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Performance comparison of computational methods for modeling alpha-helical structures

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Abstract Geometry optimization results are reported for secondary structural elements of small proteins and polypeptides. Emphasis is placed on how well molecular mechanics as well as semiempirical, ab initio, and density functional methods describe α -helical and related structures in purely theoretical models (Gly10, Ile10) as well as in realistic models (an α -helical region of calmodulin, and the complete structure of a small protein). Many of the methods examined here were found to provide unsatisfactory descriptions of the hydrogen-bonding interactions within polypeptide-type structures, as the α -helical canonical secondary structure motif was not reproduced accurately. Ab initio and DFT methods provided reasonable results only when solvation models were included, although Hartree-Fock failed even with solvation in one of the test cases; among the semiempirical methods, one of the PM6 implementations performed very well.

Keywords Peptide \cdot Protein \cdot Alpha helix \cdot Hydrogen bond \cdot DFT \cdot PM6

Introduction

Computational examinations of enzyme mechanisms utilizing methods such as density functional theory (DFT) or Hartree–Fock (HF) have traditionally been restricted—

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especially in metalloproteins containing transition metal ions-to smaller-size models of the active sites, due to the substantial computing resources required [1-5]. On the other hand, the reliability of empirical methods for describing protein structures has long been established [4, 6-9]. In recent years, the use of accurate solvation models and QM/MM-type techniques has allowed the influence of the protein environment to be accounted for more explicitly, although the less-relevant parts of the polypeptide are treated at levels of theory that are inferior to those applied to the active site itself [1-5]. Additionally, there have been constant efforts to improve the performance of lower levels of theory (semiempirical, molecular mechanics), so that they can be applied not just to the distant polypeptide regions but also to the actual active site [10-13]. At the current rate of evolution in computer performance, it may well be possible in the near future to apply DFT or (post-)HF methods to wholeprotein enzyme models. Indeed, single-point energies have already occasionally been computed for large-and even complete-protein models at higher levels of theory, such as DFT [13-25]. It is in this context that the study described in this paper sought to gauge how well some of the commonly used computational models predicted the geometries of polypeptide chains. This was considered to be a particularly challenging situation for several of these models. First, polypeptide architectures tend to rely primarily on weak, noncovalent interactions (mainly hydrogen bonding, although recent results have suggested significantly more important contributions from other factors, such as van der Waals dispersion interactions [26-28]). Second, methods requiring parametrization may not necessarily perform efficiently if they have not been parameterized, especially when they are applied to proteins. The present study sought to examine the degree to which α -helical geometries can be described by computational methods; it did not aim to predict the relative stability of this type of secondary structure compared to others.

Methods

The present study employed the following α -helical models: Gly₁₀, Gly₅₀, Ile₁₀, and Ile₅₀. Additionally, two experimentally known structures that were extracted from the Protein Data Bank (3CLN and 1ALG) were also examined computationally. These models were built within the Builder module of the Spartan software package [29], and were either capped at the N- and C-termini with hydrogen atoms (i.e., C-terminal COOH, N-terminal –NH₂) or left in ionized forms (carboxylate and ammonium). A distinctive feature of these structures is the regularity imposed by hydrogen bonds between the C=O and NH groups of different peptide bonds; the tables provided in this manuscript show the hydrogen-bond lengths extracted from these starting α -helical canonical structures.

Geometry optimizations were performed either in vacuum or with the CPCM continuum solvent model as implemented in the Gaussian09 software package [30]. Molecular mechanics calculations utilized the force fields Amber and UFF, as implemented in Gaussian09 [30] and Hyperchem [31]. The semiempirical PM6 method was employed as implemented in the Gaussian09 [30] and MOPAC [32] software packages, and the PM3 implementation from Hyperchem [31] was also used.

HF/3-21G* and density functional theory (M062X/6-31G**, M062/6-311+G**, BP86/6-31G**) computations were performed in Gaussian09 [31]. The CPCM solvation model was employed for HF and DFT calculations [30, 33]. Standard convergence criteria as defined in the respective software packages were employed.

Calculations performed with the Gaussian09 package utilized the default settings for the SCF cycles and geometry optimization, namely the fine grid (75,302) to numerically evaluate the integrals, a self-consistent field convergence of 10^{-8} hartrees, a maximum force of 0.000450 hartrees/bohr, an RMS force of 0.000300 hartrees/bohr, a maximum displacement of 0.001800 bohrs, and an RMS displacement of 0.001200 bohrs. The default SCF procedure uses a combination of EDIIS and CDIIS [34], with no damping or Fermi broadening. The nature of each stationary point after optimization was checked by calculating the harmonic vibrational frequencies to ensure that the stationary points found were genuine minima.

We recently reported a technical evaluation of the various optimization algorithms as employed with the PM6 method in the MOPAC2009 software package, using an α -helix of decaglycine as a test case [35]. Based on this extensive study, we chose to use the PM6-D2 Hamiltonian [36] and cpcm continuous solvent model (using a conductor-like screening model [37], a dielectric constant of water of 78.39, and a solvent radius of 1.3 Å) with the L-BFGS optimization method. The default optimizer in MOPAC2009

is Baker's eigenvector-following method [38]. The Davidon– Fletcher–Powel algorithm [39, 40] was the first quasi-Newton generalized method implemented in MOPAC; this was subsequently improved upon by the Broyden–Fletcher–Goldfarb– Shanno (BFGS) procedure [41–44]. An advanced variant of the BFGS optimizer is the "limited-memory BFGS" function minimizer, which calculates the inverse Hessian as needed [45–47], and is thus suitable for optimizing larger systems.

Note that, although we examined the geometries of α -helices, this does not imply that we expect them to be the only—or even the most stable—forms of secondary structure available for the peptides examined in the present study. It is, nevertheless, a basic biochemical principle that all simple peptides are able to form the two main types of secondary structure (assuming that less usual amino acids such as proline are not present), including the α helix. Geometry optimizations (M062x/ 6-31G**) reveal for instance that the 3₁₀ and α -helix geometries are within~5 kcal mol⁻¹ of each other energetically, and are close in energy to other forms as well. These issues relating to the relative stabilities of secondary structural elements are, however, a topic for ongoing investigations, and are not discussed further here.

Results and discussion

Gly₁₀

Geometry optimization results obtained for the neutral Gly_{10} model are shown in Table 1 and Fig. 1. Among the molecular mechanics methods, UFF leads to a strongly distorted structure (cf. Fig. 1), where the length of the helix drops from 13.9 Å in the initial canonical structure to 10.8 Å; further data on this result are therefore not shown in Table 1. The Amber force field (Hyperchem implementation) preserves the helix length to within 1 Å of the starting structure, but yields average hydrogen bonds that are ~0.25 Å longer than in the starting canonical model. Among the semiempirical methods, by far the best performance is provided by the MOPAC implementation of PM6: the values of hydrogen bonds remain very close to each other (1.9–2.1 Å) throughout the model, in contrast to the Gaussian PM6 and PM3 methods, which yield values as high as 4.5. The MOPAC PM6 also appears to perform well in terms of the total length of the helix: it is within 0.6 Å of the canonical geometry.

Helix lengths predicted by ab initio and DFT methods fall within ± 0.1 Å of the canonical geometry, with slightly better values obtained from solvated models. The lengths of the hydrogen bonds appear to be optimally modeled by HF in the solvated model: an average of 1.88 Å was obtained, compared to values of >2 Å obtained with the other methods

Neutral Gly ₁₀ ^a	Initial	Amber Hyperchem	PM3 Hyperchem	PM6 Gaussian	PM6 MOPAC	HF/ 2.21C*	M062X/	HF/ 2.21C*	M062X/
	Vacuum	Vacuum	Vacuum	Vacuum	Water	Vacuum	Vacuum	Water	Water
1–5	1.74	2.13	1.85	2.18	1.99	2.21	2.51	1.96	2.12
2–6	1.74	1.97	1.89	2.04	1.92	1.93	2.13	1.81	1.97
3–7	1.74	2.04	1.89	3.45	1.94	2.03	2.15	1.9	2.02
4-8	1.74	2	1.86	2.12	1.93	2.05	2.23	1.86	1.99
5–9	1.74	2.05	3.14	2.09	1.99	1.92	2.68	1.87	2.11
6–10	1.74	2.1	4.56	4.47	2.12	2.12	3.08	1.86	2.26
(CO…HN) _{avg}	1.74	2.05	2.53	2.72	1.98	2.04	2.46	1.88	2.08
Helix length	13.91	14.53	14.52	12.08	14.49	12.94	14.9	14.03	14.61

Table 1 CO...NH hydrogen-bond lengths and helix lengths for the neutral Gly₁₀ models

^a Amino acids whose NH/CO groups engage in hydrogen bonds are indicated in this column; numbering starts from the N-terminus of the peptide

^b Length of the helix, measured between the α -carbon atoms of the first and last amino acids, respectively

of its class. The M06 functional predicts two unusually long hydrogen bonds at the two termini of the helix in the unsolvated model but not in the solvated one. On the other hand, Table S1 of the "Electronic supplementary material" (ESM) shows that another density functional, BP86, completely fails to reproduce an α -helical structure in vacuum, even though it performs reasonably well when used together with a solvation model (average hydrogen bond length -1.95 Å).

One general observation is that all methods predict asymmetry in the helix, with the hydrogen bonds at the extremities of the chain displaying larger values than those in the middle of the chain. Furthermore, the lengths of the two hydrogen bonds at the two ends are not identical according to all of the methods employed here.

For the zwitterionic α -Gly₁₀ structure, Fig. 2 and Table 2 illustrate the starting point and selected optimized geometries. The Amber force field leads to loss of the α -helical character of the structure, with hydrogen-bonding distances increasing to as much as 7.8 Å. The UFF force field was found to give similar results to Amber (data not shown).

The zwitterionic Gly_{10} HF structure obtained in vacuum is entirely nonhelical (cf. Fig. 2). Far better results are



Fig. 1 Graphical representation of the optimized neutral Gly₁₀ geometries provided by various methods employed in the present study





obtained when solvation is included with the HF and DFT methods. The M06 results also illustrate the effect of using a larger basis set on the results: hydrogen bonds contract by 0.1 Å upon changing from $6-31G^{**}$ to $6-311+G^{**}$ —a change that is expected, since the larger basis set leads to a more accurate description of weak interactions such as hydrogen bonds. In contrast to the M06 data shown in Table 2, DFT data obtained for the Gly10 zwitterion with the BP86 functional show that even solvation cannot produce an intact

helical structure during geometry optimization, as the two terminal hydrogen bonds are elongated to more than 4 Å (Table S1 of the ESM).

Among the methods employed here, HF/3-21G* with water as solvent appeared to offer the best performance for the Gly₁₀ zwitterion, as it maintained an α -helical structure that had regular hydrogen bonds with lengths close to those expected for the canonical α -helix. However, MOPAC PM6 gave a similar level of performance (the average of

Table 2	CONH hydrogen-bond lengths and helix lengths for the zwitterionic Gly10 models. Labels are as in Table 1. The HF/3-21G**/vacuum
geometry	y is not listed, as the helix was completely destroyed upon geometry optimization

Zwitterionic Gly ₁₀ ^a	Initial Vacuum NH…OC	Amber Gaussian Vacuum NH…OC	PM6 MOPAC Water NH…OC	HF/3-21G* Water NH…OC	M062X/6-31G** Water NH…OC	M062X/6-311+G** Water NH…OC
1–5	1.74	3.75	2.06	2.04	2.25	2.24
2–6	1.74	4.6	1.94	1.84	2.04	1.99
3–7	1.74	7.75	1.95	1.89	2.02	1.97
4-8	1.74	2.01	2	1.88	2.01	1.98
5–9	1.74	2.56	1.98	1.89	2.28	2.09
6–10	1.74	6.41	2.03	1.96	2.65	2.41
(CO…HN) _{avg}	1.74	4.91	1.99	1.92	2.21	2.11
Helix length ^b	13.91	12.52	14.44	14.09	14.85	14.78

Neutral ILE ₁₀	Initial Vacuum NH…OC	Amber Hyper Vacuum NH…OC	PM3 Hyper Vacuum NH…OC	PM6 Gaussian Vacuum NH…OC	PM6 MOPAC Water NH…OC	HF/3-21G* Vacuum NH…OC	M062X/6-31G** Vacuum NH…OC	M062X/6-31G** Water NH…OC	HF/3-21G* Water NH…OC
1–5	1.72	2.13	2.43	2.16	1.94	4.09	2.05	1.97	2.28
2-6	1.72	2.01	2.53	2.11	2.01	2.06	2.16	2.07	1.99
3–7	1.72	2.26	2.55	2.21	2.05	2.57	2.36	2.2	2.28
4-8	1.72	2.18	2.56	2.14	2.03	4.84	2.26	2.14	2.32
5–9	1.72	2.13	2.8	2.84	2.06	3.52	3.09	2.18	2.11
6–10	1.72	2.05	3	3.27	2.04	3.49	3.32	2.14	2.2
(CO···HN) _{avg}	1.72	2.12	2.65	2.45	2.02	3.42	2.54	2.11	2.2
Helix length	13.86	14.43	14.65	15.05	14.26	15.89	14.96	14.42	14.77

hydrogen-bond length was only 0.07 Å longer than that for $HF/3-21G^*$ with water as solvent).

α -Gly₅₀

Figure S2 and Table S2 of the ESM illustrate results obtained for a neutral Gly₅₀ α -helix, in order to gauge the extent to which the results obtained for the decaglycine model can be used to derive trends observable in larger models. The HF/3-21G* approach again yields results close to those expected for the canonical α -helix, with the MOPAC PM6 method performing slightly worse than for the decaglycine model (average hydrogen-bond length of 2.22 Å in Gly₅₀ vs. 1.98 Å in Gly₁₀). The Gaussian09 implementation of the PM6 method reshapes the helix so that a hydrogen bond is formed in the optimized geometry between amino acids 1 and 6, as opposed to the hydrogen bond between amino acids 1 and 5 in the starting geometry (with this trend conserved throughout the helix: 2-7, 3-8, 4-9 instead of 2-6, 3-7, 4-8). Results for the UFF method are also listed in Table 1; in agreement with what was seen in the decaglycine model, hydrogen bonds are predicted to be too long (2.7 Å, almost 1 Å longer than the expected value), and the helix length is also distinctly different from the canonical value as well as the value computed with HF/ 3-21G*. Just as was seen for the decaglycine models, all of the methods indicated that the hydrogen bonds were asymmetric throughout the helix.

Model data for the zwitterionic Gly₅₀ are shown in Table S3 and Fig. S3 of the ESM. The MOPAC PM6 method again performs remarkably well both in terms of hydrogen bond length and overall helix length. The Amber and UFF force fields each predict unreasonably long hydrogen bonds at the ends of the helix (2.6–6 Å); also, the helix lengths predicted by these force fields differ by more than 10 Å from the canonical value. The Gaussian09 implementation of the semiempirical method PM6 rearranges the α -helical structure in the same way as it did for the neutral model, so that 1–6 hydrogen bonds are preferentially formed over 1–5; furthermore, the N- and C- termini of the chain are eliminated from the molecule in the form of NH₃ and CO₂ during geometry optimization.

α-Ile10

Isoleucine-containing α -helices were also examined to check the extent to which side chains larger than that of glycine affect the performance of each computational method in describing

Table 4 CO···NH hydrogen bond lengths and helix lengths for the Ile_{10} zwitterion. Labels are as in Table 1

Zwitterionic Ile ₁₀	Initial Spartan	Amber Hyper Vacuum NH…OC	HF/3-21G* Gaussian Vacuum NH…OC	PM6 MOPAC Water NH…OC	HF/3-21G* Gaussian Water NH…OC	M062X/6-31G** Gaussian Water NH…OC
1–5	1.72	2.06	1.86	2.01	3.76	2.04
2–6	1.72	2.05	4.43	2.01	1.99	2.09
3–7	1.72	3.58	5.88	2.04	2.34	2.19
4-8	1.72	3.9	4.38	2.09	3.96	2.16
5–9	1.72	3.13	2	2.09	2.09	2.34
6–10	1.72	3.24	3.43	2.03	2.46	2.37
(CO…HN) _{avg}	1.72	-	-	2.04	2.76	2.19
Helix length	13.86	14.34	-	14.4	15.62	14.56





this element of secondary structure. Results for Ile_{10} (Tables 3 and 4 and Figs. 3 and 4) and Ile_{50} (Table S4) were generally similar to those obtained for the glycine models.

Table 3 illustrates that for the neutral Ile_{10} model, the Amber force field performs as well as it did for Gly_{10} . The relative performance levels of the semiempirical





 Table 5
 Alpha helix CO···NH

 bond lengths for the calmodulin
 fragment

Long loop of 3CLN	PDB	Amber Gaussian Vacuum	PM6 MOPAC Water	HF/3-21G* Gaussian Vacuum
	NH…OC	NH…OC	NH…OC	NH…OC
Phe65-Leu69	1.79	1.8	1.96	2.29
Pro66–Thr70	2.19	2.01	2.02	2.01
Glu67-Met71	2.06	1.82	1.95	2.34
Phe68-Met72	1.79	1.92	1.79	1.89
Leu69–Ala73	2.04	1.99	2	1.92
Thr70–Arg74	2.74	1.72	1.99	2.23
Met71-Lys75	2.21	2.72	1.87	1.92
Met72-Met76	2.32	5.44	1.98	2.09
Ala73–Lys77	2.6	6.5	1.96	1.98
Arg74–Asp78	2.67	3.26	1.88	1.84
Lys75–Thr79	2.94	4.02	1.82	3.37
Met76–Asp80	2.89	1.86	1.92	1.77
Lys77-Ser81	2.67	2.02	1.99	1.94
Asp78–Glu82	2.31	6.29	1.92	4.1
Thr79–Glu83	2.84	3.26	1.88	2.64
Asp80–Glu84	3	2.39	2.01	1.83
Ser81–Ile85	2	2	2.03	1.96
Glu82–Arg86	2.04	2.71	1.93	1.96
Glu83–Glu87	2.45	1.98	1.86	1.73
Glu84–Ala88	2.12	3.41	1.93	1.93
Ile85–Phe89	2.31	3.67	1.92	2.58
Arg86–Arg90	2.71	3.11	1.88	1.66
Glu87–Val91	2.27	-	2.19	1.85
Ala88–Phe92	1.89	-	2.5	3.87
(CO…HN) _{avg}	2.09	2.18	1.97	1.95
RMSD	-	4.98	1.23	3.74

methods are also the same: the only reasonable result (monotonous hydrogen-bond lengths that are well below 2.5 Å) is provided by MOPAC PM6. On the other hand, the HF method, which described Gly10 reasonably well, fails here, yielding hydrogen-bond lengths as long as 4.8 Å. The M06 functional appears to be more reliable in this respect, as the average hydrogen-bond length is ~2.5 Å in both Gly₁₀ and Ile₁₀.

Table 4 illustrates geometry optimization for the Ile_{10} zwitterion. The Amber force field fails to provide reasonable hydrogen-bonding geometries, while the MOPAC PM6 once again provides reasonable results. Unexpectedly, the HF method fails not only in vacuum but also in solvent: in the latter case, hydrogen bonds as long as 4 Å are predicted, as opposed to the reasonable geometry predicted for Gly₁₀ (see Table 2). By contrast, the M06 functional with solvation provides a distinctly more reasonable geometry and is, in this respect, the one method—alongside Mopac PM6— that provides reasonable descriptions for all of the models examined thus far.

α -Ile-50

Table S4 and Figs. S4 and S5 of the ESM show data for neutral and zwitterionic α -Ile₅₀ structures. With the Amber force field, the two ends of the helix are distorted (with hydrogen-bond lengths as long as 5.4 Å), but the rest of the chain retains reasonable hydrogenbond lengths for an α -helix. For the neutral structure, the UFF force field yields hydrogen bonds of >3 Å (data not shown).

Calmodulin model

Calmodulin shows α -helical regions, and thus represents a reasonable test case to use to check the extent to which the results obtained for models such as Gly10/50 and Ile10/50 can be extrapolated to experimentally known peptide/protein structures. The central part of calmodulin features a long α -helix (28 residues, overall charge –2). This portion of the protein was extracted from the X-ray diffraction structure,

Fig. 5 Graphical representation of the optimized 3CLN longloop geometries obtained by various methods



pdb code 3CLN [48], and used in computations as detailed in Table 5. The N- and C-termini were modeled as ammonium and carboxylate groups, respectively, and an overall charge of -2 was assigned (as determined by the number of Asp/Glu and Lys/Arg amino acids, whose side chains were modeled as charged -1 and +1, respectively). Three methods with reasonable computational costs were employed in this case. As shown in Table 5 and Fig. 5, the Amber method does not model the helical structure properly, predicting a well-defined bending point midway through the structure (cf. Fig. 5) and several unacceptably long (>3 Å) hydrogen-bonding distances (amino acids 74-85). A bent helix is also predicted by the HF approach, which unfortunately also eliminates the C-terminal carboxylate as CO₂ upon geometry optimization. The MOPAC PM6 method is the only method that completely preserves the α -helical structure; however, in this case, the tendency to give a canonical structure is exaggerated: even though there are hydrogen-bond distances as long as 3 Å in the experimental structure, the PM6 method predicts that all of the hydrogen bonds in the structure are in the \sim 1.9–2.0 Å range.

1ALG

The PDB structure with a code of 1ALG is a protein composed of 24 residues with a global α -helix structure (although some of the terminal amino acids are not part of the helix) [49]. Computations were performed on the complete structure of this protein without any truncation; data are shown in Table 6 and Fig. 6. Interestingly, methods that gave less impressive results for the smaller/simpler models discussed above (e.g., the Gaussian implementation of PM6, or HF/3-21G* in vacuum) are found to reproduce the regularity of the helical structure of the complete, experimentally known, protein 1ALG reasonably well. In fact, even where the experimental structure shows deviations from an α -helix, such as for the

 Table 6
 Alpha helix CO···NH hydrogen-bond lengths for the 1ALG structure

	1ALG	PDB	Amber Gaussian Vacuum	PM6 Gaussian Vacuum	PM6 MOPAC Water	HF/3-21G* Vacuum	HF/321G* Water
	Charge -1	NH…OC	NH…OC	NH…OC	NH…OC	NH…OC	NH…OC
1	Met8–Gly4	2.58	-	2.28	1.84	1.88	1.93
2	Leu9–Cys5	2.29	2.95	2	1.97	2.13	2.13
3	Glu10-Asp6	1.94	1.82	2	1.95	2.02	1.83
4	Gly11–Glu7	2.14	2	2.27	2.21	1.94	1.93
5	Phe12-Met8	2.33	1.8	2.02	1.82	1.85	1.9
6	Ala13-Leu9	2.02	1.88	2.25	1.88	1.9	1.84
7	Val14–Gln10	1.94	1.84	2.03	2.02	1.91	1.81
8	Ala15–Gly11	2.14	1.93	2.01	2	1.94	2.17
9	Val16–Phe12	1.83	1.85	2	2.01	1.89	1.97
10	Lys17–Ala13	2.25	2.05	2.21	1.93	2.34	1.95
11	Met18-Val14	1.91	3.78	1.85	1.8	1.85	1.97
12	Gly19-Ala15	1.74	2.04	1.89	2.14	1.74	1.91
	(CO…HN) _{avg}	2.09	2.17	2.07	1.96	1.95	1.95
	RMSD	-	2.74	2.38	1.18	1.59	1.30

first two hydrogen bonds (which are longer) and the last hydrogen bond (which is distinctly shorter), the computational methods fail to predict these deviations.

Summary

Several computational methods for performing the geometric optimization of simple α -helical protein models were tested.

Methods that performed well with larger protein structures appeared to be unable to describe simpler α -helices properly. One of the implementations of the semiempirical method PM6, alongside density functional calculations including solvation (with the M06 functional performing better than the others), appear to provide reasonable results for all of test cases studied here, although the hydrogen-bonding distances obtained using these methods were typically within 0.2–0.6 Å of those expected for the canonical structure.

Fig. 6 Graphical representation of the optimized 1ALG geometries obtained by various methods



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