Article

# COSMOsim3D: 3D-Similarity and Alignment Based on COSMO Polarization Charge Densities

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Ⓢ *Supporting Information*

ABSTRACT: COSMO σ-surfaces resulting from quantum chemical calculations of molecules in a simulated conductor, and their histograms, the so-called σ-profiles, are widely proven to provide a very suitable and almost complete basis for the description of molecular interactions in condensed systems. The COSMOsim method therefore introduced a global measure of molecular similarity on the basis of similarity of σ-profiles, but it had the disadvantage of neglecting the 3D distribution of molecular polarities, which is crucially determining all ligand−receptor binding. This disadvantage is now overcome by COSMOsim3D, which is a logical and physically sound extension of the COSMOsim method, which uses local σ-profiles on a spatial grid. This new method is used to measure intermolecular similarities on the basis of the 3D representation of the surface polarization charge densities σ of the target and the probe molecule. The probe molecule is translated and rotated in space in order to maximize the sum of local σ-profile similarities between target and probe. This sum, the COSMOsim3D similarity, is a powerful descriptor of ligand similarity and allows for a good discrimination between bioisosters and random pairs. Validation experiments using about 600 pharmacological activity classes in the MDDR database are given. Furthermore, COSMOsim3D represents a unique and very robust method for a field-based ligand−ligand alignment.
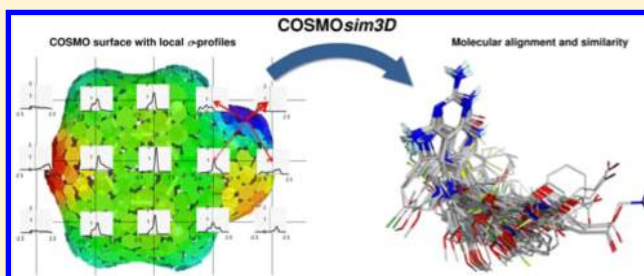
■ INTRODUCTION

In the framework of the Conductor-Like Screening Method for Realistic Solvation (COSMO-RS[1−3]), which allows for the prediction of a broad range of fluid phase thermodynamic properties on the basis of quantum chemical calculations for solutes and solvents, the COSMO[4] surface polarization charge density σ, and its molecular surface histograms, the so-called σ-profiles, have been proven to be excellent descriptors for the quantification of the most important kinds of molecular interactions in the liquid phase, such as polar interactions, hydrogen bonding, and hydrophobicity.[5−8] For the reader not familiar with the concept of σ-profiles, detailed introductions can be found in any of the COSMO-RS publications[1−3] and in special detail in ref 1, pages 85ff. The exceptional suitability of σ for the quantification of hydrogen bond interactions has been further confirmed in a recent quantum chemical study.[9] Because the same intermolecular interaction modes that govern fluid phase thermodynamics are also responsible for ligand−receptor interactions, it is most plausible that a σ-based description of ligand−ligand similarity or ligand−receptor interactions should be very promising. On the basis of these considerations, we presented COSMOsim,[10,11] which uses the molecule-specific global σ-profile disregarding the spatial distribution of the polarization charge density to measure ligand−ligand similarities. This method was shown to provide useful discrimination of bioisosteric and random ligand pairs,

especially for smaller molecules. Besides speed, one of the major advantages of COSMOsim is that it naturally supports scaffold hopping by using the molecular COSMO-RS σ-surface instead of the molecular structure. Further, analogy-based QSPR based on COSMOsim delivers powerful models for properties that are mainly governed by isotropic interactions, like logS, logP, logBB, etc.

However, the selective binding of ligands to receptors is known to be based on multiple strong interactions, and the 3D arrangement of the interaction sites of a ligand thus plays a crucial role. Such information is not included in ordinary σ-profiles. As a result, in anisotropic protein−ligand interactions, COSMOsim tends to retrieve false positives along with true positives. This clearly results from the fact that the global σ-profiles do not contain any information about the spatial distribution of the polarization charge densities on the molecular surface. The use of a grid of local σ-profiles should overcome this deficiency. In order to generate 3D σ-profiles, we project the COSMO-σ-surface onto a 3D grid of a certain resolution and thus generate local σ-profiles (LSPs) for the grid points (see Figure 1 and Materials and Methods section). While the target molecule is fixed on the grid, the probe molecule is represented in different orientations and is
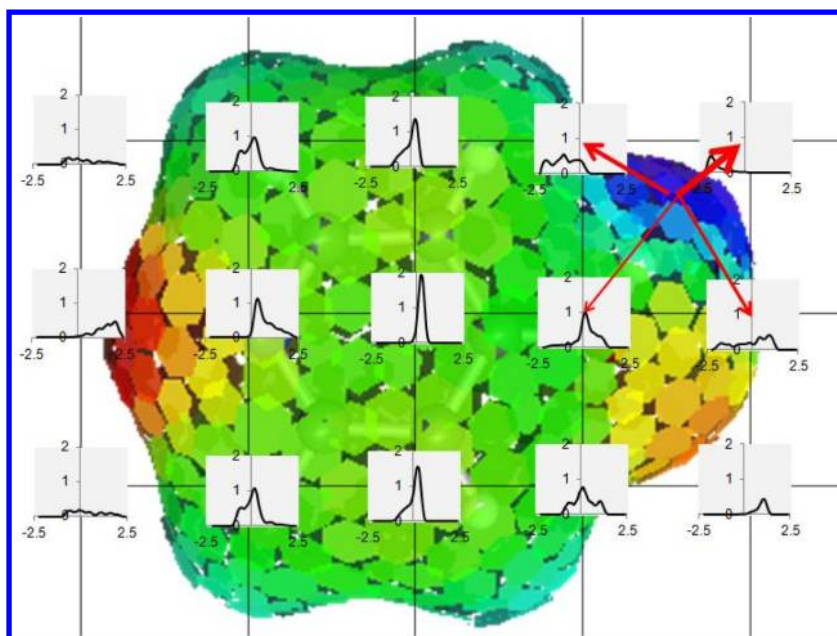
Figure 1. Schematic visualization of the construction of local σ-profiles, in 2D for clarity. Each surface segment gets assigned to the local σ-profiles of the neighboring grid points.

translated and rotated to achieve the maximum overlap of the local σ-profile similarities. In this way, the optimal alignment and the COSMOsim3D similarity are obtained.

COSMOsim3D requires COSMO-σ-surfaces of target and probe molecules that can be obtained from a DFT calculation using different quantum chemical programs with suitable COSMO implementations. The TURBOMOLE/COSMO implementation[7,8] is used throughout this paper. It is advisable to take multiple conformers of each ligand into account, as their spatial σ-profiles are much more sensitive to conformational changes than the global σ-profiles used in the original COSMOsim method.

DFT calculations of large scale databases can still be very time consuming, in particular with geometry optimization. We found that proper conformation analysis at MM level or at semi-empirical QM-level followed by a single-point COSMO calculation at BP-SVP-COSMO-SP level delivers good-quality σ-surfaces in reasonable time, i.e., within a few minutes per conformer. With a recent update, the COSMOfrag program[12,13] is able to provide approximate COSMO σ-surfaces (CF-COSMO files) on the basis of just a 3D geometry of the ligand and the COSMOfrag database of precalculated σ-profiles. This shortcut reduces the computation time for the COSMO file generation to less than a second per conformer and, thus, makes the COSMOsim3D method presented here suitable for high-throughput screening applications.

In the following, we first present the COSMOsim3D method in detail. Then we describe the computational methods used throughout three computational experiments. Then we discuss the results of a broad validation study of the COSMOsim3D method on the discrimination of true and random bioisosteric pairs from the BioSter database.[14] This is followed by a large scale study on identification of compounds with identical activity classification in the MDDR database.[15] As a final validation, we present the results of an automated alignment of ligands using COSMOsim3D, followed by a summary and outlook.

MATERIALS AND METHODS

COSMOsim3D Method Description. The starting point of the COSMOsim3D method is the COSMO files generated by quantum-chemical calculation combined with the continuum solvation model COSMO (with ε = infinity, i.e., in the conductor limit). These files contain the information about the position (x_i, y_i, z_i), the areas a_i and the COSMO polarization charges q_i for all COSMO surface segments. For each segment i, a locally averaged polarization charge density σ_i is calculated according to the standard procedure,[1] using an averaging radius of 0.5 Å. For standard σ-profiles, as used in the COSMOsim method, a histogram with σ-bin width of $\delta^\sigma = 0.1$ e/nm^2 is generated from all segments of a molecule. For the generation of a σ-profile, the area of each segment i is associated to the two neighboring σ-grid centers of the actual value σ_i, denoted as σ_{i+} and σ_{i-}, according to σ-distance weights

$$w_{i+}^\sigma = (\sigma_{i+} - \sigma_i)/\delta^\sigma$$
$$w_{i-}^\sigma = (\sigma_i - \sigma_{i-})/\delta^\sigma \qquad (1)$$

This weighting generates a smooth and charge-conserving assignment of the COSMO surface segments to the histogram, ensuring that the integral of the σ-profile is the total surface of the molecule and that the σ-weighted integral, i.e., the first moment of the histogram, is the sum of the original COSMO charges, i.e., the negative of the total charge of molecule.

Instead of generating just one such one-dimensional σ-profile for the entire molecules, in COSMOsim3D we generate a local one-dimensional σ-profile at each position of a regular 3D grid, i.e., altogether a four-dimensional histogram, with three Cartesian dimensions (x,y,z) and σ as fourth dimension. The grid size in space is set to $\delta^x = \delta^y = \delta^z = 1$ Å by default and used this way throughout this paper. For the 16 neighboring grid points of segment i with coordinates (x_i,y_i,z_i,σ_i) in the 4D space, the weights w^x, w^y, w^z, and w^σ are computed in complete analogy to eq 1, and the segment area is assigned to the 16 neighbor grid points according to the product of the four weights. This weighting ensures a smooth linear interpolation
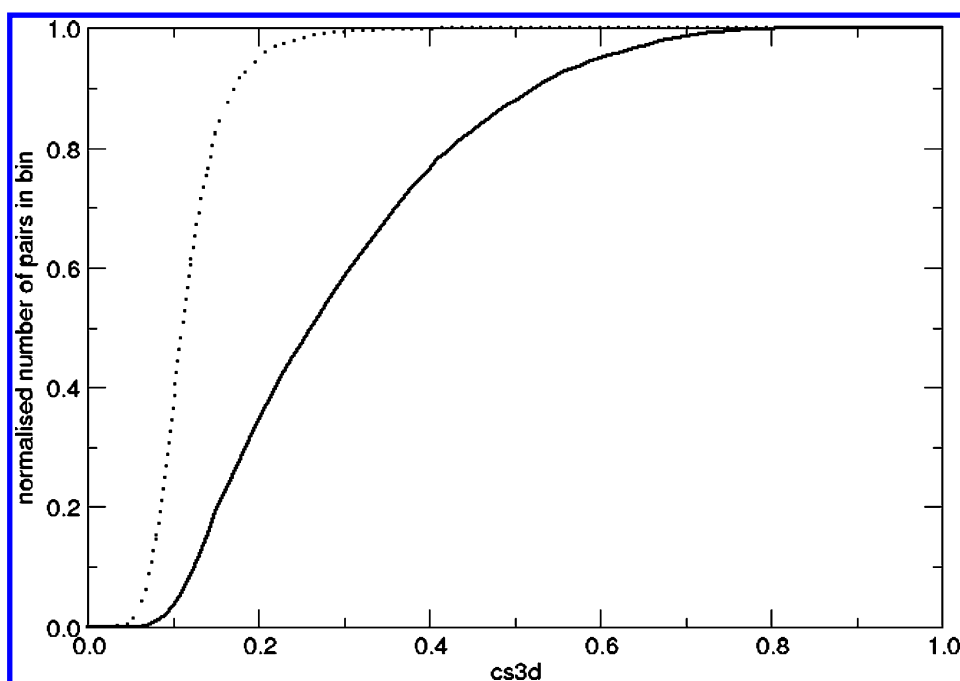
Figure 2. cs3d separates random pairs (dotted line) from bioisosteric pairs (solid line).

of the segment assignment if a surface segment is moved between the grid points.

In COSMOsim3D, the first molecule, called target further on, is initially shifted in a way that its COSMO surface center is located on the center of the cubic grid, followed by the calculation of the 3D-σ-profile (sp3d1) on the grid points. Then the same is done for the second molecule, called probe further on, resulting in a 3D-σ-profile sp3d2. Next the 3D-σ-similarity COSMOsim3D is calculated as a weighted sum over the σ-similarities of the local σ-profiles according to

$$\text{COSMOsim3D}(\text{sp3d1, sp3d2}) = \frac{\sum_{ix,iy,iz}(a1(ix,\,iy,\,iz) + a2(ix,\,iy,\,iz))\text{SMS}(\text{sp3d1}(ix,\,iy,\,iz),\,\text{sp3d1}(ix,\,iy,\,iz))}{\sum_{ix,iy,iz}(a1(ix,\,iy,\,iz) + a2(ix,\,iy,\,iz))} \tag{2}$$

where $a1(ix,iy,iz)$ and $a2(ix,iy,iz)$ are the total intensities of the local target and probe σ-profiles at the grid point, respectively. The denominator is identical to the sum of the total surface areas of target and probe. The SMS (sigma-match similarity) calculations are performed according to ref 10, with the default parameters derived in that paper.[12]

After the initial evaluation of COSMOsim3D, the position and orientation of the probe is optimized in order to maximize COSMOsim3D. This is done by a trial-and-error line search in the direction of each of the three unit-translation and unit-rotations, with minimum steps of 0.01 Å and 0.1°, respectively. After each translational or rotational step, COSMOsim3D is re-evaluated and the step is accepted if it leads to an increase in COSMOsim3D. After convergence of the optimization, more optimizations are performed, each starting from another start position and start orientation of the probe. During our test, we found that 25 reasonably chosen rotations were sufficient to find the optimal superposition and the maximum COSMO-sim3D value in essentially all cases. After tuning the optimization algorithm (for more details, see Appendix A), the typical time demand for the evaluation of the optimal COSMOsim3D similarity for a pair of drug-like molecules thus is on the order of 5 s on a single 2.5 GHz CPU.

COSMO File Generation. COSMO files were generated in two ways: In the conventional way of COSMO file generation, we used the TURBOMOLE program[16,17] for a single-point BP-SVP-COSMO density functional calculation in combination with the COSMO solvation model implemented in TURBO-MOLE. These calculations typically take about 3−5 min for one drug molecule on a single 2.5 GHz CPU. Alternatively, we used the novel CF-COSMO capability of COSMOfrag,[13,14] which generates approximate COSMO files from fragments taken from the COSMOfrag database (CFDB) of precalculated COSMO files. Technically, this is done in the same way as COSMOfrag usually generates approximate σ-profiles, but with the difference that for the CF-COSMO file generation a 3D-input geometry is required. The CF-COSMO module takes for each atom of the input structure the COSMO surface segments from the most suitable precalculated compound stored in the CFDB and translates and rotates this into the local coordinate system of the target atom. In this way, COSMO files are generated that do not represent a perfectly closed molecular surface but which have roughly the right surface segments and polarization charge densities in roughly the right spatial position. For more details of the CF-COSMO method and a comparison of CF-COSMO files with original COSMO files, see Appendix A.

Tautomer and Conformer Generation. SDF versions of the BioSter and MDDR databases were converted into SMILES using OpenBabel.[18] The chemical structures were desalted and neutralized, if possible. Tautomers were generated for each compound on the basis of the canonical SMILES using an in-house program. For each tautomer, up to three main
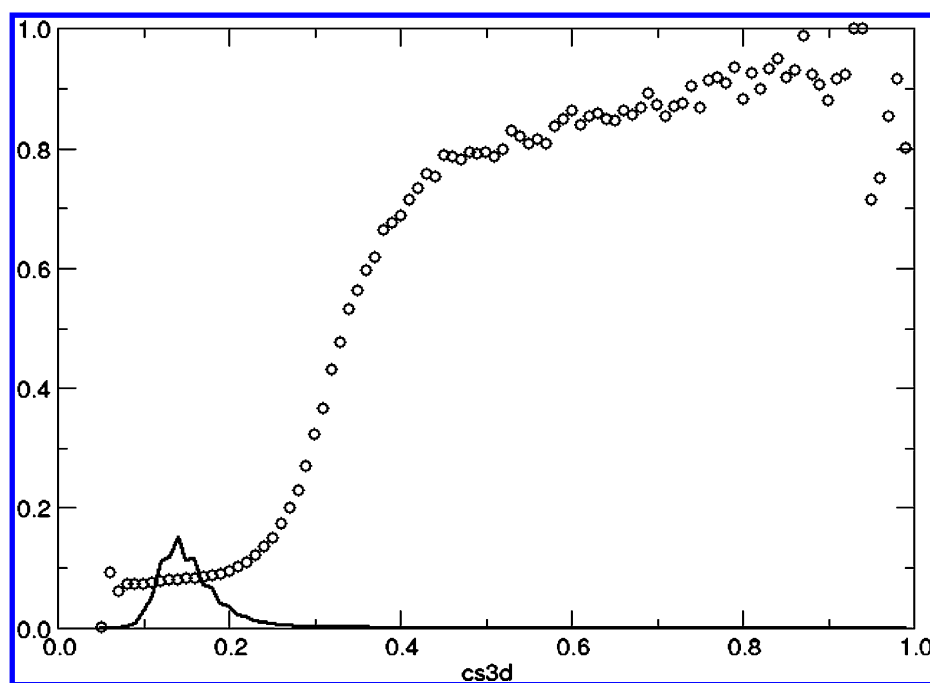
Figure 3. Fraction of pairs sharing at least one MDDR activity class as function of cs3d after filtering with the limit COSMOsim >0.9.

conformers were generated using the Msmab and Mcnf options of the MOLOC program.[19]

COSMOsim Calculations. The COSMOsim module in the COSMOfrag program[10,13] was used to compute the complete similarity matrix of 214,513 × 214,513 compounds. For each compound, a list with the potential bioisosters was generated using a cutoff for the COSMOsim value of >0.9, which corresponds to a reasonable cutoff derived from our previous work.[10] This preselection was then used to run COSMOsim3D on these tentative bioisosteric pairs of compounds.

COSMOsim3D Calculations. COSMOsim3D was used as described above to calculate the σ-surface-based similarity of the tentative bioisosteric pairs, whereby all conformers of all tautomers of the target compound were compared with all conformers of all tautomers of the probe compound. The highest COSMOsim3D value obtained from this procedure was assigned as COSMOsim3D value to the corresponding compound pair. These calculations were done in two ways on the basis of the original COSMO files and with the approximate CF-COSMO files.

■ RESULTS AND DISCUSSION

Validation Study on Bioisosteric Pairs. A screening experiment on the separation of true and false (random) bioisosteric pairs was performed using the 5089 bioisosteric pairs and the 5589 random pairs provided in the BioSter database.[14] In this test, only the main conformation was used for each compound. It should be noted that the random pairs are not proven to be nonbioisosteric. Hence, a small percentage of true bioisosteric pairs may be contained in this subset.

The COSMOsim3D similarity (cs3d) has been calculated for all pairs of the two subsets. Figure 2 shows the accumulated fraction of pairs as a function of cs3d. For the random pairs, the maximum slope is at ∼0.1, and 95% of the bioisosteric pairs have cs3d <0.2. The curve for the true bioisosteric pairs is clearly shifted toward higher cs3d values. Its maximum slope is at ∼0.25. A separation at a value of 0.2 would leave ∼5% false

positives in the set of random pairs at the cost of ∼30% false negatives in the bioisoster pair subset.

Validation Study on Recognition of Activity Classes. The MDDR database[15] was used to extract 214,513 chemical structures and their assignment to 580 different activity classes (ACTIV_INDEX). Each compound may be assigned to more than one class. Canonical SMILES strings were generated using OpenBabel (version 2.2.3).[18] Approximate σ-profiles, one per compound, were generated using COSMOfrag on the basis of the canonical SMILES. In addition, tautomers and conformers have been generated for each compound, and their CF-COSMO files were generated as described in the Materials and Methods section.

As a first step, for all 214,513 × 214,512/2 pairs, i.e., for 23 × 10⁹ pairs, the COSMOsim similarity has been calculated. Pairs with COSMOsim <0.9 have been skipped. A total of 59,144,819 of the pairs passed this threshold. In the original data set, the probability that a pair shares at least one activity class was 6.4%. In the remaining data set, this probability increased to 24.2%, which corresponds to an overall enrichment factor of 3.8 resulting from the COSMOsim >0.9 filter.

Originally, we planned to calculate the COSMOsim3D similarity (cs3d) for all of the 59,144,819 remaining pairs. Because an average 2.8 tautomeric/conformeric forms of each compound, i.e., ∼8 combinations per compound pair, had to be considered, this would have corresponded to ∼480 × 10⁶ cs3d evaluations. Finally, this turned out to be too time consuming. Observing no significant changes in the fraction of activity pairs anymore (Figure 3), and having spent already 180 days on ∼50 CPUs, we stopped the experiment after evaluation of ∼30% of these pairs. It is worth noting that the average time demand per cs3d evaluation for a pair thus was only ∼5 s on one CPU.

Figure 3 shows the overall statistics from this experiment. The solid line shows the distribution of cs3d values within all tested pairs. Although the COSMOsim similarity for all pairs was >0.9, the cs3d for most pairs is only in the range of 0.1−0.2 . The dots mark the percentage of pairs sharing at least one
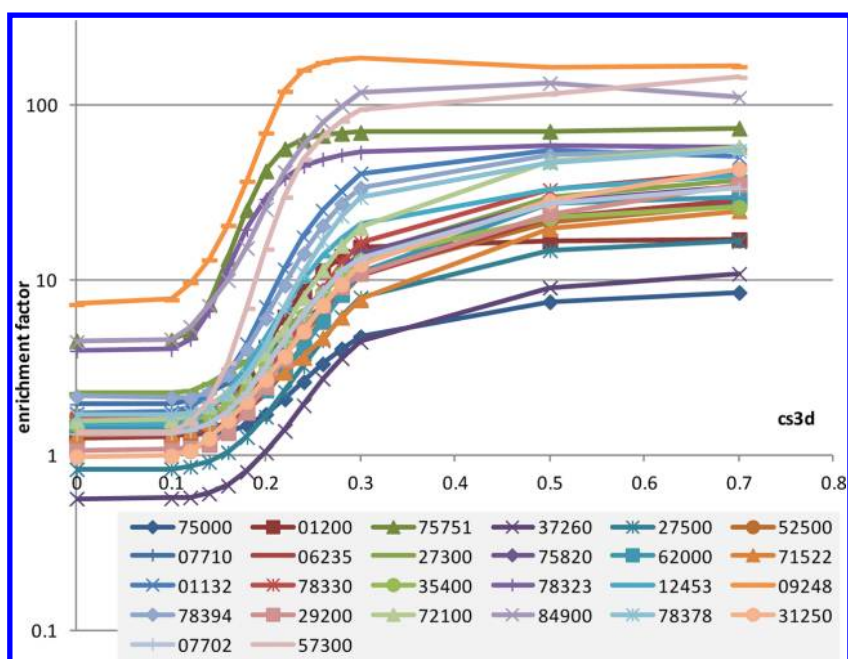
Figure 4. Enrichment factors of pairs belonging to the same MDDR activity class as a function of cs3d for the 26 most populated classes, after filtering with the limit COSMO*sim* >0.9. The MDDR activity class code is given in the inset. Explanations of the MDDR activity class codes are given in Table S1 of the Supporting Information.

activity class at each binned value of cs3d. At low cs3d values, i.e., for the largest part of the pairs, this value is only in the range of 7−10%, i.e., essentially not much higher than the value of 6.4% for random pairs from the MDDR database. A strong increase can be observed at cs3d ~0.3. A 50% chance for sharing at least one activity class is reached at cs3d = 0.33 and 80% are reached at cs3d = 0.45, further increasing to 100% at higher values of cs3d, with increasing statistical noise because the number of pairs having such large values of cs3d becomes very small.

Figure 4 shows the cs3d dependence of the individual enrichment factors for the 26 most populated activity classes in MDDR compared to the population of these pairs in the complete MDDR. All enrichment factors strongly increase in the range of 0.1−0.3 and mainly reach a plateau at higher values. In detail, interesting differences among the different classes can be observed with respect to the initial value, i.e., the enrichment factor already achieved through the COSMO*sim* >0.9 preselection, with respect to the exact point of maximum increase and the maximum achievable enrichment factor. However, an investigation and discussion of these details are beyond the scope this paper.

COSMO*sim3D* Alignment Study. Obviously, COSMO-*sim3D* can be used not only for pairwise alignment but also for the alignment of larger data sets. For such, we have additionally implemented a self-consistent alignment mode in which the first m of a list of n compounds can be defined as targets. In this mode, COSMO*sim3D* starts with the alignment of compound C2 vs compound C1. Then it simultaneously aligns compound C3 vs (C1 + C2)/2, which means that on each grid point the average of the similarities to the two previously aligned compounds is used. This is continued to the last target, i.e., Cm, which is aligned vs the average of the $m − 1$ previous targets. Then we start again at C1, align it to the other $m − 1$ targets and continue until no significant increase in the overall similarity is observed throughout an entire loop. Finally, the

remaining $m − n$ compounds get aligned vs the average of the n target compounds. Furthermore, in order to avoid any potential bias resulting from starting with the C1−C2 pair, a superself-consistent alignment mode is introduced, in which the procedure of the self-consistent alignment is repeated, starting with all subsequent pairs instead of starting just with C2 vs C1 alignment. The best of the resulting self-consistent alignments of the targets is then used in order to align the remaining compounds.

As an example, we applied this automated unsupervised alignment to the ACE data set from the Sutherland data collection.[20] For all 114 molecules, COSMO files were generated by single-point DFT/COSMO calculations, conserving the geometries given by Sutherland. After randomization of the positions and orientations of all compounds, the alignment was carried out as a superself-consistent alignment with respect to the three molecules considered as targets by Sutherland on this data set, i.e., with MOL_09, MOL_18, and THIOL_14. We used a grid size of 1.0 Å and 80 starting orientations per molecule.

The resulting alignment is shown in Figure 5 (left), with the Sutherland alignment given for comparison on the right. At
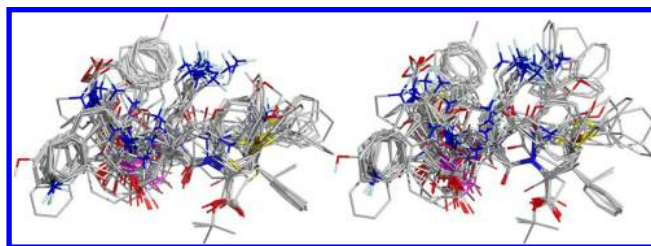

Figure 5. Alignment of the 114 ACE ligands from the Sutherland data set: (left) generated by COSMO*sim3D* with superself-consistent treatment of the three target ligands and (right) alignment as generated by Sutherland.

least visually, the cs3d alignment appears to be more consistent than the alignment provided by Sutherland. The performance of these alignments in the context of 3D-QSAR will be described in a forthcoming paper.[21]

On the example of 12 compounds from this cs3d aligned ACE data set, starting with the three targets and completed with representatives of the different compound classes, we show in Figure 6 a novel visualization mode of the 3D similarity
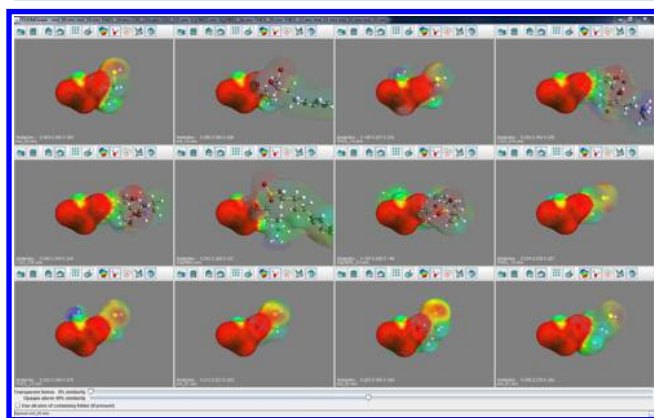

Figure 6. Screenshot of the cs3d-multiview mode in COSMOview. The same 3D-rotation and zoom operations apply to all visualized molecules; the degree of surface transparency depends on the local values of the cs3d similarity.

of molecules, called cs3d-multiview mode further on. The COSMO surfaces with embedded ball&stick structures get visualized for k compounds in k separate windows, which are subject to the same 3D-rotation and zoom operations. The local values of the cs3d similarity are used in order to control the surface transparency. By default regions with high similarity are shown with low transparency and thus have intensive colors, while areas with low similarity get high transparency, so that in these regions mainly the ball&stick structure can be seen. We consider this visualization mode of molecular similarity as an interesting auxiliary tool for investigating sets of aligned compounds, although we are aware that the number of compounds that can be visualized simultaneously in this way is limited by the considerable memory requirements.

CONCLUSIONS

By the presented COSMO*sim3D* method, the rich and consistent information content of the COSMO surface polarization charge densities with respect to all physiologically relevant intermolecular interaction modes can be used for a 3D-alignment and 3D-similarity measure, which allows for a very good separation of true bioisosteric pairs from random pairs. Hence, COSMO*sim3D* can be used as a powerful novel tool to search for bioisosteric analogues of known biologically active compounds. A special strength is the sound theoretical and almost ab initio foundation of the underlying COSMO σ-surfaces, which allow for their application to very different and novel chemical situations. Another advantage is the exclusive dependence of COSMO*sim* on the surface polarization charges, allowing for the detection of physiological similarity of chemically very different structures and hence for scaffold hopping.

OUTLOOK

In this article, we have restricted ourselves to report the application of COSMO*sim3D* to the alignment and similarity measure for pairs of potential ligand molecules. On the basis of the simple rule that the best interaction partner for a molecular surface segment with polarization charge density σ is a segment with the opposite polarization charge density, or simply the rule "σ likes $-\sigma$", it is straightforward to extend COSMO*sim3D* to the alignment and similarity measure of ligand candidates to the inverted sigma profiles of enzyme receptor areas. The achieved COSMO*sim3D* similarity can in this way be considered as a measure of the potential interaction free energy, which corresponds to the pK_i value for the enzyme inhibition. Such applications will be investigated in future. Another straightforward extension to COSMO*sim3D* method is the usage of the 3D-σ-profiles of a set of aligned ligands as basis for a 3D-QSAR analysis, e.g., by the molecular field analysis approach, in order to generate a ligand-based model for the prediction of pK_i, where the alignment may be performed with the COSMO*sim3D* method. Details and benchmark results for this method, which we call COSMO*sar3D*, will be reported soon in a forthcoming paper.[21]

APPENDIX A: COSMO*frag*

The COSMO*frag* method[11] has been described as a fast shortcut for COSMO-σ-profile-based calculations, which replaces the relatively time-consuming quantum chemical DFT/COSMO calculation with a fast generation of an approximate σ-profile, composed of partial, atom-wise σ-profiles taken from structurally most similar atoms, which are retrieved from a database of ∼60,000 chemically most diverse drug-like compounds with precalculated DFT-COSMO σ-profiles, the so-called COSMO*frag* database (CFDB). On the basis of just a chemical structure input like SMILES, such COSMO*frag* σ-profile generation takes only a fraction of a second.

For the purpose of COSMO*sim3D* calculations, we have extended the COSMO*frag* technique toward the generation of approximate 3D-COSMO files, marked by the file extension fcos. On the basis of an input 3D-structure of a new molecule M, this CF-COSMO extension of COSMO*frag* for each atom i of M selects the most appropriate representation j from the database in the usual way of COSMO*frag*. It opens the compressed COSMO file (ccf) of the respective molecule containing atom j, which is in the CFDB, and takes the surface segments belonging to j out of the local coordinate system of atom j and moves them into the corresponding local coordinate system of the target atom i by a linear transformation, consisting of a translation and rotation. Because the i and j have a similar chemical neighborhood, the two local coordinate systems of i and j have a meaningful relation to each other, if they are based on the local directions to most similar nearest neighbors. As a result, the surface segments taken from the reference atom j end up approximately at the correct position in the surrounding of the target atom i. In this way, the surface of molecule M will at the end be filled with COSMO-σ-surface segments that have roughly the correct σ-values at roughly the right positions. In contrast to original COSMO files, these segments will not be strictly surface filling but leave some smaller open regions, while in other regions segments may be overlapping (Figure A1). Therefore, it is important that the local σ-averaging is performed on the reference molecules

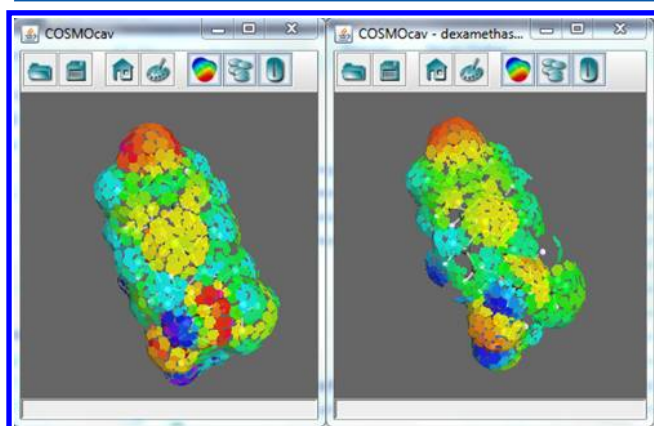because σ-averaging on the CF-COSMO files might yield artifacts.

Figure A1. Comparison between a true COSMO surface (a) and a fcos surface generated by COSMO*frag* (b) for dexamethasone. While the two surfaces are overall remarkably similar, a few small open regions can be observed in the fcos surface

In Figure A2, we show the results of a self-similarity test of CF-COSMO files vs the corresponding original COSMO files for the 11,280 compounds considered in the BioSter study described above. Almost all cs3d values are in the range 0.4–0.65, with the maximum at about cs3d = 0.55. This is clearly separated from the critical cs3d range of 0.2–0.3, which according to the examples considered above appears to be essential for the identification of bioisosterism. Therefore, bioisosterism can almost equally well be identified on the basis of CF-COSMO files as on original COSMO files.
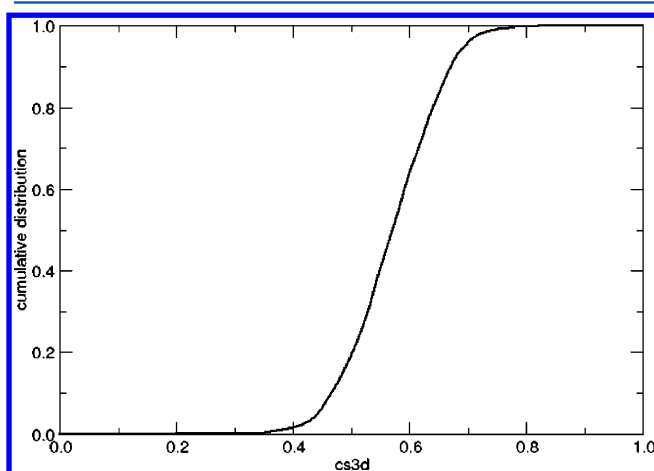

Figure A2. Self-similarity test of CF COSMO files vs the corresponding original COSMO files for the 11,280 compounds of the BioSter study (see text for details).

We confirmed this finding by repeating the bioisosteric pair experiment described above with CF-COSMO files instead of the original COSMO files. The corresponding curves for the accumulated fractions of pairs are very close (within 2%) to those obtained with the full COSMO files.

■ ASSOCIATED CONTENT

Ⓢ Supporting Information

Table S1: More detailed information about the 26 MDDR activity classes considered in Figure 4. This information is available free of charge via the Internet at http://pubs.acs.org.

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Notes
The authors declare the following competing financial interest(s): A.K. is CEO and K.W. is employee of COSMO*logic*. COSMO*logic* distributes COSMO*sim3D* on a commercial basis. M.T. is managing director of Origenis. Origenis will receive royalties from COSMO*sim3D* sales.

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