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# ARCIMBOLDO_LITE: single-workstation implementation and use 

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ARCIMBOLDO solves the phase problem at resolutions of around $2 \AA$ or better through massive combination of small fragments and density modification. For complex structures, this imposes a need for a powerful grid where calculations can be distributed, but for structures with up to 200 amino acids in the asymmetric unit a single workstation may suffice. The use and performance of the single-workstation implementation, ARCIMBOLDO_LITE, on a pool of test structures with 40-120 amino acids and resolutions between 0.54 and $2.2 \AA$ is described. Inbuilt polyalanine helices and iron cofactors are used as search fragments. $A R C I M B O L D O \_B O R G E S$ can also run on a single workstation to solve structures in this test set using precomputed libraries of local folds. The results of this study have been incorporated into an automated, resolution- and hardware-dependent parameterization. $A R C I M B O L D O$ has been thoroughly rewritten and three binaries are now available: ARCIMBOLDO_LITE, ARCIMBOLDO_SHREDDER and ARCIMBOLDO_BORGES. The programs and libraries can be downloaded from http://chango.ibmb.csic.es/ ARCIMBOLDO_LITE.

## 1. Introduction

The phase problem is the central problem of crystallography and is a bottleneck in the determination of macromolecular crystal structures (Hendrickson, 2013). Several methods exist to provide starting phases for structure factors to approximate the correct phases, which are missed in the diffraction experiment. In the case of macromolecules, experimental phasing through heavy-atom derivatives or anomalous scattering at particular wavelengths (Hendrickson, 1991) is used if no previous structural knowledge from a related structure is available for phasing by molecular replacement (Rossman, 1972; Navaza, 1994).
$A b$ initio phasing from the native diffraction intensities alone is accomplished by direct methods in the case of molecules with less than 200 atoms that diffract to atomic resolution (Karle \& Hauptman, 1956). Dual-space recycling has succeeded in extending the use of direct methods to small proteins of up to 1000 atoms, maintaining the atomicity restriction (Miller et al., 1993; Sheldrick et al., 2012). Approaches to relax this restriction have involved sophisticated use of the Patterson function (Caliandro et al., 2008), data extrapolation to extend the resolution limit beyond the experimentally measured data (Caliandro et al., 2005; Usón et al., 2007) and density-modification algorithms tailored to the high-resolution case, such as low-density elimination (Shiono \& Woolfson, 1992; Refaat \& Woolfson, 1993), the sphere of influence (Sheldrick, 2002) and VLD (Burla et al., 2010, 2011,
2012). Starting phasing from a small but highly accurate substructure in the context of $A C O R N$ has been reported to be remarkably effective (Yao et al., 2005 , 2006) and our own tests have corroborated this finding: as little as $10 \%$ of the main-chain atoms suffice to solve a structure at $2 \AA$ resolution through density modification. This provides a way to escape the need for atomic resolution that restricts direct methods and real-space atom selection. The atomicity constraints can be substituted by enforcing secondary- or tertiary-structure stereochemistry. A related proof of principle has been established using small fragments such as $\alpha$-helices (Glykos \& Kokkinidis, 2003) and nucleotides (Robertson \& Scott, 2008; Robertson et al., 2010) as search fragments to seed phasing. This is accomplished in ARCIMBOLDO (Rodríguez et al., 2009, 2012) by combining a search for small polyalanine model fragments with Phaser (McCoy et al., 2007) with density modification and autotracing with SHELXE (Thorn \& Sheldrick, 2013). Extremely successful approaches based on more complete models of lower accuracy (Rigden et al., 2008) have been developed and include the enhancement of poor search models and de novo prediction through modelling with Rosetta (Qian et al., 2007) or $\operatorname{QUARK}$ (Xu \& Zhang, 2012) linked to molecular replacement with Phaser. This design underlies methods such as MR-Rosetta (DiMaio et al., 2011) and AMPLE (Bibby et al., 2012, 2013; Keegan et al., 2015), and other implementations (Shrestha et al., 2011; Shrestha \& Zhang, 2015).

The original ARCIMBOLDO procedure relies on the assumption that the main chain of small secondary-structural elements, such as $\alpha$-helices, is conserved among unrelated protein structures. Because of the generality of such partial models, which are predictable from the sequence (Cole et al., 2008) or data (Caliandro et al., 2012; Morris et al., 2004), this method is in fact an ab initio approach. Recently, it has been shown how unrelated structures can share sufficiently similar local folds comprising the main chain of a few secondarystructure elements. Libraries of such folds can be extracted with BORGES (Sammito et al., 2013) from the entire PDB (Bernstein et al., 1977; Berman et al., 2000) and then input into $A R C I M B O L D O_{-} B O R G E S$ as search models. If a the structure of a distant homologue is known, fragments from this template can also be extracted and evaluated against the rotation function, as implemented in ARCIMBOLDO_ SHREDDER (Sammito et al., 2014).

As small models account for a low fraction of the total scattering, the figures of merit characterizing a correct solution may fail to identify it as such, making it indistinguishable from a large pool of incorrect solutions. This has required the implementation of $A R C I M B O L D O$ as a multisolution method, in which all possible solutions have to be explored

Table 1
Performance of ARCIMBOLDO_LITE on a pool of 12 workstations for two test structures (PDB entries 2iu1 and 4k82).

| Hardware | OS | RT, 2iu1 | RT, 4k82 |
| :---: | :---: | :---: | :---: |
| Mean for pool of 12 workstations |  | 7 h 37 min | 2 h 1 min |
| Xeon X5560 at $2.80 \mathrm{GHz}, 8$ cores sharing 24 GB | Debian 6.0 | 4 h 26 min | 1 h 18 min |
| i7 2600 at $3.40 \mathrm{GHz}, 4$ cores sharing 12 GB | Debian 6.0 | 5 h 34 min | 1 h 36 min |
| i7 3930 K at $3.20 \mathrm{GHz}, 6$ cores sharing 16 GB | Debian 6.0 | 4 h 10 min | 1 h 21 min |
| Xeon E5410 at $2.33 \mathrm{GHz}, 8$ cores sharing 32 GB | Debian 6.0 | 7 h 46 min | 1 h 34 min |
| 2 Duo E4500 at $2.20 \mathrm{GHz}, 2$ cores sharing 512 MB | Debian 6.0 | 32 h 19 min | 8 h 45 min |
| i7 950 at $3.07 \mathrm{GHz}, 4$ cores sharing 6 GB | Ubuntu 10.04.2 | 6 h 41 min | 2 h 3 min |
| Xeon E5440 at $2.83 \mathrm{GHz}, 8$ cores sharing 16 GB | Ubuntu 10.04.3 | 6 h 58 min | 1 h 56 min |
| i7 X980 at $3.33 \mathrm{GHz}, 6$ cores sharing 12 GB | Ubuntu 10.04.4 | 5 h 1 min | 1 h 21 min |
| Xeon E5-2650 v2 at $2.66-4.00 \mathrm{GHz}, 16$ cores sharing 125 GB | Centos 7.1.1503 | 1 h 55 min | 32 min |
| Xeon E5-2680 v3 at $2.50 \mathrm{GHz}, 24$ cores sharing 16 GB | openSUSE 13.2 | 4 h 34 min | 1 h |
| I7-3720QM at $2.60 \mathrm{GHz}, 8$ cores sharing 8 GB | Mac OSX Yosemite 10.10 | 3 h 16 min | 1 h 10 min |
| I7-3720QM at $2.80 \mathrm{GHz}, 8$ cores sharing 4 GB | Mac OSX Mavericks 10.9 | 8 h 54 min | 1 h 40 min |

independently up to the step of density modification and autotracing, where a correct density map is unequivocally recognizable at $2 \AA$ resolution. Supercomputing facilities are required to manage, distribute and check this magnitude of computation, and databases to aid in storing and retrieving the amount of data generated. Improvement in the identification of correct solutions allows the pool of solutions to be narrowed in some cases (Oeffner et al., 2013). The accumulated experience in using this method and advances in the software involved (Phaser and SHELXE) have made it possible to deploy a standalone version which will just require multiprocessing facilities and particular filters to reduce computation depending on the characteristics of the available hardware.

In the present work, we have developed and assessed singlemachine implementations of the $A R C I M B O L D O$ methods described by Millán et al. (2015). ARCIMBOLDO_LITE performs ab initio phasing using polyalanine helices or other single search fragments. $A R C I M B O L D O \_B O R G E S$ performs $a b$ initio phasing with nonspecific libraries of small folds. For the sake of completeness, although not discussed in this paper, ARCIMBOLDO_SHREDDER uses fragments derived from a distant homologue template. The new software has been thoroughly rewritten and dimensioned to run on a single, multiprocessor machine, thus eliminating the limiting condition of access to a supercomputer or grid cluster. No database is required either. The binary standalone version is distributed for Linux and Mac OS at http://chango.ibmb.csic.es/ ARCIMBOLDO_LITE. The performance of $A R C I M B O L D O_{-}$ LITE and ARCIMBOLDO_BORGES on a pool of test structures and systems is described and the conclusions of these tests have been used to set up data-dependent default parameterization.

## 2. Materials and methods

### 2.1. Computing setup

Structure solution was run on the eight identical eight-core machines of an HP ProLiant BL460c blade system, using them as single workstations with dual quad-core Xeon E5440
processors (2.83 GHz and 16 GB RAM). The Linux distribution installed was Ubuntu Server 10.04 LTS.

### 2.2. Software versions

The $A R C I M B O L D O$ binary is deployed for Linux and Macintosh (some examples of compatible distributions and versions can be found in Table 1). It was generated with PyInstaller 2.1 and Python 2.6, with version 2014 upwards of SHELXE and version 2.5.6 upwards of Phaser from the PHENIX (Adams et al., 2010) or CCP4 (Winn et al., 2011) distributions. Model and maps were examined with Coot (Emsley et al., 2010). Figures were prepared with PyMOL (v.1.5.0.4; Schrödinger) and gnuplot 4.4. F-weighted mean phase errors were calculated with SHELXE (Sheldrick, 2002) against the final models deposited in the PDB (Bernstein et al., 1977; Berman et al., 2000).

### 2.3. Test set used

In this study, the structure of EIF5 (Bieniossek et al., 2006; PDB entry 2 iu 1 ) at $1.7 \AA$ resolution and the structure of Lv-ranaspumin (Hissa et al., 2014; PDB entry 4 k 82 ) at $1.7 \AA$ resolution previously solved with $A R C I M B O L D O$ were used to benchmark performance on different hardware setups. Furthermore, we used a pool of 294 structures with resolutions between 2.2 and $0.54 \AA$, sizes of between 44 and 120 residues in the asymmetric unit and 47 different space groups, approximately following the frequency distribution seen in the PDB. In particular, the most frequent 36 space groups are represented, with $P 2_{1} 2_{1} 2_{1}$ predominating. The PDB codes for these 294 structures are as follows: 1ejg, 1en2, 1ew4, 1ezj, 1f94, $1 \mathrm{fk} 5,1 \mathrm{~g} 2 \mathrm{r}, 1 \mathrm{~g} 6 \mathrm{u}, 1 \mathrm{gk} 6,1 \mathrm{gmx}, 1 \mathrm{gvd}, 1 \mathrm{gxu}, 1 \mathrm{i} 2 \mathrm{t}, 1 \mathrm{i} 71,1 \mathrm{iqz}, 1 \mathrm{j} 2 \mathrm{l}$, $1 \mathrm{j} 8 \mathrm{~b}, 1 \mathrm{j} 8 \mathrm{e}, 1 \mathrm{kth}, 1 \mathrm{kwi}, 1191,11 \mathrm{sl}, 11 \mathrm{xj}, 1 \mathrm{~m} 1 \mathrm{q}, 1 \mathrm{mg} 4,1 \mathrm{mk} 0$, 1ne8, $1 \mathrm{njh}, 1 \mathrm{nnx}, 1 \mathrm{ap}, 1 \mathrm{oks}, 1 \mathrm{x} 3,1 \mathrm{p} 9 \mathrm{~g}, 1 \mathrm{pz} 4,1 \mathrm{q} 8 \mathrm{~d}, 1 \mathrm{r} 6 \mathrm{j}, 1 \mathrm{r} 7 \mathrm{j}$, 1riy, 1rlk, 1rw1, 1rwj, 1sbx, 1t07, 1tg0, 1tgr, 1ts9, 1ttz, 1tuk, 1tuw, 1u2h, 1u3y, 1u84, 1u9p, 1ub9, 1ucs, 1uj8, 1uoy, 1use, 1usm, $1 \mathrm{v} 05,1 \mathrm{v} 2 \mathrm{z}, 1 \mathrm{v} 70,1 \mathrm{vbw}, 1 \mathrm{vjk}, 1 \mathrm{vyi}, 1 \mathrm{whz}, 1 \mathrm{wpa}, 1 \mathrm{wri}, 1 \mathrm{xak}, 1 \mathrm{xbi}$, 1xe1, 1xg8, 1xw3, 1y0n, 1y6x, 1y91, 1ygt, 1yib, 1yu5, 1yzm, 1z0p, 1z21, 1z96, 1zld, 1zt3, 1zva, 1zzk, 2asc, 2b1y, 2b8i, 2bkf, 2brf, $2 \mathrm{c} 60,2 \mathrm{cbo}, 2 \mathrm{cg} 7,2 \mathrm{cmp}, 2 \mathrm{cwr}, 2 \mathrm{cwy}, 2 \mathrm{cyj}, 2 \mathrm{~d} 3 \mathrm{~d}, 2 \mathrm{~d} 9 \mathrm{r}, 2 \mathrm{e} 3 \mathrm{~h}$, 2ea9, 2efv, 2es9, 2ewh, 2ewk, 2ewt, 2f60, 2fb6, 2fht, 2fi0, 2fq3, $2 \mathrm{fu} 2,2 \mathrm{~g} 7 \mathrm{o}, 2 \mathrm{gkr}, 2 \mathrm{gpi}, 2 \mathrm{gyz}, 2 \mathrm{~h} 8 \mathrm{e}, 2 \mathrm{~h} 9 \mathrm{u}, 2 \mathrm{haz}, 2 \mathrm{hc} 8,2 \mathrm{hdz}, 2 \mathrm{hl} 7$, 2hlr, 2hpj, 2i4a, 2i5f, 2i6v, 2iay, 2igp, 2ip6, 2ivy, 2j6b, 2j8b, 2j97, $2 \mathrm{jku}, 2 \mathrm{nml}$, 2nqw, 2ns0, 2nsc, 2nuh, 2o0q, 2o1k, 2o37, 2o4t, $2 \mathrm{o} 9 \mathrm{u}, 2 \mathrm{od} 5,2 \mathrm{ooa}, 2 \mathrm{qqk}, 2 \mathrm{qqq}, 2 \mathrm{ouf}, 2 \mathrm{ovg}, 2 \mathrm{oxo}, 2 \mathrm{p} 5 \mathrm{k}, 2 \mathrm{p} 6 \mathrm{v}$, $2 \mathrm{pk} 8,2 \mathrm{pnd}, 2 \mathrm{ppn}, 2 \mathrm{pst}, 2 \mathrm{py} 0,2 \mathrm{q} 2 \mathrm{f}, 2 \mathrm{q} 79,2 \mathrm{qff}, 2 \mathrm{qmt}, 2 \mathrm{qsb}$, 2qsk, 2qtd, 2qvo, 2qyw, 2r39, 2r4q, 2rff, 2rh2, 2rhf, 2ril, 2uux, $2 \mathrm{v} 75,2 \mathrm{vc} 8,2 \mathrm{vkl}, 2 \mathrm{vq} 4,2 \mathrm{vsd}, 2 \mathrm{wbx}, 2 \mathrm{wj} 5,2 \mathrm{wkd}, 2 \mathrm{wuj}, 2 \mathrm{x} 3 \mathrm{~g}$, $2 x f d, 2 x n q, 2 y v 4,2 y v i, 2 y x f, 2 y z t, 2 z q e, 2 z q m, 2 z x y, 3 a 0 s, 3 a 38$, 3a4c, 3adg, 3agn, 3b64, 3bjo, 3bn0, 3bn7, 3bri, 3bt4, 3bv8, 3c0f, 3 ca 7 , 3ce7, 3cec, 3ci9, 3cq1, 3ctr, 3cw3, 3df8, 3dml, 3dqy, 3e0e, 3e21, 3e56, 3e9v, 3eaz, 3efg, 3emi, 3enu, 3ewg, 3exy, 3f14, 3f2e, $3 f 40,3 \mathrm{fb} 1$, 3fdr, 3ff2, 3ff5, 3ffy, 3fkc, 3fmy, 3ft7, 3g21, 3g2b, 3 ghf , 3 goe, 3 gv 3 , $3 \mathrm{~h} 01,3 \mathrm{~h} 36$, 3h8h, 3h9w, 3hgl, 3hms, 3 hnx , $3 \mathrm{hqx}, 3 \mathrm{hrl}$, 3hro, 3hz7, 3i8z, 3idw, 3im3, 3iv4, 3jsc, 3jsr, 3jtz, 3ju3, 3jvl, 3k0x, 3k3v, 3kkf, 3kp8, 3kw6, 3kzd, 3132, 314h, 319a,

3lax, 3lbj, 3ldc, 3le4, 3lwc, 3lyg, 3lyw, 3msh, 3mwz, 3mxz, 3n3f, $3 \mathrm{nbm}, 3 \mathrm{npd}$, 3nrw, 3nx6, 3nzl, 3oiz, 3oou, 3osh and 3phn.

The program described has also been tested on a pool of 15 structures (Millán et al., 2015) first solved with previous ARCIMBOLDO implementations.

### 2.4. Libraries of local folds

For the $A R C I M B O L D O_{-} B O R G E S$ tests, six precomputed libraries containing geometrically clustered variations of a particular fold were used. Two of them contain pairs of contiguous helices: parallel and antiparallel, respectively. Three correspond to 20 amino acids arranged in threestranded $\beta$-sheets in antiparallel, parallel and parallelantiparallel dispositions. The last contains a sample of disulfide bridges linking two tetrapeptides in all possible conformations. These libraries of models superimposed to match the generating template were computed with BORGES (Sammito et al., 2013) and are available for download from our website, along with libraries of 24 amino acids arranged in four strands with different relative arrangements (http://chango.ibmb.csic.es/ download). A library with four antiparallel $\beta$-strands has been used redundantly on six cases previously solved with the described library of three-stranded antiparallel sheets to confirm its validity. This and other new libraries will be incorporated into our public repository. Besides the libraries used in this study, a library of DNA-binding motifs is also downloadable (Pröpper et al., 2014).

## 3. Results and discussion

### 3.1. Implementation

ARCIMBOLDO_LITE, ARCIMBOLDO_SHREDDER and ARCIMBOLDO_BORGES have been thoroughly rewritten in the course of the last year. The $A R C I M B O L D O_{-}$ LITE implementation is deployed as a binary file and has been designed to run on a single multicore machine. The need for a MySQL database present in previous implementations has been removed, as it is intended to solve easier target structures ranging from 90 to 200 amino acids with complete diffraction data reaching $2.0 \AA$ resolution or beyond. Moreover, particular stress has been placed on simplifying the input, reducing it to minimal requirements and producing helpful output, in particular direct troubleshooting instructions and clear identification of solution files. The program provides default values for all parameters except for the basic specification of experimental data: paths to the reflection data files in .hkl (Sheldrick, 2008) and .mtz (Winn et al., 2011) format have to be input, as well as the LABELS identifying the data in the .mtz file and the contents of the asymmetric unit (molecular weight and number of copies). Paths to Phaser and SHELXE have to be set unless they are already defined for the user within the system. The default search is exemplified as two copies of a model helix composed of 14 alanines, but the user should select the most appropriate hypothesis according to secondary-structure prediction or previous knowledge about the target structure. For model polyalanine helices, an inbuilt
library is provided and only the length needs to be specified. Model fragments for some cofactors and ligands with high scattering power are also provided internally, in particular for iron-sulfur clusters and the haem group. Information on defaults and usage for all mandatory and optional parameters can be obtained by typing ARCIMBOLDO_LITE -b in the terminal window.

Fig. 1 displays a diagram summarizing the flow. The program starts by reading and validating the instruction file. It will substitute default values by user-input choices while checking for formal correctness and an acceptable parameter range. Next, checks are performed to customize the job, adapting it to the experimental data and available hardware, as well as to prevent common errors that cause run termination at a more advanced stage.
(i) The number of physical cores in the computer is read from /proc/cpuinfo (in Mac OS this information is extracted with sysctl -n hw.ncpu).
(ii) External search models are confirmed to be accessible files in valid PDB format containing at least one atom. Format problems are corrected if possible.
(iii) Paths to files and work directories are verified in terms of existence, accessibility and user permissions.
(iv) If a sequence is provided to compute the molecular weight, its validity is checked.
(v) Python, Phaser and SHELXE versions are checked for compatibility with the deployed binary.
(vi) A Phaser job is launched to validate configuration and data compatibility while generating a corrected data file with anisotropically scaled amplitudes for use in subsequent jobs.
(vii) A $S H E L X E$ job is launched to test the input line and data format, identifying amplitudes or intensities.
(viii) The space group, if specified, is tested to be a Sohncke space group in a standard setting.

Failure of any of these tests will cause the program to stop, with an error message printed to the standard output indicating the cause. Furthermore, if the resolution of the experimental data does not reach $2 \AA$ a warning will be issued, while below $2.5 \AA$ resolution the program will terminate.

Results are output in xml and html format and are constantly updated for the user to monitor the program flow. The html file (Fig. 2) echoes all of the parameters used in the run and a table of results and figures of merit characterizing final and partial solutions. The table can be sorted by any of its entries. The html page provides links to the files containing the best map and trace.

ARCIMBOLDO_LITE runs in two macrocycles. During the first one Phaser sequentially places all requested copies of the search model(s) specified through the configuration file and partial structures are ranked and possibly optimized by trimming against the CC (Fujinaga \& Read, 1987) with SHELXE (Sheldrick \& Gould, 1995). In the second macrocycle, $A R C I M B O L D O$ selects from all of the partial solutions produced a number of solutions one less than the number of physical cores in the machine to be expanded through density modification and autotracing with SHELXE. To render calculations manageable, hard limits are imposed on the
maximum number of solutions after each step. These limits are set in relation to the number of cores available. Therefore, a more powerful machine will by default pursue more partial solutions and may solve a structure that a more limited machine would fail to phase unless a longer job is customized. By default, negative LLG values are discarded and, even though this parameter can be overridden through the .bor file, the alternative of revising the expected asymmetric unit composition and/or data resolution appears to be preferable.

Phaser jobs are executed as independent, monitored processes and input is given through keyword files for compatibility with all environments and middleware. Firstly, a rotation search (Storoni et al., 2004) is computed for the input model and solutions within $75 \%$ of the top LLG are selected. These are clustered within an angular difference of $15^{\circ}$, taking symmetry into account, after converting the Euler angles to quaternions or rotation matrices. If, exceptionally, the number of rotation peaks exceeds 10000 , numerical $k$-means clustering is used as a filter prior to geometrical clustering. To save time, an ancillary Cantor pairing value is used. The rotation characterized by the highest LLG is used as a reference to determine similar rotations. The process is iterated over the nonclustered rotations until the list has been scanned.

Translation performance (McCoy et al., 2005) is evaluated jointly for each cluster of rotation solutions, limiting the maximum number of solutions to be sent to packing and refinement. In this way, variability is preserved if a given fragment can be placed in several parts of a structure. For the placed fragments, an initial CC is computed with SHELXE, optionally eliminating residues to increase this figure of merit. PDB optimization (Sheldrick \& Gould, 1995) is usually not


Figure 1
ARCIMBOLDO_LITE flow scheme. ARCIMBOLDO operations are labelled in red. Input data and program settings are validated and the run is configured with parameterization defaults depending on data resolution and limiting calculations by the number of cores in the workstation. In a first macrocycle all copies of the search fragments are placed with Phaser and optimized with SHELXE. Similar solutions are clustered. A number of substructure solutions equal to the number of physical cores minus one is subject to density modification and autotracing with SHELXE, prioritizing according to top LLG and CC.

SPACEGROUP: P 1211
CELL DIMENSIONS: $23.26,60.44,23.59,90.00,90.99,90.00$
RESOLUTION: 0.92
NUMBER OF UNIQUE REFLECTIONS: 35634.00
2. STEP: Locating Sequentially the Fragments

| Fragment 1 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Cluster | Rotation Function |  |  |  |  | Translation Function |  |  |  |  | Packing |  |  |  |  | Rigid Body Refinement |  |  | Initial CC <br> After Refinement <br> CC | Best Trace CC/aa |  |  |
|  | \#Rots. | $\begin{aligned} & \hline \text { Top } \\ & \text { LLG } \\ & \hline \end{aligned}$ | $\begin{aligned} & \hline \text { Mean } \\ & \text { LLG } \\ & \hline \end{aligned}$ | $\begin{gathered} \text { Top } \\ \text { Zscore } \\ \hline \end{gathered}$ | Mean Zscore | \#Trans. | $\begin{gathered} \hline \text { Top } \\ \text { LLG } \end{gathered}$ | Mean <br> LLG | $\begin{gathered} \hline \text { Top } \\ \text { Zscore } \end{gathered}$ | $\begin{aligned} & \text { Mean } \\ & \text { Zscor } \end{aligned}$ | \#Sol. | $\begin{gathered} \hline \text { Top } \\ \text { LLG } \\ \hline \end{gathered}$ | $\begin{aligned} & \hline \text { Mean } \\ & \text { LLG } \end{aligned}$ | $\begin{gathered} \text { Top } \\ \text { Zscore } \end{gathered}$ | Mean Zscore | \#Sol. | $\begin{aligned} & \hline \text { Top } \\ & \text { LLG } \\ & \hline \end{aligned}$ | Mean LLG |  |  | CC | \#Res. traced |
| 0 | 20 | 18.99 | 14.21 | 4.62 | 3.73 | 2 | 35.51 | 32.35 | 6.90 | 5.71 | 2 | 35.51 | 32.35 | 6.90 | 5.71 | 2 | 39.20 | 37.75 | 7.56 |  |  |  |
| 1 | 1 | 16.68 | 16.68 | 4.19 | 4.19 | 4 | 25.36 | 23.54 | 4.01 | 3.50 | 4 | 25.36 | 23.54 | 4.01 | 3.50 | 4 | 34.80 | 28.17 | 7.08 |  |  |  |
| 2 | 3 | 16.67 | 14.46 | 4.19 | 3.78 | 7 | 22.00 | 21.12 | 3.62 | 3.00 | 7 | 22.00 | 21.12 | 3.62 | 3.00 | 7 | 24.00 | 22.34 | 6.20 |  |  |  |
| 3 | 1 | 16.33 | 16.33 | 4.12 | 4.12 | 4 | 22.49 | 21.67 | 3.46 | 3.18 | 4 | 22.49 | 21.67 | 3.46 | 3.18 | 4 | 24.60 | 23.20 | 6.56 |  |  |  |
| 4 | 3 | 13.79 | 13.54 | 3.65 | 3.61 | 2 | 26.09 | 25.99 | 5.18 | 5.15 | 2 | 26.09 | 25.99 | 5.18 | 5.15 | 2 | 28.90 | 28.80 | 3.84 |  |  |  |
| 5 | 1 | 13.71 | 13.71 | 3.64 | 3.64 | 1 | 21.86 | 21.86 | 4.04 | 4.04 | 1 | 21.86 | 21.86 | 4.04 | 4.04 | 1 | 22.90 | 22.90 | 7.59 |  |  |  |
| Fragment 2 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| Cluster | Rotation Function |  |  |  |  | Translation Function |  |  |  |  | Packing |  |  |  |  | Rigid Body Refinement |  |  | Initial CC | Best Trace CC/aa |  |  |
|  | \#Rots. | $\begin{gathered} \text { Top } \\ \text { LLG } \\ \hline \end{gathered}$ | $\begin{aligned} & \text { Mean } \\ & \text { LLG } \\ & \hline \end{aligned}$ | $\begin{gathered} \text { Top } \\ \text { Zscore } \end{gathered}$ | Mean <br> Zscore | \#Trans. | $\begin{aligned} & \text { Top } \\ & \text { LLG } \\ & \hline \end{aligned}$ | Mean <br> LLG | $\begin{gathered} \text { Top } \\ \text { Zscore } \end{gathered}$ | Mean <br> Zscore | \#Sol. | $\begin{aligned} & \text { Top } \\ & \text { LLG } \\ & \hline \end{aligned}$ | Mean <br> LLG | $\begin{gathered} \text { Top } \\ \text { Zscore } \end{gathered}$ | $\begin{aligned} & \text { Mean } \\ & \text { Zscore } \end{aligned}$ | \#Sol. | $\begin{aligned} & \text { Top } \\ & \text { LLG } \\ & \hline \end{aligned}$ | $\begin{aligned} & \hline \text { Mean } \\ & \text { LLG } \\ & \hline \end{aligned}$ | After Refinement CC | Cycle | CC | \#Res. traced |
| (0,6) | 11 | 56.31 | 50.27 | 4.43 | 3.58 | 7 | 108.16 | 93.75 | 10.25 | 9.18 | 7 | 108.16 | 93.75 | 10.25 | 9.18 | 4 | 124.90 | 118.85 | 13.38 |  |  |  |
| (0,9) | 33 | 52.94 | 49.95 | 3.79 | 3.51 | 21 | 117.06 | 98.86 | 12.30 | 10.05 | 21 | 117.06 | 98.86 | 12.30 | 10.05 | 14 | 125.00 | 106.64 | 12.16 |  |  |  |
| (1,9) | 18 | 48.93 | 40.73 | 3.85 | 3.73 | 92 | 99.10 | 52.27 | 10.29 | 5.14 | 3 | 99.10 | 91.80 | 10.29 | 9.31 | 3 1 | 103.90 | 98.40 | 9.99 |  |  |  |
| (1, 11) | 96 | 54.19 | 42.45 | 4.86 | 3.88 | 163 | 119.42 | 59.00 | 12.64 | 5.70 | 13 | 119.42 | 94.96 | 12.64 | 9.67 | 5 | 124.60 | 111.36 | 13.08 | 4 | 45.79 | 81 |
| (4, 6) | 2 | 45.49 | 45.49 | 4.42 | 4.41 | 2 | 91.19 | 88.89 | 9.53 | 9.15 | 2 | 91.19 | 88.89 | 9.53 | 9.15 | 2 | 103.90 | 101.30 | 10.00 |  |  |  |
| (4, 11) | 42 | 47.61 | 42.52 | 4.80 | 3.85 | 28 | 99.58 | 83.12 | 10.81 | 8.91 | 28 | 99.58 | 83.12 | 10.81 | 8.91 | 12 | 105.70 | 95.93 | 10.30 |  |  |  |
| Show All Hide Not Relevant |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| The curre file is: /h <br> - FR <br> - RE <br> - FT <br> - Pa <br> - IN <br> - EX | nt best s me/daw <br> F: Pos. in FINEMEN <br> Fos. in <br> CK: Pos. i <br> P: Pos. in <br> TIALCC <br> - After <br> PANSION <br> le 4: <br> Final | solution is: id/1191/8 <br> Rank: 48 L NT ROTAT <br> Rank: 1 LL in Rank: 1 Rank: 19 Refinemen: CC: 45.79 | th10_0_0x EXP_LIBR <br> LG: $\mathbf{5 2 . 2 7}$ ZS ON AND M <br> G: $\mathbf{1 1 9 . 4 2}$ ZS <br> LG: 119.42 <br> LG: 124.60 <br> Pos. in Rank | 2FR1_2-1.pd RY/4/0/h10 <br> CORE: 4.46 To DEL <br> ORE: 12.64 Top SCORE 12.64 CORE: 0.00 T 5 INITCC: 12 <br> Traced: $\mathbf{8 1 . 0 0}$ | b with FINAL _0_0xx2FR1 LLG in Cluste p LLG in Clus Top LLG in Cl op LLG in Clus <br> . 67 Top INITCC | CC: 45.79 <br> -1.pdb <br> (1, 11): 56.3 <br> (1, 11): 119 <br> ter $(\mathbf{1}, \mathbf{1 1 )}$ : <br> ( $\mathbf{( 1 , 1 1 ) : 1 2 5}$ <br> in Cluster (1, | and n . res <br> 31 Top ZSC <br> 9.42 Top Z 119.42 Top 25.00 Top <br> 1, 11): 13.3 | dues traced 8 <br> ORE in Cluste <br> CORE in Clus ZSCORE in Cl SCORE in Clu | 81 <br> ( $\mathbf{1 , 1 1 ) : ~} 4.94$ <br> ter (1, 11): 12. <br> uster $(1,11)$ : 1 <br> ster (1, 11): $\mathbf{0 . 0}$ |  |  |  |  |  |  |  |  |  |  |  |  |  |

(a)



| \#Cluster | Rotation and Model Refinement |  |  |  |  | Translation Function |  |  |  |  | Packing |  |  |  |  | Rigid Body Refinement |  |  | Initial CC |  | Best Trace CC/aa |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | \#Rots. | $\begin{gathered} \hline \text { Top } \\ \text { LLG } \end{gathered}$ | $\begin{aligned} & \text { Mean } \\ & \text { LLG } \\ & \hline \end{aligned}$ | $\begin{gathered} \text { Top } \\ \text { Zscore } \end{gathered}$ | $\begin{gathered} \text { Mean } \\ \text { Zscore } \\ \hline \end{gathered}$ | \#Trans. | $\begin{aligned} & \hline \text { Top } \\ & \text { LLG } \\ & \hline \end{aligned}$ | $\begin{aligned} & \hline \text { Mean } \\ & \text { LLG } \\ & \hline \end{aligned}$ | $\begin{gathered} \text { Top } \\ \text { Zscore } \end{gathered}$ | $\begin{aligned} & \text { Mean } \\ & \text { Zscore } \end{aligned}$ | \#Sol. | $\begin{aligned} & \hline \text { Top } \\ & \text { LLG } \\ & \hline \end{aligned}$ | Mean LLG | $\begin{gathered} \text { Top } \\ \text { Zscore } \end{gathered}$ | $\begin{aligned} & \text { Mean } \\ & \text { Zscore } \end{aligned}$ | \#Sol. | $\begin{array}{\|c\|} \hline \text { Top } \\ \text { LLG } \\ \hline \end{array}$ | $\begin{gathered} \hline \text { Mean } \\ \text { LLG } \end{gathered}$ | Before Refinement CC | After <br> Refinement CC | MODE | Cycle | CC | \#Res. traced |
| 0 | 7310 | 29.81 | 17.99 | 5.33 | 3.64 | 700 | 57.20 | 39.86 | 8.07 | 5.54 | 267 | 57.20 | 40.11 | 8.07 | 5.43 | 267 | 63.60 | 42.44 | 7.49 | 7.85 | FAST | 1 | 45.4 | 103 |

The current best solution is: 11 es_ 0 _115.pdb with FINALCC: 45.40 and n. residues traced 103
file is: /home/alfotest/1v70/11_EXP/0/1/0/1les_0_115_rotra.pdb

- FRF: Pos. in Rank: $\mathbf{6 7 5 8}$ LLG: $\mathbf{2 6 . 5 0}$ ZSCORE $\mathbf{4 . 0 4}$ Top LLG in Cluster 0: $\mathbf{3 3 . 1 1}$ Top ZSCORE in Cluster 0 : 5.79
- REFINEMENT ROTATION AND MODEL
- FTF: Pos in Rank 1LLG 57.20 ZSCORE: 7.94 Top LLG in Cluster 0. 57.20 Top ZSCORE in Cluster 0. 8.07
- PACK. Pos in RNP Pos in Rank 5 LLG 58.50 ZSCORE 0.00 Top UG in Cluster 0.63 .60 Top ZSCORE in Cluster 0 0. 0.00
- INITIALCC $\begin{aligned} & \text { - Before Refinement. Pos. in Rank: } 1 \text { INITCC: } 7.49 \text { Top INITCC in Cluster 0. 7.49 }\end{aligned}$
- After Refinement: Pos. in Rank: 5 INITCC: 7.25 Top INITCC in Cluster 0: 7.85
- EXPANSION

Cycle 1:
Final CC: $\mathbf{4 5 . 4 0 \%} \mathrm{N}$. Residues Traced: $\mathbf{1 0 3 . 0 0}$
It seems you have a good solution!
Here you can find the best solution and map for further refinement.
BORGES will end now.
(b)

Figure 2
Html output for (a) the $A R C I M B O L D O \_L I T E$ solution of PDB entry 1191 and (b) the $A R C I M B O L D O \_B O R G E S$ solution of PDB entry 1v70.

Table 2
Performance of $A R C I M B O L D O \_L I T E$ and $A R C I M B O L D O \_B O R G E S$ on a pool of 294 test structures.

A detailed table with PDB codes and parameterizations is available as Supporting Information.

|  | Pool of <br> structures | Solved by <br> ARCIMBOLDO_LITE | Solved by <br> ARCIMBOLDO_BORGES | Total <br> solved |
| :--- | :---: | :--- | :---: | :---: |
| Tetal | 29 | 143 | 38 | 181 |
| Resolution |  |  |  |  |
| $1-0.54 \AA$ | 24 | 14 | 5 | 19 |
| $1.3-1.0 \AA$ | 43 | 25 | 8 | 33 |
| $1.6-1.3 \AA$ | 78 | 38 | 12 | 50 |
| $2.2-1.6 \AA$ | 149 | 66 | 13 | 79 |
| Fragments |  |  |  |  |
| Helices | $230 \dagger$ | 139 | 3 | 142 |
| Fe clusters, haem | 5 | 4 | 0 | 4 |
| $\beta$-Sheets | 147 | 0 | 31 | 34 |
| Antiparallel (udu) | 131 | - | 2 | 31 |
| Parallel-antiparallel (uud) | 24 | - | 1 | 2 |
| Parallel (uuu) | 13 | - | 6 | 1 |
| Antiparallel (udud) | $6(70)$ | - | 1 | 6 |
| Disulfide bridges | 20 | 0 |  | 1 |

$\dagger 1$ curved; 2 antiparallel (ud).
critical if the search fragment is a helix, but is the default for libraries of folds.

Subsequent searches for additional copies of the same or different fragments are also subjected to clustering of rotation solutions. Clusters are identified by combinations of numbers corresponding to the original and the new rotation peaks. Equivalent combinations are kept apart for simplicity when calculating the translation search but are joined for the refinement stage so that they may be fused if equal. Clustering
and evaluation results are written into formatted text files, which can be used by the program to continue an interrupted run. Relaunching an interrupted program will read the present folders and continue from the latest stage completed, which implies that a change in parameterization will only be effective if all previous output for this and subsequent steps has been deleted.

After all fragment-placement operations have been performed, selected solutions are subjected to density modification and autotracing with SHELXE (Sheldrick, 2010). Selected solutions may correspond to intermediate fragment-placement rounds as they are chosen to produce the highest CC and LLG. The default number of solutions probed is equal to the number of physical cores in the machine minus one.

If the user specifies a PDB format file with the extension .ent in the configuration file, analysis of results compared with this reference is performed. Goodness of partial solutions is evaluated through the r.m.s.d. of rotations after optimal translation and mean phase errors versus the reference for placed fragments and final solutions. This information is not output to the standard .html file, but has been used in the present study of test structures to evaluate results and to guide


ARCIMBOLDO_LITE solution of PDB entry 1191. (a) Secondary-structure prediction. (b) Structure in rainbow-coloured cartoon representation. (c) SHELXE electron-density map of a solution; the trace is shown as a coil and both placed helical fragments are shown as sticks, with residues eliminated during PDB optimization displayed in red.
default parameterization. The procedure starts by extracting all of the helices (or any other search fragment given as input) found in the .ent file. For each one of them a rotation LLG at their current position is computed in Phaser through the GYRE mode to provide an estimate of the value range for a correct solution. For the placed fragments, the $F$-weighted mean phase difference to the reference structure (Lunin \& Woolfson, 1993) is computed with SHELXE.

### 3.2. Timing benchmarks on various hardware

The structure of the C-terminal domain of human antitermination factor 5 (PDB entry 2iu1; Bieniossek et al., 2006) was used on 12 machines with hardware ranging from four to 24 physical cores and a minimum of 1 GB of RAM per core. A case is included in which 512 MB of RAM was shared by two cores, showing that the associated increase in run time (RT) makes such a configuration unsuitable. Conversely, a machine endowed with CPU overclocking up to 4 GHz and almost 8 GB of RAM per core rendered an RT that was half that of a typical workstation. Table 1 shows the time taken for structure solution on the various machines. It ranged from one to several hours in the default mode that runs as many processes as there are physical cores detected, reserving one thread for program calculations, mainly clustering and combination of solutions, and running as many Phaser and SHELXE jobs as there are cores minus one. It is worth remarking that the job computed is tailored to the hardware and thus a greater or lesser number of solutions are pursued depending on the number of cores. Accordingly, the time required by a more powerful machine may not be much less, but it may succeed in solving a structure where a machine with fewer cores fails.

### 3.3. Performance of $A R C I M B O L D O$ tests

Table 2 summarizes the single-workstation performance of ARCIMBOLDO_LITE and ARCIMBOLDO_BORGES on a set of test structures. An initial baseline was set by running ARCIMBOLDO_LITE blindly on the pool of 294 structures with a fragment search configured to find two polyalanine helices of 14 residues and using a common SHELXE parameterization. This resulted in the solution of 100 of the 294 structures. SHELXE was set to perform five iterations involving 30 cycles of density modification and autotracing, increasing the time for random search 20 -fold and taking into account the presence of helices, a solvent content of $55 \%$ and fragment optimization against the CC. In many cases, three iterations and a tenfold time increase to seed autotracing led to structure solution. Accounting for the particularities of individual structures opens a way to improve the success rate with minimum intervention.
3.3.1. Ultrahigh resolution (beyond $1 \AA$ ). Within the pool of 294 test cases, 24 are atomic resolution structures diffracting to $1 \AA$ resolution or beyond and, of these, 11 contain at least one helix of six amino acids or longer. Such short helices are useful only in this resolution context, where some of the structures are very small. All such helical structures in this group were eventually solved. The behaviour of this group will be illu-
strated by the case of PDB entry 1191 in space group $P 2_{1}$ with diffraction to $0.92 \AA$ resolution. Fig. 2(a) displays the html output from the $A R C I M B O L D O \_L I T E$ run. The all-helical structure displays three long helices of 18,15 and 25 residues, as expected from the secondary-structure prediction displayed in Fig. 3(a). The long third helix is divided into two stretches by a sharp kink. This scenario would appear to be deceptively simple, as the data quality is excellent, but the blind parameterization that was initially used failed to solve it. The same happened with a number of other structures in this group during the first, blind test. Instead, search trials with different helix lengths and parameterization tests led to a solution when searching for two helices of ten residues and increasing the number of density-modification cycles, turning on low-density elimination (Yao et al., 2005), decreasing the solvent content and increasing the rounds of autotracing but lowering the time seed (SHELXE command line -m200 -v0.5-a10 -t1 -q -f -s0. 25 -o).

This result suggested the convenience of taking into account particularities anticipated from data resolution to automate default parameterization. Some common settings are adopted, for instance the presence of helices is always assumed for tracing whenever a helical search model is used and default memory allocation is increased to fit the hardware. The model helix of 14 amino acids fits most helical stretches within an r.m.s.d. of $0.3 \AA$ for the $\mathrm{C}_{\alpha}$ atoms. The availability of extremely high resolution data allows smaller fragments to be extended, but suffers more severely from the lack of accuracy in the model. As can be seen in Figs. 3(b) and 3(c), they contained errors as one of the helices is misplaced. A solution from an imperfect structure is frequently found, and wrong residues may be initially eliminated to increase the CC. Thus, PDB optimization is switched on and, given sufficient cycles and high resolution, the remaining errors are slowly erased. Starting from shorter helices may be more effective as accuracy is enhanced. More density-modification cycles are beneficial and low-density elimination should be turned on, whereas tracing of the atomic map is simpler and can be accelerated with a smaller time parameter. The solvent content is typically extremely low for this resolution group.
3.3.2. High resolution ( $1.3-1 \AA$ ). In the group of 33 structures with helices and resolutions between 1 and $1.3 \AA, 25$ could be solved when searching for predicted helices and running eight iterations of density modification and autotracing; the number of density-modification cycles was not increased so much this time, setting it to 100 , and low-density elimination was given a lower weight of 0.25 . A solvent content of $35 \%$ was assumed and a time seed of 10 was used. As the resolution becomes lower, more starting information is needed for successful expansion; thus, longer helices of 14 amino acids rather than the extremely small helices of less than ten residues were used as search fragments.
3.3.3. Medium resolution ( $1.6-1.3 \AA$ ). In the pool of 62 structures with helices and resolutions between 1.3 and $1.6 \AA$, 38 could be solved when searching for predicted helices and running eight autotracing cycles interspersed with 50 densitymodification cycles, with low-density elimination set to
0.1 and a time seed of 10 and assuming $45 \%$ solvent content.
3.3.4. Low resolution ( $2.2-1.6 \AA$ ). At the lower resolution end, 66 of 149 structures were solved. 124 of the structures contained at least $8 \%$ helices, which is the lowest limit found to be accessible at ultrahigh resolution, but in this context such a low percentage is clearly not sufficient to extend helical fragments. More fragments of longer size but tolerating higher deviations provide successful starting substructures. The lower the resolution, the more difficult it becomes to successfully extend the partial structures, even if they are located. The number of density-modification cycles is gradually decreased from 15 at the higher resolution end of this span to ten below $2 \AA$, a solvent content of $50-60 \%$ is assumed and low-density elimination is entirely switched off. The time seed for autotracing is increased.
3.3.5. Alternative search fragments. Five of the structures in the test pool contained cofactors, in particular $\mathrm{Fe}_{4} \mathrm{~S}_{4}, \mathrm{Fe}_{2} \mathrm{~S}_{2}$ and a haem group. Despite the presence of helices, solving such structures without first accounting for the electron-rich group failed at the stage of fragment location. As the presence of cofactors is easily predictable, their use as search fragments appeared to be indicated. The search for them may be
combined with that for other fragments, but in the cases tested the location of model fragments from the REFMAC5 dictionary (Murshudov et al., 2011; Vagin et al., 2004) was sufficient for a successful expansion to the full structure in four of the five cases. The parameterization was set to match the resolution group, but the initial fragment was kept throughout the first autotracing cycles (SHELXE -K flag) as only polypeptide main chain is built into the electron-density map (Sheldrick, 2010). These fragments may be specified by their three-letter library name in the instruction .bor file as they have been included along with the model helix in the internal library.

The straight model helix built into $A R C I M B O L D O$ may not be the optimal search fragment for helical structures containing markedly curved or otherwise distorted helices. This is the case for the 90 -amino-acid, $1.5 \AA$ resolution structure with PDB code 3132 in space group $I 4_{1} 22$. Solution with $A R C I M B O L D O \_L I T E$ failed, whereas using a collection of 162 models for differently distorted helices of 18 amino acids extracted from the PDB and gathered in a library rendered a solution. PDB entry 3132 is one of the 38 cases solved with libraries of similar models with a single-computer implementation of $A R C I M B O L D O \_B O R G E S$.
Conf: $\mathfrak{l l}$ )

Pred: CCEEECCCEEECCCCCEEEEEEEECCCEEEEEEEECCCCC CCCCCCCCCCEEEEEEECEEEEEECCEEEEECCCCEEEEC CCCCEEEEECCCCCEEEEEEECCC
AA: MEIKDLKRLARYNPEKMAKIPVFQSERMLYDLYALLPGQA QKVHVHEGSDKVYYALEGEVVVRVGEEEALLAPGMAAFAP AGAPHGVRNESASPALLLVVITAPRP


(b)

(c)

Figure 4
$A R C I M B O L D O \_B O R G E S$ solution of PDB entry 1v70. (a) Secondary-structure prediction. (b) Structure in rainbow-coloured cartoon representation. (c) SHELXE electron-density map of a solution; the trace is shown as a coil and the placed antiparallel three-stranded fragments are shown as sticks, with residues eliminated during PDB optimization displayed in red.

### 3.4. Performance of BORGES libraries on test structures

The pool of test structures contains 147 cases in which $\beta$-strands are the only secondary-structure fragments or make up most of the fold. In addition, 20 structures feature four or more disulfide bridges anchoring a peculiar fold, as is typical in small toxin inhibitors. Such cases are outside the scope of an ARCIMBOLDO_LITE secondary-structure fragment search, but can be solved with libraries of small folds in ARCIMBOLDO_BORGES (Sammito et al., 2013). Although the method is better suited to large computing resources, the new implementation can run either connecting to a local or a remote grid or on a single multicore machine. Again, automatic parameterization is related to the available hardware, and only four rotation clusters will be sequentially evaluated by default when the program is run on a single machine. The test structures in this pool are small enough to be computed on the eight-core workstations used for the $A R C I M B O L D O_{-}$ LITE tests. The typical run time was $2-5 \mathrm{~d}$. The same precomputed libraries were used in all tests. They are available for download from the $A R C I M B O L D O$ web page. Three libraries contain $\beta$-sheets with three strands in parallel, parallel-antiparallel and antiparallel arrangements. The latter have been the most successful, being more frequent; 131 of the cases in the group of $\beta$-sheet-containing structures display this arrangement of strands. In addition to the library of single distorted helices mentioned, two libraries contain two contiguous helices in parallel and antiparallel dispositions, respectively. They are especially indicated for coiled coils, as single helices tend to be placed in the same position in fragment searches. Finally, a library of two tetrapeptides linked by a disulfide bridge has also been tested.

Two test cases were solved with the library of helices, and 31 with the antiparallel, two with the parallel-antiparallel and one with the parallel three-stranded library, respectively. One structure was solved with the disulfide-linked peptide library. The optimal parameterization has been found to be different for structures mainly composed of $\beta$-strands and is accordingly set as default for $A R C I M B O L D O \_B O R G E S$ as exemplified in Fig. 4. If the program is launched with one of the nonhelical libraries, the default expansion stage will not search for helices, will use a lower solvent content than in the helical case, will keep the original fragment for $K$ cycles and will perform more autotracing iterations. As in ARCIMBOL$D O_{-} L I T E$, a line in the . bor file setting the parameterization for $S H E L X E$ will override any of the inbuilt defaults, but leaving this line unset will apply the resolution-tailored and hardware-tailored values described.

### 3.5. Statistics and performance

Table 2 summarizes the single-workstation performance of ARCIMBOLDO_LITE and ARCIMBOLDO_BORGES on the 294 test structures. Beyond the 181 structures ( $62 \%$ ) solved as described, some could be solved if run on a computer with more cores or if the parameterization was further adapted to run a more exhaustive attempt or even with less restrictions on a grid of computers. For example, the 114-
amino-acid structure with PDB code 3 f 40 in space group $C 222_{1}$ at $1.27 \AA$ resolution renders correct substructures composed of two helices of 12 and eight alanines, respectively, but these are not prioritized among the seven expanded helices. Running $A R C I M B O L D O \_L I T E$ on a machine with more cores or a grid would solve it. Alternatively, the number of expansions to try may be increased by the user. Regarding ARCIMBOLDO_BORGES, the 105 -amino-acid structure with PDB code 2 haz at $1.7 \AA$ resolution is solved automatically on a double Xeon workstation with similar features to those used for benchmarking performance in this study but endowed with 3 GB rather than 2 GB of RAM per core. This case has not been counted as solved in our results.

High-resolution cases yield a higher success rate: $78 \%$ for atomic resolution structures, decreasing to $64 \%$ for data in the 1.6-1.3 A resolution range and to $53 \%$ below $1.6 \AA$ resolution. For structures mainly composed of $\beta$-strands, phasing with the secondary-structure fragments has not found any practical use within ARCIMBOLDO, whereas precomputed libraries of $\beta$-sheets have been successful in $22 \%$ of the identified cases. Not all have been tested, as structures previously solved with helices were eliminated from the pool. In the same manner, the more frequent antiparallel disposition of three strands was tested first and only structures where this fold would not work were further considered for phasing with parallel-antiparallel or parallel libraries. Six structures were phased both with libraries containing three-stranded and four-stranded models. They are counted only once in the total of 181 structures. The four-stranded libraries are also available for download and might be useful for larger structures. In the same way, two structures were phased with the library of disulfide bridges and one with a library of polyalanine helices extracted from real structures and presenting different deviations from the regular model helix. This structure, PDB entry 3132, could not be solved with a model helix, and the use of the library illustrates that a starting model with a lower r.m.s.d may succeed where the model fails and that the more exhaustive run needed may be worthwhile to solve an unknown structure.

## 4. Concluding remarks

ARCIMBOLDO_LITE succeeds in solving 143 out of a pool of 294 test structures on an eight-core workstation, with sizes ranging from 40 to 120 residues and resolutions between 2.2 and $0.54 \AA$. The fragments placed are predominantly straight polyalanine helices and, in three cases, iron cofactors. ARCIMBOLDO_BORGES solves a further 38 structures using precomputed libraries of small folds consisting of contiguous helices, three-stranded $\beta$-sheets and disulfidelinked tetrapeptides. The run times for $A R C I M B O L D O_{-}$ LITE jobs are typically a couple of hours, whereas ARCIM$B O L D O_{-} B O R G E S$ requires one or two days on a single machine.

The results of this study have been used to incorporate default parameterization as a function of data resolution and search fragment. Leaving the SHELXE line unset in the input .bor file will activate $A R C I M B O L D O$ defaults, which may differ from SHELXE defaults.

The binaries for Linux and MAC OS have been validated for the latest distributions and can be downloaded from http://chango.ibmb.csic.es/ARCIMBOLDO_LITE. The same binaries that run on a single workstation can be run to distribute jobs on a local or remote Condor, SGE/Opengrid or Torque/MOAB grid. The precomputed libraries can be downloaded together with the ARCIMBOLDO_BORGES binary.

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