# The Discretizable Molecular Distance Geometry Problem 

Antonio Mucherino

Laboratorie d'Informatique, Ecole Polytechnique<br>joint work with: L. Liberti, C. Lavor and N. Maculan

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## Outline

1 Introdution
■ The MDGP
■ The Discretizable MDGP
2 The Branch \& Prune algorithm
■ The algorithm

- Computational experiments

3 Open problems and future work
■ Always a power of 2 number of solutions?
■ Accuracy of the instances

The MDGP

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## Introduction to the problem

$\square$ Let $X=\left\{x_{1}, x_{2}, \ldots, x_{n}\right\}$ be a set of points in a three-dimensional space

- Let us suppose that the coordinates of the points $x_{i}$ are unknown
- Iet us sumpose that some of the distances $d_{i j}$ between couples of points $x_{i}$ and $x_{j}$ are known
- Problem: can we find all the coordinates $x_{i}$ from the known distances $d_{i j}$ ??


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## Introduction to the problem

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\left\{d_{i j}: x_{i}, x_{j} \in X\right\} \supseteq D \longrightarrow X=\left\{x_{1}, x_{2}, \ldots, x_{n}\right\} ? ? ?
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If $X$ is a molecular conformation, and each $x_{i}$ represent an atom, this problem is known as

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## The MDGP

Formulation I

A molecular conformation is solution for the MDGP if and only if

$$
\left\|x_{i}-x_{j}\right\|=d\left(x_{i}, x_{j}\right)
$$

This is a constraint satisfation problem

The MDGP

## The MDGP

## Formulation II

Let us define the function

$$
g(X)=\sum_{\{i, j\}}\left(\left\|x_{i}-x_{j}\right\|^{2}-d_{i j}^{2}\right)^{2} .
$$

■ A conformation $X$ is solution for the MDGP if and only if $g(X)=0$

- Solutions can be found by minimizing the function $g(X)$

■ This is an unconstrained global optimization problem

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The MDGP

## The MDGP

## Formulation III

■ Let $G=(V, E, d)$ be a weighted undirected graph, where

- $V$ is the set of indeces $i$ of the atoms in $X$ (the vertices of G)

■ $E$ is the subset of couples of atoms whose distance is known (the edges of G);

- $d$ is the set of known distances (the weights of $G$ )
$\square X$ is a solution for the MDGP if there is a function $x$ such that



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$\square X$ is a solution for the MDGP if there is a function $x$ such that
$\square X=\{x(v): v \in V\}$, where $x: G \rightarrow \mathbb{R}^{3}$
$\square\|x(u)-x(v)\|=d(u, v)$ for each $\{u, v\} \in E$


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## Our approach

## Assumption I

## $E$ contains all cliques on quadruplets of consecutive vertices:

$$
\forall i \in\{4, \ldots, n\} \forall j, k \in\{i-3, \ldots, i\} \quad\{j, k\} \in E
$$



## Our approach

## Assumption II

The following strict triangular inequality holds:

$$
d_{i-1, i+1}<d_{i-1, i}+d_{i, i+1}, \quad i=2, \ldots, n-1
$$



## The DMDGP

If Assumptions I and II hold, each atom can be placed in two possible positions only.


By exploiting the known distances and angles, the cosine of the torsion angle among 4 consecutive atoms can be computed.

## The DMDGP

If Assumptions I and II hold, each atom can be placed in two possible positions only.


The cosine corresponds to two torsion angles $\omega$, having opposite sign.

## The DMDGP

We call Discretizable MDGP (DMDGP) the problem of finding a function

$$
x: G \rightarrow \mathbb{R}^{3}
$$

such that

$$
\|x(u)-x(v)\|=d(u, v) \quad\{u, v\} \in E
$$

where $G=(V, E, d)$ is a weighted undirected graph and Assumptions I and II hold.

## Symmetry

It is possible to prove that:
■ If the DMDGP problem has a solution, then it has at least another symmetric solution
■ Couples of solutions of the DMDGP are symmetric
$\square$
Theorem
Let $x: G \rightarrow \mathbb{R}^{3}$ be a solution for the DMDGP, defined by the torsion angles $\omega_{1,4}, \ldots, \omega_{n-3, n}$. If we invert the sign of $\sin \omega_{i-3, i}$ for $i=4, \ldots, n$, then we obtain a new solution $x^{\prime}: G \rightarrow \mathbb{R}^{3}$ for the DMDGP.

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The Discretizable MDGP

## Symmetry



The two conformations look the same, but they are actually different!

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## Basic idea

- The BP algorithm mimicks the structure of the problem closely: we build a tree of possible choices for the atomic positions.
- At each step, the $i$-th atom can be placed in two possible positions $x_{i}, x_{i}^{\prime}$
- the two positions are both feasible: the search is branched;
- only one position is feasible: the search procedees torwards one direction only;
- both the positions are infeasible: the branch is pruned and the search is backtracked.
- How to check the feasibility of atomic positions? Using pruning tests.


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## Two pruning tests

■ Direct Distance Feasibility (DDF)
■ Known and obtained distances are compared and, if they differ, the new atom position is infeasible
$\square\left\|x_{j}-x_{i}\right\| \neq d_{i j} \Longrightarrow j$ is infeasible
■ Dijkstra Shortest-Paths (DSP)
■ If $D(i, k)<\left\|x_{h}-x_{i}\right\|-d_{h k}$ for all feasible $x: G \rightarrow \mathbb{R}^{3}$, then the BP search node for atomic position $x_{i}$ can be pruned, where

- $h<i<k$

■ $D(i, k)$ is the shortest path between $i$ and $k$ in the graph $G$;

- \| $x_{h}-x_{i} \|$ is the computed distance between $x_{h}$ and $x_{i}$;
- $d_{h k}$ is the known distance between $x_{h}$ and $x_{k}$.
- DDF: simple and efficient;
- DSP: more complex and more efficient.


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## A software procedure

## A software procedure has been developed in C programming language.

Legend of the following Tables:
■ Name - instance name;
■ $n$ - instance dimension;
■ $|E|$ - number of known distances;

- \#Sol - number of found solutions;

■ \#DDF - number of times DDF pruned atoms;
■ \#DSP - number of times DSP pruned atoms;

- CPU - CPU time;

■ LDE - Largest Distance Error:

$$
L D E=\frac{1}{|E|} \sum_{(i, j) \in E} \frac{\left|\left\|x_{i}-x_{j}\right\|-d_{i j}\right|}{d_{i j}}
$$

## Experiments on artificial instances

## Only one pruning test

Experiments have been carried out on a set of artificially generated instances [Lavor, 2006]. Only the pruning test DDF is used here.

| Instance |  |  | BP-One |  |  |  |  | $B P-A / I$ |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Name | $n$ | $\|E\|$ | $C P U$ | $\# D D F$ | $L D E$ | $C P U$ | \#DDF | \#Sol |  |  |
| lavor10 | 10 | 24 | 0.00 | 0 | $1.63 \mathrm{e}-16$ | 0.00 | 0 | 64 |  |  |
| lavor15 | 15 | 70 | 0.00 | 7 | $1.08 \mathrm{e}-09$ | 0.00 | 11 | 1 |  |  |
| lavor20 | 20 | 103 | 0.00 | 9 | $1.28 \mathrm{e}-09$ | 0.00 | 16 | 1 |  |  |
| lavor25 | 25 | 106 | 0.00 | 10 | $1.62 \mathrm{e}-09$ | 0.00 | 49 | 2 |  |  |
| lavor30 | 30 | 219 | 0.00 | 15 | $3.86 \mathrm{e}-09$ | 0.00 | 381 | 2 |  |  |
| lavor35 | 35 | 166 | 0.00 | 13 | $1.22 \mathrm{e}-09$ | 0.00 | 169 | 16 |  |  |
| lavor40 | 40 | 306 | 0.00 | 16 | $5.61 \mathrm{e}-06$ | 0.00 | 136 | 2 |  |  |
| lavor45 | 45 | 351 | 0.00 | 30 | $4.79 \mathrm{e}-09$ | 0.00 | 58 | 1 |  |  |
| lavor50 | 50 | 203 | 0.00 | 49 | $6.50 \mathrm{e}-10$ | 0.07 | 23364 | 512 |  |  |
| lavor55 | 55 | 224 | 0.00 | 26 | $1.43 \mathrm{e}-09$ | 3.45 | 1304580 | 262144 |  |  |
| lavor60 | 60 | 227 | 0.00 | 17 | $2.05 \mathrm{e}-09$ | 0.62 | 262428 | 8192 |  |  |
| lavor65 | 65 | 455 | 0.00 | 1165 | $7.89 \mathrm{e}-09$ | 0.02 | 5184 | 8 |  |  |
| lavor70 | 70 | 331 | 0.00 | 25 | $1.23 \mathrm{e}-08$ | 16.30 | 2798220 | 4194304 |  |  |

Note that the symmetric solutions that can be found by the Theorem on symmetry are not included.

## Experiments on artificial instances

## Two pruning tests

Both the pruning tests DDF and DSP are here used.

| Instance |  | $B P-O n e$ |  |  |  |  |  |  |  | $B P-A / I$ |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Name | $n$ | $\|E\|$ | $C P U$ | \#DDF | \#DSP | $L D E$ | $C P U$ | \#DDF | \#DSP | \#Sol |  |  |
| lavor10 | 10 | 24 | 0.00 | 0 | 0 | $1.63 \mathrm{e}-16$ | 0.00 | 0 | 0 | 64 |  |  |
| lavor15 | 15 | 70 | 0.00 | 7 | 0 | $1.08 \mathrm{e}-09$ | 0.00 | 11 | 0 | 1 |  |  |
| lavor20 | 20 | 103 | 0.00 | 9 | 0 | $1.28 \mathrm{e}-09$ | 0.00 | 16 | 0 | 1 |  |  |
| lavor25 | 25 | 106 | 0.00 | 8 | 1 | $1.62 \mathrm{e}-09$ | 0.00 | 24 | 5 | 2 |  |  |
| lavor30 | 30 | 219 | 0.00 | 15 | 0 | $3.86 \mathrm{e}-09$ | 0.00 | 277 | 10 | 2 |  |  |
| lavor35 | 35 | 166 | 0.00 | 9 | 1 | $1.22 \mathrm{e}-09$ | 0.00 | 145 | 8 | 16 |  |  |
| lavor40 | 40 | 306 | 0.00 | 16 | 0 | $5.61 \mathrm{e}-06$ | 0.00 | 64 | 4 | 2 |  |  |
| lavor45 | 45 | 351 | 0.00 | 17 | 2 | $4.79 \mathrm{e}-09$ | 0.00 | 39 | 3 | 1 |  |  |
| lavor50 | 50 | 203 | 0.00 | 24 | 8 | $6.50 \mathrm{e}-10$ | 0.05 | 5924 | 5648 | 512 |  |  |
| lavor55 | 55 | 224 | 0.00 | 22 | 2 | $1.43 \mathrm{e}-09$ | 7.02 | 1173508 | 65536 | 262144 |  |  |
| lavor60 | 60 | 227 | 0.00 | 13 | 2 | $2.05 \mathrm{e}-09$ | 1.22 | 221462 | 16387 | 8192 |  |  |
| lavor65 | 65 | 455 | 0.00 | 720 | 115 | $7.89 \mathrm{e}-09$ | 0.01 | 3200 | 528 | 8 |  |  |
| lavor70 | 70 | 331 | 0.00 | 25 | 0 | $1.23 \mathrm{e}-08$ | 38.50 | 2732664 | 32774 | 4194304 |  |  |

DDF+DSP are more efficient, because they prune atoms earlier on the search tree, but the computational time usually increases.

## Comparisons to DGSOL

We compared the BP algorithm to DGSOL [Moré and Wu, 1999].

| Instance |  |  | DGSOL |  |
| :---: | :---: | :---: | :---: | :---: |
| Name |  | $n$ | $\|E\|$ | $C P U$ |
| lavor10 | 10 | 24 | 0.03 | $3.01 \mathrm{e}+01$ |
| lavor15 | 15 | 70 | 0.05 | $0.00 \mathrm{e}+00$ |
| lavor20 | 20 | 103 | 0.08 | $0.00 \mathrm{e}+00$ |
| lavor25 | 25 | 106 | 0.24 | $2.62 \mathrm{e}-02$ |
| lavor30 | 30 | 219 | 1.02 | $4.43 \mathrm{e}-07$ |
| lavor35 | 35 | 166 | 1.38 | $1.00 \mathrm{e}-01$ |
| lavor40 | 40 | 306 | 0.57 | $1.94 \mathrm{e}-06$ |
| lavor45 | 45 | 351 | 1.33 | $9.47 \mathrm{e}-07$ |
| lavor50 | 50 | 203 | 1.55 | $7.43 \mathrm{e}-02$ |
| lavor55 | 55 | 224 | 2.06 | $2.31 \mathrm{e}-03$ |
| lavor60 | 60 | 227 | 0.41 | $1.50 \mathrm{e}+03$ |
| lavor65 | 65 | 455 | 2.94 | $1.27 \mathrm{e}-01$ |
| lavor70 | 70 | 331 | 5.10 | $9.60 \mathrm{e}-02$ |

The computational time and quality of the solutions are not comparable with the ones of the BP algorithm.

Computational experiments

## Creating real instances

## PDB file $\Longrightarrow$ MDGP instance

- the three-dimensional conformation of a protein molecule can be downloaded from the PDB
- all the distances between its atoms can be computed
- all the distances smaller than 6A can be used for generating an instance for the MDGP
- this simulates data obtained from NMR experiments, which are able to detect only distances within such a range
- in these experiments, only the backbone atoms $N, C_{\alpha}$ and $C$ are considered


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## Experiments on real instances

| Instance |  |  |  |  | BP-One |  | BP-AII |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Name | $n$ | $\|E\|$ | $C P U$ | $L D E$ | $C P U$ | \#Sol | DGSOL |  |
| 1aqr | 120 | 929 | 0.00 | $1.86 \mathrm{e}-09$ | 0.00 | 2 | 7.77 | $4.88 \mathrm{e}-01$ |
| 2erl | 120 | 1136 | 0.00 | $1.33 \mathrm{e}-14$ | 0.00 | 1 | 9.38 | $2.92 \mathrm{e}-01$ |
| 1crn | 138 | 1250 | 0.00 | $2.24 \mathrm{e}-13$ | 0.00 | 1 | 9.47 | $2.24 \mathrm{e}-01$ |
| 1ahl | 147 | 1205 | 0.00 | $1.50 \mathrm{e}-09$ | 0.00 | 8 | 6.95 | $1.46 \mathrm{e}-01$ |
| 1brz | 159 | 1394 | 0.00 | $3.53 \mathrm{e}-13$ | 0.00 | 2 | 11.39 | $4.66 \mathrm{e}-01$ |
| 1hoe | 222 | 1995 | 0.00 | $3.18 \mathrm{e}-13$ | 0.00 | 1 | 16.83 | $2.06 \mathrm{e}-01$ |
| 1lfb | 232 | 2137 | 0.00 | $5.31 \mathrm{e}-14$ | 0.00 | 1 | 38.94 | $2.88 \mathrm{e}-01$ |
| 1pht | 249 | 2283 | 0.00 | $2.73 \mathrm{e}-12$ | 0.00 | 1 | 42.50 | $2.00 \mathrm{e}-01$ |
| 1jk2 | 270 | 2574 | 0.00 | $2.09 \mathrm{e}-13$ | 0.00 | 1 | 86.98 | $4.05 \mathrm{e}-01$ |
| 1f39a | 303 | 2660 | 0.00 | $2.68 \mathrm{e}-12$ | 0.00 | 1 | 37.24 | $2.80 \mathrm{e}-01$ |
| 1acz | 324 | 3060 | 0.00 | $3.15 \mathrm{e}-12$ | 0.02 | 4 | 35.97 | $3.97 \mathrm{e}-01$ |
| 1poa | 354 | 3193 | 0.00 | $1.36 \mathrm{e}-13$ | 0.00 | 1 | 64.03 | $4.67 \mathrm{e}-01$ |
| 1fs3 | 378 | 3443 | 0.00 | $8.08 \mathrm{e}-13$ | 0.01 | 1 | 54.68 | $2.69 \mathrm{e}-01$ |
| 1mbn | 459 | 4599 | 0.00 | $1.36 \mathrm{e}-09$ | 0.00 | 1 | 124.24 | $4.46 \mathrm{e}-01$ |
| 1rgs | 792 | 7626 | 0.00 | $4.22 \mathrm{e}-13$ | 0.01 | 1 | 237.93 | $4.69 \mathrm{e}-01$ |
| 1m40 | 1224 | 20382 | 0.02 | $1.00 \mathrm{e}-12$ | 5.26 | 1 | 1142.49 | $4.89 \mathrm{e}-01$ |
| 1bpm | 1443 | 14292 | 0.02 | $2.85 \mathrm{e}-13$ | 0.02 | 1 | 398.29 | $5.06 \mathrm{e}-01$ |
| 1n4w | 1610 | 16940 | 0.02 | $1.19 \mathrm{e}-12$ | 0.02 | 1 | 994.51 | $5.26 \mathrm{e}-01$ |
| 1mqq | 2032 | 19564 | 0.02 | $4.90 \mathrm{e}-12$ | 0.06 | 1 | 451.58 | $5.40 \mathrm{e}-01$ |
| 1rwh | 2265 | 21666 | 0.02 | $2.08 \mathrm{e}-13$ | 0.06 | 1 | 934.29 | $5.38 \mathrm{e}-01$ |
| 3b34 | 2790 | 29188 | 0.07 | $1.17 \mathrm{e}-11$ | 0.07 | 1 | 940.95 | $6.47 \mathrm{e}-01$ |
| 2e7z | 2907 | 42098 | 0.08 | $4.26 \mathrm{e}-12$ | 0.09 | 1 | 915.39 | $6.40 \mathrm{e}-01$ |
| 1epw | 3861 | 35028 | 0.16 | $3.19 \mathrm{e}-12$ | 0.25 | 1 | 2037.86 | $4.92 \mathrm{e}-01$ |

## The instance 1aqr

The sequence of signs of the sine values in the 4 solutions of the instance 1aqr.
There are only 2 non-symmetric solutions.
Their only difference stands in the $8^{\text {th }}$ torsion angle.

| from atom 1 to atom 20 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| + | + | + | + | + | + | - | + | + | - | - | + | + | - | + | + | - | + | - |  |
| + | + | + | + | + | + | - | - | + | - | - | + | + | - | + | + | - | + | - | - |
| + | + | + | - | - | - | + | + | - | + | + | - | - | + | - | - | + | - | + | + |
| + | + | + | - | - | - | + | - | - | + | + | - | - | + | - | - | + | - | + | + |
| from atom 21 to atom 40 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| + | - | - | + | - | + | + | - | - | + | - | + | + | + | - | + | + | - | + |  |
| + | - | - | + | - | + | + | - | - | + | - | + | + | + | - | + | + | - | + | + |
| - | + | + | - | + | - | - | + | + | - | + | - | - | - | + | - | - | + | - | - |
| - | + | + | - | + | - | - | + | + | - | + | - | - | - | + | - | - | + | - | - |
| from atom 41 to atom 60 lloll |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| - | + | + | - | + | + | - | + | + | - | + | - | - | + | - | - | + | + | - | + |
| - | + | + | - | + | + | - | + | + | - | + | - | - | + | - | - | + | + | - | + |
| + | - | - | + | - | - | + | - | - | + | - | + | + | - | + | + | - | - | + | - |
| + | - | - | + | - | - | + | - | - | + | - | + | + | - | + | + | - | - | + | - |

Only the first 60 backbone atoms are shown.

Computational experiments

## The instance 1aqr

## All the solutions of the instance 1aqr.



## Outline

1 Introdution

- The MDGP
- The Discretizable MDGP

2 The Branch \& Prune algorithm

- The algorithm
- Computational experiments

3 Open problems and future work
■ Always a power of 2 number of solutions?

- Accuracy of the instances


## An instance with 27 solutions

■ You might have noted that the BP algorithm always found a power of 2 solutions for each considered instance.
■ However, there are instances whose number of solutions is not a power of 2.
■ For example, we have an instance with 27 solutions.
■ Might the power of 2 rule hold only for protein conformations?

Accuracy of the instances

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## What if the accuracy lowers?

■ Our final goal is to use data from NMR for generating protein conformations;
$\square$ data from the NMR are affected by experimental errors:
■ systematic uncertainty on each distance;

- a certain percentage of completely wrong distances.

■ the BP algorithm needs to be improved for managing experimental errors.

## Summary

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- We presented a Branch \& Prune algorithm for DMDGP
- We showed that the Branch \& Prune algorithm is able to solve instances of DMDGP related to protein conformations (artificially generated or real)
- We showed the effeciency and effectiveness of two possible pruning tests
- We showed the open problems on which we are still working


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