SABBAC: online Structural Alphabet-based protein BackBone reconstruction from Alpha-Carbon trace

Julien Maupetit, R. Gautier¹ and Pierre Tufféry*

Equipe de Bioinformatique Génomique et Moléculaire, INSERM U726, Université Paris 7, case 7113, 2, place Jussieu, 75251 Paris cedex 05, France and ¹Institut de Pharmacologie Moléculaire et Cellulaire, UMR 6097 CNRS/UNSA, 660 route des Lucioles, 06560 Sophia Antipolis, France

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ABSTRACT

SABBAC is an on-line service devoted to protein backbone reconstruction from alpha-carbon trace. It is based on the assembly of fragments taken from a library of reduced size, selected from the encoding of the protein trace in a hidden Markov model-derived structural alphabet. The assembly of the fragments is achieved by a greedy algorithm, using an energybased scoring. Alpha-carbon coordinates remain unaffected. SABBAC simply positions the missing backbone atoms, no further refinement is performed. From our tests, SABBAC performs equal or better than other similar on-line approach and is robust to deviations on the alpha-carbon coordinates. It can be accessed at http://bioserv.rpbs.jussieu.fr/SABBAC. html.

INTRODUCTION

There are various fields where the structure of proteins is expressed using a coarse grained model that needs to be expanded in an all atom model. Such preoccupation occurs for instance in fields such as ab-initio or de novo protein fold generation, comparative modelling or the refinement of experimental data obtained at low resolution. It might also occur for structures generated from normal modes simulations. Starting from a coarse grained model of a protein structure, which might provide only some information about where the residues are roughly located in 3D space, the full protein structure generation is often decomposed in a two-part process: (i) generation of the backbone coordinates and (ii) side chain positioning. The alpha-carbon reconstruction process is one way to address the former process, and several studies have tackled this problem of the protein all-atom peptide reconstruction from alpha-carbon trace (1-19).

Among the different approaches that have been proposed, some explore the conformational space of the peptidic units to produce a complete backbone. For instance, the approach developed by Payne (16) attempts to identify the optimal rotation of peptide units using a potential of mean force depending on adjacent residues. Most approaches, such as Ref. (17,19), rely on the assembly of fragments extracted from libraries derived from known structures. Such fragments are then assembled to produce a peptidic chain fitting as best as possible the alpha-carbon trace, usually using energy or geometry criteria to drive the search. Fragment based approaches are in general confronted with the limitation that, in order to improve the accuracy of the reconstruction, it is necessary to maintain large and up to date collections of fragments. The MaxSprout server (17) has been one of the first proposed. It can still be considered as a reference. It is based on a fragment library regularly updated from the protein structures available.

Here, we introduce SABBAC, a procedure that relies on a new approach to fragment selection—among a reduced set—and assembly. It uses the encoding of the alpha-carbon trace using a hidden Markov model derived structural alphabet (20) to select at each position in the structure a small set of candidates among a complete set of only 155 candidates fragments describing all the letters of the structural alphabet—i.e. to describe all the conformations of all protein structures. It then assembles the fragments using a greedy algorithm, searching for the 'optimal' combination of fragments compatible with the alpha-carbon trace and produce a full-protein backbone reconstruction.

METHODS

Structure dependent fragment library

To select fragments that will be used for the structure generation we encode the alpha-carbon trace in the Hidden Markov Model derived SA-27 structural alphabet described in Ref. (20). Here, we use the optimal trajectory as produced by

*To whom correspondence should be addressed. Tel: 331 44 27 77 33; Fax: 331 43 26 38 30; Email: tuffery@ebgm.jussieu.fr

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the viterbi algorithm. Then at each position, we retain as candidate fragments that represent the letter of the structural alphabet. As described in Ref. (20–22), we have only a number of 155 fragments to describe the 27 letters of the alphabet. Having selected only one letter to describe the conformation of each fragment of four residue length, the average number of fragments used, is only on the order of 5–6 per position.

Peptide units coordinates generation

To quickly generate atoms coordinates for the N, HN, C' and O atoms of the peptide units, we follow a procedure described previously by Milik et al. (18). We use coordinates precomputed in a local reference defined from three consecutive alpha carbons. Since SA-27 describes the conformations of fragments of four residue length, two different sets of atoms can be used to define such local reference. For each prototype fragment associated to each letter of SA-27, we have precomputed the atom coordinates at the first, second and third peptidic bond in these two possible references. We have assessed the stability and accuracy of using each set of coordinate (data not shown). Here, we use the reference associated with the last three alpha-carbons of the fragments of four residue length to position the atoms between the second and third alphacarbon. For the extremities of the polypeptidic chain only, we consider the first and third peptidic bonds of the fragments.

Search for an optimal combination of fragments

Given the collection of candidate fragments at each position of the structure, we use a greedy algorithm, as described in Ref. (21,22) to search for their optimal combination. At each position, given a collection of reconstruction up to the previous position, we generate all the new possible assemblies of size increased by one residue. Each of them is scored (see below), and we retain only a limited subset of assemblies to iterate to the next position (heap size). Here, we use a heap size of only 10, a number learnt from the reconstruction of series of structures. Compared with Ref. (21,22), we use a simpler strategy: the rebuild process is only achieved by growing from N- towards C-terminal. The process is not iterated.

Energy evaluation

To drive the search, we use an energy criterion combining some of the OPEP force field (23) with some terms assessing the quality of the geometry of the alpha-carbons.

$$E = E_{\text{SC, SC}} + E_{\text{HB}} + E_{\text{VdW}} + E_{\text{PhiP}} + E_{\text{BB}} + E_{\text{Trans}},$$

where $E_{SC,SC}$ describes side chain–side chain interactions; E_{HB} , hydrogen bonds, E_{VdW} the Van der Waals backbone– backbone and backbone–side chain interactions, E_{PhiP} the phi positive contribution, E_{BB} the alpha carbon valence angle distortion and E_{Trans} is a pseudo energy term related to the transitions between consecutive fragments. Backbone is described using an all-atom representation (N, HN, CA, C, O). Side chains are represented using a single particle, described by a centroid and a radius for each residue type. Centroid coordinates and ray values have been computed and optimized on a non-redundant Protein Data Bank (PDB) of <30% identity (J. Manupetit, P. Tufféry and P. Derreumaux, manuscript in preparation). All these terms are computed as described in Ref. (23) except for the phi positive term, that is set to a constant of +0.5 kcal/mol if the phi angle value is positive, 0 otherwise, except for glycines (-0.3 kcal/mol for phi positive values). For $E_{\rm BB}$, we simply use the square of the deviation to the canonical angle value of 110°. For $E_{\rm Trans}$, we use

$$E_{\text{Trans}}^{i \rightarrow i+1} = \sum_{i=1}^{N} -\log(p(i \rightarrow i+1))$$

where $p(i \rightarrow i + 1)$ is the probability that the fragment selected at position i + 1 follows the fragment at position i. $p(i \rightarrow i + 1)$ have been evaluated from a non-redundant set of proteins of the PDB with <30% sequence identity.

PERFORMANCES

We have assessed the performances of SABBAC on various test sets. Some results are reported Table 1. Its upper section

 Table 1. Backbone reconstruction performance. Comparison with other methods

PDB n Residues SABBAC MaxSprout bb Adcock subset 111M 154 0.29 0.42 0.91 ICTF 68 0.43 0.73 0.85 IIGD 61 0.36 0.44 0.68 IOMD 107 0.35 0.41 0.77 ISEMA 58 0.48 0.34 1.00 ITIMA 247 0.59 0.60 0.97 IUBQ 76 0.35 0.38 0.44 0.98 2LYM 129 0.38 0.44 0.98 2MHR 118 0.50 0.54 0.88 2PCY 99 0.42 0.54 0.81 SNLL 138 0.37 0.46 0.85 Mean 0.41 0.47 0.89 SD 0.09 0.10 0.08 PDB newcomers subset 1 1 1 0.47 0.89 0.55 0.54 0.96			Main chain RMSd		
Adcock subset 111M 154 0.29 0.42 0.91 ICTF 68 0.43 0.73 0.85 IIGD 61 0.36 0.44 0.68 IOMD 107 0.35 0.41 0.77 ISEMA 58 0.48 0.34 1.00 ITIMA 247 0.59 0.60 0.97 IUBQ 76 0.35 0.38 0.96 2CTS 437 0.40 0.45 0.86 2LYM 129 0.38 0.44 0.98 2MHR 118 0.50 0.54 0.88 2PCY 99 0.42 0.54 0.91 2WRP 104 0.30 0.42 0.87 SD 0.09 0.10 0.08 PDB newcomers subset I I NXA 0.86 IT70A 255 0.42 0.50 0.95 ITXOA 235 0.41 <th>PDB</th> <th>n Residues</th> <th>SABBAC</th> <th>MaxSprout</th> <th>bb</th>	PDB	n Residues	SABBAC	MaxSprout	bb
111M 154 0.29 0.42 0.91 ICTF 68 0.43 0.73 0.85 IIGD 61 0.36 0.44 0.68 IOMD 107 0.35 0.41 0.77 ISEMA 58 0.48 0.34 1.00 ITIMA 247 0.59 0.60 0.97 IUBQ 76 0.35 0.38 0.44 0.98 2CTS 437 0.40 0.45 0.86 2LYM 129 0.38 0.44 0.98 2MHR 118 0.50 0.54 0.88 2PCY 99 0.42 0.54 0.91 2WRP 104 0.30 0.42 0.87 4PTI 58 0.53 0.44 0.88 SD 0.09 0.10 0.08 PDB newcomers subset I I NZA 346 0.55 0.54 0.96 ITXOA 235	Adcock subset				
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	111M	154	0.29	0.42	0.91
IIGD610.360.440.68IOMD1070.350.410.77ISEMA580.480.341.00ITIMA2470.590.600.97IUBQ760.350.380.962CTS4370.400.450.862LYM1290.380.440.982MHR1180.500.540.882PCY990.420.540.912WRP1040.300.420.874PTI580.530.440.815NLL1380.370.460.85Mean0.410.470.89SD0.090.100.08PDBnewcomers subset1IPXZA3460.550.540.961RKIA1010.580.440.881S7LA1770.290.360.861T70A2550.420.500.95IXOA2350.410.380.96IV0ED6660.480.450.89IV7BA1750.300.410.87IVRSB2550.340.420.84IVKCA1490.280.330.82IWRA880.430.360.79IWMHA830.270.280.82IWPBG1680.370.350.86IWRA880.430.360.76IX6JA88 <td< td=""><td>1CTF</td><td>68</td><td>0.43</td><td>0.73</td><td>0.85</td></td<>	1CTF	68	0.43	0.73	0.85
10MD 107 0.35 0.41 0.77 $1SEMA$ 58 0.48 0.34 1.00 $1TIMA$ 247 0.59 0.60 0.97 $1UBQ$ 76 0.35 0.38 0.96 $2CTS$ 437 0.40 0.45 0.86 $2LYM$ 129 0.38 0.44 0.98 $2MHR$ 118 0.50 0.54 0.88 $2PCY$ 99 0.42 0.54 0.87 $4PTI$ 58 0.53 0.44 0.81 $5NLL$ 138 0.37 0.46 0.85 Mean 0.41 0.47 0.89 SD 0.09 0.10 0.08 PDB newcomers subset V V $1PXZA$ 346 0.55 0.54 $1T70A$ 255 0.42 0.50 0.95 $1TXOA$ 235 0.41 0.38 0.96 $1V0ED$ 666 0.48 0.45 0.89 $1V7BA$ 175 0.30 0.41 0.89 0.28 0.33 0.82 $1VRA$ 103 0.47 0.59 1.00 $1VRA$ 103 0.47 0.59 0.09 $1TXOA$ 255 0.34 0.42 0.88 $1VB5B$ 255 0.34 <td>1IGD</td> <td>61</td> <td>0.36</td> <td>0.44</td> <td>0.68</td>	1IGD	61	0.36	0.44	0.68
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	10MD	107	0.35	0.41	0.77
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	1SEMA	58	0.48	0.34	1.00
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	1TIMA	247	0.59	0.60	0.97
2CTS 437 0.40 0.45 0.86 2LYM 129 0.38 0.44 0.98 2MHR 118 0.50 0.54 0.88 2PCY 99 0.42 0.54 0.91 2WRP 104 0.30 0.42 0.87 4PTI 58 0.53 0.44 0.81 5NLL 138 0.37 0.46 0.85 Mean 0.41 0.47 0.89 SD 0.09 0.10 0.08 PDB newcomers subset 1PXZA 346 0.55 0.54 0.96 1RKIA 101 0.58 0.44 0.88 1STLA 177 0.29 0.36 0.86 1T70A 255 0.42 0.50 0.95 1XCA 175 0.30 0.41 0.87 1VB5B 255 0.34 0.42 0.84 1VKCA 149	1UBQ	76	0.35	0.38	0.96
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	2CTS	437	0.40	0.45	0.86
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	2LYM	129	0.38	0.44	0.98
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	2MHR	118	0.50	0.54	0.88
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	2PCY	99	0.42	0.54	0.91
4PTI 58 0.53 0.44 0.81 5NLL 138 0.37 0.46 0.85 Mean 0.41 0.47 0.89 SD 0.09 0.10 0.08 PDB newcomers subset	2WRP	104	0.30	0.42	0.87
5NLL 138 0.37 0.46 0.85 Mean 0.41 0.47 0.89 SD 0.09 0.10 0.08 PDB newcomers subset 1PXZA 346 0.55 0.54 0.96 1RKIA 101 0.58 0.44 0.88 1S7LA 177 0.29 0.36 0.86 1T70A 255 0.42 0.50 0.95 1TXOA 235 0.41 0.38 0.96 1V0ED 666 0.48 0.45 0.89 1V7BA 175 0.30 0.41 0.87 1VB5B 255 0.34 0.42 0.84 1VKCA 149 0.28 0.33 0.82 1VR4A 103 0.47 0.59 1.00 1VR9A 121 0.42 0.45 0.79 1WMHA 83 0.27 0.28 0.82 1WPBG 16	4PTI	58	0.53	0.44	0.81
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1S7LA 177 0.29 0.36 0.86 1T70A 255 0.42 0.50 0.95 1TXOA 235 0.41 0.38 0.96 1V0ED 666 0.48 0.45 0.89 1V7BA 175 0.30 0.41 0.87 1VB5B 255 0.34 0.42 0.84 1VKCA 149 0.28 0.33 0.82 1VR4A 103 0.47 0.59 1.00 1VR9A 121 0.42 0.45 0.79 1WMHA 83 0.27 0.28 0.82 1WPBG 168 0.37 0.35 0.86 1WMIA 88 0.41 0.42 0.81 1X6JA 88 0.43 0.36 0.76 1XB9A 108 0.46 0.51 0.81 1XE0B 107 0.61 0.62 0.90 Mean 0.42 0.44 0.88 SD<	1RKIA	101	0.58	0.44	0.88
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$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	1V0ED	666	0.48	0.45	0.89
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	1V7BA	175	0.30	0.41	0.87
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IVR4A 103 0.47 0.59 1.00 IVR9A 121 0.42 0.45 0.79 IWMHA 83 0.27 0.28 0.82 IWPBG 168 0.37 0.35 0.86 IWMIA 88 0.41 0.42 0.81 IX6JA 88 0.43 0.36 0.76 IXB9A 108 0.46 0.51 0.81 IXE0B 107 0.61 0.62 0.90 Mean 0.42 0.44 0.88 SD 0.10 0.09 0.07 GLOBAL 0.41 0.45 0.87 SD 0.09 0.10 0.07	1VKCA	149	0.28	0.33	0.82
1VR9A 121 0.42 0.45 0.79 1WMHA 83 0.27 0.28 0.82 1WPBG 168 0.37 0.35 0.86 1WMIA 88 0.41 0.42 0.81 1X6JA 88 0.43 0.36 0.76 1XB9A 108 0.46 0.51 0.81 1XE0B 107 0.61 0.62 0.90 Mean 0.42 0.44 0.88 SD 0.10 0.09 0.07 GLOBAL Wean 0.41 0.45 0.87 SD 0.09 0.10 0.07	1VR4A	103	0.47	0.59	1.00
1WMHA 83 0.27 0.28 0.82 1WPBG 168 0.37 0.35 0.86 1WMIA 88 0.41 0.42 0.81 1X6JA 88 0.43 0.36 0.76 1XB9A 108 0.46 0.51 0.81 1XE0B 107 0.61 0.62 0.90 Mean 0.42 0.44 0.88 SD 0.10 0.09 0.07 GLOBAL Mean 0.41 0.45 0.87 SD 0.09 0.10 0.07	1VR9A	121	0.42	0.45	0.79
1WPBG 168 0.37 0.35 0.86 1WMIA 88 0.41 0.42 0.81 1X6JA 88 0.43 0.36 0.76 1XB9A 108 0.46 0.51 0.81 1XE0B 107 0.61 0.62 0.90 Mean 0.42 0.44 0.88 SD 0.00 0.07 GLOBAL Mean 0.41 0.45 0.87 SD 0.09 0.10 0.07	1WMHA	83	0.27	0.28	0.82
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	1WPBG	168	0.37	0.35	0.86
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	1WMIA	88	0.41	0.42	0.81
1XB9A 108 0.46 0.51 0.81 1XE0B 107 0.61 0.62 0.90 Mean 0.42 0.44 0.88 SD 0.10 0.09 0.07 GLOBAL	1X6JA	88	0.43	0.36	0.76
1XE0B 107 0.61 0.62 0.90 Mean 0.42 0.44 0.88 SD 0.10 0.09 0.07 GLOBAL	1XB9A	108	0.46	0.51	0.81
Mean 0.42 0.44 0.88 SD 0.10 0.09 0.07 GLOBAL	1XE0B	107	0.61	0.62	0.90
SD 0.10 0.09 0.07 GLOBAL University Mean 0.41 0.45 0.87 SD 0.09 0.10 0.07	Mean		0.42	0.44	0.88
GLOBAL Mean 0.41 0.45 0.87 SD 0.09 0.10 0.07	SD		0.10	0.09	0.07
Mean0.410.450.87SD0.090.100.07	GLOBAL				
SD 0.09 0.10 0.07	Mean		0.41	0.45	0.87
	SD		0.09	0.10	0.07

Backbone reconstruction performance is evaluated compared with the native structure (backbone heavy atoms r.m.s.d). Two test sets are presented, the first one is a subset of those presented by SA Adcock (19), and the second one is composed of recent newcomers of the PDB (24).

presents SABBAC results obtained for a collection of proteins discussed in Ref. (19), which provides some elements of comparison with earlier approaches. For bb, we present the results as obtained with the bb program available for download, without any minimization. For MaxSprout, we have used the on-line server. As shown, SABBAC gives on average better results than both MaxSprout and bb. The lower part reports some results obtained for recent entries of the PDB, not introduced for the identification of our collection of fragments. In this subset, the performances of SAB-BAC are unaffected. We have also checked how SABBAC performs for perturbed alpha-carbon trace. Figure 1 presents distributions of deviations of the peptide unit planes between native and rebuilt structures of the proteins of Table 1; for perturbed alpha-carbon traces. Even for traces randomly perturbed by over 1 Å on alpha-carbon coordinates, SABBAC results are only marginally affected. We emphasize that correct peptide unit plane orientation also implies correct side chain orientation. Figure 2 shows the rebuilt structures obtained for the oncomodulin (10MD PDB code), from native trace (A) and from highly degenerate alpha carbons trace (B). Finally, we have also considered the reconstruction for a series of models from the CASP6 experiment (25). We have retained all the targets corresponding to complete structures (not domain-only targets) and removed targets having missing fragments. For each, we have considered the best and the rank 5 models (as classified by CASP GDT-TS). Incomplete models were discarded. This resulted in 31 targets-60 models-including homology modelling (14 targets), fold recognition (13 targets) and new fold (4 targets) categories. We have rebuilt the backbone using SABBAC and MaxSprout. For MaxSprout, we obtained results for only 57 models. Figure 3 plots, for all the models, for SABBAC and MaxSprout, the fractions of the peptide unit planes that deviate by $<10^{\circ}$ or 40° to that of the native structures. It is expressed as a function of the deviation of the model to the native structure, quantified using the TM-Score (26). A value of TM-Score of 1 is reached for perfect correspondence between modelled and native main chain. The lowest the value, the worse the model. Such measure is more relevant than the classical root mean square deviation (r.m.s.d.) since it assess the parts of the structures that correspond, not necessarily resulting in extremely low values if a domain undergoes a collective motion relative to another, for instance the alpha-carbon RMS deviations are between 1.6 and over 10 Å; for the 60 models. As can be seen, SABBAC performs overall better than MaxSprout for a 10° deviation, implying a more accurate reconstruction. The performance differences get smaller for 40°. Compared with the CASP models, SABBAC tends to propose better results for the lower TM scores, which suggest a good robustness to main chain perturbation. For values of TM score close to 1, CASP models



Figure 1. Peptidic bonds angle distribution. Distribution of the peptide unit planes angular deviations between native and rebuilt structures for perturbed alpha-carbon traces. Each deviation is calculated for locally fitted backbones. Left, average alpha-carbon trace perturbation of 0.2 Å. Right, average perturbation of 1.0 Å.



Figure 2. SABBAC rebuilding example, 10MD. (A) Native and SABBAC rebuilt structures. (B) Native and SABBAC rebuilt structures from an alpha-carbon trace perturbed by 0.8 Å on average. The native structure is represented in blue.



Figure 3. SABBAC and MaxSprout reconstruction performance for 31 CASP6 targets. For each of target, the best and the rank 5 models have been considered. The fraction of the peptide unit planes deviating by $<10^{\circ}$ (left) and 40° (right) are plotted as a function of the TM-Scores of the models. Lines correspond to regressions. Red + CASP6 models versus native structure; black x SABBAC reconstruction; blue triangles MaxSprout reconstruction.

are in general better. One must however consider that the processes of model generation and SABBAC reconstruction largely differ, SABBAC not performing any refinement.

IMPLEMENTATION

The SABBAC server integrates all the steps required for the complete protein reconstruction from an alpha-carbon trace specified in the PDB format. In the present version, atoms present in the PDB file other than alpha-carbons are discarded, as well as hetero groups. As output, SABBAC will

return a file with all the backbone atoms, and the detail of the energy scores, for the complete structure and by residue. Since the greedy algorithm heap size is of 10, it is possible to ask for a number of models varying between 1 and 10. The models are sorted according to their scores, by decreasing order—the best is at rank 1. Side chains are not positioned by default, but it is possible to chain SABBAC to SCit (27) fast side chain positioning method in order to obtain all atom models. This fast version of SCit positions side chains by selecting the most probable side chain conformation given backbone conformation, removing side chain conformations having clashes with the backbone. Side chain–side chain clashes are then considered. Using a greedy algorithm, the calculation time depends on the size of the protein and is expected to increase linearly with the size. However, this is presently not true for the energy calculation. For small proteins, typical calculation times are in the order of few seconds. For larger size proteins, calculation time can increase up to several minutes or tens of minutes, depending on the server load.

DISCUSSION AND FUTURE WORK

In its present form, the SABBAC seems a reliable alternative to the reference MaxSprout server that pioneered the field. SABBAC provides, on average, a more accurate reconstruction for accurate alpha-carbon trace, and its performances for perturbed traces are still relevant. One strong feature of SABBAC is to provide an answer even in cases where the trace is degenerated. In such cases, the MaxSprout may return structures having missing parts, if the conformations are too far from those observed in its bank of fragments. In addition, SABBAC performances rely on a very small set of conformations that are selected using a structural alphabet encoding, and that rely much less than previous approaches on updates from new structures resolved. Finally, SABBAC performs reconstruction in reasonable time, although work is under progress to improve that point.

Future direction for the SABBAC service are in terms of introducing flexibility. Firstly, it could be of interest to implement some filter to rebuild only parts of the structure specified by the user. Secondly, we are also considering the possible automation of an interface with the side chain positioning facility using a more accurate positioning method.

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