Discrete differential geometry of tetrahedrons and encoding of local protein structure

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February 2, 2008

Abstract
Local protein structure analysis is informative to protein structure analysis and has been used successfully in protein structure prediction and others. Proteins have recurring structural features, such as helix caps and beta turns, which often have strong amino acid sequence preferences. And the challenges for local structure analysis have been identification and assignment of such common short structural motifs.

This paper proposes a new mathematical framework that can be applied to analysis of the local structure of proteins, where local conformations of protein backbones are described using differential geometry of folded tetrahedron sequences. Using the framework, we could capture the recurring structural features without any structural templates, which makes local structure analysis not only simpler, but also more objective. Programs and examples are available from http://www.genocript.com.

AMS Subject Classification: 52C99, 92B99
Key Words and Phrases: Discrete differential geometry – Tetrahedron sequence – Local protein structure

1 Introduction
Protein is a sequence of amino acids, which folds into a unique three-dimensional structure in nature. And one could identify proteins with polygonal chains obtained by connecting the center of adjacent amino acids. Since the functional properties of proteins are largely determined by the structure, protein structure analysis is crucial to the study of proteins.

Local protein structure analysis is informative to protein structure analysis and has been used successfully in protein structure prediction and others. Proteins have recurring structural features, such as helix caps and beta turns, which often have strong amino acid sequence preferences. And the challenges for local structure analysis have been identification and assignment of such common
short structural motifs ([1], [2], [3], [4], [5]). Identification involves description of protein backbone conformation and definitions of the structural motifs. And assignment is not a trivial task, due to the variations observed in nature when compared to ideal ones.

In this paper, we introduce a new differential geometrical approach for local structure analysis. As for differential geometrical description, a lot of works on the surface of protein molecules are known (to name a few, [6], [7]). But protein backbone structure is usually studied via classification ([8], [9]) and differential geometrical approach has been rarely taken so far.

One of the few is the early work of [10] which described protein backbones as polygonal chain, where each line segment corresponds to the virtual-bond between consecutive $\alpha$-carbons. In contrast, we describe local conformation of protein backbones using folded tetrahedron sequences (Figure 1 (b)).

As for the shape of protein backbones, [11] proposed the notion of alpha-shape and [12] examined geometric restrictions on polygonal protein chains. Moreover, [13] reviewed topological knots in protein structure.

2 Differential geometry of triangles

For simplicity, we first consider the differential geometry of triangles.

2.1 Basic ideas

Let’s consider unit cube $[0,1]^3$ in the three-dimensional Euclidean space $\mathbb{R}^3$ and divide each of the three facets which contain $(0,0,0)$ into two triangles along diagonal, as shown in figure 2 (a). Then, if we pile the cubes up in the direction of $(-1,-1,-1)$, we would obtain “peaks and valleys” of cubes, where the division of the facets of each cube makes up a division of the surface of the peaks and valleys (figure 2 (b) top). And a “flow” of triangles in $\mathbb{R}^2$ is obtained.
by projecting the surface onto a hyperplane, (figure 2(b) bottom). For example, the grey “slant” triangles on the surface specify the closed trajectory of the grey “flat” triangles on the hyperplane.

In the following, we use monomial notation to denote points and triangles in $\mathbb{R}^3$. That is, we denote point $(l, m, n) \in \mathbb{R}^3$ by monomial $x^l x^m x^n \in \mathbb{Z}[x_1, x_2, x_3]$. And the triangle of vertices $(l, m, n), (l + 1, m, n), (l + 1, m + 1, n) \in \mathbb{R}^3$ are denoted by $x^l x^m x^n [x_1 x_2]$. For example, $a[x_i x_j]$ is the triangle of vertices $a, ax_i, and ax_j \in \mathbb{Z}[x_1, x_2, x_3]$ (figure 2(c)).

2.2 Tangent bundle over flat triangles

Let $\pi$ be the projection of the collection $S$ of all the slant triangles onto the collection $B$ of all the flat triangles along direction $(-1, -1, -1)$, where the image of $a[x_i x_j] \in S$ is denoted by $|a[x_i x_j]|$ (figure 2(c)). Then, projection $\pi$ induces tangent bundle-like structure $TB$ over $B$, where the gradient of slant triangles are defined as follows:

**Definition 1.** The gradient $Da[x_i x_j]$ of $a[x_i x_j] \in S$ is monomial $x_i x_j \in \mathbb{Z}[x_1, x_2, x_3]$. In particular, there is a one-to-one correspondence between $TB$ and $\{x_1 x_2, x_1 x_3, x_2 x_3\} \times B$. And we indicate the gradient value over a flat triangle by a bold edge as shown in figure 3(a).

For example, slant triangles $a[x_1 x_2], ax_1[x_2 x_3]$, and $a/x_3[x_3 x_1] \in S$ are projected onto the same flat triangle $|a[x_1 x_2]| \in B$ and their gradients are $x_1 x_2, x_2 x_3$, and $x_1 x_3$ respectively (figure 3(a)).

Then, a gradient value over a flat triangle specifies a local trajectory at the flat triangle as follows:

**Definition 2.** The local trajectory defined by $a[x_i x_j] \in S$ at $|a[x_i x_j]| \in B$ is the three consecutive flat triangles $\{|ax_i x_j x_1|, |a[x_i x_j]|, |a/x_j x_3 x_1|\} \subset B$. As figure 3(b) shows, these are the adjacent triangles connected along the direction of the bold edge of $|a[x_i x_j]|$. And the local trajectory is specified uniquely by
whose gradient is a trajectory specified by Figure 3: Differential structure. (a): Gradient values of slant triangles over a specified by ax1[x2x3] whose gradient is x2x3, and a/x3[x3x1] whose gradient is x1x3. (b): The local trajectory specified by a[x1x2] ∈ S at |a[x1x2]| ∈ B, |ax1[x2x1]| (downward) and |a/x2[x2x1]| (upward) ∈ B. (c): Smoothness condition at |a[x1x2]| ∈ B (white) specified by a[x1x2] ∈ S. The next triangle |ax1[x2x1]| ∈ B (grey) could assume either x1x2 or x2x3 as gradient.

the gradient of a[xi xj].

Now we impose a kind of “smoothness condition” as shown in figure 3 (c). That is, each flat triangle assume one of two gradient values, which are determined naturally by the gradient of the preceding triangle. Suppose that the gradient at current triangle |a[x1x2]| ∈ B is x1x2 and the gradient at next triangle |ax1[x2x1]| ∈ B is x1x3. Then, two flat triangles |a[x1x2]| and |ax1[x2x1]| are separated by the bold edge of |ax1[x2x1]| (figure 3 (c) right). In this case, we permit either x1x2 or x2x3 as gradient of |ax1[x2x1]|.

As an example, let’s consider the peaks and valleys shown in figure 2 (b), which is specified by three peaks a, b = a/x2, and c = ax2^2x2/x3 ∈ Z[x1, x2, x3] (figure 4). Peaks and valleys define a “smooth” vector field V on B by the following mapping:

\[ V : B \rightarrow \{x_1x_2, x_1x_3, x_2x_3\}, \quad V(|a[xi xj]|) := x_i x_j, \]

where |a[xi xj]| ∈ S is the slant triangle on the surface of the peaks and valleys over |a[xi xj]| ∈ B.

Let’s start from triangle |a[x1x2]| (grey) and move downward: t[0] = |a[x1x2]| and V(t[0]) = x1x2. Then, gradient V(t[0]) specifies local trajectory \{|ax1[x2x1]|, |a[x1x2]|, |a[x1x3]|\} at t[0] (Note that |a/x2[x2x1]| = |a[x1x3]|). Since we move downward, next triangle t[1] is |ax1[x2x1]| and we obtain V(t[1]) = x1x2. Then, gradient V(t[1]) specifies local trajectory \{|ax1[x2x1]|, |ax1[x2x1]|, |a[x1x2]|\} at t[1]. And next triangle t[2] is |ax1[x2x1]|. Continuing the process, we obtain the closed trajectory of length 10.
Figure 4: Closed trajectory of the vector field specified by three peaks $a$, $b = a/x_2$, and $c = ax_1^2x_2/x_3 \in \mathbb{Z}[x_1, x_2, x_3]$.

### 2.3 Encoding of the shape of trajectories

Finally, let’s consider variation of gradient along a trajectory. Thanks for the smoothness condition, variation of gradient, i.e., the “second derivative”, along a triangle trajectory is given as binary valued sequence.

**Definition 3.** The derivative $DV$ of vector field $V$ along trajectory $\{t[i]\}$ is defined as follows:

\[
DV : B \rightarrow \{U, D\}, \quad DV(t[i]) := \begin{cases} 
DV(t[i-1]) & \text{if } V(t[i]) = V(t[i-1]) \\
-DV(t[i-1]) & \text{otherwise},
\end{cases}
\]

where $-U := D$ and $-D := U$. In words, **change value if the gradient changes**.

As an example, let’s consider the trajectory of figure 4 again. First, set any initial value: $DV(t[0]) = D$. Then, since the first two triangles $t[0]$ and $t[1]$ have the same gradient, $DV(t[1])$ is also $D$. The value of the second derivative is $D$ until $t[3]$, where it changed to $U$ since the gradient of $t[2]$ is different from that of $t[3]$.

Continuing the process, we obtain a binary sequence of length 10, $DDDU DDUUUDU$, which describes the shape of the trajectory.

### 3 Differential geometry of tetrahedrons

Similarly we obtain a flow of tetrahedrons in $\mathbb{R}^3$ by considering peaks and valleys of 4-cubes in $\mathbb{R}^4$. In this case, each trajectory of tetrahedrons could be obtained by folding a tetrahedron sequence which satisfies the following conditions (figure 1(b)) : (i) Each tetrahedron consists of four short edges and two long edges, where the ratio of the length is $\sqrt{3}/2$ and (ii) Successive tetrahedrons are connected via a long edge and have a rotational freedom around the
arrows of slant tetrahedrons indicate the direction of “down” in gradient value over a flat triangle by a bold edge as shown in figure 5 (b), where $\pi \in B$. From left to right, $a/x_4[x_1x_2x_3]$ whose gradient is $x_1x_2x_4$, $a[x_1x_2x_3]$ whose gradient is $x_1x_2x_3$, $a[x_1x_2x_3x_4]$ whose gradient is $x_2x_3x_4$, and $ax_1x_2[x_1x_3x_4]$ whose gradient is $x_1x_3x_4$. The arrows of slant tetrahedrons indicate the direction of “down” in $R^4$.

edge. In particular, we could compute the differential structure on a trajectory without considering 4-cubes.

### 3.1 Tangent bundle over flat tetrahedrons

Let’s consider 4-cube $[0,1]^4$ in the four-dimensional Euclidean space $R^4$. Then, the facets of 4-cubes are three-dimensional unit cubes and we divide each of the four facets which contain $(0,0,0,0)$ into six tetrahedrons along diagonal, as shown in figure 5 (a) top.

In the following, we denote point $(k, l, m, n) \in R^4$ by monomial $a^k x_1^l x_2^m x_3^n \in Z[x_1, x_2, x_3, x_4]$. And the tetrahedron of vertices $(k, l, m, n), (k+1, l, m, n), (k+1, l+1, m, n), (k+1, l+1, m+1, n) \in R^4$ are denoted by $a^k x_1^l x_2^m x_3^n [x_1x_2x_3]$. For example, $a[x_i, x_j, x_k]$ is the tetrahedron of vertices $a, ax_i, ax_j, ax_k \in Z[x_1, x_2, x_3, x_4]$.

Let $\pi$ be the projection of the collection $S$ of all the slant tetrahedrons onto the collection $B$ of all the flat tetrahedrons along direction $(-1, -1, -1, -1)$, where the image of $a[x_i, x_j, x_k] \in S$ is denoted by $[a[x_i, x_j, x_k]]$ (figure 5 (b)). Then, projection $\pi$ induces tangent bundle-like structure $TB$ over $B$, where the gradient of slant tetrahedrons are defined as follows:

**Definition 4.** The gradient $Da[x_i, x_j, x_k]$ of $a[x_i, x_j, x_k] \in S$ is monomial $x_i x_j x_k \in Z[x_1, x_2, x_3, x_4]$. In particular, there is a one-to-one correspondence between $TB$ and $\{x_1 x_2 x_3, x_1 x_2 x_4, x_1 x_3 x_4, x_2 x_3 x_4\} \times B$. And we indicate the gradient value over a flat triangle by a bold edge as shown in figure 5 (b), where arrows of slant tetrahedrons indicate the direction of “down” in $R^4$.

For example, slant tetrahedrons $a/x_4[x_1x_2x_3], a[x_1x_2x_3], ax_1[x_2x_3x_4]$, and $ax_1x_2[x_3x_4x_1] \in S$ are projected onto the same flat tetrahedron $[a[x_1x_2x_3]] \in B$.
Figure 6: Local trajectory. (a): The local trajectory specified by $a[x_3x_4x_1] \in S$ at $a[x_3x_4x_1] \in B$, $a[x_3x_4x_1] \in S$ (downward) and $a[x_3x_4x_1] \in S$ (upward) $\in B$. (b): Smoothness condition at $a[x_3x_4x_1] \in B$ (white) specified by $a[x_3x_4x_1] \in S$. The next tetrahedron $a[x_3x_4x_1] \in B$ (grey) could assume either $x_1x_2x_4$ or $x_1x_3x_4$ as gradient.

and their gradients are $x_1x_2x_4$, $x_1x_2x_3$, $x_2x_3x_4$, and $x_1x_3x_4$ respectively (figure 6(b)).

Then, a gradient value over a flat tetrahedron specifies a local trajectory at the flat tetrahedron as follows:

**Definition 5.** The local trajectory defined by $a[x_i,x_j,x_k] \in S$ at $a[x_i,x_j,x_k] \in B$ is the three consecutive flat tetrahedrons $\{a[x_i,x_j,x_k], [a[x_i,x_j,x_k]], [a[x_i,x_j,x_k]]\}$. $a[x_i,x_j,x_k] \in B$. As figure 6(a) shows, these are the adjacent tetrahedrons connected along the direction of the bold edge of $a[x_i,x_j,x_k]$. And the local trajectory is specified uniquely by the gradient of $a[x_i,x_j,x_k]$.  

Now we impose a kind of “smoothness condition” as shown in figure 6(b). That is, each flat tetrahedron assume one of two gradient values, which are determined naturally by the gradient of the preceding tetrahedron. Suppose that the gradient at current tetrahedron $a[x_3x_4x_1] \in B$ is $x_1x_3x_4$ and the gradient at next tetrahedron $a[x_3x_4x_1] \in B$ is either $x_2x_3x_4$ or $x_1x_2x_3$. Then, the bold edges of the two flat tetrahedrons $a[x_3x_4x_1]$ and $a[x_3x_4x_1]$ are not connected smoothly as shown in figure 6(b). In this case, we permit either $x_1x_3x_4$ or $x_1x_2x_4$ as gradient of $a[x_3x_4x_1]$.

As an example, let’s consider a closed trajectory of peaks and valleys specified by three peaks $a = x_1x_2x_4$, $b = x_1x_3x_4$, and $a = x_2x_3x_4 \in Z[x_1, x_2, x_3, x_4]$ (figure 6). Peaks and valleys define a “smooth” vector field $V$ on $B$ by the following mapping:

$$ V: B \rightarrow \{x_1x_2x_3, x_1x_2x_4, x_1x_3x_4, x_2x_3x_4\}, \quad V([a[x_i,x_j,x_k]]) := x_i,x_j,x_k, $$

where $a[x_i,x_j,x_k] \in S$ is the slant tetrahedron on the surface of the peaks and valleys over $a[x_i,x_j,x_k] \in B$.

Let’s start from tetrahedron $a[x_3x_4x_1]$ (grey) and move downward: $t[0] = a[x_3x_4x_1]$ and $V(t[0]) = x_1x_3x_4$. Then, gradient $V(t[0])$ specifies local trajectory $\{V(x_2x_4x_1), a[x_3x_4x_1], a[x_3x_4x_2]\}$ at $t[0]$. Since we move downward,
next tetrahedron $t[1]$ is $b[x_2x_4x_1]$, and we obtain $V(t[1]) = x_1x_2x_4$. Then, gradient $V(t[1])$ specifies local trajectory \{ $b[x_2x_4x_3]$ $|$ $b[x_2x_4x_1]$, $a[x_3x_4x_1]$ \} at $t[1]$. And next tetrahedron $t[2]$ is $b[x_2x_4x_3]$. Continuing the process, we obtain the closed trajectory of tetrahedrons.

Note that the trajectory could be obtained by folding the tetrahedron sequence mentioned above (figure 1 (b)).

3.2 Encoding of the shape of trajectories

Thanks for the smoothness condition again, variation of gradient, i.e., the “second derivative”, along a tetrahedron trajectory is also given as binary valued sequence.

**Definition 6.** The derivative $DV$ of vector field $V$ along trajectory \{ $t[i]$ \} is defined as follows:

$$DV : B \to \{U, D\}, \quad DV(t[i]) := \begin{cases} DV(t[i-1]) & \text{if } V(t[i]) = V(t[i-1]) \\ -DV(t[i-1]) & \text{otherwise,} \end{cases}$$

where $-U := D$ and $-D := U$. In words, change value if the gradient changes.

As an example, let’s consider the trajectory of figure 1 again. First, set any initial value: $DV(t[0]) = D$. Then, since the first two tetrahedrons $t[0]$ and $t[1]$ have different gradient values, $DV(t[1])$ is $U$. The third tetrahedron $t[2]$ assumes yet another gradient and value of the second derivative changes to $U$. Continuing the process, we obtain a binary sequence of length six, $DUDUDU$, which describes the shape of the trajectory.
Figure 8: Encoding of local structure. Left: A polygonal chain which represents the structure of an amino acid fragment to be encoded. Middle: Approximation by a folded tetrahedron sequence. Right: Variation of gradient along the folded tetrahedron sequence.

### 4 Encoding of local protein structure

Now let’s encode local protein structure using variation of gradient along a trajectory of tetrahedrons.

To study the local structure of a protein, i.e., polygonal chain, we consider all the amino acid fragments of length five occurred in the protein. (It will turn out that length five is enough to detect local features.) And polygonal chains are approximated by folded tetrahedron sequences to detect their local features, where we permit translation and rotation during the folding process to absorb irregularity of the structure (figure 8).

Each fragment is approximated by a folded tetrahedron sequence of length five, starting from the middle point amino acid, say A. And variation of gradient along the sequence is computed to encode its structural features. We call the resulting \{D, U\}-valued sequence of length five the 5-tile code of A.

#### 4.1 Encoding algorithm

In the following, we will explain the algorithm of “tetrahedron folding with translation and rotation.” As an example, let’s consider the polygonal chain A[-2]-A[-1]-A[0]-A[1]-A[2] of figure 9(a) and compute the 5-tile code of A[0] using a sequence of five tetrahedrons T[−2]-T[−1]-T[0]-T[1]-T[2].

##### 4.1.1 Step 1

Align tetrahedron T[0] (white) with amino acid A[0] and set initial values (figure 9(b)). In this example, the gradient and second derivative of T[0] is \(x_1 x_2 x_4\) and \(D\) respectively.

Then, the initial positions of adjacent tetrahedrons T[±1] (grey) are also determined, which are moved to the positions of A[±1] respectively later.

##### 4.1.2 Step 2

Assign gradient to adjacent tetrahedrons T[±1] considering the direction of A[±2] respectively (Figure 9(c)). For example, tetrahedron T[1] could assume \(x_1 x_2 x_4\) or \(x_2 x_3 x_4\) as its gradient. And the next tetrahedron (grey) becomes closer to A[2] if \(x_2 x_3 x_4\) is assumed. Thus, the gradient of T[1] is \(x_2 x_3 x_4\) and
Figure 9: Algorithm of the 5-tile coding. (a): Polygonal chain \(AA[-2] \cdots AA[2]\) which represents the structure of an amino acid fragment to be encoded. (b): Step 1. (c): Step 2. (d): Step 3. (e): Step 4. (f): Step 5. (g) Step 6. The character strings show the corresponding sequences of gradients (left) and second derivatives (right), where top are the those of \(T[-2]\) and bottom are those of \(T[2]\). \(K\), \(L\), \(M\), and \(N\) stand for \(x_2x_3x_4\), \(x_1x_2x_3\), \(x_1x_2x_4\), and \(x_1x_3x_4\) respectively.

its second derivative is \(U\) since the gradients of \(T[0]\) and \(T[1]\) are different. In the same way, \(T[-1]\) is assigned \(x_1x_2x_4\) and \(D\) as its gradient and second derivative respectively.

Note that the initial positions of adjacent tetrahedrons \(T[\pm2]\) (grey) are also determined, which are moved to the positions of \(AA[\pm2]\) respectively later.

4.1.3 Step3

Translate tetrahedrons \(T[\pm1]\) to the positions of \(AA[\pm1]\) respectively (Figure 9 (d)). Adjacent tetrahedrons \(T[\pm2]\) (grey) are also moved with \(T[\pm1]\) respectively.

4.1.4 Step4

Rotate tetrahedrons \(T[\pm1]\) at the positions of \(AA[\pm1]\) so that the bold edges become parallel to the direction from \(AA[0]\) to \(AA[\pm2]\) respectively (Figure 9 (e)). Adjacent tetrahedrons \(T[\pm2]\) (grey) are also moved with \(T[\pm1]\) respectively.

4.1.5 Step5

Assign gradient to adjacent tetrahedrons \(T[\pm2]\) considering the direction of \(AA[\pm2]\) respectively (figure 9(f)). For example, tetrahedron \(T[2]\) could assume \(x_1x_3x_4\) or \(x_2x_3x_4\) as its gradient. And the next tetrahedron (not shown) becomes closer to \(AA[2]\) if \(x_1x_3x_4\) is assumed. Thus, the gradient of \(T[2]\) is \(x_1x_3x_4\)
and its second derivative is $D$ since the gradients of $T[1]$ and $T[2]$ are different. In the same way, $T[−2]$ is assigned $x_{1}x_{2}x_{3}$ and $U$ as its gradient and second derivative respectively.

4.1.6 Step6

Translate tetrahedrons $T[±2]$ to the positions of $AA[±2]$ respectively (figure 9 (g)). And we have obtained binary sequence $UDDUD$, the 5-tile code of $A[0]$, which describes the shape of the amino acid fragment shown in figure 9 (a).

4.2 One-letter representation of 5-tile codes

To save space, we use numerals and alphabets to denote 5-tile code $C_{1}C_{2}C_{3}C_{4}C_{5}$. First, compute the value $Y$ of the code which is defined as follows: $Y = 2^{4} * C'_{1} + 2^{3} * C'_{2} + 2^{2} * C'_{3} + 2 * C'_{4} + C'_{5}$, where $C'_{i} = 1$ if $C_{i}$ is equal to $U$ and $C'_{i} = 0$ if not. Then, assign the number to the code if the value $Y$ is less than 10. Otherwise, assign the $(Y−9)$-th alphabet to the code.

For example, $DDDUU$ corresponds to binary number 00011 and $Y = 3$. Thus, 3 is assigned to the code. On the other hand, $DUDUD$ corresponds to binary number 01010 and $Y = 10$. Thus, the first alphabet $A$ is assigned to the code.

4.3 Example: transferase 1RKL

The local structure of transferase 1RKL shown in figure 11 (a) is encoded as follows:

\[
\text{MISDEQLNSL AITFGIVMT LIVIHAVDS TMSPKN} \quad \cdot \cdot 000RQAAA \quad \text{AAAHA} \quad \text{AAAAA} \quad \text{BORO} \quad 000 \cdot \cdot
\]

where the top row shows the amino acid sequence of 1RKL and the bottom shows the corresponding 5-tile codes. As you see, we could capture recurring structural features without any structural templates (figure 1(c)).

In previous works, common short structural motifs (structural templates) of proteins are often identified by clustering a set of representative protein fragments, using unsupervised machine learning. Thus, identification and assignment of such motifs has been the challenges for local structure analysis. And, as a result, their methods could not recognize new local structural features nor structural distortions.

On the other hand, there is no need for identification and assignment of structural templates in our method since we don’t use them at all. And the 5-tile codes could detect both new local features and structural distortions because they are computed directly from atomic coordinates.
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