

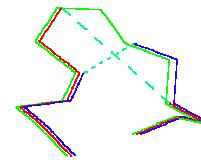
# Structural Alignment of Proteins

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Princeton University  
CS597A, Fall 2007

## Goal

Align protein structures

1 2 3 4 5 6 7 8 9 10 11 12 13 14  
PHE ASP ILE CYS ARG LEU PRO GLY SER ALA GLU ALA VAL CYS  
PHE ASN VAL CYS ARG THR PRO --- --- --- GLU ALA ILE CYS  
PHE ASN VAL CYS ARG --- --- --- THR PRO GLU ALA ILE CYS



[Marian Novotny]

## Terminology

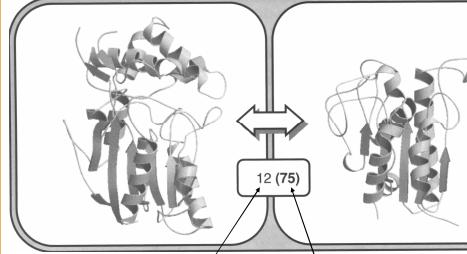
### Superposition

- Given correspondences, compute optimal alignment transformation, and compute alignment score

### Alignment

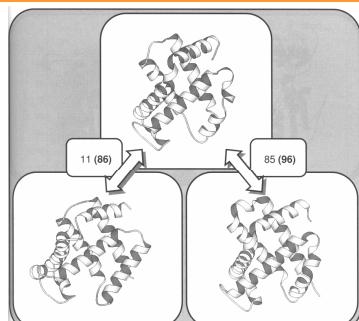
- Find correspondences, and then superpose structures

## Structure vs. Sequence



[Orengo04, Fig 6.2]

## Structure vs. Sequence



[Orengo04, Fig 6.1]

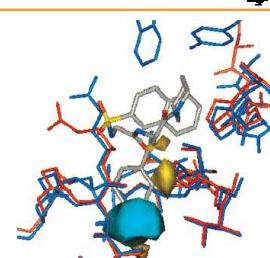
## Applications

Fundamental step in:

- Analysis
- Visualization
- Comparison
- Design

Useful for:

- Structure classification
- Structure prediction
- Function prediction
- Drug discovery



Comparison of S1 binding pockets of thrombin (blue) and trypsin (red).

[Katzenhofer09]

## Goals

Desirable properties:

- Automatic
- Discriminating
- Fast

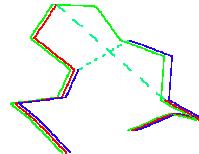


## Theoretical Issues

NP-complete problem

- Arbitrary gap lengths
- Global scoring function

1 2 3 4 5 6 7 8 9 10 11 12 13 14  
PHE ASP ILE CYS ARG LEU PRO GLY SER ALA GLU ALA VAL CYS  
PHE ASN VAL CYS ARG THR PRO --- --- --- GLU ALA ILE CYS  
PHE ASN VAL CYS ARG --- --- --- THR PRO GLU ALA ILE CYS



## Methodological Issues



Choices:

- Representation
- Scoring function
- Search algorithm

## Methodological Issues



Factors governing choices:

?

## Methodological Issues



Factors governing choices:

- Application: homology detection, drug design, etc.
- Granularity: atom, residue, fragment, SSE
- Representation: inter-molecular, intra-molecular
- Scoring: geometric, gaps, chemical, structural, etc.
- Correspondences: sequential, non-sequential
- Gap penalty: expect gaps near loops, etc.
- Flexibility: rigid, flexible
- Target: single protein, representative proteins, PDB

## Methodological Issues



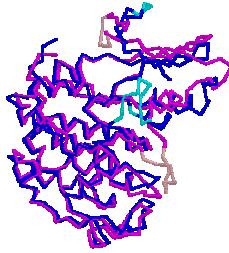
Representations:

- Residue positions
- Local geometry
- Side chain contacts
- Distance matrices (DALI)
- Properties (COMPARER)
- SSEs (SSM, VAST)
- Geometric invariants

## Methodological Issues

Scoring functions:

- Distances (RMSD)
- Substitutions
- Gaps



## Methodological Issues

Search algorithms:

- Heuristics (CE)
- Monte Carlo (DALI, VAST)
- Dynamic programming (STRUCTAL, SSAP)
- Graph matching (SSM)

## Outline

Alignment issues

Example alignment methods ←

Fold prediction experiment

Function prediction experiment

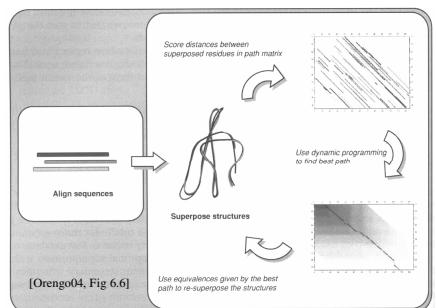
## Example Methods

SSAP	Taylor & Orengo, 1989
STRUCTAL	Subbiah, Laurents & Levitt, 1993 Gerstein & Levitt 1998
DALI	Holm & Sander, 1993 Holm & Park, 2000
DEJAVU /LSQMAN	Kleywegt, 1996
CE	Shindyalov & Bourne, 1998
SSM	Krissinel & Henrick, 2003

+ 30 others!

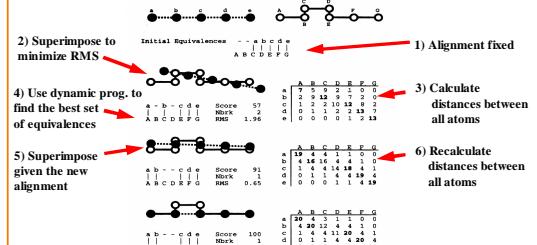
Slide by Rachel Kolodny

## STRUCTAL



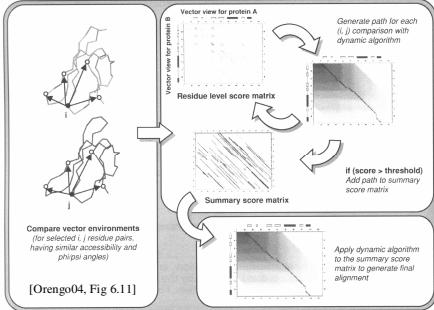
[Subbiah93, Gerstein98]

## STRUCTAL



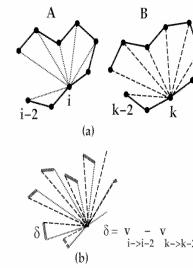
[Subbiah93, Gerstein98]

## SSAP



[Orengo96]

## SSAP

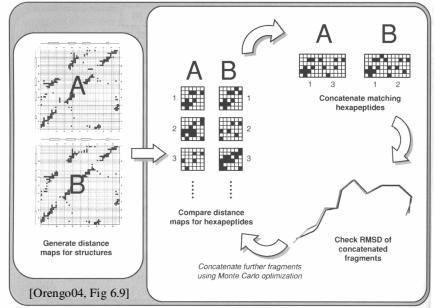


Protein A							
H	S	E	R	R	H	V	F
G	28	10	10	10	10	10	10
Q	10	28	10	10	10	10	10
V	10	10	28	10	10	10	10
G	10	10	10	28	10	10	10
M	10	10	10	10	28	10	10
A	10	10	10	10	10	28	10
C	10	10	10	10	10	10	28

Protein B							
D	I	L	R	K	H	V	F
G	10	10	10	10	10	10	10
Q	10	10	10	10	10	10	10
V	10	10	10	10	10	10	10
G	10	10	10	10	10	10	10
M	10	10	10	10	10	10	10
A	10	10	10	10	10	10	10
C	10	10	10	10	10	10	10

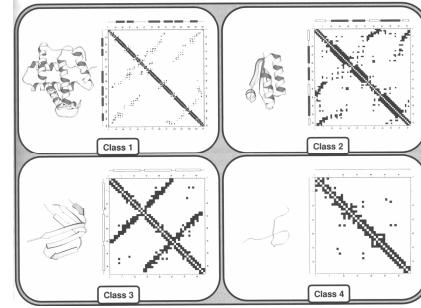
[Orengo96]

## DALI



[Holm93]

## DALI

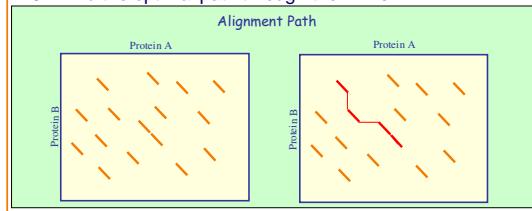


[Orengo04, Fig 6.7]

## CE

### Basic steps:

1. Compare octameric fragments to create candidate aligned fragment pairs (AFP)
2. Stitch together AFPs according to heuristics
3. Find the optimal path through the AFPs



## SSM

### Two-step solution:

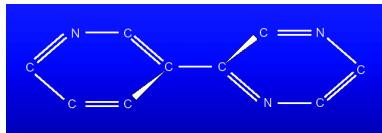
1. Graph representation of structures
2. Graph matching



## SSM

Slide by Eugene Krissel

### Graph representation of molecular structures



- Simple and intuitive, however results in intractably large graphs for proteins
- Solution: build graphs over stable substructures, such as secondary structure elements (SSEs). Having a correspondence between SSEs, one may use that for the 3D alignment of all core atoms.

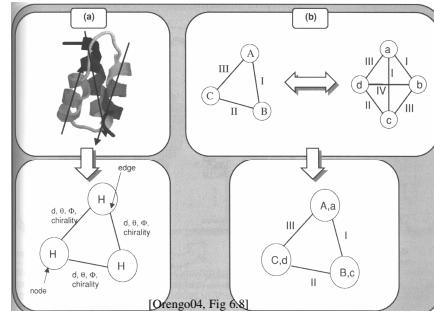
Slide by Eugene Krissel

## SSM



Slide by Eugene Krissel

### SSM



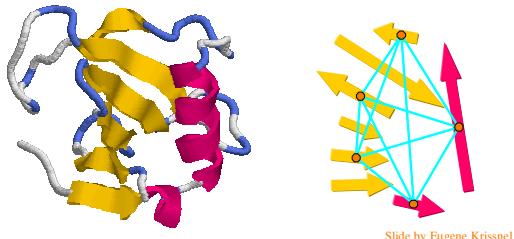
[Orengo04, Fig 6.8]

## SSM

Slide by Eugene Krissel

### Graph representation of protein SSEs

E. M. Mitchell et al. (1990) J. Mol. Biol. 212:151  
A. P. Singh and D. L. Brutlag (1997) ISMB-97 4:284



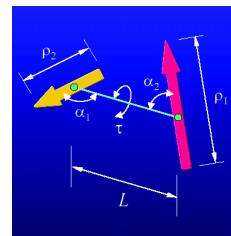
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## SSM



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### Protein graph labeling



#### Composite label of a vertex

- type - helix or strand
- length r

#### Composite label of an edge

- length L (directed if connects vertices from the same chain)
- vertex orientation angles  $\alpha_1$  and  $\alpha_2$
- torsion angle  $\tau$

Vertex and edge labels are matched with thresholds on particular quantities

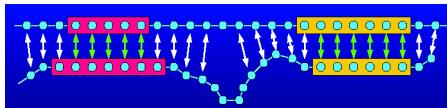
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## SSM

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### $C_\alpha$ alignment

- $C_\alpha$ -alignment is used as an initial guess for  $C_\alpha$ -alignment
- $C_\alpha$ -alignment is an iterative procedure based on the expansion of shortest contacts at best superposition of structures



- $C_\alpha$ -alignment is a compromise between the alignment length  $N_a$  and r.m.s.d. The optimised quantity is

$$Q = \frac{N_a^2}{(1 + (r.m.s.d./R_0)^2) N_1 N_2}$$

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## SSM



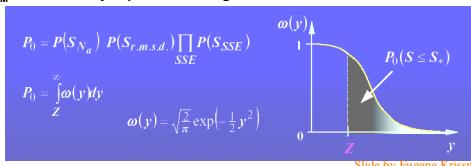
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### Statistical significance of match

- The overall probability of getting a particular match score by chance is the measure of the statistical significance of the match

$$P_{value} = 1 - \left( 1 - P(S_{N_a}) P(S_{r.m.s.d.}) \prod_{SSE} P(S_{SSE}) \right)^{N_{combinations}}$$

- $P_M$  is traditionally expressed through so-called Z-characteristics



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## SSM

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### SSM output

- Table of matched Secondary Structure Elements (SSE alignment)
- Table of matched core atoms ( $C_a$ - alignment) with dists between them
- Rotational-translation matrix of best structure superposition
- R.M.s.d. of  $C_a$ - alignment
- Length of  $C_a$ - alignment  $N_a$
- Number of gaps in  $C_a$ - alignment  $N_g$
- Quality score Q
- Probability estimate for the match  $P_M$
- Z-characteristics
- Sequence identity

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## SSM

### List of matches

#### Structure Alignment Results

Selection of output

Query: pdb entry 1ldc chain A : 479 residues  
L-LACTATE DEHYDROGENASE: CYTOCHROME C OXIDOREDUCTASE 1LDc 3 (FLAVOCYTOCHROME B+D)  
(E.C.1.1.2.3) MUTANT WITH TYR143 1LDc replaced by PHE (Y143P) COMPLEXED WITH PYROVATE 1LDc 5

Examined 18295 entries (39511 chains),  
Matches 1–14 of 14.

Back to query...  
reset results

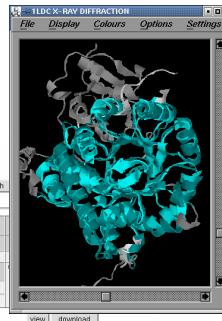
#	Scoring	Query	Target (PDB entry)
	Q P Z	Rmsd N <sub>seq</sub> N <sub>g</sub> % <sub>seq</sub> % <sub>seq</sub>	Match % <sub>seq</sub> N <sub>res</sub> Title
1	1.00 82.6 27.4	0.00 478 0 100 100	1ldc:A 100 478 L-LACTATE DEHYDROGENASE: CYTOCHROME C OXIDOREDUCTASE 1LDc 3 (FLAVOCYTOCHROME B+D) (E.C.1.1.2.3) MUTANT WITH TYR143 1LDc replaced by PHE (Y143P) COMPLEXED WITH PYROVATE 1LDc 5
2	0.99 62.7 23.8	0.30 478 1 100 100	1lcoo:A 91 480 FLAVOCYTOCHROME B+D (E.C.1.1.2.3)
3	0.98 59.5 23.1	0.41 478 1 100 100	1ltdd:A 91 481 FLAVOCYTOCHROME B+D (E.C.1.1.2.3)FCB 3
4	0.94 59.1 23.1	0.56 478 1 100 97	1fob:A 91 494 CRYSTALLOGRAPHIC STUDY OF THE RECOMBINANT L-LACTATE DEHYDROGENASE CYTOCHROME C OXIDOREDUCTASE 1LDc 5 (FLAVOCYTOCHROME B+D) COMPLEXED WITH PYROVATE 1LDc 5
5	0.91 55.4 22.3	0.51 474 2 98 97	1kb1:A 86 504

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## SSM

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### Match details



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## SSM

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### SSE alignment

#### Secondary Structure Alignment

Query PDB 1ldc:A Target PDB 1thw:A

101 HL 131 A ASN 124  L ER 136	<=> 2 HL 131 A ASN 7  L ER 19
111 HL 101 A THR 137  I SER 146	<=> 2 HL 101 A PRO 20  I LEU 29
121 HL 101 A PRO 152  I ALA 141	<=> 3 HL 101 A PRO 34  I VAL 43
131 HO 101 A SER 153  I SER 142	<=> 3 HO 101 A SER 35  I SER 43
141 HO 101 A GDX 208  GLY 217	<=> 9 HL 141 A LYS 89  GLY 102
151 HO 101 A SER 209  GLY 218	<=> 10 HL 151 A SER 90  GLY 103
161 HL 8 A SER 234  ALA 241	<=> 11 HL 9 A SER 114  CTD 122
171 BD 5 A GIM 249  I ER 253	<=> 12 BD 5 A LRR 126  LRR 130
181 BD 5 A LRR 250  I ER 254	<=> 13 BD 5 A LRR 127  LRR 131
191 BD 4 A LBD 277  THR 280	<=> 14 BD 5 A THR 152  THR 156
201 HS 8 A ARG 289  LYS 296	<=> 15 H1 7 A ARG 165  ASN 171
211 HO 8 A ARG 290  LYS 297	<=> 16 HO 8 A ARG 166  ASN 172
221 BD 6 A ILE 346  VAL 351	<=> 19 BD 7 A LYS 227  ILE 233
231 HO 11 A ARG 353  I LEU 363	<=> 20 HO 11 A ARG 235  I GLD 245
241 BD 16 A ILE 364  VAL 375	<=> 21 BD 16 A ILE 364  VAL 375
251 HO 16 A ALA 383  I PR 398	<=> 23 H1 10 A VAL 261  GLY 278
271 BD 11 A ILE 404  VAL 415	<=> 24 BD 11 A ILE 404  VAL 415
281 BD 11 A ILE 414  I ER 424	<=> 25 H1 11 A ARG 289  LRR 299
291 BD 3 A GLY 426  I ER 431	<=> 26 BD 4 A ALA 303  LRR 306
301 HO 3 A GLY 427  I ER 432	<=> 27 HO 3 A GLY 427  I ER 432

OCA SCOP domain | SCOP family

GeneCensus | FSSP | 3Dex | CATH | PDBsum

SWISS-PROT | PROTEIN ATLAS | BRENDA

ProteinNet | MDP | PDB2GO | GOX | SGD

view download sequence view superposed view view download sequence

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## SSM

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### C - alignment

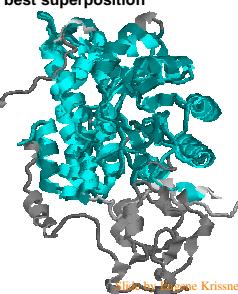
Rotation-translation matrix:  
(to be applied to the query)

$$-0.710 \quad 0.423 \quad -0.563 \quad \boxed{X} \quad \boxed{Y} \quad \boxed{Z} \quad \boxed{98.778}$$

$$-0.471 \quad 0.309 \quad 0.826 \quad \times \quad + \quad \boxed{96.485}$$

$$0.524 \quad 0.852 \quad -0.020 \quad \boxed{Z} \quad + \quad \boxed{-59.445}$$

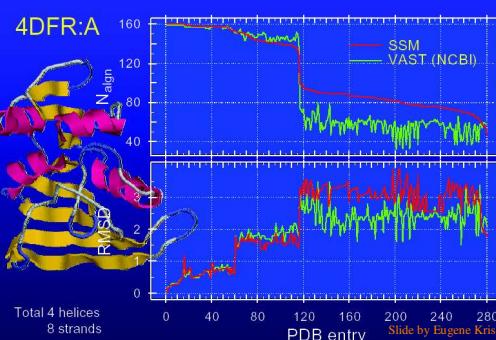
### Rotational-translation matrix of best superposition



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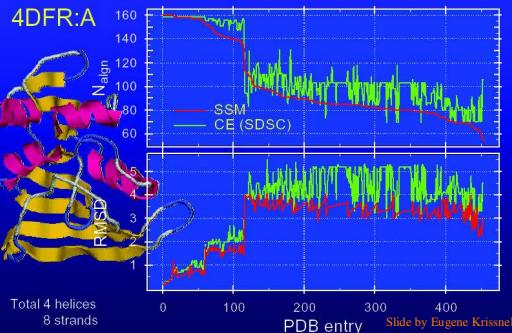
## SSM Results

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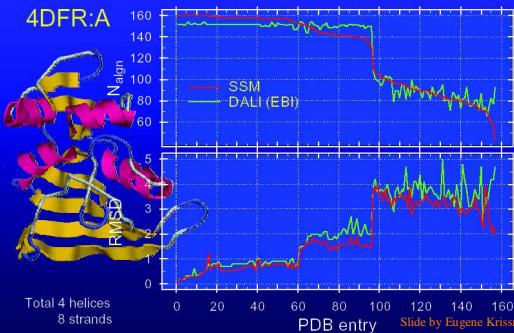
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## SSM Results



Slide by Eugene Krissinel

## SSM Results



Slide by Eugene Krissinel

## Outline

- Alignment issues
- Example alignment methods
- Fold prediction experiment ←
- Function prediction experiment

## Fold Prediction Experiments

Evaluate how useful alignment algorithms are for predicting a protein's fold

How?

## Fold Prediction Experiments

- Kolodny, Koehl, & Levitt [2005]
  - ROC curves and geometric measures using CATH
- Sierk & Pearson [2004]
  - ROC curves using CATH
- Novotny et al. [2004]
  - Checked a few dozen cases using CATH
- Leplae & Hubbard [2002]
  - ROC curves using SCOP

## Fold Prediction Experiments

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  - Checked a few dozen cases using CATH
- Leplae & Hubbard [2002]
  - ROC curves using SCOP

## Kolodny, Koehl, & Levitt [2005]



Large scale alignment study

- 2,930 structures (all pairs)
- 6 structural alignment algorithms
- 4 geometric scoring functions
- Evaluation with respect to CATH topology level
- 20,000 hours of compute time

## Tested Methods



SSAP	Taylor & Orengo, 1989
STRUCTAL	Subbiah, Laurents & Levitt, 1993 Gerstein & Levitt 1998
DALI	Holm & Sander, 1993 Holm & Park, 2000
DEJAVU /LSQMAN	Kleywegt, 1996
CE	Shindyalov & Bourne, 1998
SSM	Krissinel & Henrick, 2003
Best-of-All	Best of above methods

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## Scoring Functions



Consider # aligned residues & geometric similarity:

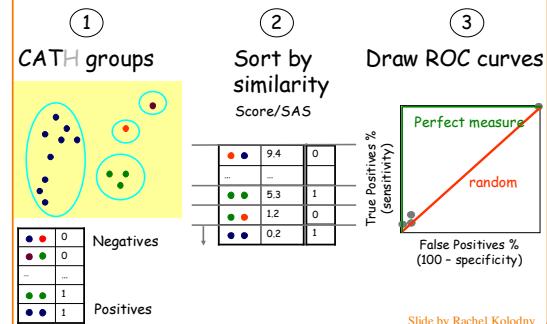
$$SAS = \frac{RMSD \times 100}{N_{mat}}$$

Also penalize gaps:

$$GSAS = \begin{cases} if(N_{mat} > N_{gap}) & \frac{RMSD \times 100}{N_{mat} - N_{gap}} \\ else & 99.9 \end{cases}$$

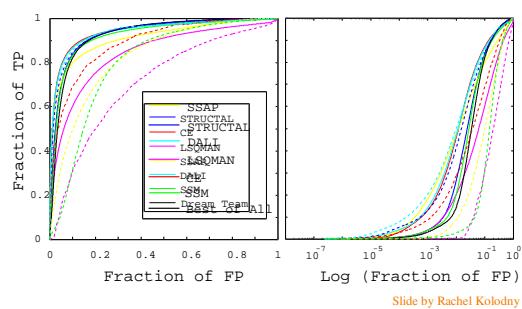
[Kolodny05]

## Evaluation Using ROC Curves



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## SAS & Native ROC Curves



## ROC Curve Issues



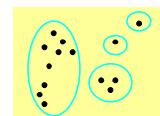
Uses only internal ordering

- Estimation of similarity can be very wrong

● ●	9.4	—
—	—	—
● ●	5.3	1
● ●	1.2	0
● ●	0.2	1

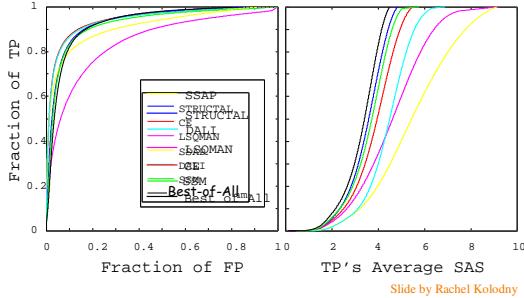
Native scores or SAS

Converts a classification gold standard into binary truth

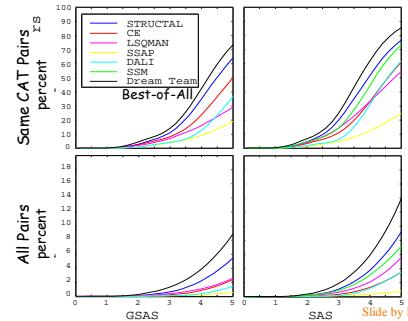


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## Comparing SAS Values Directly



## GSAS & SAS Distributions



## Contributions to “Best-of-All”

	Total	SSAP	STRUCTAL	DALI	LSQMAN	CE	SSM
GSAS≤5 Å	275,547 (100%)	832 (0.3%)	<b>199,871</b> (69%)	5868 (2.1%)	54,606 (20%)	24,370 (8.8%)	—
SAS≤5 Å	530,225 (100%)	498 (0.09%)	<b>286,972</b> (53%)	15,648 (2.9%)	103,408 (19.2%)	15,844 (2.9%)	117,385 (21.8%)
SI≤5 Å	978,531 (100%)	3745 (0.4%)	<b>497,330</b> (51%)	24,767 (2.5%)	201,202 (21%)	17,142 (1.8%)	234,345 (24%)
MI≤0.8	880,503 (100%)	4579 (0.5%)	<b>373,542</b> (65%)	31,402 (3.6%)	63,088 (7.2%)	72,974 (8.3%)	134,918 (15.3%)

The absolute number of alignments contributed by each method is listed and the percentage of alignments is given in parentheses. The largest contributor is shown in bold.

[Kolodny05]

## Outline

### Alignment issues

### Example alignment methods

### Fold prediction experiment

### Function prediction experiment



## Function Prediction Experiment

Evaluate how useful alignment methods are for predicting a protein's molecular function

How?

## Data Set

Proteins crystallized with bound ligands

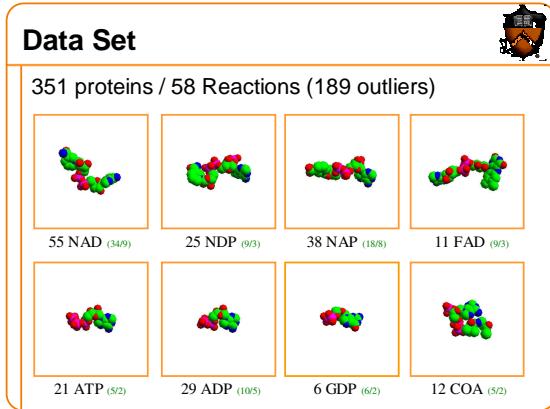
- PDB file must have resolution  $\leq 3$  Angstroms
- Ligands must have  $\geq 20$  HETATOMS

Classified by reaction/reactant

- PDB file must have an EC number (enzymes only)
- EC number must have a KEGG reaction with a reactant whose graph closely matches ligand in PDB file

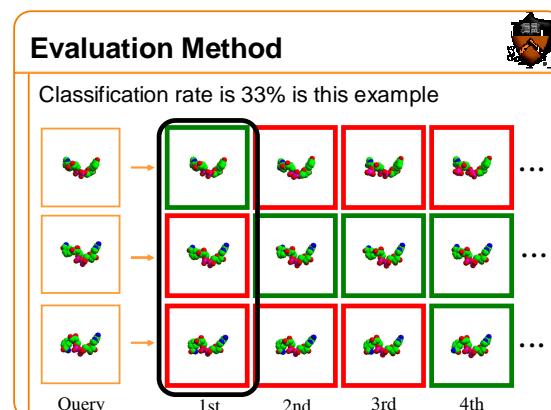
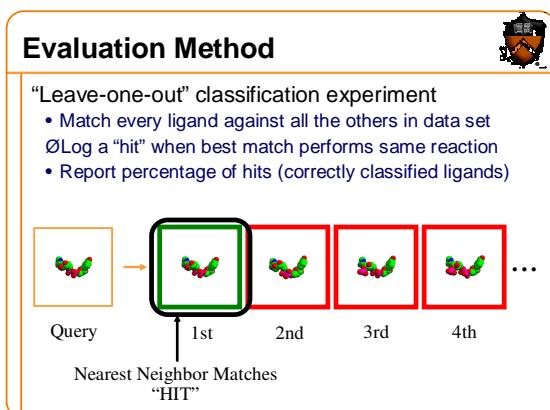
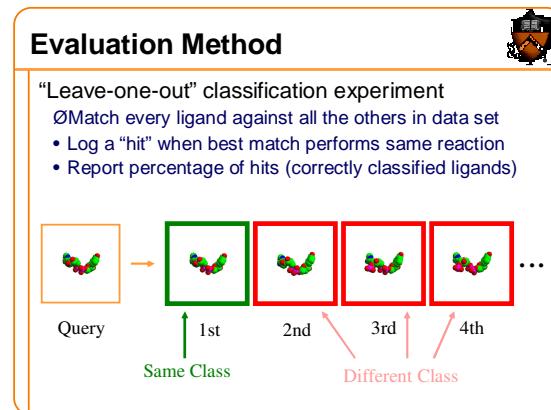
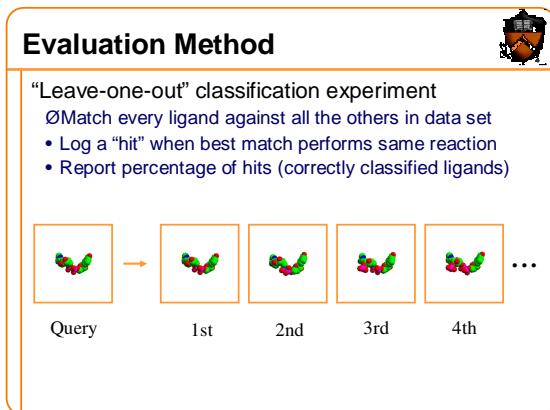
Non-redundant

- No two ligands contacting domains with same CATH S95
- No two ligands contacting domains with same SCOP SP
- No two ligands from same PDB file



### Data Set

REACTION	NAME	#	REACTION	NAME	#	REACTION	NAME	#
R00145	NAD	2	R00162	ATP	3	R00408	FAD	5
R00214	NAD	2	R003647	ATP	2	R00924	FAD	2
R00342	NAD	7	R00124	ADP	2	R00115	FAD	2
R00538	NAD	3	R00497	ADP	2	MISC	FAD	2
R00623	NAD	5	R00756	ADP	2	R00351	CQA	3
R00703	NAD	5	R01512	ADP	2	R03521	COA	2
R00741	NAD	1	R00757	ADP	2	R00321	COA	2
R01403	NAD	1	R00847	AMP	2	R02961	SAM	3
R01778	NAD	3	R00330	GDP	2	MISC	SAM	3
R00012	NAP	2	R01135	GDP	4	R03552	ACO	2
R00343	NAP	2	R01130	IMP	3	R00291	GDU	2
R00625	NAP	2	R02094	TMP	2	R03522	GTT	12
R00939	NAP	2	R02101	UMP	6	R01146	PQQ	3
R01046	NAP	4	R00365	IMP	2	R00190	U	2
R00258	NAP	1	R00362	USP	1	R00320	MTA	2
R01195	NAP	2	R01229	SGP	2	R03435	IPF	1
R02477	NAP	2	MISC	ATP	16	R02886	CBI	4
R00703	NAI	2	MISC	ADP	19	R01590	ACD	2
R00939	NDP	5	MISC	AMP	10	R00529	ADX	2
R01063	NDP	2	MISC	A3P	5	R03491	SIA	2
R01195	NDP	2	MISC	GDP	2	R00137	NMN	3
R01200	NDP	1	MISC	UDP	4	R00222	MIA	2
MISC	NAH	20	MISC	UMP	1	R03609	13T	2
MISC	NAI	2	MISC	SGP	1	MISC	etc	etc
	NDP	16						



## Sequence Alignment Method

Use FASTA to compute Smith-Waterman score for every pair of SCOP domains contacting ligand

```
> fasta34 dlgv0a diguya
  10      20      30      40      50      60
dlgv0a AGVLDSARISFIAIMELGQVQDVTAQVLCGHDAMVYKVTTFVAGIEADLISABRA
.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.
diguya AGVLDAAARYRTPIAMEAGVSVEDVQAMLGGHDEMVPLPFPSTISGIPFVSEFIAPRLA
  10      20      30      40      50      60
  70      80      90      100     110     120
dlgv0a ELVERTRITGAEIVNHLKGGSAYFSPATSVEMEVESIVLDKRVLTCAVSLDQGVGDGT
.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.
diguya QIVERTRKGGEBIVNLTKTGSAAYYAPAAAATQMVVEAVLKDKKRVMPPAAYLTQQYGLNDI
  70      80      90      100     110     120
  130     140     150     160
dlgv0a FVGVPVKLGKNGVREHYEIKLDQSDLLQKSAKIVDENCKML
.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.
diguya YFGVFLVILGAGGVEKILELPNEERMALLNASAKAVRATLDTL
  130     140     150     160
54.48% identity
156 out of 163 amino acids overlap
Smith-Waterman score: 588
```

## Sequence Alignment Method

Use FASTA to compute Smith-Waterman score for every pair of SCOP domains contacting ligand

```
> fasta34 dlgv0a diguya
  10      20      30      40      50      60
dlgv0a AGVLDSARISFIAIMELGQVQDVTAQVLCGHDAMVYKVTTFVAGIEADLISABRA
.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.
diguya AGVLDAAARYRTPIAMEAGVSVEDVQAMLGGHDEMVPLPFPSTISGIPFVSEFIAPRLA
  10      20      30      40      50      60
  70      80      90      100     110     120
dlgv0a E
diguya D
  130     140     150     160
dlgv0a FVGVPVKLGKNGVREHYEIKLDQSDLLQKSAKIVDENCKML
.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.:.
diguya YFGVFLVILGAGGVEKILELPNEERMALLNASAKAVRATLDTL
  130     140     150     160
54.48% identity
156 out of 163 amino acids overlap
Smith-Waterman score: 588
```

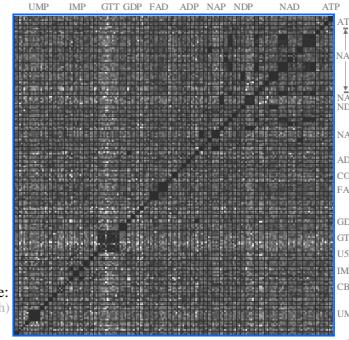
## Sequence Alignment Method

Use FASTA to compute Smith-Waterman score for every pair of SCOP domains contacting ligand

$$D(A, B) = 1 / \max_{A_i \in A, B_j \in B} \text{SmithWaterman}(A_i, B_j)$$

## Sequence Alignment Results

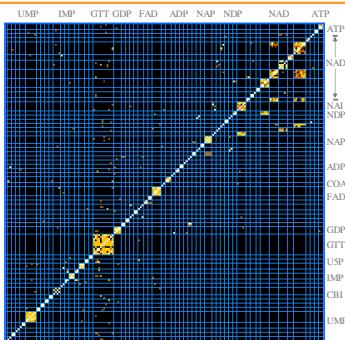
Similarity matrix:



1/SmithWaterman Score:  
(Darker means better match)

## Sequence Alignment Results

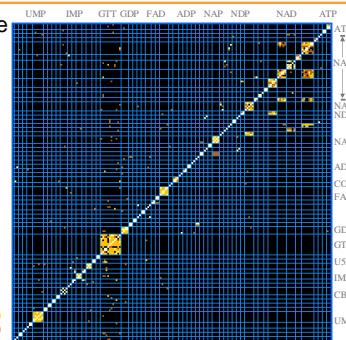
Tier matrix:



Best Matches:  
(Beige = 0th tier)  
(Yellow = 1st tier match)  
(Orange = 2nd tier match)

## Sequence Alignment Results

Classification rate  
FASTA = 68%  
Random = <1%



Best Matches:  
(Beige = 0th tier)  
(Yellow = 1st tier match)  
(Orange = 2nd tier match)

## Structure Alignment Method

Use CE to compute similarity of protein structures

```
CE - ~/ebi/data/pdbs/1jsu.pdb A ~/ebi/data/pdbs/lhcl.pdb _ scratch
Structure Alignment Calculator, version 1.02, last modified: Jun 15, 2001.
```

CE Aligner  
Aligner  
Rmsd =  
Z-Score =  
CPU =  
Seque...

$$\begin{aligned} X2 &= (0.997420)*X1 + (-0.071548)*Y1 + \\ &\quad (0.005923)*Z1 + (-0.936873)*B1 \\ Y2 &= (1.000000)*X1 + (-0.000000)*Y1 + \\ &\quad (-0.626397)*Z1 + (-119.695427) \\ Z2 &= (-0.040214)*X1 + (0.625133)*Y1 + \\ &\quad (-0.779482)*Z1 + (84.334198) \end{aligned}$$

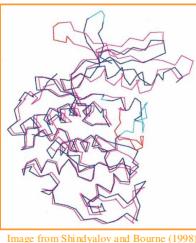
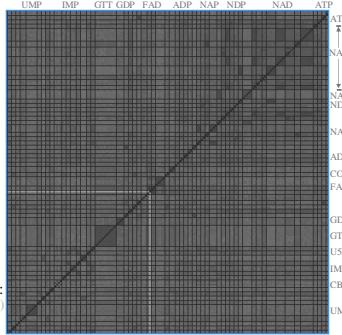


Image from Shindyalov and Bourne (1998)

## Structure Alignment Results

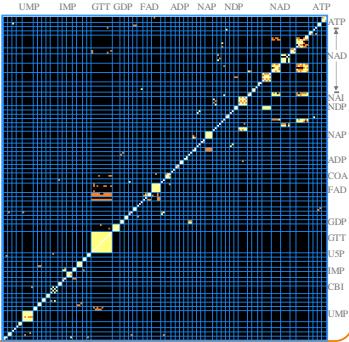
Similarity matrix:



1/CE -Z-Score:  
(Darker means better match)

## Structure Alignment Results

Tier matrix:



Best Matches:  
(Blue = 1st tier match)  
(Yellow = 1st tier match)  
(Orange = 2nd tier match)

## Structure Alignment Results

Classification rate:

FASTA = 68%  
CE = 65%  
Random = <1%

## Structure Alignment Results

Classification rate: When Smith-Waterman  $\geq 500$ :  
FASTA = 68% Sequence = 80%  
CE = 65% CE = 72%  
Random = <1% Random = <1%

When Smith-Waterman  $< 500$ :  
CE = 53%  
FASTA = 44%  
Random = <1%

## Conclusion

Many algorithms for structural alignment, differing according to

- Application: homology detection, drug design, etc.
- Granularity: atom, residue, fragment, SSE
- Representation: inter-molecular, intra-molecular
- Scoring: geometric, gaps, chemical, structural, etc.
- Correspondences: sequential, non-sequential
- Gap penalty: expect gaps near loops, etc.
- Flexibility: rigid, flexible
- Target: single protein, representative proteins, PDB

None seems best for all situations

All probably provide some benefit over sequence