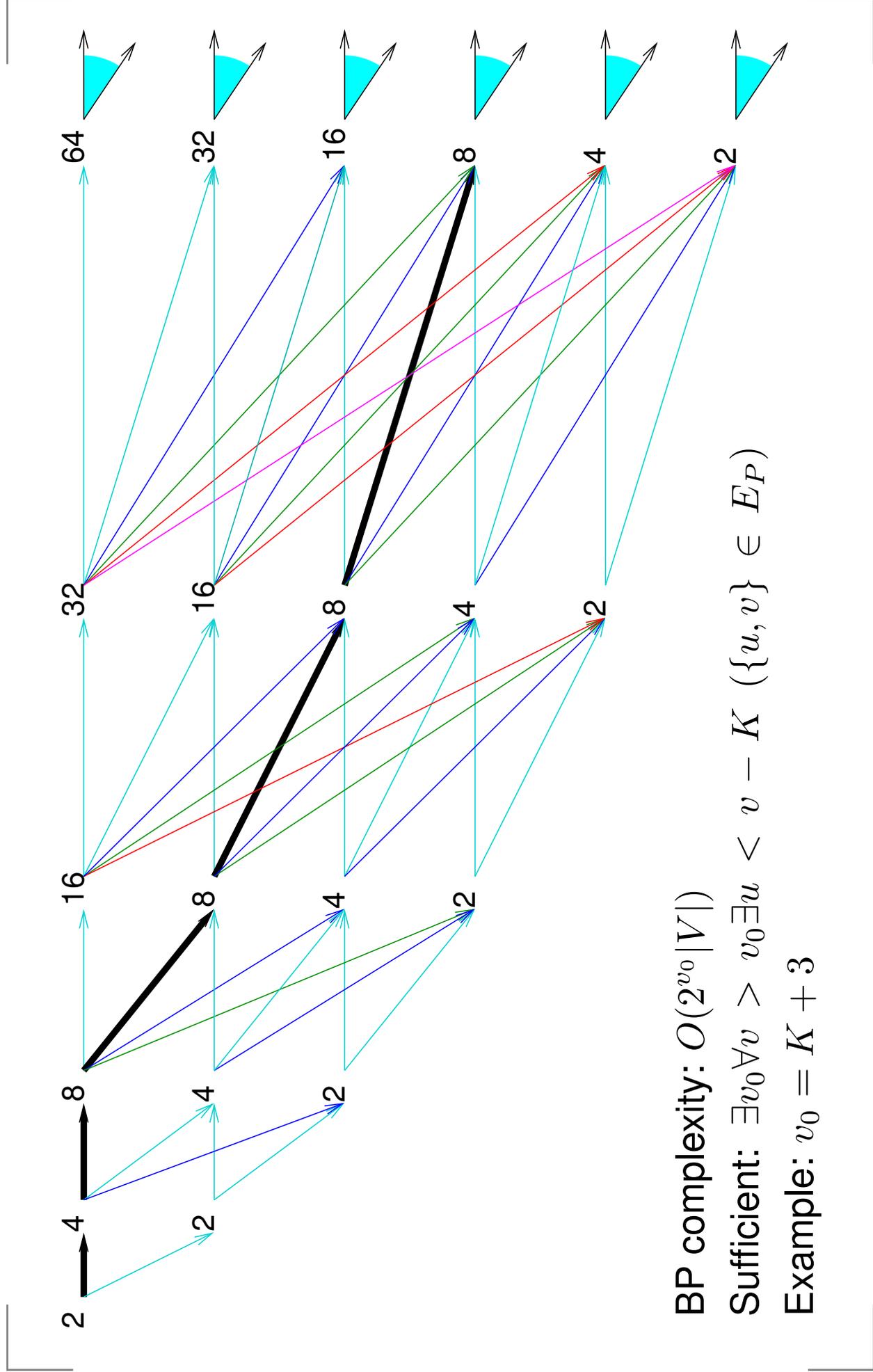


# A polynomial BP?

- Empirically: never an exponential-time increase behaviour in our experiments (instances generated from PDB files)
- Embed 10000-atom protein backbones in 10-15s on one core
- Easy to show that BP has worst-case exponential complexity
- Are proteins a polynomial case of the DMDGP?
- Complexity depends on BP nodes; since  $\text{height} \leq |V|$ , only need to consider treewidth
- A pruning edge  $\{u, v\}$  with  $u < v - K$  reduces the number of nodes at level  $v$  from  $2^{v-K}$  to  $2^{v-K-(u-1)}$  (by symmetry)



# Constant treewidth

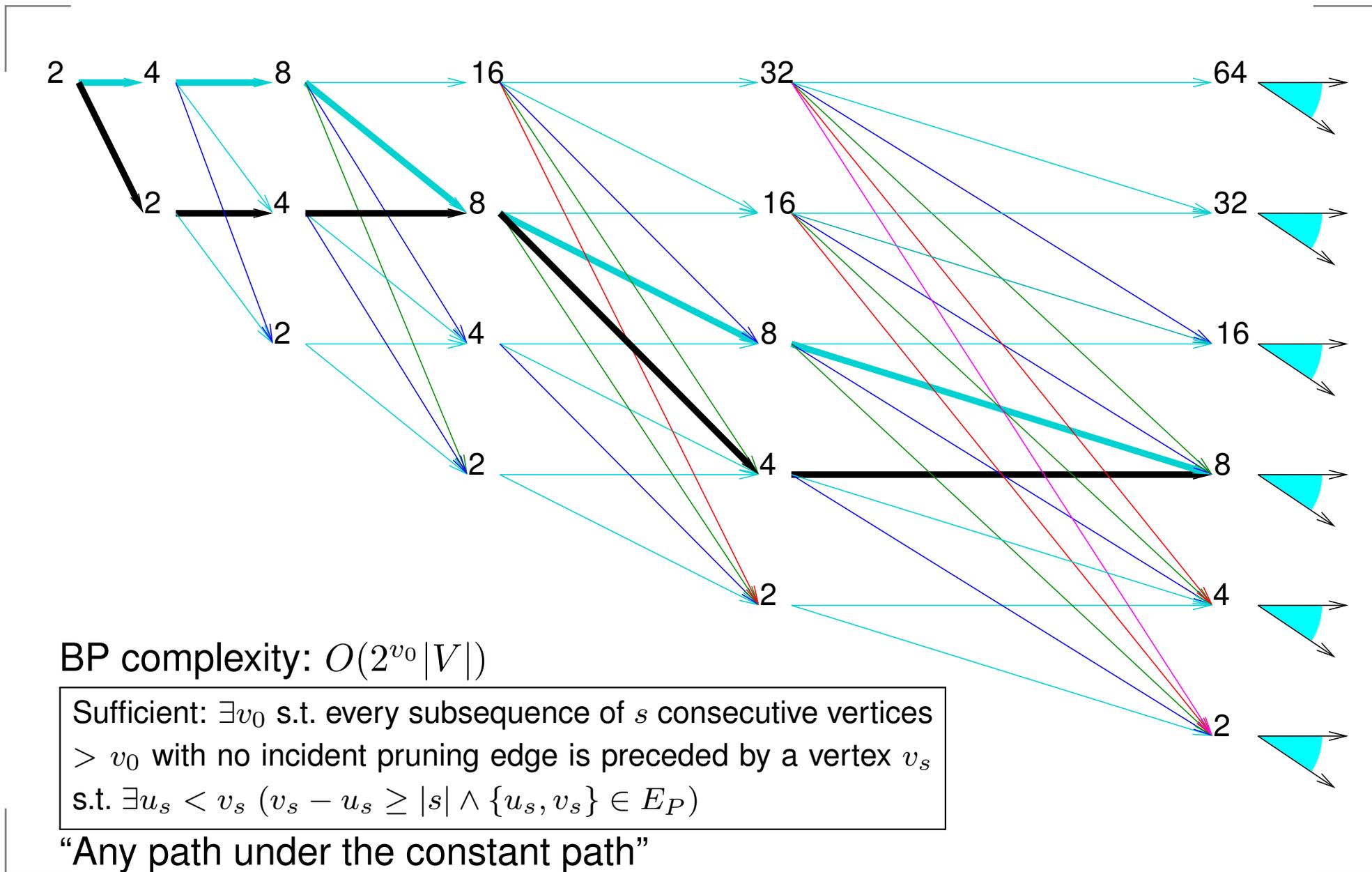


BP complexity:  $O(2^{v_0} |V|)$

Sufficient:  $\exists v_0 \forall v > v_0 \exists u < v - K (\{u, v\} \in E_P)$

Example:  $v_0 = K + 3$

# Constant-bounded treewidth





# Fixed parameter tractability

- We can also allow treewidth growth as long as it's logarithmic in  $n$
- This yields a fixed-parameter tractable behaviour for BP (w.r.t.  $v_0$ )

We tested all our protein instances: all display either constant or const-bounded treewidths with very low  $v_0$  (i.e.  $v_0 = 4$ )

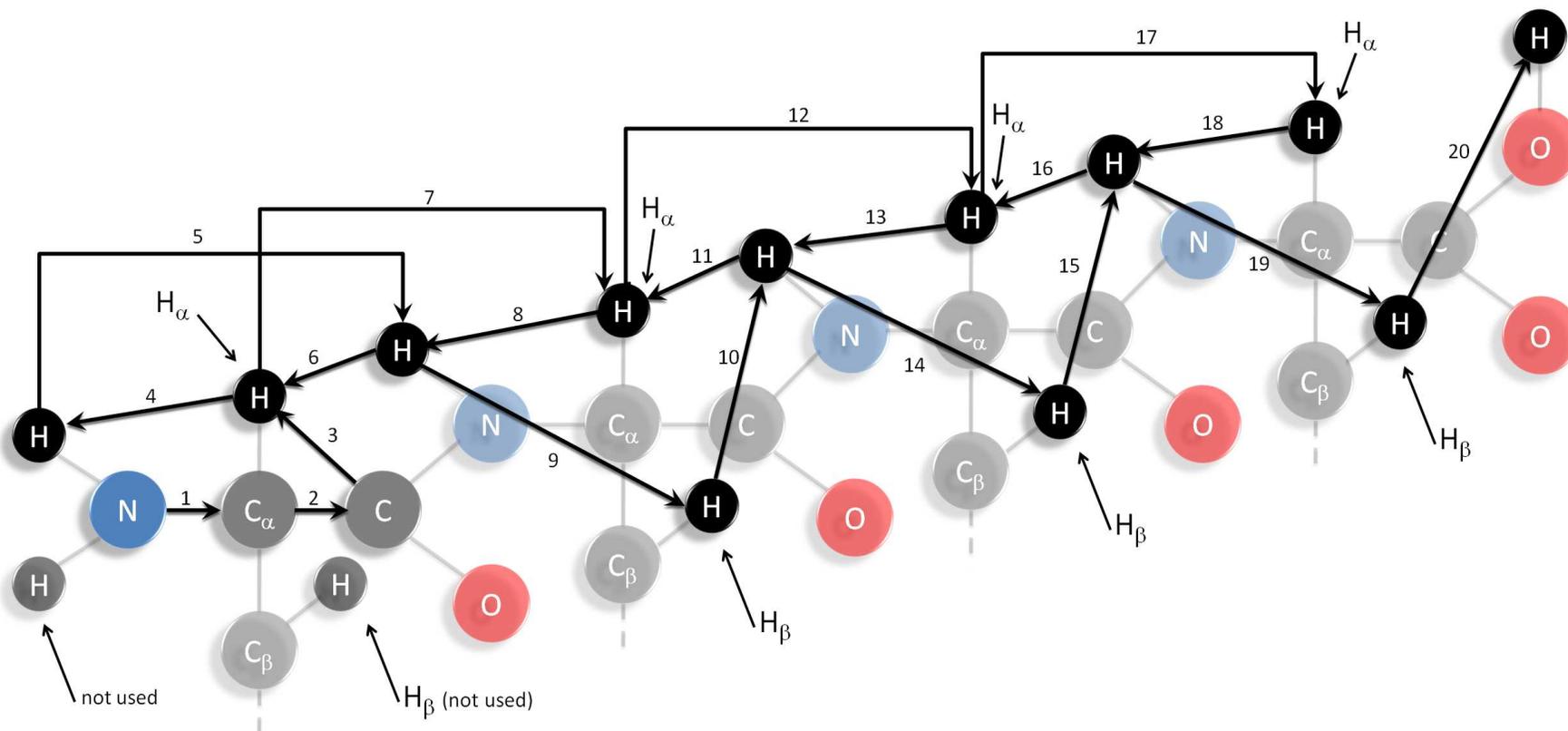
**BP is polynomial on proteins (?)**



# Application to proteomics

# Virtual hydrogen backbone

- The most accurate NMR distances are between hydrogen atoms only, but the actual backbone is a chain of N-C<sub>α</sub>-C groups
- So find a *virtual* backbone composed of hydrogens only, and such that its order satisfies the DMDGP requirements



Certain hydrogens must be enumerated twice



# Listing atoms twice

- If a hydrogen is listed twice, then there are  $i \neq j \in V$  indexing the same atom
- Thus  $x_i = x_j$  and  $d_{ij} = 0$
- For all  $k$  such that  $\{i, k\} \in E$ , we have that  $\{j, k\} \in E$  as  $d_{jk} = d_{ik} + 0$ , and

$$d_{ij} + d_{jk} = 0 + d_{jk} = d_{ik}$$

SO STRICT TRIANGULAR INEQUALITIES do not hold for all atom triplets

- However, it only fails on *nonconsecutive* triplets

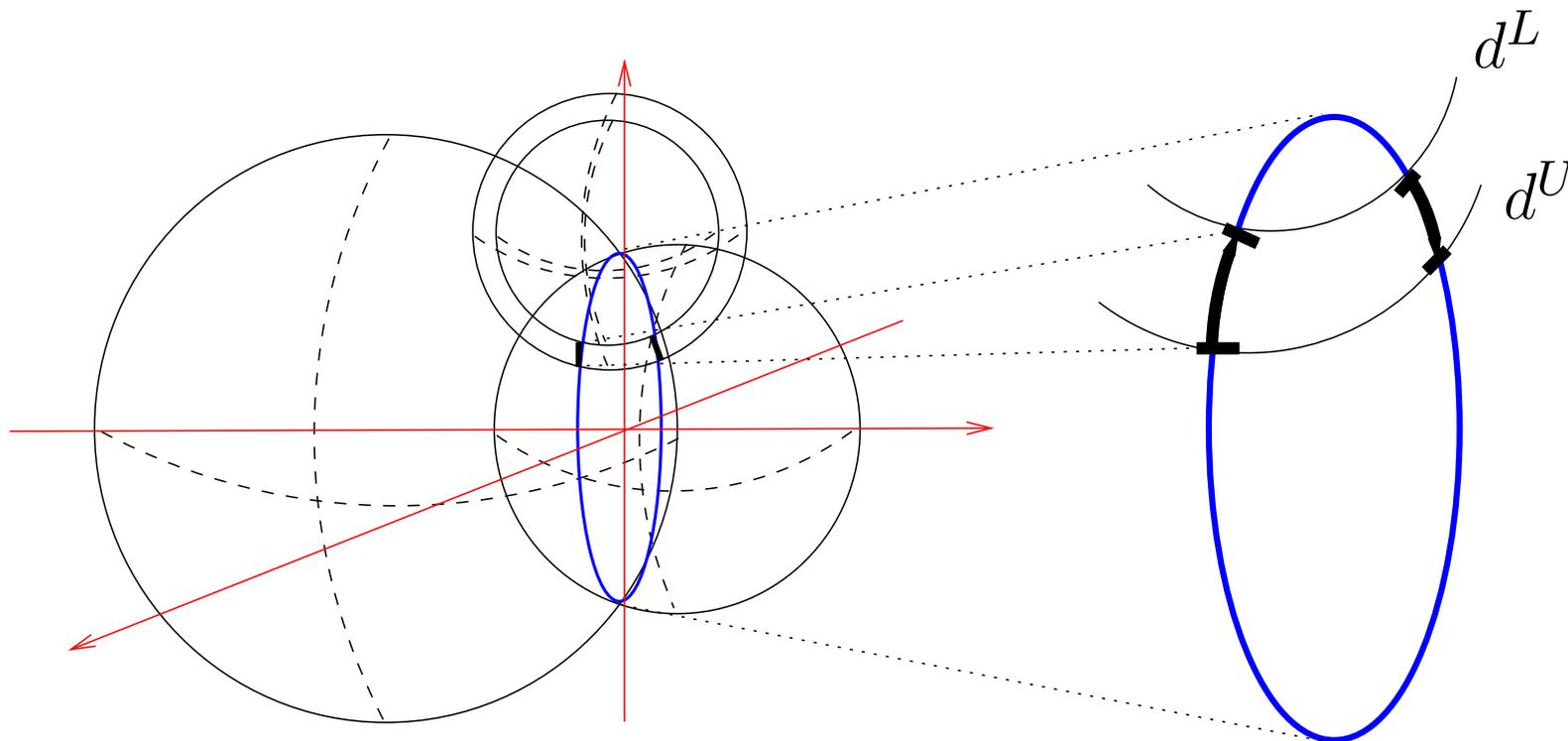
Hence, BP still applies

- Also, zero pruning distances help keeping floating point errors under control



# Uncertain distances

- Typically, NMR provides uncertain distances, modelled by intervals  $[d_{uv}^L, d_{uv}^U]$
- Cannot be used for discretization



Two precise distances and an uncertain one



# The actual situation

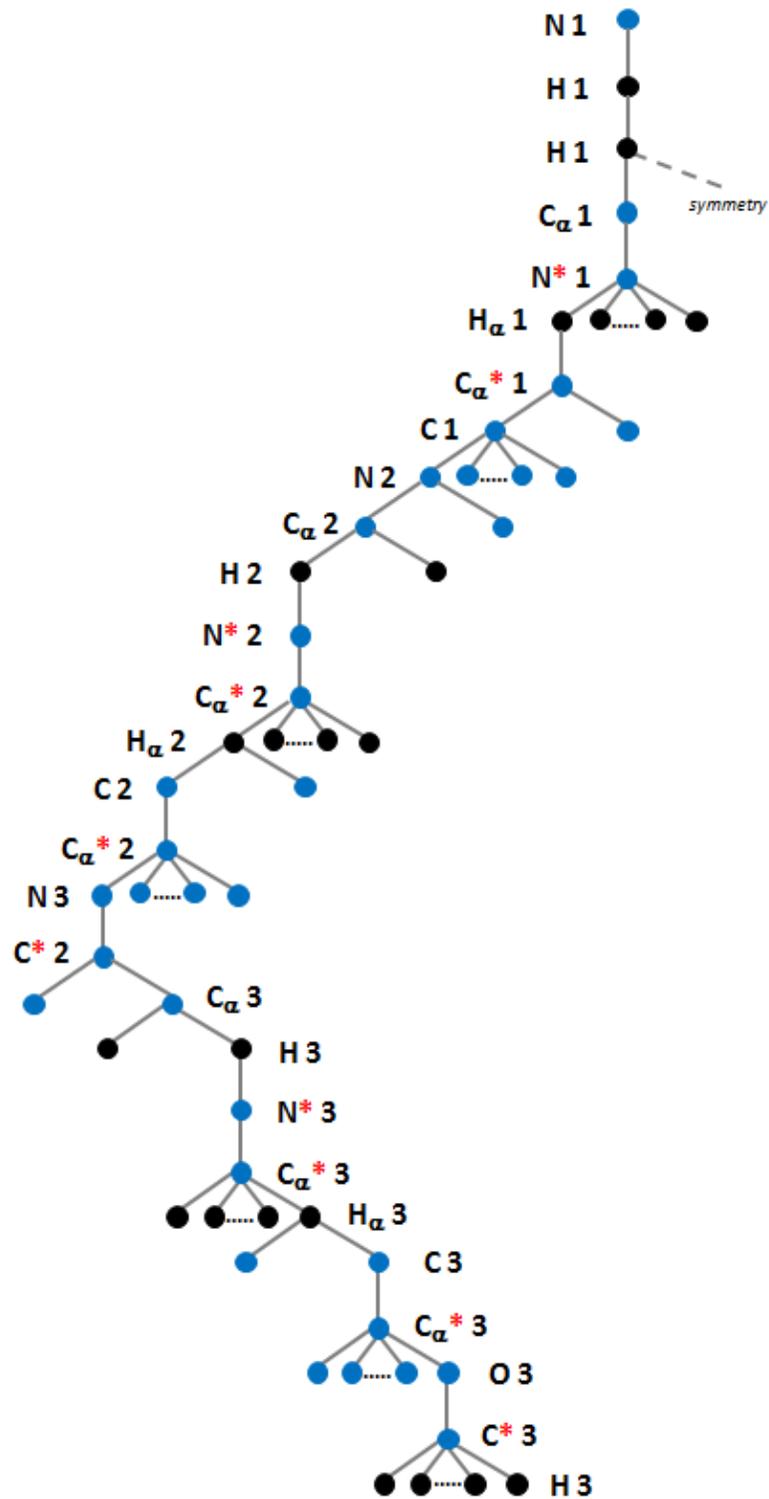
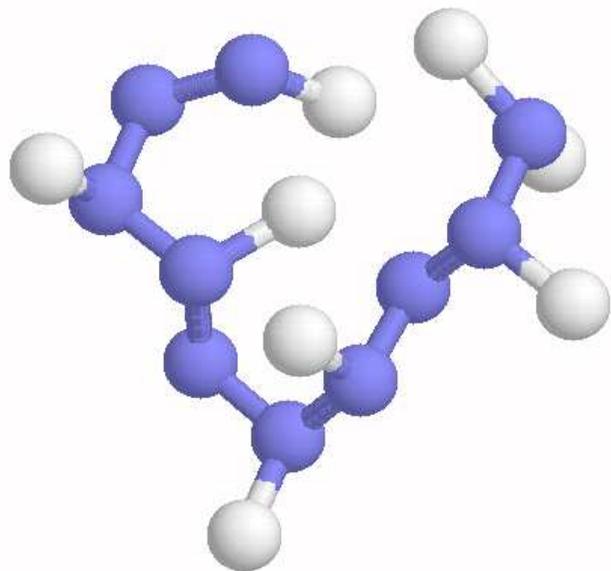
- We know several distances  $d_{uv}$  precisely because of chemical properties
- Some distances take values in a finite set  $D_{uv}$
- The distribution of **precise/discrete/uncertain** distances on the protein backbone does not satisfy the DMDGP requirements

*Re-orders provide a solution: use all **precise** distances for discretization, plus a few of the **discrete** whenever needed; **uncertain** distances are used for pruning*

- Pruning with intervals is easy: if the current point  $x_v$  is s.t.  $\|x_v - x_u\| \in [d_{uv}^L, d_{uv}^U]$  for all  $u \in \alpha(v)$  accept it, otherwise prune it
- Discrete distances  $D_{uv}$  simply give rise to BP nodes at level  $v - 1$  with potentially  $2|D_{uv}|$  subnodes



# iBP





# Implementations



# Sequential code

- The code is available in open source
- Download:  
`http://www.antoniomucherino.it/en/mdjeep.php`
- Any doubt, ask the MASTER (Antonio Mucherino)



# Parallel code

*Seconds of user CPU on Grid5000 ([www.grid5000.fr](http://www.grid5000.fr))*

$ V $	CPUs			
	1	2	8	64
5000	3.21	1.30	0.54	0.36
7500	4.73	3.15	1.25	0.93
10000	13.38	5.49	2.49	1.57

Embed subgraphs then glue embeddings (rigidity  $\Rightarrow$  exact)



# A selection of current work

- Work with biochemists/bioinformaticians at Institut Pasteur to access and treat real NMR data
- Use  $\mathcal{G}_P x = X$  result from symmetry to obtain all solutions from just one
- Extend complexity study to actual problem with discrete/uncertain distances
- Progress on “MDGP  $\in$  NP?” question

See <http://www.lix.polytechnique.fr/~liberti/publications.html> for more papers



# Surveys

- **Survey 1:** Liberti, Lavor, Mucherino, Maculan, *Molecular distance geometry methods: from continuous to discrete*, International Transactions in Operational Research, 18:33-51, 2010
- **Survey 2:** Lavor, Liberti, Maculan, Mucherino, *Recent advances on the discretizable molecular distance geometry problem*, European Journal of Operational Research, 219:698-706, 2012
- **Survey 3:** Liberti, Lavor, Maculan, Mucherino, *Euclidean distance geometry and applications*, SIAM Review, to appear (meanwhile: arXiv 1205.0349v1)



# End of course



# Appendix

# Continuous formulation

- Solving the system

$$\forall \{i, j\} \in E \quad \|x_i - x_j\| = d_{ij}, \quad (10)$$

is numerically challenging

LHS involves  $\sqrt{\arg}$ , floating point ops  $\Rightarrow \arg < 0 \Rightarrow$  error and abort

$\Rightarrow$  square both sides

- Usually, cast as a penalty objective to be minimized

$$\min_x \sum_{\{i,j\} \in E} (\|x_i - x_j\|^2 - d_{ij}^2)^2. \quad (11)$$

- Unconstrained minimization of a polynomial of fourth degree

# General-purpose methods

- sBB (exact): OK on small and medium-sized instances

because we know the optimal value of the objective (0), lower bound is tight at the initial tree levels

- VNS (heur): good for large(ish) instances

- MultiLevel Single Linkage (heur) [Kucherenko et al. '06]: so-so

Atoms	Variables	sBB		VNS		MLSL	
		OF Value	Time	OF Value	Time	OF Value	Time
cube8	24	0	0.22	0	1.21	0	13.56
cube27	81	0	30.39	0	34.01	0	300.285
cube64	192	0	2237.73	0	398.875	0	2765.13
lavor5	15	0	0.02	0	0.48	0	0.57
lavor10	30	0	1.12	0	7.06	0	69.71
lavor20	60	0	2.25	0	49.99	0	411.152
lavor30	90	0	488.87	0	352.06	0	1634.09
lavor40	120	-	-	0.09	1258.13	0.547	2376.01
lavor50	150	-	-	0	673.48	0	3002.88



# MDGP-specific methods



## Smoothing-based:

- Continuation method (heur) [Moré, Wu '97]
- Double VNS with smoothing (heur) [L. et al. '09]
- DC optimization with smoothing (heur) [An et al. '03]
- Hyperbolic smoothing (heur) [Xavier '08]



## Alternating projections algorithm (heur) [Glunt et al. 90]:

*iterative updating of a dissimilarity matrix*



## Geometric build-up (exact/heur) [Dong, Wu '03 and '07]: *triangulation*



## GNOMAD (heur) [Williams et al. '01]

*iterative updating of atomic ordering minimizing error contribution*



## Monotonic Basin Hopping (heur) [Grosso et al. '09]

*funnel-based population heuristic*



## Self-organization heuristic (heur) [Xu et al. '03]

*pairwise atomic position modification heuristic*



## SDP-based formulation [Ye et al. '09]