

# **6th Permanent European School on Bioinformatics**

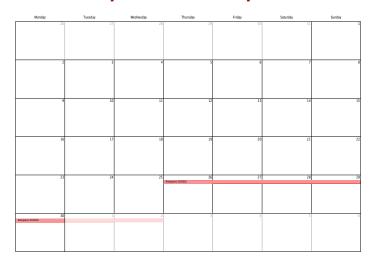


http://bioinfo.cipf.es/6ESB/

#### Preliminary program for the BioSapiens 6th ESB

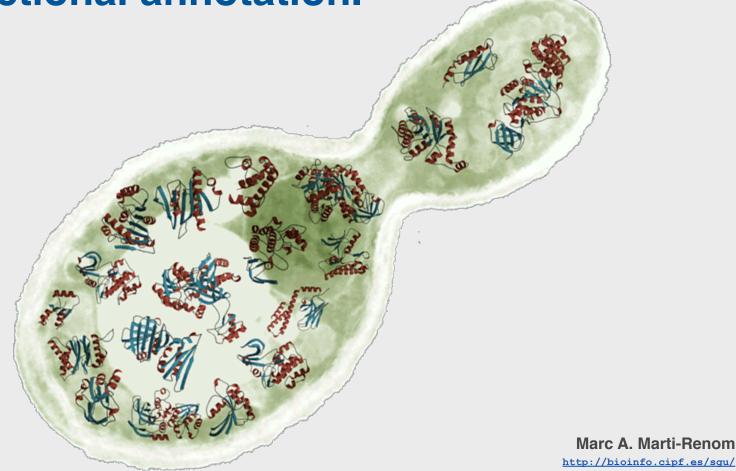
- Day 1 (April 26<sup>th</sup>) Databases
   Trainees from the European Bioinformatics Institute (EBI).
- Day 2 (April 27<sup>th</sup>) Analysis of microarray data Trainees to be confirmed (Brazma's group)
- Day 3 (April 28<sup>th</sup>) Proteins and protein families
   Trainee Dr. Yaniv Lowenstein (Linial's group)
- Day 4 (April 29<sup>th</sup>) Protein structure prediction
   Trainees to be confirmed (Marti-Renom's group)
- Day 5 (April 30<sup>th</sup>) Systems biology
   Trainee Dr. Ildefonso Cases (Valencia's groups)

#### 26th April to 30th of April 2007





Comparative protein structure models for functional annotation.



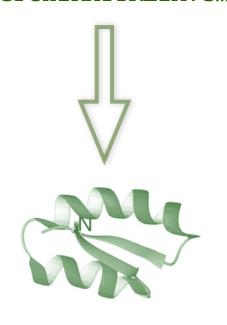
Structural Genomics Unit Bioinformatics Department Prince Felipe Resarch Center (CIPF), Valencia, Spain





### Principles of protein structure

GFCHIKAYTRLIMVG...



Anacystis nidulans

Anabaena 7120

GECHIKAYTRLIMVG...

Condrus crispus

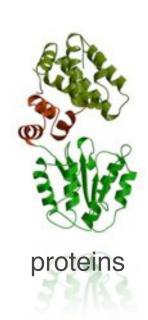
Desulfovibrio vulgaris

Folding (physics) *Ab initio* prediction

Evolution (rules)
Threading
Comparative Modeling

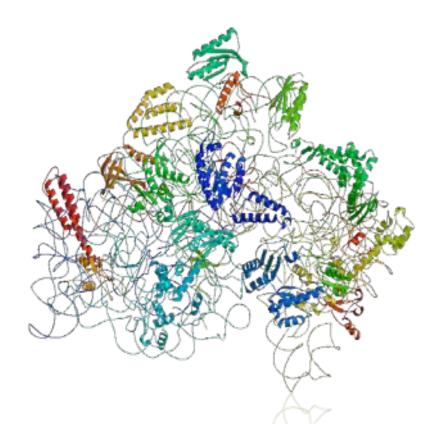
### From domains to assemblies





~2.5 domains in a protein a few domain partners per domain

#### assemblies



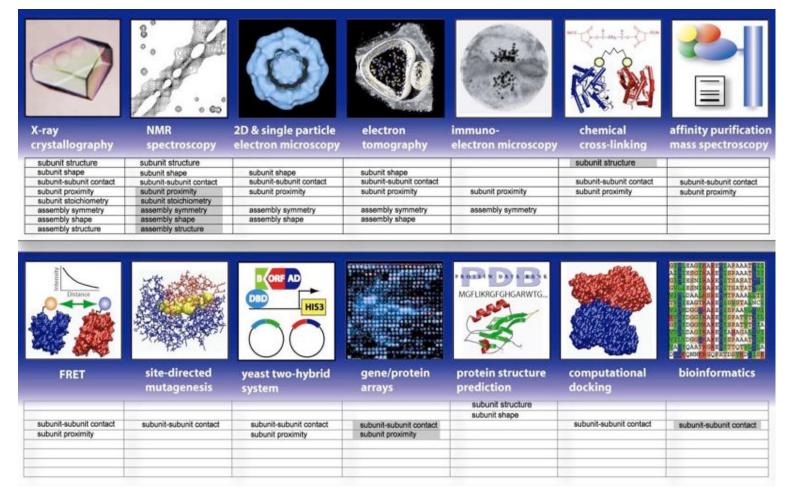
#### Determining the structures of proteins and assemblies

Use structural information from any

source: measurement, first principles, rules,

resolution: low or high resolution

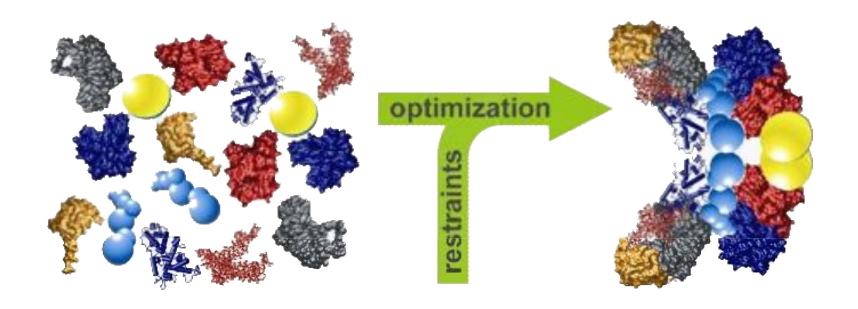
to obtain the set of all models that are consistent with it.



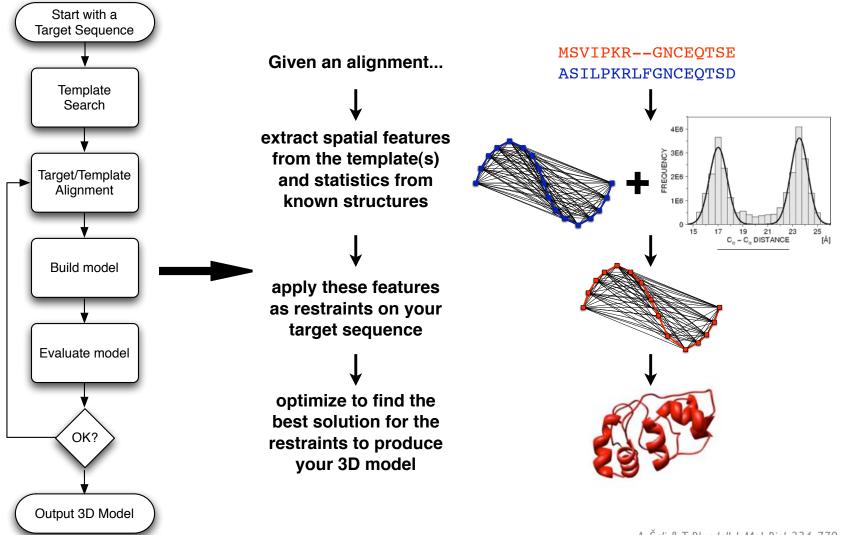
### Modeling by satisfaction of spatial restraints

- 1) Representation of a system.
- 2) Scoring function (spatial restraints).
- 3) Optimization.

There is nothing but points and restraints on them.

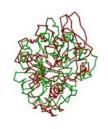


### Comparative modeling by satisfaction of spatial restraints MODELLER

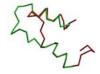


A. Šali & T. Blundell. J. Mol. Biol. 234, 779, 1993. J.P. Overington & A. Šali. Prot. Sci. 3, 1582, 1994. A. Fiser, R. Do & A. Šali, Prot. Sci., 9, 1753, 2000.

# Comparative modeling by satisfaction of spatial restraints Types of errors and their impact



Wrong fold



Miss alignments



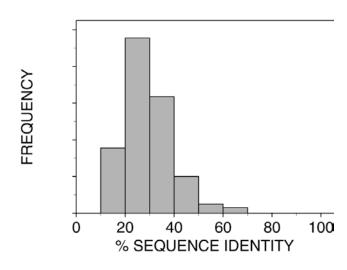
**Loop regions** 

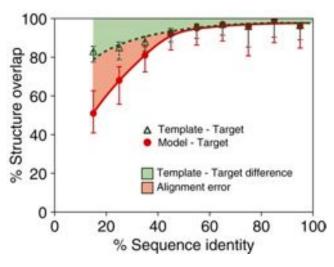


**Rigid body distortions** 



Side-chain packing



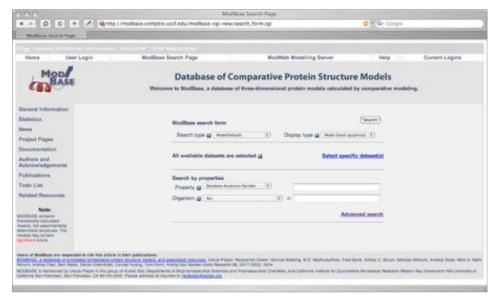


### **ModBase Statistics**

#### Large-scale modeling of the TrEMBL-SWISSPROT databases

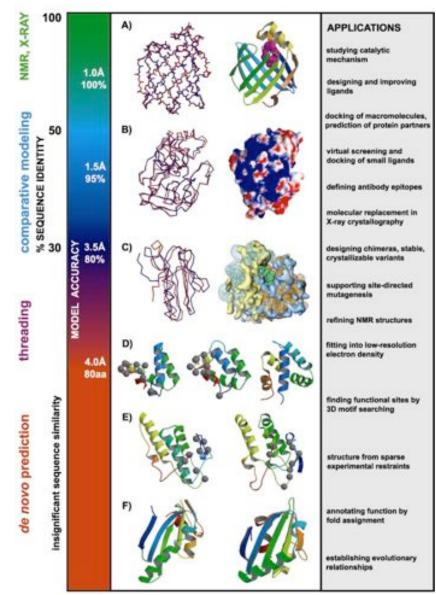
http://www.salilab.org/modbase/

Sequences (total)	1,930,692
Sequences (modeled)	1,084,784
Models	3,094,542





#### Utility of protein structure models, despite errors



# For many protein structures function is *unknown*

	Structural Genomics*	Traditional methods
Annotaated**	654	28,342
Not Annotaated	506 (43.6%)	6,815 (19,4%)
Total deposited	1,160	35,157

\* annotated as STRUCTURAL GENOMICS in the header of the PDB file \*\*annotated with either CATH, SCOP, Pfam or GO terms in the MSD database 36,317 protein structures, as of August 8th, 2006

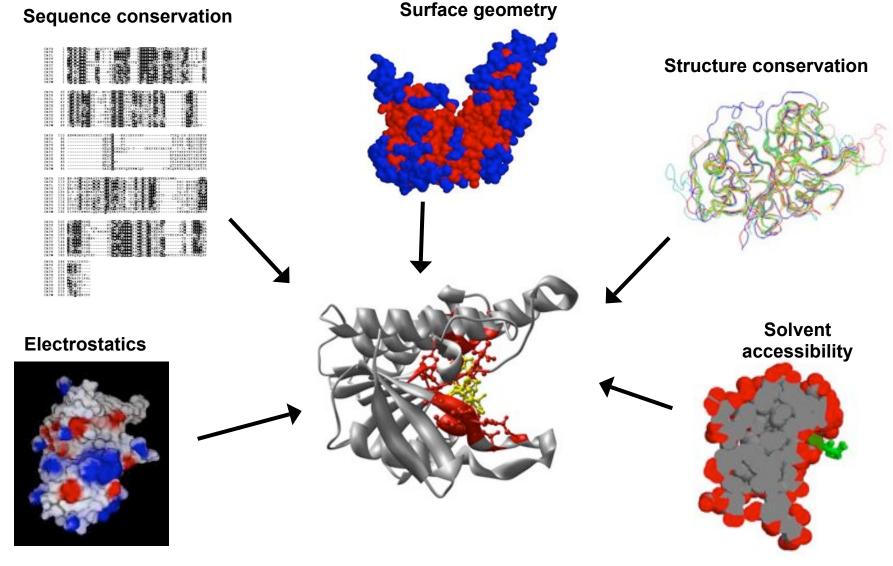
# For 20% protein structures function is *unknown*

	Structural Genomics*	Traditional methods
Annotaated**	654	28,342
Not Annotaated	506 (43.6%)	6,815 (19,4%)
Total deposited	1,160	35,157

\* annotated as STRUCTURAL GENOMICS in the header of the PDB file \*\*annotated with either CATH, SCOP, Pfam or GO terms in the MSD database 36,317 protein structures, as of August 8th, 2006

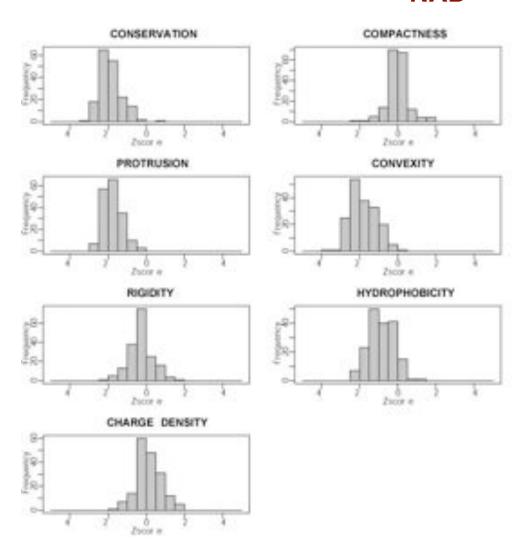


### Representation



# Scoring

#### **NAD**



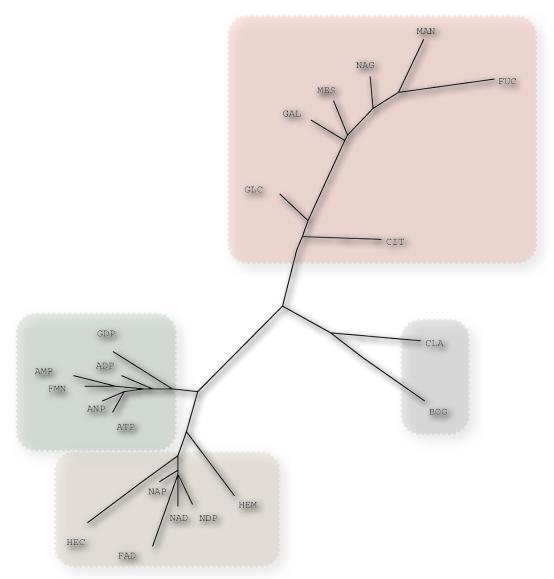
$$\longrightarrow w_k = \frac{1}{M} \sum_{\alpha=1}^M \tilde{f}_k^{(\alpha)}$$

M = number of proteins in training set

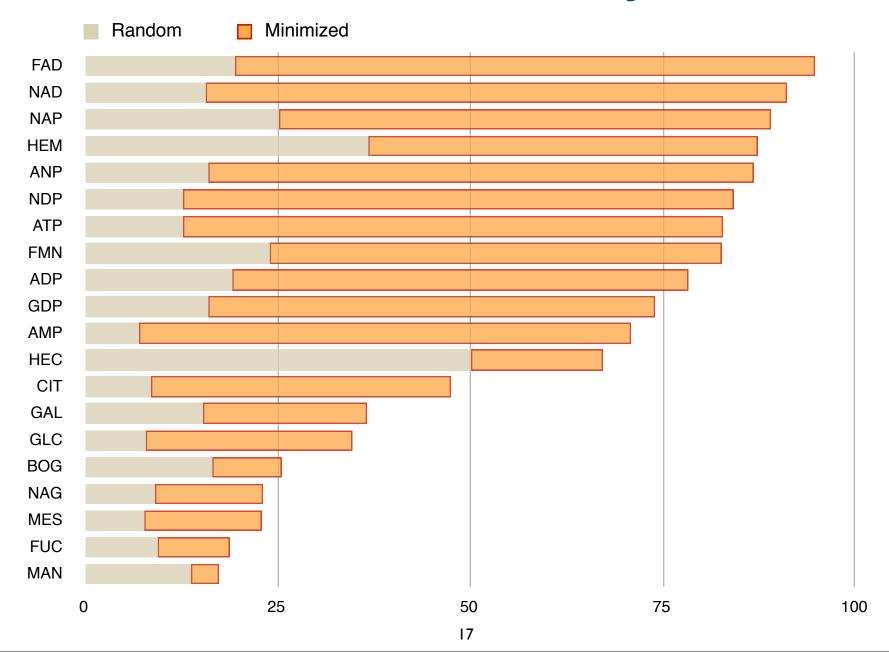
## Ligand fingerprints

	Compactness	Conservation	Charge density	B-factor	Protrusion coefficient	Convexity score	Hydrophobicity
ADP	-1.266	-2.009	0.447	-0.414	-1.521	-1.388	-0.118
AMP	-1.62	-1.962	0.341	-0.381	-1.909	-1.944	-0.518
ANP	-1.007	-2.227	0.176	-0.392	-1.706	-1.595	-0.14
ATP	-1.122	-2.156	0.228	-0.274	-1.845	-1.768	0.038
BOG	-2.067	-0.012	0.552	-0.465	-0.356	-0.49	-0.781
CIT	-2.948	-1.58	0.563	-0.527	-0.922	-0.838	-0.113
FAD	0.505	-2.108	0.366	-0.702	-1.735	-1.725	-0.75
FMN	-1.132	-1.98	0.382	-0.387	-1.803	-1.886	-0.695
FUC	-3.43	0.016	-0.295	-0.123	0.002	0.132	0.459
GAL	-3.186	-0.538	-0.234	-0.068	-0.906	-0.987	0.298
GDP	-1.061	-1.471	0.409	-0.81	-1.472	-1.423	0.182
GLC	-2.813	-1.247	-0.207	-0.399	-1.247	-1.337	-0.089
HEC	-0.172	-0.912	0.286	-0.325	-1.153	-1.27	-1.282
HEM	-0.65 I	-1.571	0.683	-0.5 I	-1.797	-1.937	-1.47
MAN	-3.72	0.131	0.105	-0.52	-0.605	-0.509	0.405
MES	-3.049	-0.24	-0.338	-0.479	-0.714	-0.926	0.296
NAD	-0.005	-1.852	0.156	-0.232	-1.775	-1.804	-0.858
NAG	-3.419	-0.46	-0.126	-0.154	-0.341	-0.523	-0.078
NAP	-0.009	-1.898	0.612	-0.321	-1.587	-1.656	-0.336
NDP	0.217	-1.741	0.535	-0.312	-1.463	-1.562	-0.498

# Ligand fingerprints



### **Prediction accuracy**

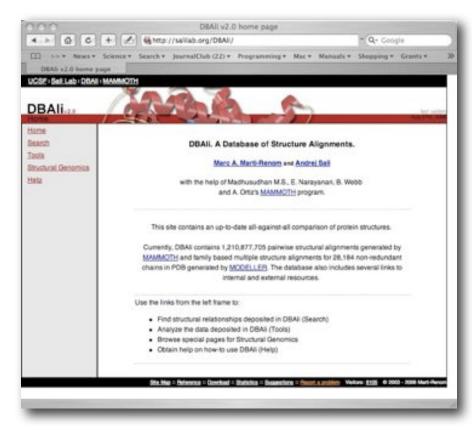




### DBAliv2.0 database

http://bioinfo.cipf.es/squ/services/DBAli/

http://www.salilab.org/DBAli/



- Uses MAMMOTH for similarity detection
- √ VERY FAST!!!
- √ Good scoring system with significance

Ortiz AR, (2002) Protein Sci. 11 pp2606

- √ Fully-automatic
- ✓ Data is kept up-to-date with PDB releases
- ✓ Tools for "on the fly" classification of families.
- ✓ Easy to navigate
- ✓ Provides tools for structure analysis

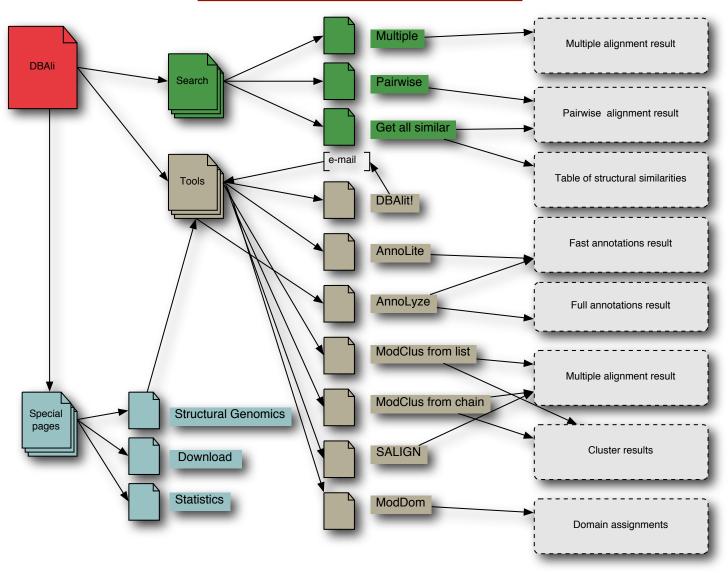
Does not provide a stable classification similar to that of CATH or SCOP

Pairwise structure alignn	nents
Last update:	February 15th, 2007
Number of chains:	88,276
Number of structure-structure comparisons:	1,425,479,365
Multiple structure alignm	ents
Last update:	January 23rd, 2007
Number of representative chains:	30,900
Number of families:	11,615

### DBAliv2.0 database

http://bioinfo.cipf.es/squ/services/DBAli/

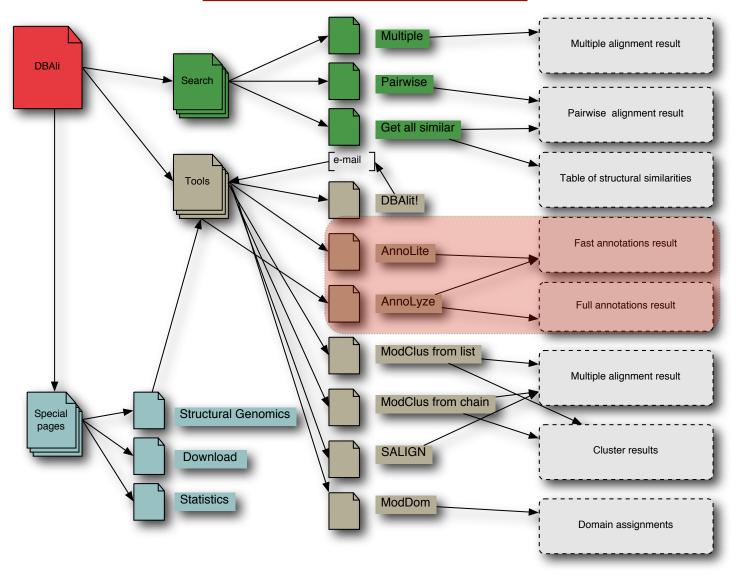




### DBAliv2.0 database

http://bioinfo.cipf.es/squ/services/DBAli/





### **AnnoLite**

PFAM:		0.00		Glycoeyi hydrolasa family 7
interPro:	0	1.30-99	PR001722	Glycoside hydrolase, family ?
	0	6.08-61	PR008985	Concension A-like lectrivglucanese
	0	1.00-42	EB000254	Cellulose-binding region, fungal
DC Number:	0	1.20-44	3.2.1.91	Celulose 1,4-beta-celicbiosidase.
		6.0e-41	3.2.1.4	Cellulase.
GO Morecular Function:		6.00-36	0030248	celulose binding [
	0	0.44-36	0016162	cellulose 1,4-beta-cellobiosidase activity ;
	0	1.0e-35	0004553	hydrolase activity, hydrolyzing O-glycosyl compounds:
		1,49-30	0008810	celulase activity (
	0	3.10-20	0016798	hydrolase activity, acting on glycosyl bonds. ¿
		1.00+0	0016787	hydrolase activity ;
GO Biological Process:		1,16-63	0030245	cellulose catabolism 4
		1.28-54	0000272	polysacchande catabolism (
	0	3.69-20	9995975	carbohydrate metabolism (
GO Cettular Component		1.20-23	0005576	extracelular region ;

- Information annotated in the MSD database.
- a High, a medium and a low confidence annotations not annotated in the MSD database.
- @ High, @ medium and @ low confidence annotations already annotated in the MSD database.

### **Benchmark set**

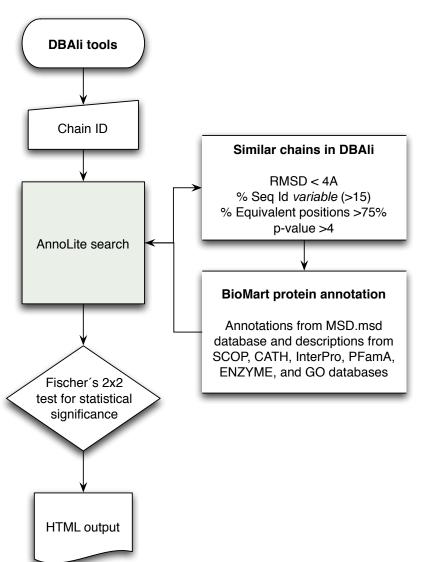
	Number of chains
Initial set*	50,223
FULL annotation**	10,997
Non-redundant set***	1,879

\*data from BioMart MSD.3 (release February 2005)

<sup>\*\*</sup>annotated with CATH, SCOP, Pfam, EC, InterPro, and GO terms in the MSD database

<sup>\*\*</sup>not two chains can be structurally aligned within 2A, superimposing more than 60% of their C atoms and have a length difference inferior to 30aa

### Method



	Con	f.P-value	Link	Description
CATH:	0	7.54-99	2,70,100,10	1,4-Beta-O-Glucan Celebohydrolase I, subunit A
SCOP:	0	0.00	5.29.1.10	Glycoeyi hydrolase family 7 catalytic core
PFAM:	0	0.00	PF00840	Glycoeyi hydrolase family 7
InterPro:	0	1.30-99	PR001722	Glycoside hydrolase, family ?
	0	6.08-51	PR008985	Concensive in A like lectiviguosnese
	0	1.00-42	EB000254	Cellulose-binding region, fungal
EC Number:		1.20-44	3.2.1.91	Celulose 1,4-beta-celobiosidase.
		6.0e-41	3.2.1.4	Cellulase.
GO Molecular Function:		6.0e-36	0030248	celulose binding (
	0	8.40-36	0016162	cellulose 1,4 beta-cellobiosidase activity 2
	0	1.0e-35	0004553	hydrolase activity, hydrolyzing O-glycosyl compounds: ;
		1,40-30	0008810	contrase activity (
	0	3.10-20	0016798	hydrolese activity, acting on glycosyl bonds. ¿
		1.00+0	0016787	hydrolase activity:
GO Biological Process:		1,16-63	0030245	cellulose catabolism 4
		1.20-54	0000272	polysaccharde catabolism (
	0	3.69-20	0005875	carbohydrate metabolism ¿
GO Cetiular Component		1,20-23	0005576	extracelular region ;
Information annotated in the	MSO	OUTSTANK.		
High, a medium and a low	e pent	dence anno	ations not arrested	ed in the MSD database.

## Scoring function

#### Fisher's 2x2 contingency test

	Non- similar	Similar	Total
Annotated	а	b	a+b
Not Annotated	С	d	c+d
Total	a+c	b+d	n

1b78A SCOP c.51.4.1	Similar	Not similar	Total
Annotated	4	2	6
Not Annotated	0	71,096	71,096
Total	4	71,098	71,102

$$p = \binom{a+b}{a} \binom{c+d}{c} / \binom{n}{a+c}$$
$$= \frac{(a+b)!(c+d)!(a+c)!(b+d)!}{n!a!b!c!d!}$$

$$p = 1.78e^{-19}$$

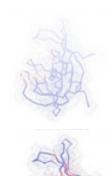
# Sensitivity .vs. Precision

	Optimal cut-off	Sensitivity (%) Recall or TPR	Precision (%)
SCOP fold	1e-6	92.7	88.4
CATH fold	1e-3	95.7	90.1
InterPro	1e-3	88.4	78.2
PFam family	1e-4	90.5	82.8
EC number	1e-4	93.3	79.7
GO Molecular Function	1e-1	84.3	80.9
GO Biological Process	1e-3	85.5	74.8
GO Cellular Component	1e-2	77.6	58.6

Sensitivity = 
$$\frac{TP}{TP + FN}$$
 Precision =  $\frac{TP}{TP + FP}$ 

# AnnoLyze

d.113.1.1	23.60 0.00	50 51 52 53 54 55 56 57 58 77 78 79 80 33 84 85 93 95 97 99 134 135 138 142 145
Patter	Av. binding Ax rest after congenie seq.id.	Residues in predicted binding site (size proportions) to the local conservation)
sherited p	artners:	
ACY	10.67	29 29 31 37 44 45 81 83 85 94 96 98 103 121 135
		19 20 21 48 49 51 96 98 136



### Benchmark

Number of chains	
Initial set*	78,167
<b>LigBase**</b> 30,126	
Non-redundant set*** 4,948 (8,846 ligands)	

\*all PDB chains larger than 30 aminoacids in length (8th of August, 2006)

\*\*annotated with at least one ligand in the LigBase database

\*\*\*not two chains can be structurally aligned within 3A, superimposing more than 75% of their C atoms, result in a sequence alignment with more than 30% identity, and have a length difference inferior to 50aa

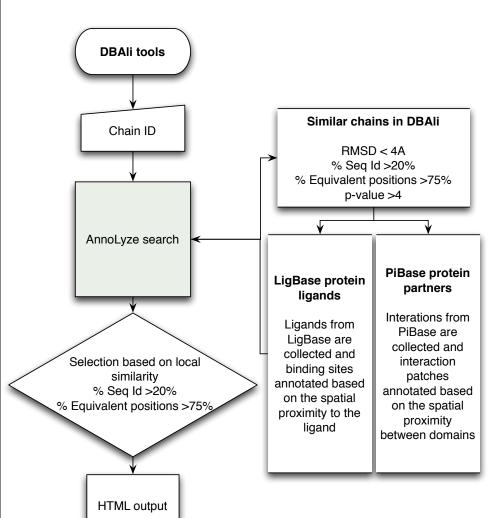
	Number of chains		
Initial set*	78,167		
πBase**	30,425		
Non-redundant set***	4,613 (11,641 partnerships)		

\*all PDB chains larger than 30 aminoacids in length (8th of August, 2006)

\*\*annotated with at least one partner in the Base database

\*\*\*not two chains can be structurally aligned within 3A, superimposing more than 75% of their C atoms, result in a sequence alignment with more than 30% identity, and have a length difference inferior to 50aa

### Method



Lipane	As binding site	Av. tesique	Residues in predicted binding site
-	860.10	conservation	save proportional to the local consensition)
MO2	59.03	0.105	48 49 52 62 63 66 67 113 116
CRY	20.00	2.111	23 29 31 37 44 48 49 83 85 84 96 103 121
800	20.00	9.111	19 20 21 48 49 51 96 98 136
ACY	15.87	0.163	23 29 31 37 44 45 81 83 85 94 96 98 103 121 135

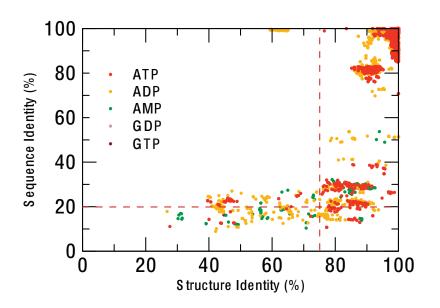


Patie	ate seq st	conservation	19 20 50 51 52 53 54 55 56 57 58 77 78 79 80
1.113.1.1	02000		81 82 83 84 85 93 95 97 99 134 135 138 14

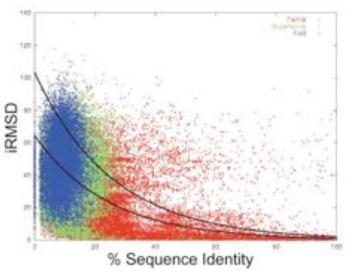


## Scoring function

#### Ligands



#### **Partners**



Aloy et al. (2003) J.Mol.Biol. 332(5):989-98.

## Sensitivity .vs. Precision

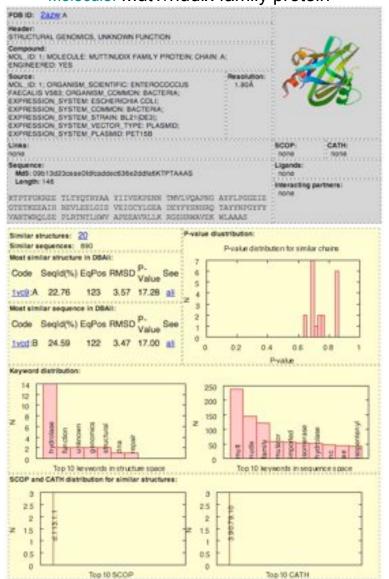
	Optimal cut-off	Sensitivity (%) Recall or TPR	Precision (%)
Ligands	30%	71.9	13.7
Partners	Partners 40%		55.7

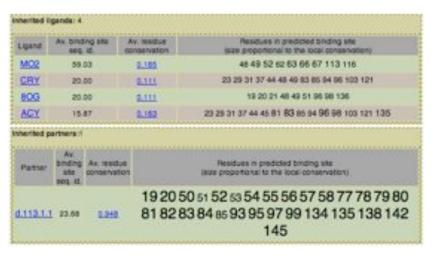
Sensitivity = 
$$\frac{TP}{TP + FN}$$
 Precision =  $\frac{TP}{TP + FP}$ 

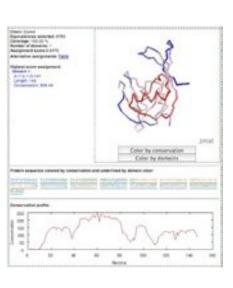
### **Example (2azwA)**

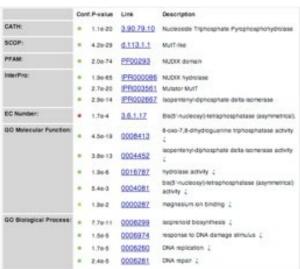
#### Structural Genomics Unknown Function

Molecule: MutT/nudix family protein

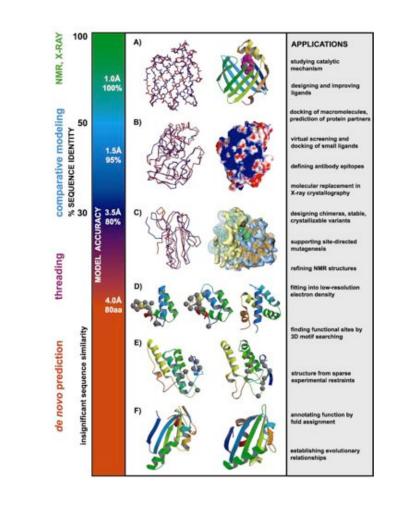


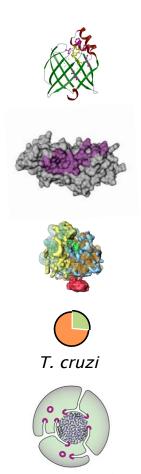






#### Can we use models to infer function?





# What is the physiological ligand of Brain Lipid-Binding Protein?

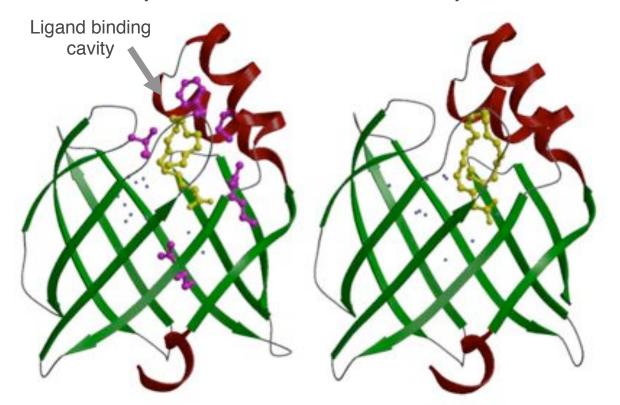
Predicting features of a model that are not present in the template

BLBP/oleic acid

BLBP/docosahexaenoic acid

Cavity is not filled

Cavity is filled



- 1. BLBP binds fatty acids.
  - 2. Build a 3D model.
- 3. Find the fatty acid that fits most snuggly into the ligand binding cavity.

# Structural analysis of missense mutations in human BRCA1 BRCT domains

Nebojsa Mirkovic, Marc A. Marti-Renom, Barbara L. Weber, Andrej Sali and Alvaro N.A. Monteiro

Cancer Research (June 2004). 64:3790-97

Cannot measure the functional impact of every possible SNP at all positions in each protein! Thus, prediction based on general principles of protein structure is needed.



### Missense mutations in BRCT domains by function

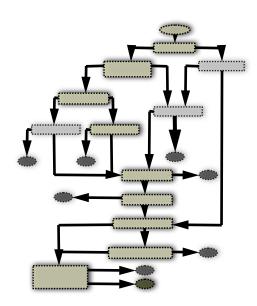
cancer not cancer associate associated

?

no transcription activation

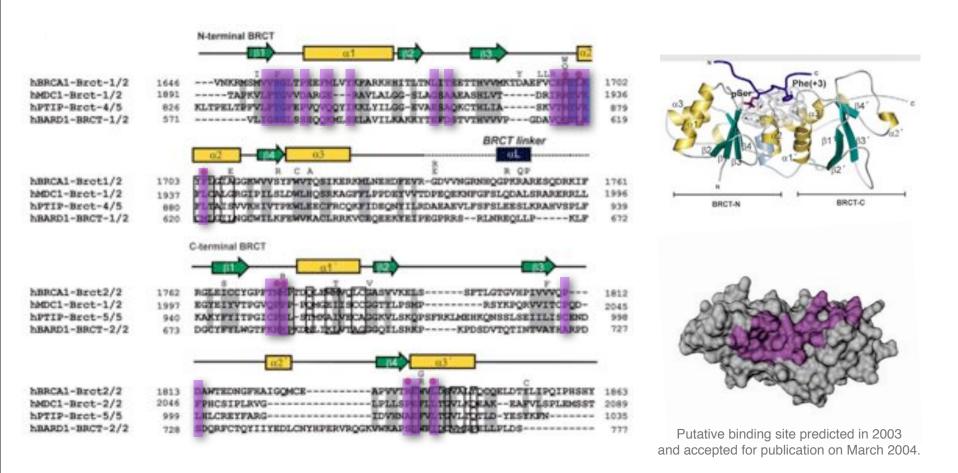
transcription activation

M1652K F1761S L1705PS L1657P C1697R M1775E 1715NS1 E1660G R1699W M1775K H1686Q A1708E 722FF17 L1780P S1715R R1699Q 34LG173 **I1807S** K1702E P1749R V1833E 8EG1743 M1775R Y1703HF A1843T RA1752 1704S PF1761I V1665M D1692N G1706A M1652I D1733G A1669S M1775V P1806A R1751P C1787S W1718S A1823T R1751Q G1788D V1833M V1653M T1720A L1664P G1788V W1837R W1730S R1758G F1734S G1803A W1837G T1685A L1764P V1804D S1841N E1735K T1685I **I1766S** V1736A V1808A A1843P M1689R P1771L G1738R V1809A T1852S D1692Y D1739E V1809F T1773S P1856T F1695L V1810G D1739G P1776S P1859R V1696L Q1811R D1739Y **D1778N** R1699L P1812S V1741G D1778G G1706E N1819S H1746N D1778H W1718C M1783T



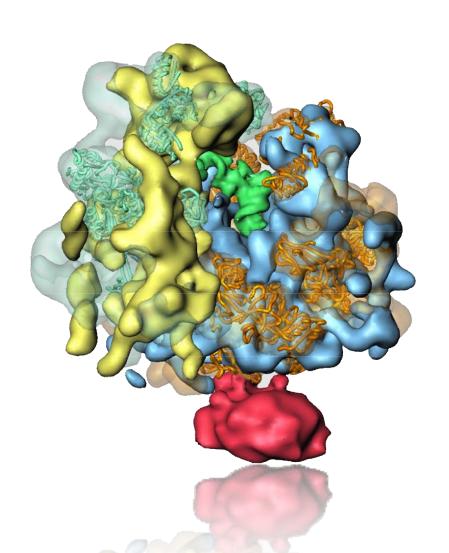
?

## Putative binding site on BRCA1



Williams *et al.* 2004 Nature Structure Biology. **June 2004 11**:519 Mirkovic *et al.* 2004 Cancer Research. **June 2004 64**:3790

### S. cerevisiae ribosome

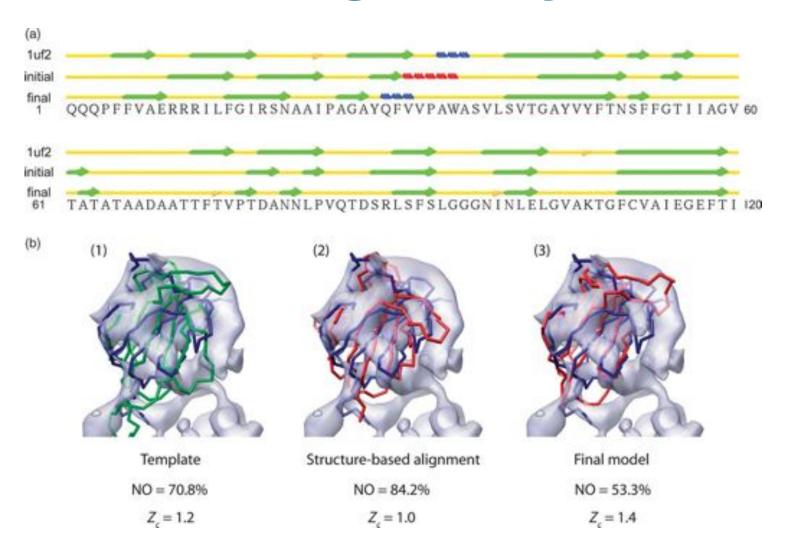


Fitting of comparative models into 15Å cryo-electron density map.

43 proteins could be modeled on 20-56% seq.id. to a known structure.

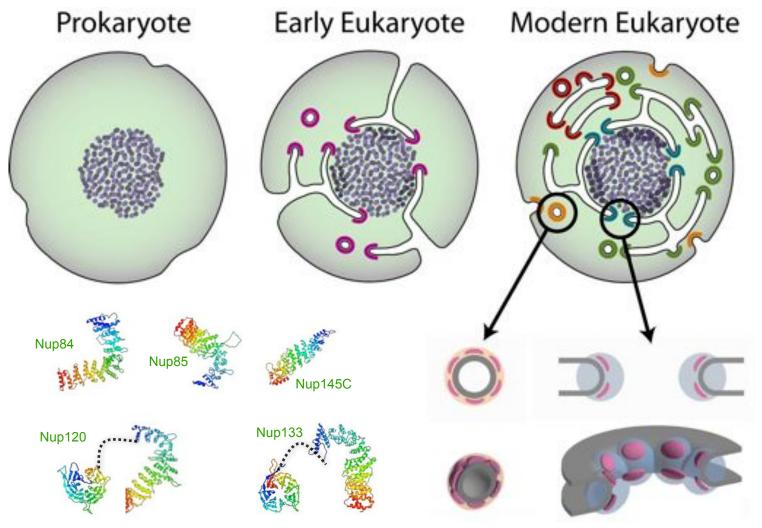
The modeled fraction of the proteins ranges from 34-99%.

## Modeling & cryoEM



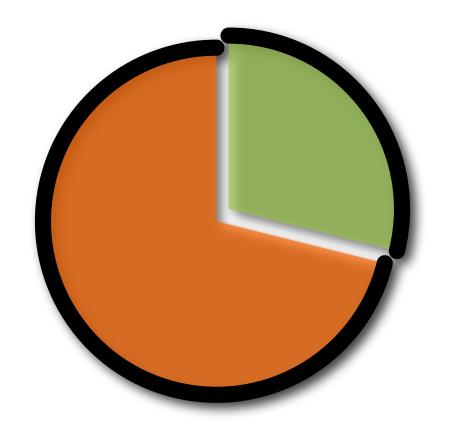
Topf etal. JMB, 357, 1655 (2006)

# The Nucleopore complex Cell evolution (?)



### **Tropical Disease Initiative (TDI)**

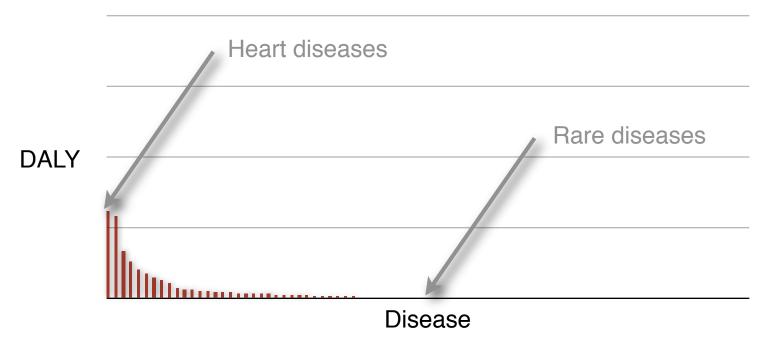
Predicting binding sites in protein structure models.



http://www.tropicaldisease.org

## Need is High in the Tail

- DALY Burden Per Disease in Developed Countries
- DALY Burden Per Disease in Developing Countries



Disease data taken from WHO, World Health Report 2004

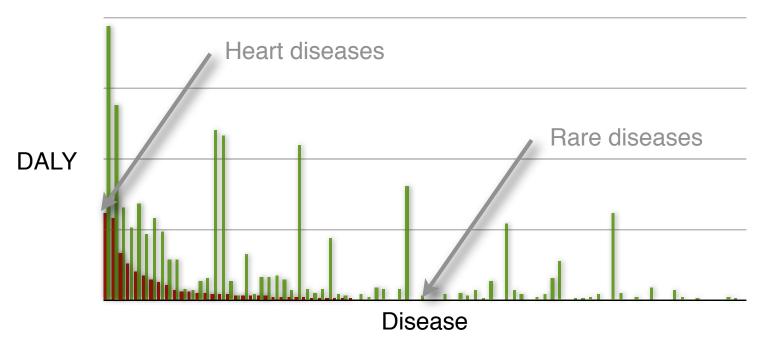
DALY - Disability adjusted life years

DALY is not a perfect measure of market size, but is certainly a good measure for importance.

DALYs for a disease are the sum of the years of life lost due to premature mortality (YLL) in the population and the years lost due to disability (YLD) for incident cases of the health condition. The DALY is a health gap measure that extends the concept of potential years of life lost due to premature death (PYLL) to include equivalent years of 'healthy' life lost in states of less than full health, broadly termed disability. One DALY represents the loss of one year of equivalent full health.

## Need is High in the Tail

- DALY Burden Per Disease in Developed Countries
- DALY Burden Per Disease in Developing Countries



Disease data taken from WHO, World Health Report 2004

DALY - Disability adjusted life years

DALY is not a perfect measure of market size, but is certainly a good measure for importance.

DALYs for a disease are the sum of the years of life lost due to premature mortality (YLL) in the population and the years lost due to disability (YLD) for incident cases of the health condition. The DALY is a health gap measure that extends the concept of potential years of life lost due to premature death (PYLL) to include equivalent years of 'healthy' life lost in states of less than full health, broadly termed disability. One DALY represents the loss of one year of equivalent full health.

# "Unprofitable" Diseases and Global DALY (in 1000's)

Malaria*	46,486
Tetanus	7,074
Lymphatic filariasis*	5,777
Syphilis	4,200
Trachoma	2,329
Leishmaniasis*	2,090
Ascariasis	1,817
Schistosomiasis*	1,702
Trypanosomiasis*	1,525

Trichuriasis	1,006
Japanese encephalitis	709
Chagas Disease*	667
Dengue*	616
Onchocerciasis*	484
Leprosy*	199
Diphtheria	185
Poliomyelitise	151
Hookworm disease	59

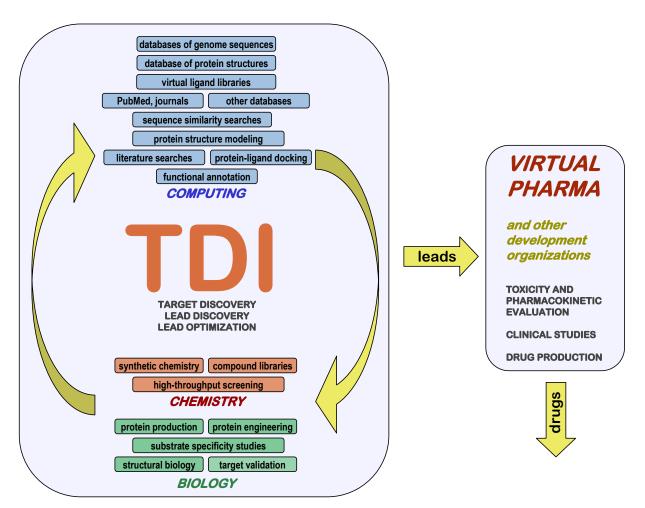
Disease data taken from WHO, World Health Report 2004

DALY - Disability adjusted life year in 1000's.

<sup>\*</sup> Officially listed in the WHO Tropical Disease Research disease portfolio.

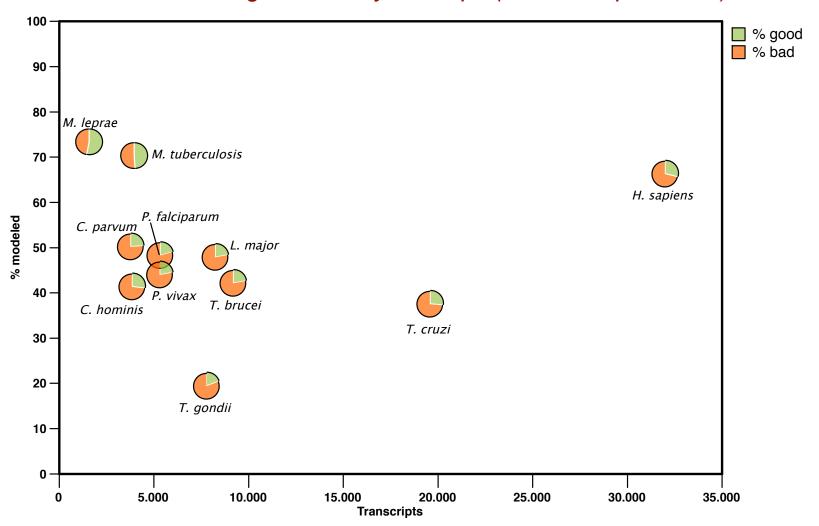
## **TDI flowchart**

http://www.tropicaldisease.org

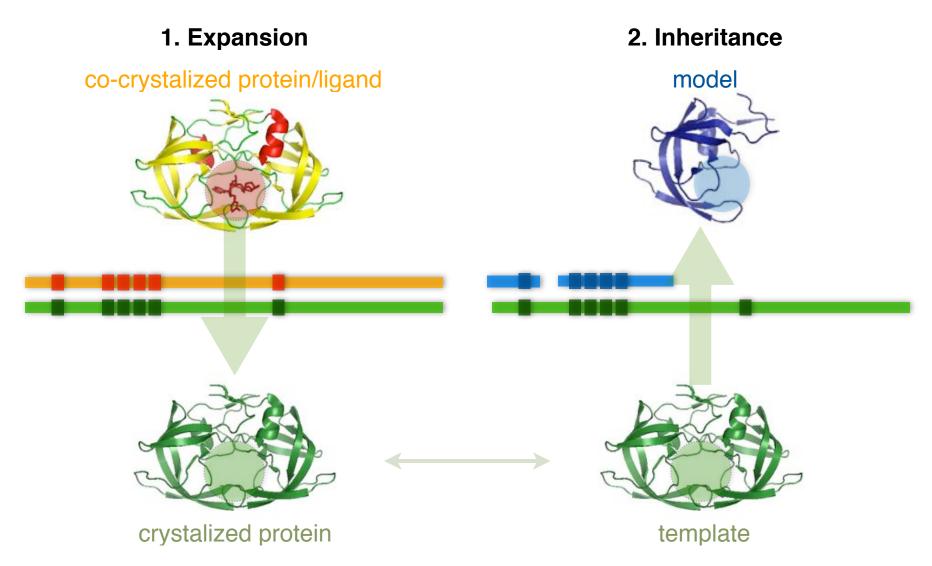


## Modeling Genomes

data from models generated by ModPipe (Eswar, Pieper & Sali)



## Comparative docking



## Ligand "expanded" space

from 6,859 templates used in "good" models

Expansion cut-off	Templates	Expanded	Unique
30%	4,639	64,800	3,178
50%	4,242	37,945	3,030
70%	3,323	20,603	2,786

## Ligand "inherited" space

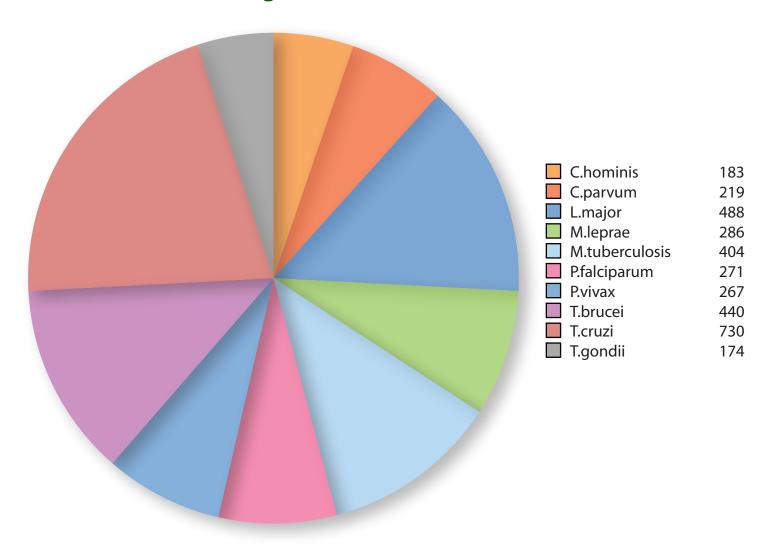
second cut-offs

#### Using a 70% "expansion" cut-off

Inheritance cut-offs	Models	Inherited	Unique
90% / 70%	5,181	23,286	1,137
90% / 80%	4,383	17,842	1,027
90% / 90%	3,462	11,803	827

### Distribution of models with inherited ligands

from 3,882 "good" models using a 90% / 90% "inherited" cut-offs



## **Summary table**

models with inherited ligands

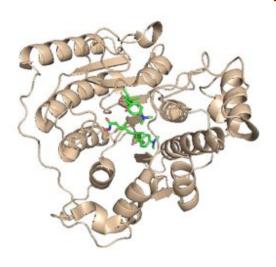
from 16,284 good models, 295 inherited a ligand/substance with at least one compound already approved by FDA and ready to be used from ZINC

	Transcripts	Good	Ligands	Lipinski	Lipinski+ZINC	FDA+ZINC
C. hominis	3,886	886	183	131	28	12 (10)
C. parvum	3,806	949	219	145	30	12 (10)
L. major	8,274	1,845	488	334	84	44 (34)
M. leprae	1,605	1,321	286	189	39	29 (25)
M. tuberculosis	3,991	2,887	404	285	71	44 (37)
P. falciparum	5,363	1,057	271	191	48	20 (16)
P. vivax	5,342	1,042	267	177	37	18 (15)
T. brucei	921	1,795	440	309	94	46 (36)
T. cruzi	19,607	3,915	730	493	127	62 (52)
T. gondii	7,793	587	174	124	28	8 (7)
TOTAL	60,588	16,284	3,462	2,378	586	295 (242)

## Example of inheritance (expansion)

LmjF21.0680 from L. major "Histone deacetylase 2" (model 1)

#### Template 1t64A a human HDAC8 protein.

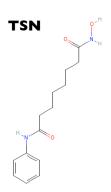


	Origen	Formula	Name	Cov.	Seq, Id. (%)
ZN	X-ray	Zn <sup>2+</sup>	Zinc ion		
NA	X-ray	Na <sup>+</sup>	Sodium ion		
CA	X-ray	Ca <sup>2+</sup>	Calcium ion		
TSN	X-ray	C <sub>17</sub> H <sub>22</sub> N <sub>2</sub> O <sub>3</sub>	Trichostatin A		
ѕнн	Expanded	C <sub>14</sub> H <sub>20</sub> N <sub>2</sub> O <sub>3</sub>	Octadenioic acid hudroxyamide phenylamide	100.00	83.8

### Example of inheritance (inheritance)

LmjF21.0680 from L. major "Histone deacetylase 2" (model 1)

	Formula	Name	Cov.	Seq, Id. (%)	Residues
TSN	C <sub>17</sub> H <sub>22</sub> N <sub>2</sub> O <sub>3</sub>	Trichostatin A	100.00	90.9	90 131 132 140 141 167
sнн	C <sub>14</sub> H <sub>20</sub> N <sub>2</sub> O <sub>3</sub>	Octadenioic acid hudroxyamide phenylamide	100.00	90.9	169 256 263 293 295



#### suberoylanilide hydroxamic acid

#### **Pharmacological Action:**

Anti-Inflammatory Agents, Non-Steroidal
Antineoplastic Agents
Enzyme Inhibitors
Anticarcinogenic Agents

Inhibits histone deacetylase I and 3

ѕнн	O H
H H	H

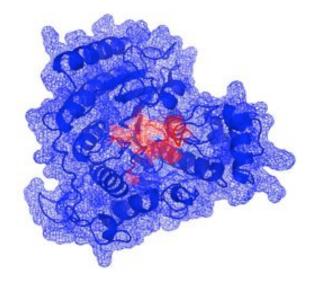
#### trichostatin A

#### **Pharmacological Action:**

Antibiotics, Antifungal
Enzyme Inhibitors
Protein Synthesis Inhibitors

chelates zinc ion in the active site of histone deacetylases, resulting in preventing histone unpacking so DNA is less available for transcription

	LmjF21.0680.1.pdb
Template	It64A
Seq. Id (%)	38.00
MPQS	1.47



### Example of inheritance (CDD-Roos-literature)

LmjF21.0680 from L. major "Histone deacetylase 2" (model 1)

Proc. Natl. Acad. Sci. USA Vol. 93, pp. 13143–13147, November 1996 Medical Sciences

### Apicidin: A novel antiprotozoal agent that inhibits parasite histone deacetylase

(cyclic tetrapeptide/Apicomplexa/antiparasitic/malaria/coccidiosis)

Sandra J. Darkin-Rattray\*†, Anne M. Gurnett\*, Robert W. Myers\*, Paula M. Dulski\*, Tami M. Crumley\*, John J. Allocco\*, Christine Cannova\*, Peter T. Meinke‡, Steven L. Colletti‡, Maria A. Bednarek‡, Sheo B. Singh§, Michael A. Goetz§, Anne W. Dombrowski§, Jon D. Polishook§, and Dennis M. Schmatz\*

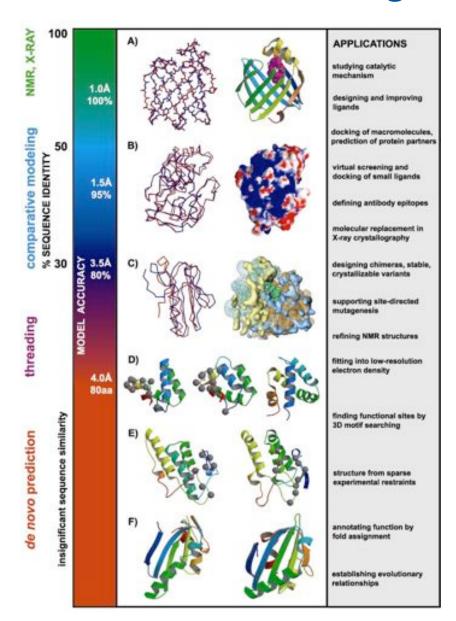
Departments of \*Parasite Biochemistry and Cell Biology, ‡Medicinal Chemistry, and §Natural Products Drug Discovery, Merck Research Laboratories, P.O. Box 2000, Rahway, NJ 07065

Antimicrobial Agents and Chemotherapy, Apr. 2004, p. 1435–1436 0066-4804/04/\$08.00+0 DOI: 10.1128/AAC.48.4.1435–1436.2004 Copyright © 2004, American Society for Microbiology. All Rights Reserved.

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Antimalarial and Antileishmanial Activities of Aroyl-Pyrrolyl-Hydroxyamides, a New Class of Histone Deacetylase Inhibitors

### "take home" message



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**Maya Topf** 

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Min-Yi Shen

**Damien Devos** 

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Alejandro Panjkovich (CU)

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John Kuriyan (UCB)

NY-SGXRC

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