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A few Applications of Pattern Recognition Techniques to Proteomics

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Overview

- Searching in a database of protein structures
 - Search for a structural "motif"
 - Pairwise comparison
 - All-to-All comparison
- Study the interaction between structures and other molecules (Protein Docking)
 - Molecular surface representation
 - Protein surface comparison
 - Geometric shape descriptors
 - Shape matching algorithms

PDB



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Protein Data Bank (PDB) <u>http://www.rcsb.org/pdb/</u>

| | 0 | 5,000 | 10,000 | 15,000 | 20,000 | 25,000 | Nur 30,000 3 | nber 15,000 40 | 0,000 45,001 | 50,000 | 55,0 | 00 60, | 000 | 65,000 | | | | | |
|---|------|-------|--------|--------|--------|--------|-----------------|-------------------|--------------|--------|------|--------|-----|--------|--------|--------|---------|------|-------|
| | 2010 | 2 | | | | | | | | _ | _ | | | 9 | | | | | |
| | 2009 | 1 | | | | | | | | | | | -10 | | | | | | |
| | 2000 | | | | | | | | - | | -1 | | | | | | | | |
| | 2008 | - | | - | - | | | | | | | | | | | | | | |
| | 2007 | 1 | | | | | | | | | | | | | | | | | |
| | 2006 | | | | | | - | - | 9 | | | | | | | | | | |
| | 2000 | | | | | | | | | | | | | | | | | | |
| | 2005 | -2 | | | | | | | ATOM | 1 | N | GLY | A | 1 | 44.842 | 51.034 | 101.284 | 0.01 | 27.20 |
| | 2004 | | | | | - | | | ATOM | 2 | CA | GLY | A | 1 | 45.640 | 50.230 | 100.389 | 0.01 | 26.99 |
| a | 2004 | | | | | | | | ATOM | 3 | С | GLY | A | 1 | 46.692 | 49.648 | 101.308 | 0.01 | 26.80 |
| × | 2003 | -2 | | | 1 | | | | ATOM | 4 | 0 | GLY | A | 1 | 46.895 | 50.222 | 102.381 | 0.01 | 26.91 |
| | | | | | -2 | | | | ATOM | 5 | N | SER | A | 2 | 47.283 | 48.516 | 100.951 | 1.00 | 26.26 |
| | 2002 | 1 | | | | | | | ATOM | 6 | CA | SER | A | 2 | 48.277 | 47.866 | 101.761 | 1.00 | 26.17 |
| | | | | -0 | | | | | ATOM | 7 | С | SER | A | 2 | 49.212 | 47.031 | 100.845 | 1.00 | 24.21 |
| | 2001 | | | | | | | | ATOM | 8 | 0 | SER | A | 2 | 49.060 | 47.195 | 99.630 | 1.00 | 19.77 |
| | 2000 | | | - | | | | | ATOM | 9 | CB | SER | A | 2 | 47.438 | 47.091 | 102.800 | 1.00 | 26.31 |
| | 2000 | e | 1 | | | | | | ATOM | 10 | OG | SER | A | 2 | 46.276 | 46.356 | 102.404 | 1.00 | 27.99 |
| | 1999 | | | | | | | | ATOM | 11 | N | HIS | A | 3 | 50.147 | 46.186 | 101.370 | 1.00 | 23.93 |
| | | - | | | | | | | ATOM | 12 | CA | HIS | A | 3 | 51.129 | 45.389 | 100.609 | 1.00 | 21.44 |
| | 1998 | 4 | | | | | | | ATOM | 13 | С | HIS | A | 3 | 50.953 | 43.905 | 100.849 | 1.00 | 20.32 |
| | | - | | | | | | | ATOM | 14 | 0 | HIS | A | 3 | 50.530 | 43.595 | 101.950 | 1.00 | 22.00 |
| | 1997 | | | | | | | | ATOM | 15 | CB | HIS | A | 3 | 52.555 | 45.674 | 100.990 | 1.00 | 19.69 |
| | 1996 | 1 | | | | | | | ATOM | 16 | CG | HIS | A | 3 | 52.940 | 47.090 | 100.611 | 1.00 | 21.44 |
| | 1990 | 1 | | | | | | | ATOM | 17 | ND1 | HIS | A | 3 | 53.371 | 47.470 | 99.422 | 1.00 | 20.87 |
| | 1995 | - | | | | | | | ATOM | 18 | CD2 | HIS | A | 3 | 52.956 | 48.175 | 101.433 | 1.00 | 21.69 |
| | | | | | | | | | ATOM | 19 | CE1 | HIS | A | 3 | 53.676 | 48.730 | 99.476 | 1.00 | 20.57 |
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Levels of protein structure representation

- Primary structure
- Secondary structure
- Tertiary structure
- Quaternary structure



Primary structure: the sequence of amino acids



MHGAYRTPRSKTDAYGCQILETRAS

Secondary structures

Three basic components:

- helix
- sheet



Loops (linear connections between the components)

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The helices



- One of the most closely packed arrangement of residues.
- ~40% of residues in globular proteins

The sheet









loosely packed arrangement of residues.

Parallel

Antiparallel Twisted

Secondary Structures Representation

 Secondary structures are represented as linear vectors (segments):

the axis for the helix and the best fit segment for a sheet

 An alignment algorithm is used to match an helix segments with known axes to determine helix axis. Direct segment fits are made to fit sheet strands. Secondary Structure Determination

- Programs: <u>DSSP</u> and <u>STRIDE</u>.
- On the average 4.8% of the target residues were differently assigned, this number reaching 12% for certain targets.

Protein Structure Comparison



Secondary structure representation

- Each secondary structure is displayed as a cylinder
- The protein is represented by and ordered sequence of cylinder with two labels: helices or sheets



GHT applied to proteins

 For every protein, the distance (p) of every secondary structure from a reference point (RP, eg the geometric center of the protein) and the angle (theta) between the direction of the secondary structure in the 3D space and the segment linking the center of that secondary structure with the RP are first calculated. (GH reference table RT)

In the way of GHT (simplified 2D representation)



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In the way of GHT



Query protein



Mapping Rule

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helices and sheets



Votes Space

GHT applied to proteins

- In the 3D space of a given "object protein", every secondary structure of a "model protein" votes a circumference of points starting from every secondary structure of the object protein.
- If the proteins are similar in shape, the circumferences will all intersect in a given point.

GHT and Motif Retrieval

 Other than shape resemblance, the algorithm might be used for motif retrieval, as for instance when the "model protein" is one or many possible motifs to be looked for in a given "object protein".

Generalized Hough Transform (SSS)



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Main characteristics

- the mapping rule, for each compatible correspondent, in 3D is a circle on a plane perpendicular to the axis of the secondary structure
- Other information can be exploited to increase the S/N ratio:
 - the length of the secondary structure
 - the residues properties contained in the SS
 - any other (biochemical, morphological, etc.) peculiarities.

The implementation

- The voting space is smoothed by accumulation of nearby votes (within a given radius) for each point
- After smoothing, the highest peaks in the voting space are detected (avoiding to pick high votes that however are not the top of a peak but lie close to one such peak)
- Only the relevant votes are stored in memory: there isn't a matrix with all the possible cells.

Smoothing Algorithm

- Smoothing is performed by accumulating votes within a given radius, for every point in the vote space.
- The classic version, i.e., checking every vote for the vicinity condition, has been proven to be too time-consuming for applications, with a time complexity of O(n²), where n is the number of votes in the vote space.
- The smoothing problem can be seen as an "orthogonal search" problem, i.e., finding points within a given cube in space.
- A particular structure has been implemented for solving this problem with a O(n log³(n)) complexity: Range Trees.

PROTEIN 1FNB



The protein contains 22 Secondary Structure. Searched motif: Greek key (4 β -sheets). The red circles are the *helices* and the blue circles are the *sheets*. The cyan blue triangles indicate the orientation of the secondary structures. The **black** point is the reference point.

PROTEIN 7FAB



The protein contains 46 Secondary Structure. Searched motif: 3 helix and 2 sheet. The red circles are the *helices* and the blue circles are the *sheets*. The cyan blue triangles indicate the orientation of the secondary structures. The **black** point is the reference point.

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SSC: Secondary Structures Co-occurrences



SST: Secondary Structures Triplets



4 SSs motif: Terns co-occurrence



5 SSs motif: Terns co-occurrence



Type A: π -helix, I_1 Type B: α -helix, I_2 Type C: α -helix, I_3 Type D: α -helix, I_4 Type E: α -helix, I_5

PROTEIN 1FNB



The protein contains 22 Secondary Structure. Searched motif: Greek key (4 β -sheets). The red circles are the *helices* and the blue circles are the *sheets*, in **bold** the motif SSs. July, 2nd 2013

PROTEIN 7FAB



The protein contains 46 Secondary Structure. Searched motif: 3 helices and 2 sheets. The red circles are the *helices* and the blue circles are the *sheets*, in **bold** the motif SSs. July, 2nd 2013 30

Searching performances

Searching a Greek Key motif (4 SSs, all β -sheets) in 1FNB

| Methods | Candidate RP | Error Rate % | SearchTime |
|---------|--------------------|--------------|------------|
| SSS | [31.41 1.16 11.94] | 0.32 | 35.2sec |
| SSC | [31.33 1.08 11.79] | 0.33 | 3.86sec |
| SST | [31.38 1.08 11.69] | 0.00 | 5.76 sec |
| MDM | [31.40 1.12 11.66] | 0.16 | 7.59sec |

Searching a motif with 5 SSs (3 helices and 2 sheets) in 7FAB

| Methods | Candidate RP | Error Rate % | Search Time |
|---------|---------------------|--------------|-------------|
| SSS | [-17.56 9.46 15.17] | 0.28 | 108sec |
| SSC | [-17.40 9.14 15.48] | 0.34 | 42.51sec |
| SST | [-17.48 9.17 15.48] | 0.00 | 48.89sec |
| MDM | [-17.45 9.16 15.50] | 0.15 | 112.17sec |

PV Benchmark (20 proteins)

| PDB ID | Description | Number of SSs |
|--------|--|---------------|
| 2Z98 | FMN-dependent NADH-azoreductase | 14 |
| 2Z9C | FMN-dependent NADH-azoreductase | 15 |
| 2Z9B | FMN-dependent NADH-azoreductase | 16 |
| 4GCR | GAMMA-B CRYSTALLIN | 18 |
| 3E9O | Pre-mRNA-splicing factor 8 | 21 |
| 1FNB | FERREDOXIN-NADP+ REDUCTASE | 22 |
| 3E9L | Pre-mRNA-processing-splicing factor 8 | 23 |
| 2PZN | Aldose reductase | 24 |
| 3C3U | Aldo-keto reductase family 1 member C1 | 26 |
| 2Z7G | Adenosine deaminase | 28 |
| 2Z7K | Queuine tRNA-ribosyltransferase | 33 |
| 2PRL | Dihydroorotate dehydrogenase, mitochondrial | 34 |
| 2QX8 | Ribosyldihydronicotinamide dehydrogenase [quinone] | 35 |
| 2QMY | Ribosyldihydronicotinamide dehydrogenase | 36 |
| 3C94 | Exodeoxyribonuclease I | 37 |
| 2QX9 | Ribosyldihydronicotinamide dehydrogenase [quinone] | 38 |
| 3C95 | Exodeoxyribonuclease I | 39 |
| 3DC7 | Putative uncharacterized protein lp_3323 | 43 |
| 3DHP | Alpha-amylase 1 | 44 |
| 7FAB | IGG1-LAMBDA NEW FAB (LIGHT CHAIN) | 46 |

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PV Benchmark: basic features



PV Benchmark: performances



Average performances

SSC

| Number of proteins | Number of SSs per motif | Total number of motifs | Total Searching Time (sec) | Average Searching Time per motif (msec) |
|--------------------------|----------------------------|---------------------------|----------------------------------|---|
| 20 | 3 | 105971 | 119.882 | 1.1 [0.55-1.48] |
| 20 | 4 | 918470 | 1275.585 | 1.4 [0.51-1.75] |
| 20 | 5 | 6455009 | 11261.911 | 1.7 [0.52-2.19] |

SST

| Number of proteins | Number of SSs per motif | Total number of motifs | Total Searching Time (sec) | Average Searching Time per motif (msec) |
|--------------------------|----------------------------|---------------------------|----------------------------------|---|
| 20 | 3 | 105971 | 768.508 | 7.3 [0.9-11.7] |
| 20 | 4 | 918470 | 10303.806 | 11.2 [1.2-16.9] |
| 20 | 5 | 6455009 | 111809.428 | 17.3 [1.4-24.4] |

EGI - protein 1aa01





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Future outlook

- These algorithms are extremely interesting, not to the least for its capability to be executed on parallel computing systems (thanks to the inherent parallel nature thereof).
- The accuracy of the comparison has the potential to be increased to competitive standards with respect to other approaches.

Example of match with cardinality 10 between proteins 2qx8 (source) and 2qx9 (search)



Study interaction between structures and other molecules (Protein Docking) through mathematical morphology

Surface Modeling



Mathermatical morphology: Closing Operation



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Protein-ligand docking

- A large molecule (receptor) and a small molecule (ligand) docking in a cavity.
- Key in lock



42

Docking Protein-Protein Interfaces



The docking site is more planar rather than a cavity.



Accuracy vs coverage

- Accuracy: how many of the solutions found were correct?
 A= (F ∩ T) /F
- Coverage: How many of the correct solutions were found? $C=(F \cap T)/T$



Convex Hull



Note that straight lines (facet in 3D) are represented in a point in the EGI, and their 'spike' can be easily eliminated, leaving only the true contour (surface) of the object (protein).



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A CH of a molecule (1MK5)

- The CH of a molecule is the smallest convex polyhedron that contains the molecule
- A practical *O(n log n)* algorithm for general dimensions CH computing, is Quickhull (Barber, 1996)



Propagation in concavity volume from CH





Propagation in the concavitiy volume

- The *concavity volume* is the region R between the CH and the SES
- B_{CH} is the set of the border voxels of CH
- A is the increasing set of voxels contained in R; E is to the recruited near neighbors
- d_n + w_n is the minimum distance in the near neighbors by the displacement w

```
\begin{array}{l} \mathsf{D}_{i} = \left\{ \begin{array}{l} 1 \ iff \ i \in \mathsf{B}_{\mathsf{CH}} \\ 0 \ otherwise \end{array} \right\} \\ \mathsf{A} = \mathsf{B}_{\mathsf{CH}}; \\ \mathsf{N} = (\mathsf{A} \oplus \mathsf{K}) \cap \mathsf{R}; \\ \mathsf{E} = \mathsf{N} - \mathsf{A}; \\ \mathsf{while} \ \mathsf{E} \neq \varnothing \ \mathsf{do} \\ \forall \mathsf{e} \in \mathsf{E} : d_{e} = \min_{n \in nn_{e}} \ (d_{n} + w_{n}); \\ \mathsf{A} = \mathsf{N}; \\ \mathsf{N} = (\mathsf{A} \oplus \mathsf{K}) \cap \mathsf{R}; \\ \mathsf{E} = \mathsf{N} - \mathsf{A}; \\ \mathsf{done} \end{array}
```

Propagation in the CV: 2D sketch



Back-propagation for pockets and tunnels search (1MK5)



Pockets detected on the 1MK5



Protein Inspector - finding pockets







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Protein Inspector: pockets refinement

All pockets at once with TD constraint







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Protein Inspector - pockets refinement

Single pocket with TD and ligand box constraints



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The main pocket



The second main pocket



The main pocket of 1MK5



- Adjacent atoms
 - green from $CAST_p$
 - Red and green from PPS

1MK5



Protein Inspector - finding protrusions

Segmentation of protrusions for a 2D shape







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Protein Inspector - finding protrusions





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Protein inspector - user interface



2D sketch for searching protuberances

| | ******** | | *************************************** | | *** | **** | | ***** | |
|------|--------------|---|---|--|-----|------|--|-------|--|
| **** | | • | | | | **** | | | |

Active sites matching

- Two data structures are proposed:
 - the Concavity Tree Arcelli C, Sanniti di Baja G (1978) "Polygonal covering and concavity tree of binary digital pictures", Proceeding International Conference MECO '78, Athens, pp. 292-297.
 - the Extended Gaussian Image

Hu, Z., · Chung, R., · Fung K. S. M., (2010), "EC-EGI: enriched complex EGI for 3D shape registration", Machine Vision and Applications, 2, 177-188.

Concavity Tree





Protein 1MK5





Protein 1MK5 concavity tree



Concavity tree



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CT: node content

- Volume
- Surface to Volume Ratio
- Skewness of Height Distribution
- Kurtosis of Height Distribution
- Mouth Aperture

- Travel Depth
- Top Five Peaks and Valleys
- Summit Density
- Mean Summit Curvatures
- Interfacial Area Ratio
- Residue Conservation

Extended Gaussian Image

The EGI of a 3D object or shape is an orientation histogram that records the distribution of surface area with respect to surface orientation.



Extended Gaussian Image





Hierarchical Search



Funk et al., SIGGRAPH 2004




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Funk et al., SIGGRAPH 2004







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77

Funk et al., SIGGRAPH 2004









Extended Gaussian Image

 EGI The EGI of a 3D object or shape is an orientation histogram that records the distribution of surface area with respect to surface orientation.

Horn, B.K.P.: Extended Gaussian images. In: Proceedings of the IEEE 72, 1671-1686 (1984).

• CEGI The Complex EGI encodes each surface patch's signed perpendicular distance from the reference coordinate center.

Kang, S.B., Ikeuchi, K.: Determining 3-D object pose using the complex extended Gaussian image. In: IEEE Computer Society Conference on Computer Vision and Pattern Recognition, pp. 580–585 (1991).

 MEGI The More Extended Gaussian Image (MEGI) model consists of a set of position vectors for surfaces and their normal vectors.

H. Matsuo and A. Iwata, "3-D Object Recognition Using MEGI Model from Range Data." Proc. 12th Int'l Conf. Pattern Recognition, pp. 843-846, Jerusalem, Israel, Oct. 1994.

• MSEGI The multi shell EGI

Dingwen Wang1,2, Jiqi Zhang2, Hau-San Wong2, and Yuanxiang Li, 3D Model Retrieval Based on Multi-Shell Extended Gaussian Image, G. Qiu et al. (Eds.): VISUAL 2007, LNCS 4781, pp. 426–437, 2007.

ECEGI The Enriched CEGI encodes each surface patch's signed with its 3D position.

Zhaozheng Hu · Ronald Chung · Kenneth S. M. Fung: EC-EGI: enriched complex EGI for 3D shape registration, Machine Vision and Applications (2010) 21:177–188.

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Complex EGI (C-GEI)

CEGI encodes each surface patch's signed perpendicular distance from the reference coordinate center Gauss mapping





Hu, Z., · Chung, R., · Fung K. S. M., (2010), "EC-EGI: enriched complex EGI for 3D shape registration", Machine Vision and Applications, 2, 477-188.

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Enriched C-E.G.I.





EGI in progress



Alessandro Gaggia

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SNORMALS SELECT COMBOBOX FOR EGI 16 QUIT PROGRAM

15 CHECK FOR MODELS' SIMILARITY

7 MODEL COLOUR COMBOBOX

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23 NUMBER OF FACES OF THE CURRENT EGI

Protein-Ligand Interaction



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CAPRI: Critical Assessment of PRediction of Interactions

 CAPRI (Critical Assessment of PRedicted Interaction) is an international effort, aimed at objectively assessing the performance of these methods by inviting developers to test their algorithms on the same protein targets and objectively evaluating the results.

