Supporting Information for "Extracting Structural Information from Residual Chemical Shift Anisotropy: Analytic Solutions for Peptide Plane Orientations and Applications to Determine Protein Structure"¹

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Below is the supporting information (SI) for the following paper [\[12\]](#page-14-0):

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Appendix

In [A](#page-0-0)ppendix A, we show that solving the system of equations from $N-H^N RDC$, and ¹³C-RCSA or ¹⁵N-RCSA to compute the peptide plane orientations leads to solving a degree 32 univariate polynomial equation. In contrast, our algorithm RDC-CSA-ANALYTIC uses a novel hybrid approach, and avoids solving such a high-degree polynomial equation. In Appendix [B](#page-2-0), we describe the algorithmic modules of RDC-CSA-ANALYTIC, and their inner workings. In **Appendix [C](#page-4-0)**, we give a proof of Proposition 2, and describe a method to construct all possible peptide plane orientations that satisfy RCSA and/or RDC data measured for a planar structural motif such as a peptide plane. Finally, in Appendix [D](#page-12-0), we give the sixteen oriented peptide planes computed for the peptide plane defined by Ala28 and Lys29 of ubiquitin.

A note on the exact solutions. The derivations below are not short, but they must only be performed once as we have done here. After that, the simple results (the closed-form solutions) may be used without thought of the proofs of correctness given here.

A Peptide Plane Orientations from $N-H^N$ RDC, and ¹³C-RCSA or ¹⁵N-RCSA

Here we consider the most general case when two ψ -defining RCSAs or RDCs are available for residues. From NMR experiments standpoint, this case is important, since one of ¹³C'-RCSA or

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 ${}^{15}{\rm N\text{-}RCSA}$ are measured along with N-H^N RDC. Further, this includes the case when two RCSAs per residue are available. We show that given the orientation or the peptide plane P_i , and two ψ -defining RDCs and/or RCSAs for vectors on the peptide plane P_{i+1} , computing the orientations of P_{i+1} relative to P_i requires us to solve a degree 32 polynomial equation.

We assume that the peptide plane is planar. Let $\mathbf{v}_1 = \boldsymbol{\delta}_{xx} = (x_1, y_1, z_1)^T$ and $\mathbf{v}_2 = \boldsymbol{\delta}_{yy}$ $(x_2, y_2, z_2)^T$ denote the two unit vectors representing two principal components of the traceless RCSA tensor on the peptide plane in the POF. Let the unit vectors $\mathbf{v}_3 = (x_3, y_3, z_3)^T$ and $\mathbf{v}_4 =$ $(x_4, y_4, z_4)^T$ denote N-H^N and C^{α}-C' bond orientations, respectively relative to the POF.

Since $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3$ and \mathbf{v}_4 are coplanar, we have

$$
\mathbf{v}_1 = \alpha \mathbf{v}_3 + \beta \mathbf{v}_4 \tag{A.1}
$$

$$
\mathbf{v}_2 = \alpha \mathbf{v}_3 + \beta \mathbf{v}_4 \tag{A.2}
$$

where $\alpha, \beta \in \mathbb{R}$ are constants, from which we get four independent equations

$$
x_1 = \alpha_1 x_3 + \beta_1 x_4 \tag{A.3}
$$

$$
y_1 = \alpha_1 y_3 + \beta_1 y_4 \tag{A.4}
$$

$$
x_2 = \alpha_2 x_3 + \beta_2 x_4 \tag{A.5}
$$

$$
y_2 = \alpha_2 y_3 + \beta_2 y_4 \tag{A.6}
$$

where α_i, β_i for $1 \leq i \leq 2$ are constants. These four equations can be used in the reduced RCSA equation $(Eq. (9)$ of the main text $[12]$ to obtain

$$
a_1x_3^2 + a_2y_3^2 + a_3x_4^2 + a_4y_4^2 + a_5x_3x_4 + a_6y_3y_4 = a_0 \tag{A.7}
$$

where a_i for $0 \le i \le 6$ are constants.

The reduced RDC equation for v_4 can be written as

$$
b_1 x_3^2 + b_2 y_3^2 = b_0 \tag{A.8}
$$

where b_i for $0 \leq i \leq 2$ are constants.

Since the angle between \mathbf{v}_3 and \mathbf{v}_4 is fixed, we have the following equation:

$$
x_3x_4 + y_3y_4 + z_3z_4 = c_0 \tag{A.9}
$$

where the constant c_0 is the cosine of the angle between \mathbf{v}_3 and \mathbf{v}_4 .

Since the orientation of the previous peptide plane has already been determined inductively, we let the unit vector along N-C^{α} bond of the previous peptide plane be denoted by $(d_1, d_2, d_3)^T$. From the fact that the angle between \mathbf{v}_4 and $(d_1, d_3, d_3)^T$ is fixed we get one more independent equation:

$$
d_1x_4 + d_2y_4 + d_3z_4 = d_0 \tag{A.10}
$$

where d_i for $0 \leq i \leq 3$ are constants.

Since \mathbf{v}_3 and \mathbf{v}_4 are unit vectors we have two more independent equations:

$$
x_3^2 + y_3^2 + z_3^2 = 1 \tag{A.11}
$$

$$
x_4^2 + y_4^2 + z_4^2 = 1. \tag{A.12}
$$

Now all we need to do is to solve the following set of simultaneous equations:

$$
a_1x_3^2 + a_2y_3^2 + a_3x_4^2 + a_4y_4^2 + a_5x_3x_4 + a_6y_3y_4 = a_0 \tag{A.13}
$$

$$
b_1 x_3^2 + b_2 y_3^2 = b_0 \tag{A.14}
$$

$$
x_3x_4 + y_3y_4 + z_3z_4 = c_0 \tag{A.15}
$$

$$
d_1x_4 + d_2y_4 + d_3z_4 = d_0 \qquad (A.16)
$$

$$
x_3^2 + y_3^2 + z_3^2 = 1 \tag{A.17}
$$

$$
x_4^2 + y_4^2 + z_4^2 = 1. \tag{A.18}
$$

It can be shown that solving this system $(Eq. (A.13)$ $(Eq. (A.13)$ to Eq. $(A.18)$ of polynomial equations is equivalent to solving a 32 degree univariate polynomial equation, which is a difficult computational problem. Our algorithm RDC-CSA-ANALYTIC uses a novel hybrid approach, and avoids solving such a high-degree polynomial equation.

B The Four Modules of RDC-CSA-ANALYTIC

Below we give a brief description of the four core modules of RDC-CSA-ANALYTIC.

Computation of Alignment Tensor. Our alignment tensor computation module is built upon our previous implementations in RDC-CSA-ANALYTIC that uses RDC data $[17, 20, 19]$ $[17, 20, 19]$ $[17, 20, 19]$. We have extended this to incorporate the constraints from RCSAs. The data fitting is done using singular value decomposition (SVD) [\[7,](#page-14-1) [17,](#page-15-0) [18,](#page-15-3) [6\]](#page-14-2), which now uses equations for RCSAs. An ideal helix model for a helical SSE of the protein is used to bootstrap the alignment tensor computation, and using SVD the initial alignment tensor S is computed. Then S iteratively refined by using the computed helix structures by RDC-CSA-ANALYTIC. Once the values of the alignment tensor is estimated, other fragments of the protein are computed using these values. Usually 2 or 3 iterations (a user-defined parameter) are sufficient to obtain a very good estimate of S.

Determination of Conformations and Orientations of the SSEs. RDC-CSA-ANALYTIC uses a divide-and-conquer approach to partition the protein backbone of size n residues into $\Theta(n)$ fragments of bounded lengths consisting of SSEs and loops, hence it runs in $\Theta(n)$ time. For each SSE, it inductively computes the peptide plane orientations using the method described in Section 2.3. The conformations are represented by a conformation tree grown recursively as we solve for the DOFs progressively. An internal (i.e., non-leaf) node in the tree represents the conformation of a part of a candidate SSE (α -helix or β -Strand), and a leaf node represents a candidate SSE conformation computed from RDCs and RCSAs. As each node is visited in a depth-first traversal of the tree, a set of conformation filters [\[13,](#page-14-3) [14\]](#page-15-4) are applied as predicates. If the node passes all the filters, then the subtree rooted at that node is visited; otherwise, the node is designated as a *dead-end* node. and the subtree rooted at that node is pruned. Finally, all remaining unpruned conformations (leaf nodes) are evaluated for the satisfaction of available experimental data, and deviations from standard average (ϕ_a, ψ_a) values for the SSE type. The top k (user-defined parameter) conformations are output. When $k = 1$, only the maximum-likelihood (ML) SSE conformation is output.

Since every conformation consistent with the data is either pruned or evaluated, RDC-CSAanalytic guarantees completeness. The experimental uncertainties in RDC and RCSA data are modeled by adding a Gaussian noise.

The RDC/RCSA RMSD between back-computed and experimental RDCs/RCSAs is computed

using the equation

RMSD_x =
$$
\sqrt{\frac{1}{n} \sum_{i=1}^{n} (r_{x,i}^b - r_{x,i}^e)^2}
$$
, (B.1)

where x is either a ϕ -defining or a ψ -defining RDC/RCSA type, n is the number of RDCs/RCSAs, $r_{x,i}^e$ is the experimental RDC/RCSA, and $r_{x,i}^b$ is the corresponding back-computed RDC/RCSA. The RMS deviations from standard average (ϕ_a, ψ_a) is given by

$$
RMSD_{\phi,\psi} = \sqrt{\frac{1}{2n} \sum_{i=1}^{n} ((\phi_i - \phi_a)^2 + (\psi_i - \psi_a)^2)},
$$
(B.2)

where (ϕ_a, ψ_a) are the standard averages of ϕ and ψ dihedrals for the SSE type, ϕ_i and ψ_i are the dihedrals computed using RDC and RCSA data, and n is the number of residues of the SSE.

Simultaneous Structure Determination and Packing of β-Strands into β-Sheets. Let $E =$ $\{E_1,\ldots,E_n\}$ be a β -sheet, where each E_i specifies the boundaries of the β -strand E_i . We arbitrarily pick one of the β -strands in the β -sheet as the first strand, and compute an ML conformation using the previous module. Let the strand conformation be included in the β -sheet S, which is initialized to an empty set. For each strand E_i adjacent to S , all possible conformations are computed using the previous module. As each conformation is evaluated, pack (we call it the pack procedure) it with S using a sparse set T_{NOE} of NOEs (or hydrogen bond information, if available), such that E_i forms favorable hydrogen bonds with S. The best scoring strand $E_{i_{\text{best}}}$ is chosen to be paired with S. This process continues until all the strands in E are computed and packed to form a β -sheet. The pseudo-code for strand packing algorithm is given below.

COMPUTEANDPACK β SHEET (S, T_{NOE})

- 1. Let $E = \{E_1, \ldots, E_n\}$ be the β -sheet specification, where E_i s are strand specifications
- 2. Initialize the partial sheet conformation $S \leftarrow \emptyset$
- 3. Compute E_1 using RDC-CSA-ANALYTIC
- 4. $S \leftarrow S \cup \{E_1\}$
- 5. $\forall E_i \in E, 2 \leq i \leq n$ in the order of strand adjacency
- 6. $\forall E_{i_{\text{leaf}}},$ a solution (leaf node) in a RDC-CSA-ANALYTIC conformation tree
- 7. PACK($S, E_{i_{\text{leaf}}}, T_{\text{NOE}}$) and test if it scores better than the previously packed sheet
- 8. Let $E_{i_{\text{best}}}$ be the strand with best packing and RDC satisfaction scores

$$
9. \qquad S \leftarrow S \cup \left\{E_{i_{\text{best}}}\right\}
$$

10. return S

Assembly of α -Helices and β -Sheets into Backbone Global Fold. It is well known that at least three distance constraints are needed to determine the relative translation between two oriented rigid bodies. The computed SSEs from RDC and RCSA data are all oriented with a four-fold degeneracy inherent to RDCs and RCSAs. Although the orientational degeneracy can be resolved when RDCs and RCSAs can be recorded in multiple alignment media, it cannot be resolved using only RDC and RCSA data in one alignment medium, unless other types of data such as NOEs are used. It has been shown that the orientational degeneracy can be resolved and the relative translations between the SSEs can be determined using a small number of (at least three NOEs between a pair of SSEs) inter-SSE NOEs [\[2,](#page-14-4) [17,](#page-15-0) [20\]](#page-15-1). Since the NOEs are interpreted as distance intervals between atoms, a small number of NOE distance restraints can bound the conformation space into a small volume which can be discretized and enumerated using a grid search of parameterized resolution. A sparse set of inter-SSE NOEs are extracted from chemical shift information alone [\[20\]](#page-15-1). For large and perdeuterated proteins, isoleucine-leucine-valine (ILV) selective labeling method $[3, 4, 15, 11]$ $[3, 4, 15, 11]$ $[3, 4, 15, 11]$ $[3, 4, 15, 11]$ and selective labeling of alanine residues $[5, 1]$ $[5, 1]$, in which protons and ¹³C isotopes are selectively incorporated into methyl groups of Ile^{δ_1} , Leu^{δ}, Val^{γ} and Ala^{β} side-chains on a deuterated background, can be used to measure NOE distances between H^{N} - H^N , methyl- H^N and methyl-methyl protons. The inter-SSE NOEs often involve side-chain protons. Since the side-chain conformation is not know a priori, a discretized set of side-chain conformations can be used to model the possible side-chain proton positions. We used Richardson's Rotamer Library [\[8\]](#page-14-10) to model the discrete side-chain conformations.

Suppose E be the set of inter-SSE NOEs used to pack two SSEs. For a distance restraint $e_i \in E$, let Q_i be the set of possible interpretations in terms of atom positions (rotamer pairs in case of NOEs). Each element $q_{ij} \in Q_i$, $1 \leq j \leq |Q_i|$ corresponds to a pair of atom positions and the corresponding distance between them which is represented by a spherical shell with inner and outer radii that respectively represent the lower and upper bounds for the distance restraint e_i . Then the volume of translation between the two SSEs that satisfies the distance restraints in E is given by

$$
\mathcal{V} = \bigcap_{i=1}^{|E|} \bigcup_{j=1}^{|Q_i|} q_{ij}.
$$
\n(B.3)

Let $S = \{s_1, \ldots, s_m\}$ be the set of SSEs computed by RDC-CSA-ANALYTIC. Our algorithm PACK takes as input the set of SSEs S , the set of inter-SSE distance restraints E , and parameters to specify the grid resolution and ensemble size, and computes an ensemble of packed structures. It uses the distance restraints to resolve the four-fold degeneracy of SSE orientations due to the symmetry of dipolar operator. At each grid point during the grid search over the space $\mathcal V$ of relative translations, the packed structures are checked for steric clashes. The computed ensemble of packed structures can further be clustered and evaluated [\[9,](#page-14-11) [10\]](#page-14-12) to trim the ensemble size, in order to obtain a set of representative packed structures.

C Peptide Plane Orientations from Three RDCs and/or RCSAs on Coplanar Internuclear Vectors

We give a proof of Proposition 2, and describe a method to construct all possible peptide plane orientations that satisfy RCSA and/or RDC data measured for a planar structural motif such as a peptide plane.

Any combination of three measurements, each of which is either an RCSA on a nucleus on the peptide plane or an RDC on an internuclear vector on the peptide plane, measured in one alignment medium, can be used in our derivation to elucidate the peptide plane orientations. In general, our derivation works for any combination of three RDCs and RCSAs measured on a planar structural motif to compute all possible discrete orientations of the planar motif that satisfy the RDCs and RCSAs. Our derivation assumes that the alignment tensor is rhombic. That is, magnitudes of

the two smaller principal components of the alignment tensor are not equal. In other words, if S_{xx} , S_{yy} and S_{zz} are the three principal components of a traceless alignment tensor **S** with $|S_{zz}| > |S_{yy}| > |S_{xx}|$, then **S** is rhombic if and only if $|S_{xx}| \neq |S_{yy}|$. The rhombicity R of the alignment tensor is given by the expression $R = (2/3)(S_{xx} - S_{yy})/S_{zz}$. Clearly, $R \in [0, 2/3]$.

We restate Proposition 2 below and then give a proof. Then we give a procedure to construct the peptide plane orientations from the roots of the quadratic equation.

Proposition 2. Given a rhombic alignment tensor, and 3 measurements, each of which is either an RCSA on a nucleus on the peptide plane P or an RDC on an internuclear vector on P , there exist at most 16 possible orientations for P that satisfy the 3 measurements, and these orientations can be written and solved in closed form by solving a quadratic equation.

Proof. The proof has two parts. In the first part, we derive a quadratic equation for the peptide plane orientations from three RCSA and/or RDC equations. In the second part, we will solve the quadratic equation exactly in closed form and construct all possible peptide plane orientations.

Derivation of a quadratic equation from three RCSA and/or RDC equations. We assume that the peptide plane is strictly planar, and we have any combination of three RCSAs and RDCs measured in one alignment medium for three unit vectors on the plane. For example, any three data types from the set ${N-H^N, C'-N, C''-C', C'H^N} \cup {^{13}C' \text{-RCSA}, {^{15}N \text{-RCSA}, {^1H \text{-RCSA}}}$ work for our derivations.

Let P denote the peptide plane. Let $\mathbf{v}_{01} = (x_{01}, y_{01}, z_{01})^T$ and $\mathbf{v}_{02} = (x_{02}, y_{02}, z_{02})^T$ be two unit vectors on P such that $\mathbf{v}_{01} \perp \mathbf{v}_{02}$. We use the literals x, y, z, u, v and t, and their subscripted versions for unknowns, and all other literals denote constants throughout the derivation. We assume that for the set of RCSAs and RDCs we know the Saupe alignment tensor S, and we work in a principal order frame that diagonalizes S.

Since RCSAs can be interpreted as linear combinations of two virtual RDCs (see Eq. (5) of the main text [\[12\]](#page-14-0)), we work on the most general form of the equations, that is, we will show how to compute the peptide plane orientations when three RCSA (i.e., ¹³C'-RCSA, ¹⁵N-RCSA and 1 H-RCSA) values are known for the respective atoms on the plane P. Then we will show that any combination of three RCSAs and RDCs is a special case of our derivation.

For ¹³C'-RCSA $\Delta \delta_{\rm C'}$, let the unit vectors $\mathbf{v}_1 = (x_1, y_1, z_1)^T$ and $\mathbf{v}_2 = (x_2, y_2, z_2)^T$ denote, in the POF, the two principal components of the CSA tensor that are on the peptide plane. For ¹⁵N-RCSA $\Delta \delta_N$, let the unit vectors $\mathbf{v}_3 = (x_3, y_3, z_3)^T$ and $\mathbf{v}_4 = (x_4, y_4, z_4)^T$ denote, in the POF, the two principal components of the CSA tensor that are on the peptide plane. For ¹H-RCSA $\Delta \delta_{\rm H}$, let the unit vectors $\mathbf{v}_5 = (x_5, y_5, z_5)^T$ and $\mathbf{v}_6 = (x_6, y_6, z_6)^T$ denote, in the POF, the two principal components of the CSA tensor that are on the peptide plane.

Since $\mathbf{v}_{01}, \mathbf{v}_{02}, \mathbf{v}_{1}, \cdots, \mathbf{v}_{6}$ are coplanar vectors, we can write

$$
\mathbf{v}_1 = a_1 \mathbf{v}_{01} + b_1 \mathbf{v}_{02} \quad \mathbf{v}_2 = a_2 \mathbf{v}_{01} + b_2 \mathbf{v}_{02}
$$

$$
\mathbf{v}_3 = a_3 \mathbf{v}_{01} + b_3 \mathbf{v}_{02} \quad \mathbf{v}_4 = a_4 \mathbf{v}_{01} + b_4 \mathbf{v}_{02}
$$

$$
\mathbf{v}_5 = a_5 \mathbf{v}_{01} + b_5 \mathbf{v}_{02} \quad \mathbf{v}_6 = a_6 \mathbf{v}_{01} + b_6 \mathbf{v}_{02},
$$

where a_i and b_i for $1 \leq i \leq 6$ are constants, determined from the fixed kinematic relationships

between these coplanar vectors. From the above equations, we can write the following equations:

$$
x_1 = a_1x_{01} + b_1x_{02} \t y_1 = a_1y_{01} + b_1y_{02}
$$

\n
$$
x_2 = a_2x_{01} + b_2x_{02} \t y_2 = a_2y_{01} + b_2y_{02}
$$

\n
$$
x_3 = a_3x_{01} + b_3x_{02} \t y_3 = a_3y_{01} + b_3y_{02}
$$

\n
$$
x_4 = a_4x_{01} + b_4x_{02} \t y_4 = a_4y_{01} + b_4y_{02}
$$

\n
$$
x_5 = a_5x_{01} + b_5x_{02} \t y_5 = a_5y_{01} + b_5y_{02}
$$

\n
$$
x_6 = a_6x_{01} + b_6x_{02} \t y_6 = a_6y_{01} + b_6y_{02}.
$$

Substituting the above equations for $x_1, \ldots, x_6, y_1, \ldots, y_6$ in the reduced RCSA equations for $\Delta \delta_{\rm C'}$, $\Delta \delta_{\rm N}$ and $\Delta \delta_{\rm H}$, we obtain

$$
A_1x_{01}^2 + A_2y_{01}^2 + A_3x_{02}^2 + A_4y_{02}^2 + A_5x_{01}x_{02} + A_6y_{01}y_{02} + A_7 = 0
$$
 (C.1)

$$
B_1x_{01}^2 + B_2y_{01}^2 + B_3x_{02}^2 + B_4y_{02}^2 + B_5x_{01}x_{02} + B_6y_{01}y_{02} + B_7 = 0
$$
 (C.2)

$$
C_1x_{01}^2 + C_2y_{01}^2 + C_3x_{02}^2 + C_4y_{02}^2 + C_5x_{01}x_{02} + C_6y_{01}y_{02} + C_7 = 0,
$$
 (C.3)

where

$$
A_1 = a\lambda_2 a_2^2 + a\lambda_1 a_1^2, \t A_2 = b\lambda_2 a_2^2 + b\lambda_1 a_1^2, \t A_3 = a\lambda_2 b_2^2 + a\lambda_1 b_1^2, A_4 = b\lambda_1 b_1^2 + b\lambda_2 b_2^2, \t A_5 = 2a\lambda_1 a_1 b_1 + 2a\lambda_2 a_2 b_2, \t A_6 = 2b\lambda_1 a_1 b_1 + 2b\lambda_2 a_2 b_2, \t A_7 = -c_1, B_1 = a\lambda_4 a_4^2 + a\lambda_3 a_3^2, \t B_2 = b\lambda_4 a_4^2 + b\lambda_3 a_3^2, \t B_3 = a\lambda_4 b_4^2 + a\lambda_3 b_3^2, B_4 = b\lambda_3 b_3^2 + b\lambda_4 b_4^2, \t B_5 = 2a\lambda_3 a_3 b_3 + 2a\lambda_4 a_4 b_4, \t B_6 = 2b\lambda_3 a_3 b_3 + 2b\lambda_4 a_4 b_4, \t B_7 = -c_2, C_1 = a\lambda_6 a_6^2 + a\lambda_5 a_5^2, \t C_2 = b\lambda_6 a_6^2 + b\lambda_5 a_5^2, \t C_3 = a\lambda_6 b_6^2 + a\lambda_5 b_5^2, C_4 = b\lambda_5 b_5^2 + b\lambda_6 b_6^2, \t C_5 = 2a\lambda_5 a_5 b_5 + 2a\lambda_6 a_6 b_6, \t C_6 = 2b\lambda_5 a_5 b_5 + 2b\lambda_6 a_6 b_6, \t C_7 = -c_3.
$$

In the above equations, the constants a, b, c_1, c_2 and c_3 are defined as

$$
a = S_{xx} - S_{zz},
$$

\n
$$
b = S_{yy} - S_{zz},
$$

\n
$$
c_1 = \Delta \delta_{\mathcal{C}'} - (\lambda_1 + \lambda_2) S_{zz},
$$

\n
$$
c_2 = \Delta \delta_{\mathcal{N}} - (\lambda_3 + \lambda_4) S_{zz},
$$

\n
$$
c_3 = \Delta \delta_{\mathcal{H}} - (\lambda_5 + \lambda_6) S_{zz},
$$

where the constants λ_1 and λ_2 are defined by the magnitudes of the principal components ¹³C'-RCSA along \mathbf{v}_1 and \mathbf{v}_2 , the constants λ_3 and λ_4 are defined by the magnitudes of the principal components ¹⁵N-RCSA along \mathbf{v}_3 and \mathbf{v}_4 , and the constants λ_5 and λ_6 are defined by the magnitudes of the principal components ¹H-RCSA along \mathbf{v}_5 and \mathbf{v}_6 , using Eq. (6) Eq. (7) of the main text [\[12\]](#page-14-0).

We now make the following three observations (verified by expanding and substituting the values for A_i , B_i , and C_i , for $1 \leq i \leq 6$), that will be useful in transforming Eq. [\(C.1\)](#page-6-0), Eq. [\(C.2\)](#page-6-1) and Eq. [\(C.3\)](#page-6-2) into simpler forms.

Observation 1.
$$
A_1B_2 - A_2B_1 = A_1C_2 - A_2C_1 = B_1C_2 - B_2C_1 = 0.
$$
 (C.4)

Observation 2.
$$
A_3B_4 - A_4B_3 = A_3C_4 - A_4C_4 = B_3C_4 - B_4C_3 = 0.
$$
 (C.5)

Observation 3.
$$
A_5B_6 - A_6B_5 = A_5C_6 - A_6C_5 = B_5C_6 - B_6C_5 = 0.
$$
 (C.6)

We use the symbol \times for scalar multiplication in our derivations unless otherwise specified. Now

Eq.
$$
(C.1) \times B_5 - \text{Eq. } (C.2) \times A_5 = 0
$$

\n
$$
\Rightarrow (A_1B_5 - A_5B_1)x_{01}^2 + (A_2B_5 - A_5B_2)y_{01}^2
$$
\n
$$
+ (A_3B_5 - A_5B_3)x_{02}^2 + (A_4B_5 - A_5B_4)y_{04}^2 + (A_7B_5 - A_5B_7) = 0,
$$
\n(C.7)

and

Eq.
$$
(C.1) \times C_5
$$
 - Eq. $(C.3) \times A_5 = 0$
\n $\Rightarrow (A_1C_5 - A_5C_1)x_{01}^2 + (A_2C_5 - A_5C_2)y_{01}^2$
\n $+ (A_3C_5 - A_5C_3)x_{02}^2 + (A_4C_5 - A_5C_4)y_{04}^2 + (A_7C_5 - A_5C_7) = 0.$ (C.8)

Further, we observe that

$$
(A_3B_5 - A_5B_3)(A_4C_5 - A_5C_4) - (A_4B_5 - A_5B_4)(A_3C_5 - A_5C_3) = 0,
$$
 (C.9)

and

$$
(A_1B_5 - A_5B_1)(A_2C_5 - A_5C_2) - (A_2B_5 - A_5B_2)(A_1C_5 - A_5C_1) = 0,
$$
 (C.10)

which can easily be shown by expanding, and then using Eq. [\(C.4\)](#page-6-3), Eq. [\(C.5\)](#page-6-4) and Eq. [\(C.6\)](#page-6-5). Now

Eq.
$$
(C.7) \times (A_4C_5 - A_5C_4) - \text{Eq. } (C.8) \times (A_4B_5 - A_5B_4) = 0
$$

\n $\Rightarrow E_1x_{01}^2 + E_2y_{01}^2 + E_3 = 0,$ (C.11)

where

$$
E_1 = (A_1B_5 - A_5B_1)(A_4C_5 - A_5C_4) - (A_1C_5 - A_5C_1)(A_4B_5 - A_5B_4)
$$

\n
$$
E_2 = (A_2B_5 - A_5B_2)(A_4C_5 - A_5C_4) - (A_2C_5 - A_5C_2)(A_4B_5 - A_5B_4)
$$

\n
$$
E_3 = (A_7B_5 - A_5B_7)(A_4C_5 - A_5C_4) - (A_7C_5 - A_5C_7)(A_4B_5 - A_5B_4).
$$

Similarly,

Eq.
$$
(C.7) \times (A_2C_5 - C_2A_5) - \text{Eq. } (C.8) \times (A_2B_5 - B_2A_5) = 0
$$

\n $\Rightarrow F_1x_{02}^2 + F_2y_{02}^2 + F_3 = 0,$ (C.12)

where

$$
F_1 = (A_3B_5 - A_5B_3)(A_2C_5 - A_5C_2) - (A_3C_5 - A_5C_3)(A_2B_5 - A_5B_2)
$$

\n
$$
F_2 = (A_4B_5 - A_5B_4)(A_2C_5 - A_5C_2) - (A_4C_5 - A_5C_4)(A_2B_5 - A_5B_2)
$$

\n
$$
F_3 = (A_7B_5 - A_5B_7)(A_2C_5 - A_5C_2) - (A_7C_5 - A_5C_7)(A_2B_5 - A_5B_2).
$$

Now

Eq.
$$
(C.1) \times B_1 - \text{Eq. } (C.2) \times A_1 = 0
$$

\n $\Rightarrow (A_3B_1 - A_1B_3)x_{02}^2 + (A_4B_1 - A_1B_4)y_{02}^2$
\n $+ (A_5B_1 - A_1B_5)x_{01}x_{02} + (A_6B_1 - A_1B_6)y_{01}y_{02} + (A_7B_1 - A_1B_7) = 0,$ (C.13)

and

Eq.
$$
(C.1) \times C_1 - \text{Eq. } (C.3) \times A_1 = 0
$$

\n $\Rightarrow (A_3C_1 - A_1C_3)x_{02}^2 + (A_4C_1 - A_1C_4)y_{02}^2$
\n $+ (A_5C_1 - A_1C_5)x_{01}x_{02} + (A_6C_1 - A_1C_6)y_{01}y_{02} + (A_7C_1 - A_1C_7) = 0.$ (C.14)

Further we observe that

$$
(A_3B_1 - A_1B_3)(A_4C_1 - A_1C_4) - (A_4B_1 - A_1B_4)(A_3C_1 - A_1C_3) = 0,
$$
\n(C.15)

which can easily be shown by expanding, and then using Eq. [\(C.4\)](#page-6-3), Eq. [\(C.5\)](#page-6-4) and Eq. [\(C.6\)](#page-6-5). Now

Eq.
$$
(C.13) \times (A_4C_1 - A_1C_4) - \text{Eq. (C.14)} \times (A_4B_1 - A_1B_4) = 0
$$

\n $\Rightarrow G_1x_{01}x_{02} + G_2y_{01}y_{02} + G_3 = 0,$ (C.16)

where

$$
G_1 = (A_5B_1 - A_1B_5)(A_4C_1 - A_1C_4) - (A_5C_1 - A_1C_5)(A_4B_1 - A_1B_4)
$$

\n
$$
G_2 = (A_6B_1 - A_1B_6)(A_4C_1 - A_1C_4) - (A_6C_1 - A_1C_6)(A_4B_1 - A_1B_4)
$$

\n
$$
G_3 = (A_7B_1 - A_1B_7)(A_4C_1 - A_1C_4) - (A_7C_1 - A_1C_7)(A_4B_1 - A_1B_4).
$$

We make one more observation:

$$
E_1F_2 - E_2F_1 = E_1G_2 - E_2G_1 = F_1G_2 - F_2G_1 = 0,
$$
\n(C.17)

from which we obtain

$$
\frac{E_1}{F_1} = \frac{E_2}{F_2}, \quad \frac{E_1}{G_1} = \frac{E_2}{G_2}, \quad \frac{F_1}{G_1} = \frac{F_2}{G_2}.
$$
\n(C.18)

Using Eq. $(C.18)$ in Eq. $(C.12)$ and Eq. $(C.16)$, we obtain

$$
E_1 x_{02}^2 + E_2 y_{02}^2 + E_4 = 0,\t(C.19)
$$

and

$$
E_1 x_{01} x_{02} + E_2 y_{01} y_{02} + E_5 = 0, \tag{C.20}
$$

where

$$
E_4 = \frac{E_1 F_3}{F1}, \quad E_5 = \frac{E_1 G_3}{G_1}.
$$

We now want to solve Eq. [\(C.11\)](#page-7-4), Eq. [\(C.19\)](#page-8-2) and Eq. [\(C.20\)](#page-8-3). Since there are four variables, we need one more independent equation, which we get from the fact that $\mathbf{v}_{01} \perp \mathbf{v}_{02}$. In fact, even when the unit vectors \mathbf{v}_{01} and \mathbf{v}_{02} are chosen not to be perpendicular to each other, our derivation works. We let $\cos(\angle \mathbf{v}_{01}, \mathbf{v}_{02}) = \tau$, where τ is a constant.

$$
x_{01}x_{02} + y_{01}y_{02} + z_{01}z_{02} = \cos(\angle \mathbf{v}_{01}, \mathbf{v}_{02}) = \tau
$$
\n
$$
\Rightarrow z_{01}^2 z_{02}^2 = (\tau - x_{01}x_{02} - y_{01}y_{02})^2
$$
\n
$$
\Rightarrow (1 - x_{01}^2 - y_{01}^2)(1 - x_{02}^2 - y_{02}^2) = (\tau - x_{01}x_{02} - y_{01}y_{02})^2
$$
\n
$$
\Rightarrow x_{01}^2 + y_{01}^2 + x_{02}^2 + y_{02}^2
$$
\n
$$
- x_{01}^2 y_{02}^2 - x_{02}^2 y_{01}^2 + 2x_{01}x_{02}y_{01}y_{02} - 2\tau(x_{01}x_{02} + y_{01}y_{02}) - (1 - \tau^2) = 0.
$$
\n(C.22)

Using Eq. [\(C.20\)](#page-8-3) in Eq. [\(C.22\)](#page-8-4), we eliminate the terms containing factor $x_{01}x_{02}$ as follows: Eq. $(C.22) \Rightarrow 2\tau(x_{01}x_{02} + y_{01}y_{02})$ $(C.22) \Rightarrow 2\tau(x_{01}x_{02} + y_{01}y_{02})$ $(C.22) \Rightarrow 2\tau(x_{01}x_{02} + y_{01}y_{02})$

$$
= x_{01}^2 + y_{01}^2 + x_{02}^2 + y_{02}^2 - x_{02}^2 y_{01}^2 - x_{01}^2 y_{02}^2 + 2x_{01} x_{02} y_{01} y_{02} - (1 - \tau^2)
$$

\n
$$
\Rightarrow 2\tau \left(\frac{-E_5 - E_2 y_{01} y_{02}}{E_1} + y_{01} y_{02} \right)
$$

\n
$$
= x_{01}^2 + y_{01}^2 + x_{02}^2 + y_{02}^2 - x_{02}^2 y_{01}^2 - x_{01}^2 y_{02}^2 + 2 \left(\frac{-E_5 - E_2 y_{01} y_{02}}{E_1} \right) y_{01} y_{02} - (1 - \tau^2)
$$

\n
$$
\Rightarrow \frac{-2\tau E_5}{E_1} + 2\tau \left(1 - \frac{E_2}{E_1} \right) y_{01} y_{02}
$$

\n
$$
= x_{01}^2 + y_{01}^2 + x_{02}^2 + y_{02}^2 - x_{02}^2 y_{01}^2 - x_{01}^2 y_{02}^2 + 2 \left(\frac{-E_5 - E_2 y_{01} y_{02}}{E_1} \right) y_{01} y_{02} - (1 - \tau^2)
$$

\n
$$
\Rightarrow \frac{-2\tau E_5}{a} + \left(2\tau \left(1 - \frac{E_2}{E_1} \right) + \frac{2E_5}{E_1} \right) y_{01} y_{02}
$$

\n
$$
= x_{01}^2 + y_{01}^2 + x_{02}^2 + y_{02}^2 - x_{02}^2 y_{01}^2 - x_{01}^2 y_{02}^2 - \frac{2E_2}{E_1} y_{01}^2 y_{02}^2 - (1 - \tau^2)
$$

\n
$$
\Rightarrow x_{01}^2 + y_{01}^2 + x_{02}^2 + y_{02}^2 - x_{02}^2 y_{01}^2 - x_{01}^2 y_{02}^2
$$

\n
$$
- \frac{2E_2}{E_1} y_{01}^2 y_{
$$

where

$$
H_1 = -\frac{2E_2}{E_1}, \quad H_2 = -2\tau \left(1 - \frac{E_2}{E_1}\right) - \frac{2E_5}{E_1}, \quad H_3 = -1 + \tau^2 + \frac{2\tau E_5}{E_1}
$$

Rearranging the terms in Eq. $(C.20)$ and squaring, we can express $y_{01}y_{02}$ as a linear combination of terms involving $x_{01}^2 x_{02}^2$ and $y_{01}^2 y_{02}^2$ as follows:

Eq. (C.20)
\n
$$
\Rightarrow E_1 x_{01} x_{02} = -E_5 - E_2 y_{01} y_{02}
$$
\n
$$
\Rightarrow E_1^2 x_{01}^2 x_{02}^2 = (-E_5 - E_2 y_{01} y_{02})^2 = E_5^2 + 2E_2 E_5 y_{01} y_{02} + E_2^2 y_{01}^2 y_{02}^2
$$
\n
$$
\Rightarrow y_{01} y_{02} = \frac{E_1^2 x_{01}^2 x_{02}^2 - E_2^2 y_{01}^2 y_{02}^2 - E_5^2}{2E_2 E_5} = I_1 x_{01}^2 x_{02}^2 + I_2 y_{01}^2 y_{02}^2 + I_3,
$$
\n(C.24)

.

.

where

$$
I_1 = \frac{E_1^2}{2E_2E_5}
$$
, $I_2 = \frac{-E_2}{2E_5}$, $I_3 = \frac{-E_5}{2E_2}$

Substituting Eq. [\(C.24\)](#page-9-0) in Eq. [\(C.23\)](#page-9-1) we obtain

$$
x_{01}^2 + y_{01}^2 + x_{02}^2 + y_{02}^2 - x_{02}^2 y_{01}^2 - x_{01}^2 y_{02}^2 + J_1 x_{01}^2 x_{02}^2 + J_2 y_{01}^2 y_{02}^2 + J_3 = 0,
$$
 (C.25)

where

$$
J_1 = H_2 I_1 = \frac{E_1(-\tau E_1 + \tau E_2 - E_5)}{E_2 E_5}
$$

\n
$$
J_2 = H_1 + H_2 I_2 = \frac{E_2(\tau E_1 - \tau E_2 - E_5)}{E_1 E_5}
$$

\n
$$
J_3 = H_3 + H_2 I_3 = \frac{\tau^2 E_1 E_2 + \tau E_1 E_5 + \tau E_2 E_5 + E_5^2 - E_1 E_2}{E_1 E_2}.
$$

We now solve the system of equations Eq. [\(C.11\)](#page-7-4), Eq. [\(C.19\)](#page-8-2), Eq. [\(C.20\)](#page-8-3) and Eq. [\(C.25\)](#page-9-2) simultaneously.

We will now transform Eq. [\(C.20\)](#page-8-3) so that the degree of each of x_{01}, y_{01}, x_{02} and y_{02} is even.

Eq. (C.20)
\n
$$
\Rightarrow E_1x_{01}x_{02} = -E_5 - E_2y_{01}y_{02}
$$
\n
$$
\Rightarrow E_1^2x_{01}^2x_{02}^2 = E_2^2y_{01}^2y_{02}^2 + 2E_2E_5y_{01}y_{02} + E_5^2 \quad \text{(squaring both sides)}
$$
\n
$$
\Rightarrow 2E_2E_5y_{01}y_{02} = E_1^2x_{01}^2x_{02}^2 - E_2^2y_{01}^2y_{02}^2 - E_5^2
$$
\n
$$
\Rightarrow 4E_2^2E_5^2y_{01}^2y_{02}^2 = E_1^4x_{01}^4x_{02}^4 + E_2^4y_{01}^4y_{02}^4
$$
\n
$$
-2E_1^2E_2^2x_{01}^2x_{02}^2y_{01}^2y_{02}^2 - 2E_1^2E_5^2x_{01}^2x_{02}^2 + 2E_2^2E_5^2y_{01}^2y_{02}^2 + E_5^4 \quad \text{(squaring both sides)}
$$
\n
$$
\Rightarrow E_1^4x_{01}^4x_{02}^4 + E_2^4y_{01}^4y_{02}^4 - 2E_1^2E_2^2x_{01}^2x_{02}^2y_{01}^2y_{02}^2 - 2E_1^2E_5^2x_{01}^2x_{02}^2 - 2E_2^2E_5^2y_{01}^2y_{02}^2 + E_5^4 = 0. \quad \text{(C.26)}
$$

Now we solve Eq. [\(C.11\)](#page-7-4), Eq. [\(C.19\)](#page-8-2), Eq. [\(C.25\)](#page-9-2) and Eq. [\(C.26\)](#page-10-0) simultaneously. Let

$$
u = x_{01}^2 \tag{C.27}
$$

and

$$
v = x_{02}^2. \tag{C.28}
$$

Then,

Eq.
$$
(C.11) \Rightarrow y_{01}^2 = \frac{-E_3 - E_1 u}{E_2}
$$
 (C.29)

and

Eq.
$$
(C.19) \Rightarrow y_{02}^2 = \frac{-E_4 - E_1 v}{E_2}
$$
. (C.30)

Using Eq. $(C.29)$ and Eq. $(C.30)$ in Eq. $(C.25)$, we obtain

$$
K_0 + K_1 u + K_2 v + K_3 u v = 0,
$$
\n(C.31)

where

$$
K_0 = E_3E_4J_2 - E_2E_4 - E_2E_3 + E_2^2J_3
$$

\n
$$
= \frac{E_2}{E_1E_5} \left(-E_3E_4E_5 + E_3E_4\tau E_1 - E_3E_4E_2\tau - E_4E_1E_5 - E_3E_1E_5 - E_2E_1E_5 \right)
$$

\n
$$
+ \frac{E_2}{E_1E_5} \left(\tau^2 E_1E_2E_5 + \tau E_2E_5^2 + \tau E_1E_5^2 + E_5^3 \right)
$$

\n
$$
K_1 = -E_1E_2 + E_2E_4 + E_2^2 + E_1E_4J_2 = \frac{E_2(-E_1E_5 + E_2E_5 + \tau E_1E_4 - \tau E_2E_4)}{E_5}
$$

\n
$$
K_2 = E_2E_3 + E_2^2 - E_1E_2 + E_1E_3J_2 = \frac{E_2(E_2E_5 - E_1E_5 + \tau E_1E_3 - \tau E_2E_3)}{E_5}
$$

\n
$$
K_3 = E_1^2J_2 + E_2^2J_1 + 2E_1E_2 = 0.
$$

Noting that $K_3 = 0$, we can write Eq. [\(C.31\)](#page-10-3) as

$$
K_0 + K_1 u + K_2 v = 0
$$

\n
$$
\Rightarrow v = \frac{-K_0 - K_1 u}{K_2}.
$$
\n(C.32)

Using Eq. $(C.29)$ and Eq. $(C.30)$ in Eq. $(C.26)$, we obtain

$$
L_1 u^2 + L_2 u v + L_3 v^2 + L_4 u + L_5 v + L_6 = 0,
$$
\n(C.33)

where

$$
L_1 = E_1^2 E_4^2
$$

\n
$$
L_2 = 2E_1^2 E_3 E_4 - 4E_1^2 E_5^2
$$

\n
$$
L_3 = E_1^2 E_3^2
$$

\n
$$
L_4 = 2E_1 E_3 E_4^2 - 2E_1 E_4 E_5^2
$$

\n
$$
L_5 = 2E_1 E_3^2 E_4 - 2E_1 E_3 E_5^2
$$

\n
$$
L_6 = E_3^2 E_4^2 - 2E_3 E_4 E_5^2 + E_5^4.
$$

Using Eq. $(C.32)$ in Eq. $(C.33)$ we obtain

$$
M_0 + M_1 u + M_2 u^2 = 0,\t\t(C.34)
$$

where

$$
M_0 = -\frac{L_5 K_0}{K_2} + L_6 + \frac{L_3 K_0^2}{K_2^2}
$$

\n
$$
M_1 = -\frac{L_5 K_1}{K_2} - \frac{L_2 K_0}{K_2} + L_4 + \frac{2L_3 K_0 K_1}{K_2^2}
$$

\n
$$
M_2 = -\frac{L_2 K_1}{K_2} + L_1 + \frac{L_3 K_1^2}{K_2^2}.
$$

Eq. [\(C.34\)](#page-11-1) is a quadratic equation which can be solved in closed form from which the orientations of the peptide plane can be computed as detailed below.

Constructing the oriented peptide planes. We give the details of how to compute the different orientations of a peptide plane from the solutions of Eq. [\(C.34\)](#page-11-1). The construction exploits the symmetry present in the RDC equation.

Solving Eq. [\(C.34\)](#page-11-1) we get at most two real solutions. Let $\{u_1, u_2\}$ denote the set of these two solutions. For each u_i , $1 \leq i \leq 2$, we can compute a unique v_i using Eq. [\(C.32\)](#page-10-4). For a pair (u_i, v_i) , $1 \le i \le 2$, using Eq. [\(C.27\)](#page-10-5), Eq. [\(C.28\)](#page-10-6), Eq. [\(C.29\)](#page-10-1) and Eq. [\(C.30\)](#page-10-2), we compute sixteen $(x_{01}, y_{01}, x_{02}, y_{02})$ tuples, where any two tuples differ by the sign of at least one element, but the absolute values of the elements are the same in all of the sixteen tuples. Out of these sixteen tuples only four will satisfy Eq. [\(C.20\)](#page-8-3). This is because, $(x_{01}, y_{01}, x_{02}, y_{02})$ satisfies Eq. [\(C.20\)](#page-8-3) if and only if $(-x_{01}, y_{01}, -x_{02}, y_{02})$, $(x_{01}, -y_{01}, x_{02}, -y_{02})$ and $(-x_{01}, -y_{01}, -x_{02}, -y_{02})$ satisfy Eq. [\(C.20\)](#page-8-3). Since $\mathbf{v}_{01} = (x_{01}, y_{01}, z_{01})^T$ is a unit vector, we get two solutions for z_{01} which we denote by $\{z_{01}, -z_{01}\}$. Similarly, since $\mathbf{v}_{02} = (x_{02}, y_{02}, z_{02})^T$ is a unit vector we get two solutions for z_{02} which we denote by $\{z_{02}, -z_{02}\}$. Therefore, there are four possibilities, viz. $\{(z_{01}, z_{02}), (-z_{01}, z_{02}), (z_{01}, -z_{02}), (-z_{01}, -z_{02})\}$. From Eq. [\(C.21\)](#page-8-5) we observe that the product $z_{01}z_{02}$ must have the same sign which means that only two out of the above four pairs would be considered. Without loss of generality, we let (z_{01}, z_{02}) and $(-z_{01}, -z_{02})$ be the two representative pairs. Therefore, we have eight 6-tuples which can be written as the following product using set theoretic notation:

$$
\left\{ (x_{01}, y_{01}, x_{02}, y_{02}), (-x_{01}, y_{01}, -x_{02}, y_{02}), (x_{01}, -y_{01}, x_{02}, -y_{02}), (-x_{01}, -y_{01}, -x_{02}, -y_{02}) \right\} \times \left\{ (z_{01}, z_{02}), (-z_{01}, -z_{02}) \right\},
$$

Table 1: Sixteen peptide plane orientations from single set of C^a-C', C'-N, N-H^N RDCs, measured in one alignment medium, for the peptide plane between the residues Ala28 and Lys29 of ubiquitin.

Solution	Unit Vector	Unit Vector	Index to
Number	$v_{N-HN} = (x_1, y_1, z_1)$	$v_{C'-N} = (x_2, y_2, z_2)$	Figure 1
	(0.19598, 0.84846, 0.49163)	$(0.78319, 0.55612, -0.27810)$	A
3	$(0.19598, 0.84846, -0.49163)$	(0.78319, 0.55612, 0.27810)	В
3	$(-0.19598, 0.84846, 0.49163)$	$(-0.78319, 0.55612, -0.27810)$	\mathcal{C}
4	$(-0.19598, 0.84846, -0.49163)$	$(-0.78319, 0.55612, 0.27810)$	D
5	$(0.19598, -0.84846, 0.49163)$	$(0.78319, -0.55612, -0.27810)$	E
6	$(0.19598, -0.84846, -0.49163)$	$(0.78319, -0.55612, 0.27810)$	F
	$(-0.19598, -0.84846, 0.49163)$	$(-0.78319, -0.55612, -0.27810)$	G
8	$(-0.19598, -0.84846, -0.49163)$	$(-0.78319, -0.55612, 0.27810)$	H
9	$(0.31133, -0.58536, 0.74862)$	(0.80635, 0.26825, 0.52711)	
10	$(0.31133, -0.58536, -0.74862)$	$(0.80635, 0.26825, -0.52711)$	J.
11	$(-0.31133, -0.58536, 0.74862)$	$(-0.80635, 0.26825, 0.52711)$	K
12	$(-0.31133, -0.58536, -0.74862)$	$(-0.80635, 0.26825, -0.52711)$	L
13	(0.31133, 0.58536, 0.74862)	$(0.80635, -0.26825, 0.52711)$	М
14	$(0.31133, 0.58536, -0.74862)$	$(0.80635, -0.26825, -0.52711)$	N
15	$(-0.31133, 0.58536, 0.74862)$	$(-0.80635, -0.26825, 0.52711)$	O
16	$(-0.31133, 0.58536, -0.74862)$	$(-0.80635, -0.26825, -0.52711)$	Ρ

Experimental RDCdata for ubiquitin (PDB id: 1D3Z) is taken from the BioMagResBank (BMRB) [\[16\]](#page-15-6). The right-most column indicates the indexes of the corresponding sub-figure in Figure [1,](#page-13-0) that uses the two vectors in columns 2 and 3 to define the peptide plane.

where the symbol \times is used for the Cartesian product. We note that each 6-tuple in this set represents a pair of unit vectors which forms a plane. Thus, for a pair (u_i, v_i) we have at most 8 possible peptide planes. Since there are at most two (u_i, v_i) pairs when $1 \le i \le 2$, we have at most $8 \times 2 = 16$ possible peptide plane orientations.

This completes the proof.

 \Box

D The Sixteen peptide Plane Orientations

The algorithm 3PLANAR computed the peptide plane orientations for the peptide plane defined by the residues Ala28 and Lys29 of ubiquitin, from C^{α} -C', C'-N and N-H^N RDCs in one alignment medium. It first computed the peptide plane orientations represented by pairs of \mathbf{v}_{N-H^N} and $\mathbf{v}_{C'-N}$ unit vectors. The 16 possible pairs of unit vectors are listed in Table [1.](#page-12-1) Then, each pair of \mathbf{v}_{N-HN} and $\mathbf{v}_{C,N}$ unit vectors were used to compute the corresponding oriented peptide plane. In Figure [1,](#page-13-0) each of the peptide planes that corresponds to a pair of $\mathbf{v}_{N\text{-H}^N}$ and $\mathbf{v}_{C^{\prime}\text{-N}}$ unit vectors (see the right-most column of Table [1\)](#page-12-1) is shown.

Figure 1: Sixteen peptide plane orientations from single set of C^{α} -C', C'-N, N-H^N RDCs, measured in one alignment medium, for the peptide plane between the residues Ala28 and Lys29 of ubiquitin. The peptide planes were constructed using the unit vectors given in the Table [1.](#page-12-1)

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