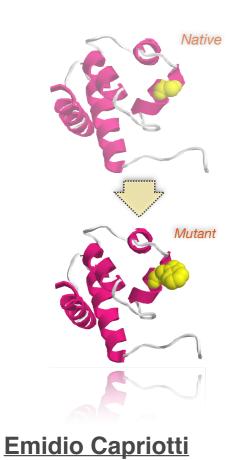
Methods for the automatic annotation of protein variants



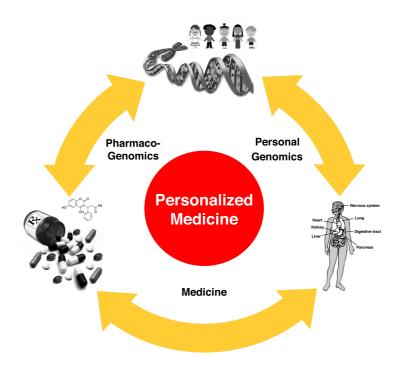
Biomolecules

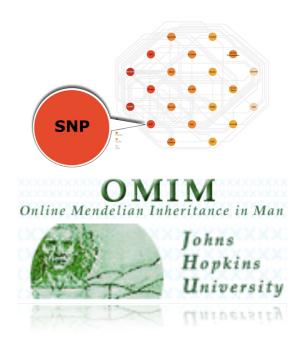
Folding and

Disease

http://biofold.org/

Meeting PRIN 2017, Bologna (Italy) September 5, 2019





Department of Pharmacy and Biotechnology (FaBiT) University of Bologna



Variant annotation methods

Since 2004 we started to develop methods for the annotation of protein variants at stability and functional levels





Home

Research

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urces Publications

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Resources

The researchers of the BioFoID Unit have developed several web server applications that are currently hosted on servers of the University of Bologna (Italy) and maintained by other collaborators. In the future a mirror of these applications will be made available on this server. Currently you can reach these web servers using the following links.

Folding and Stability

I-Mutant1.0

Neural Network based method to predict the sign of free energy change of proteins upon single point mutation.

Pub Ced

I-Mutant2.0

Support Vector Machine based method to predict the sign and the value of free energy change of proteins upon single point mutation.

Pub Ded

K-Fold

Support Vector Machine based method to predict the mechanism and rate of protein folding kinetic.

Publiced 🔼

Genomic Variations and Disease

ContrastRank

Statistical method for the classification of cancer samples using exome sequencing data.

Publicat 1

DrCancer

Support Vector Machine based method to predict cancer-causing mutations (Beta version).

Pub Ced 🔼

Fido-SNP

Machine learning method for predicting the impact of SNVs in the dog genome.

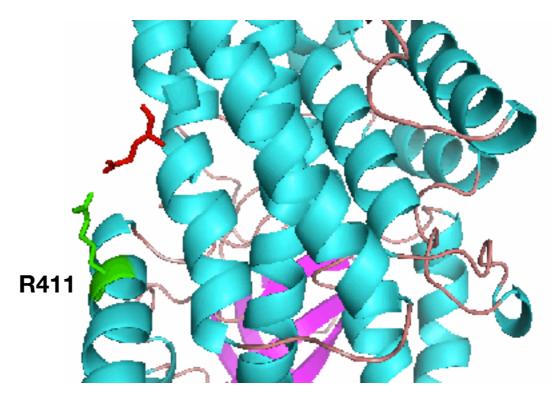
Sequence, Structure & Function

Genomic variants in sequence motifs could affect protein function. Mutation S362A of P53 affect the interaction with hydrolase USP7 and the deubiquitination of the protein.



Nonsynonymous variants responsible for protein structural changes and cause loss of stability of the folded protein.

Mutation R411L removes the salt bridge stabilizing the structure of the IVD dehydrogenase.



Conserved or not?

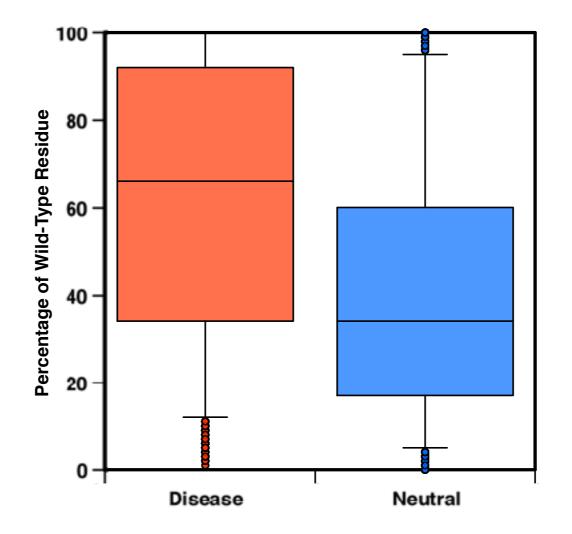
In positions 66 the Glutamic acid is highly conserved Asparagine in position 138 is mutated Threonine or Alanine

					1 [8(
	bits	E-value	N 1	100.0%	MDVGSKEVLMESPPDYSAAPRGRFGIPCCPVHLKRLLIVVVVVVLIVVVIVGALLMGLHMSOKHTEMVLEMSIGAPEAQQ	
1 P11686	400	1e-110		100.0%	MDVGSKEVLMESPPDYSAAPRGRFGIPCCPVHLKRLLIVVVVVVLIVVVIVGALLMGLHMSQKHTEMVLEMSIGAPEAQQ	
2 P15783	280	3e-74		80.6%	MDVGSKEVLMESPPDYTAVPGGRLLIPCCPVNIKRLLIVVVVVVVVVVVVVVVGALLMGLHMSQKHTEMVLEMSITGPEAQQ	
3 P21841	276	6e-73	1	78.7%	MDMSSKEVLMESPPDYSAGPRSQFRIPCCPVHLKRLLIVVVVVVVVVVVVVVVGALLMGLHMSQKHTEMVLEMSIGAPETQK	
4 P22398	270	3e-71	1	78.2%	MDMGSKEALMESPPDYSAAPRGRFGIPCCPVHLKRLLIVVVVVVVVVVVVVVVGALLMGLHMSQKHTEMVLEMSIGAPEVQQ	
5 Q1XFL5	268	1e-70	1	80.2%	MDVGSKEVLMESPPDYSAVPGGRLRIPCCPVNLKRLLVVVVVVVVVVVVVVVVGALLMGLHMSQKHTEMVLEMSLAGPEAQQ	
6 UPI0000E219B8	261	1e-68	1	89.4%	MDVGSKEVLMESPPDYSAAPRGRFGIPCCPVHLKRLLIVVVVVVVVVVVVVVVGALLMGLHMSQKHTEMVLEMSIGAPEAQQ	
7 UPI00005A47C8	259	6e-68	1	78.2%	MDVGSKEVLIESPpdYSAAPRGRLGIPCFPSSLKRLLIIVVVIVLVVVVIVGALLMGLHMSQKHTEMVLEMSMGGPEAQQ)
8 Q3MSM1	206	8e-52	1	83.4%	MDVGSKEVLMESPPDYSAVPGGRLRIPCCPVNLKRLLVVVVVVVVVVVVVVVVGALLMGLHMSQKHTEMVLEMSLAGPEAQQ	
9 Q95M82	85	3e-15	1	82.4%		
10 UPI000155C160	84	4e-15	1	48.9%		
11 UPI0001555957	82	1e-14	1	83.6%	KVRADSPPDYSVAPRGRLGIPCCPFHLKRLLIIVVVVVLIVVVVLGALLMGLHMSQKHTEM	
12 B3DM51	81	4e-14	1	34.8%	HMSQKHTETIFQMSLQD	
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• • • •						
					81 . 1	10
	bits	E-value	N 1	100.0%	RLALSEHLVTTATFSIGSTGLVVYDYQQLLIAYKPAPGTCCYIMKIAPESIPSLEAINRKVHNFQMECSLQAKPAVPTSK	
1 P11686	400			100.0%	RLALSEHLVTTATFSIGSTGLVVYDYQQLLIAYKPAPGTCCYIMKIAPESIPSLEALNRKVHNFQMECSLQAKPAVPTSK	
2 P15783		1e-110	1 1	TUU•05	KLALSERLVITATESIGSTGLVVIDIUULLIAIKPAPGICCIIMKIAPESIPSLEAHNKKVRNEUMECSLUAKPAVPISK	
	280	1e-110 3e-74				
	280 276	3e-74	1	80.6%	RLALSERVGTTATFSIGSTGTVVYDYQRLLIAYKPAPGTCCYIMKMAPQNIPSLEALTRKLQNFQAKPQVPSSK	
3 P21841(Mouse) 4 P22398	280 276 270	3e-74 6e-73	1 1	80.6% 78.7%	RLALSERVGTTATFSIGSTGTVVYDYQRLLIAYKPAPGTCCYIMKMAPQNIPSLEALTRKLQNFQAKPQVPSSK RLAPSERADTIATFSIGSTGIVVYDYQRLLTAYKPAPGTYCYIMKMAPESIPSLEAFARKLQNFRAKPSTPTSK	•
3 P21841(Mouse) 4 P22398	276	3e-74 6e-73	1 1 1	80.6% 78.7%	RLALSERVGTTATFSIGSTGTVVYDYQRLLIAYKPAPGTCCYIMKMAPQNIPSLEALTRKLQNFQAKPQVPSSK RLAPSERADTIATFSIGSTGIVVYDYQRLLTAYKPAPGTYCYIMKMAPESIPSLEAFARKLQNFRAKPSTPTSK RLALSEWAGTTATFPIGSTGIVTCDYQRLLIAYKPAPGTCCYLMKMAPDSIPSLEALARKFQANPAEPPTQ	
3 P21841(Mouse)	276 270	3e-74 6e-73 3e-71	1 1 1	80.6% 78.7% 78.2%	RLALSERVGTTATFSIGSTGTVVYDYQRLLIAYKPAPGTCCYIMKMAPQNIPSLEALTRKLQNFQAKPQVPSSK RLAPSERADTIATFSIGSTGIVVYDYQRLLTAYKPAPGTYCYIMKMAPESIPSLEAFARKLQNFRAKPSTPTSK RLALSEWAGTTATFPIGSTGIVTCDYQRLLIAYKPAPGTCCYLMKMAPDSIPSLEALARKFQANPAEPPTQ RLALSEHVGTTATFSIGSSGNVVYDYQRLLIAYKPAPGTCCYVMKMSPQSMPSLEALTKKFQNFQVSVQAKPSTPTSK	
3 P21841(Mouse) 4 P22398 5 Q1XFL5	276 270 268	3e-74 6e-73 3e-71 1e-70	1 1 1 1	80.6% 78.7% 78.2% 80.2%	RLALSERVGTTATFSIGSTGTVVYDYQRLLIAYKPAPGTCCYIMKMAPQNIPSLEALTRKLQNFQAKPQVPSSK RLAPSERADTIATFSIGSTGIVVYDYQRLLTAYKPAPGTYCYIMKMAPESIPSLEAFARKLQNFRAKPSTPTSK RLALSEWAGTTATFPIGSTGIVTCDYQRLLIAYKPAPGTCCYLMKMAPDSIPSLEALARKFQANPAEPPTQ RLALSEHVGTTATFSIGSSGNVVYDYQRLLIAYKPAPGTCCYVMKMSPQSMPSLEALTKKFQNFQVSVQAKPSTPTSK RLALSEHLVTTATFSIGSTGLVVYDYQQLLIAYKPAPGTCCYIMKIAPESIPSLEALTRKVQNFQGQWKPQGERKRPGKR	
3 P21841(Mouse) 4 P22398 5 Q1XFL5 6 UPI0000E219B8	276 270 268 261	3e-74 6e-73 3e-71 1e-70 1e-68	1 1 1 1 1	80.6% 78.7% 78.2% 80.2% 89.4%	RLALSERVGTTATFSIGSTGTVVYDYQRLLIAYKPAPGTCCYIMKMAPQNIPSLEALTRKLQNFQAKPQVPSSK RLAPSERADTIATFSIGSTGIVVYDYQRLLTAYKPAPGTYCYIMKMAPESIPSLEAFARKLQNFRAKPSTPTSK RLALSEWAGTTATFPIGSTGIVTCDYQRLLIAYKPAPGTCCYLMKMAPDSIPSLEALARKFQANPAEPPTQ RLALSEHVGTTATFSIGSSGNVVYDYQRLLIAYKPAPGTCCYVMKMSPQSMPSLEALTKKFQNFQVSVQAKPSTPTSK	
3 P21841(Mouse) 4 P22398 5 Q1XFL5 6 UPI0000E219B8 7 UPI00005A47C8 8 Q3MSM1	276 270 268 261 259	3e-74 6e-73 3e-71 1e-70 1e-68 6e-68 8e-52	1 1 1 1 1 1	80.6% 78.7% 78.2% 80.2% 89.4% 78.2% 83.4%	RLALSERVGTTATFSIGSTGTVVYDYQRLLIAYKPAPGTCCYIMKMAPQNIPSLEALTRKLQNFQAKPQVPSSK RLAPSERADTIATFSIGSTGIVVYDYQRLLTAYKPAPGTYCYIMKMAPESIPSLEAFARKLQNFRAKPSTPTSK RLALSEWAGTTATFPIGSTGIVTCDYQRLLIAYKPAPGTCCYLMKMAPDSIPSLEALARKFQANPAEPPTQ RLALSEHVGTTATFSIGSSGNVVYDYQRLLIAYKPAPGTCCYVMKMSPQSMPSLEALTKKFQNFQVSVQAKPSTPTSK RLALSEHLVTTATFSIGSTGLVVYDYQQLLIAYKPAPGTCCYIMKIAPESIPSLEALTRKVQNFQGQWKPQGERKRPGKR RLALQERVGTTATFSIGSTGIVVYDYQRLLIAYKPAPGTCCYIMKMTPENIPSLEALTRKFQDFQVKPAVSTSK	
3 P21841(Mouse) 4 P22398 5 Q1XFL5 6 UPI0000E219B8 7 UPI00005A47C8	276 270 268 261 259 206	3e-74 6e-73 3e-71 1e-70 1e-68 6e-68 8e-52	1 1 1 1 1 1 1	80.6% 78.7% 78.2% 80.2% 89.4% 78.2% 83.4%	RLALSERVGTTATFSIGSTGTVVYDYQRLLIAYKPAPGTCCYIMKMAPQNIPSLEALTRKLQNFQAKPQVPSSK RLAPSERADTIATFSIGSTGIVVYDYQRLLTAYKPAPGTYCYIMKMAPESIPSLEAFARKLQNFRAKPSTPTSK RLALSEWAGTTATFPIGSTGIVTCDYQRLLIAYKPAPGTCCYLMKMAPDSIPSLEALARKFQANPAEPPTQ RLALSEHVGTTATFSIGSSGNVVYDYQRLLIAYKPAPGTCCYVMKMSPQSMPSLEALTKKFQNFQVSVQAKPSTPTSK RLALSEHLVTTATFSIGSTGLVVYDYQQLLIAYKPAPGTCCYIMKIAPESIPSLEALTRKVQNFQGQWKPQGERKRPGKR RLALQERVGTTATFSIGSTGIVVYDYQRLLIAYKPAPGTCCYIMKMTPENIPSLEALTRKFQDFQVKPAVSTSK RLALSEHVGTTATFSIGSSGNVVYDYQRLLIAYKPAPGTCCYVMKMSPQSMPSLEALTKKFQNFQ	
3 P21841(Mouse) 4 P22398 5 Q1XFL5 6 UPI0000E219B8 7 UPI00005A47C8 8 Q3MSM1 9 Q95M82	276 270 268 261 259 206 85 84	3e-74 6e-73 3e-71 1e-70 1e-68 6e-68 8e-52 3e-15	1 1 1 1 1 1 1 1	80.6% 78.7% 78.2% 80.2% 89.4% 78.2% 83.4% 82.4%	RLALSERVGTTATFSIGSTGTVVYDYQRLLIAYKPAPGTCCYIMKMAPQNIPSLEALTRKLQNFQAKPQVPSSK RLAPSERADTIATFSIGSTGIVVYDYQRLLTAYKPAPGTYCYIMKMAPESIPSLEAFARKLQNFRAKPSTPTSK RLALSEWAGTTATFPIGSTGIVTCDYQRLLIAYKPAPGTCCYLMKMAPDSIPSLEALARKFQANPAEPPTQ RLALSEHVGTTATFSIGSSGNVVYDYQRLLIAYKPAPGTCCYVMKMSPQSMPSLEALTKKFQNFQVSVQAKPSTPTSK RLALSEHLVTTATFSIGSTGLVVYDYQQLLIAYKPAPGTCCYIMKIAPESIPSLEALTRKVQNFQGQWKPQGERKRPGKR RLALQERVGTTATFSIGSTGIVVYDYQRLLIAYKPAPGTCCYIMKMTPENIPSLEALTRKFQDFQVKPAVSTSK RLALSEHVGTTATFSIGSSGNVVYDYQRLLIAYKPAPGTCCYVMKMSPQSMPSLEALTRKFQDFQV	
3 P21841(Mouse) 4 P22398 5 Q1XFL5 6 UPI0000E219B8 7 UPI00005A47C8 8 Q3MSM1 9 Q95M82 10 UPI000155C160	276 270 268 261 259 206 85	3e-74 6e-73 3e-71 1e-70 1e-68 6e-68 8e-52 3e-15 4e-15	1 1 1 1 1 1 1 1 1	80.6% 78.7% 78.2% 80.2% 89.4% 78.2% 83.4% 82.4% 48.9% 83.6%	RLALSERVGTTATFSIGSTGTVVYDYQRLLIAYKPAPGTCCYIMKMAPQNIPSLEALTRKLQNFQAKPQVPSSK RLAPSERADTIATFSIGSTGIVVYDYQRLLTAYKPAPGTYCYIMKMAPESIPSLEAFARKLQNFRAKPSTPTSK RLALSEWAGTTATFPIGSTGIVTCDYQRLLIAYKPAPGTCCYLMKMAPDSIPSLEALARKFQANPAEPPTQ RLALSEHVGTTATFSIGSSGNVVYDYQRLLIAYKPAPGTCCYVMKMSPQSMPSLEALTKKFQNFQVSVQAKPSTPTSK RLALSEHLVTTATFSIGSTGLVVYDYQQLLIAYKPAPGTCCYIMKIAPESIPSLEALTRKVQNFQGQWKPQGERKRPGKR RLALQERVGTTATFSIGSTGIVVYDYQRLLIAYKPAPGTCCYIMKMTPENIPSLEALTRKFQDFQVKPAVSTSK RLALSEHVGTTATFSIGSSGNVVYDYQRLLIAYKPAPGTCCYVMKMSPQSMPSLEALTKKFQNFQ	

Sequence profile

The protein sequence profile is calculated running BLAST on the UniRef90 dataset and selecting only the hits with e-value < 10⁻⁹.

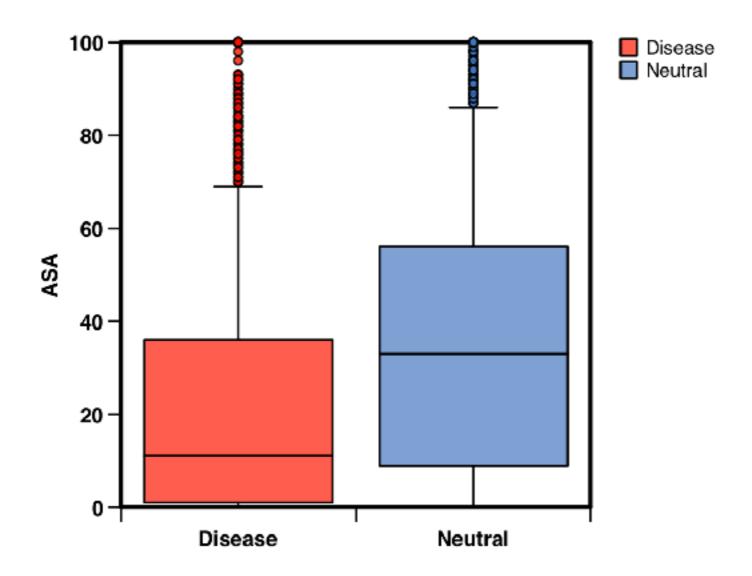
The frequency distributions of the wild-type residues for disease-related and neutral variants are significantly different (KS p-value=0).



Capriotti et al (2012). Briefings in Bioinformatics. 13; 495-512.

Structure environment

There is a significant difference (KS p-value = 2.8x10⁻⁷¹) between the distributions of the Relative Accessible Solvent Area for disease-related and neutral variants. Their mean values are respectively 20.6 and 35.7.

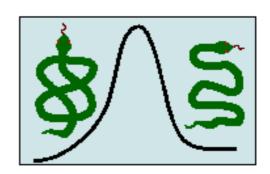


Capriotti and Altman. (2011) BMC Bioinformatics.12 (Suppl 4); S3.

Protein variant databases

Many database are available collecting data about protein variations. Some of them are not longer updated











Center for Information Biology and DNA Data Bank of Japan National Institute of Genetics

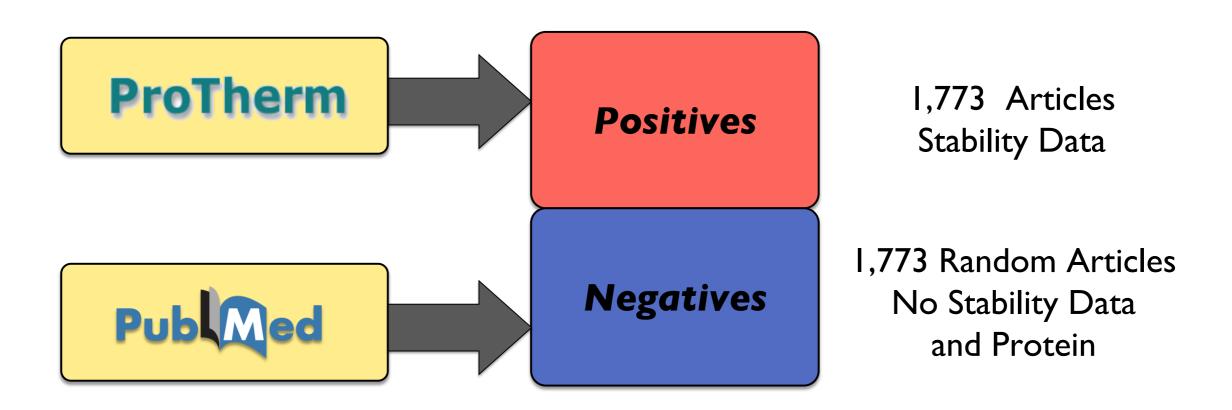
Open challenges

Automatic retrieval of the data from literature

- Integration of different sources of data in a dedicated repository
- Benchmarking of new computational methods
- Understanding the relationships between protein stability, function and disease

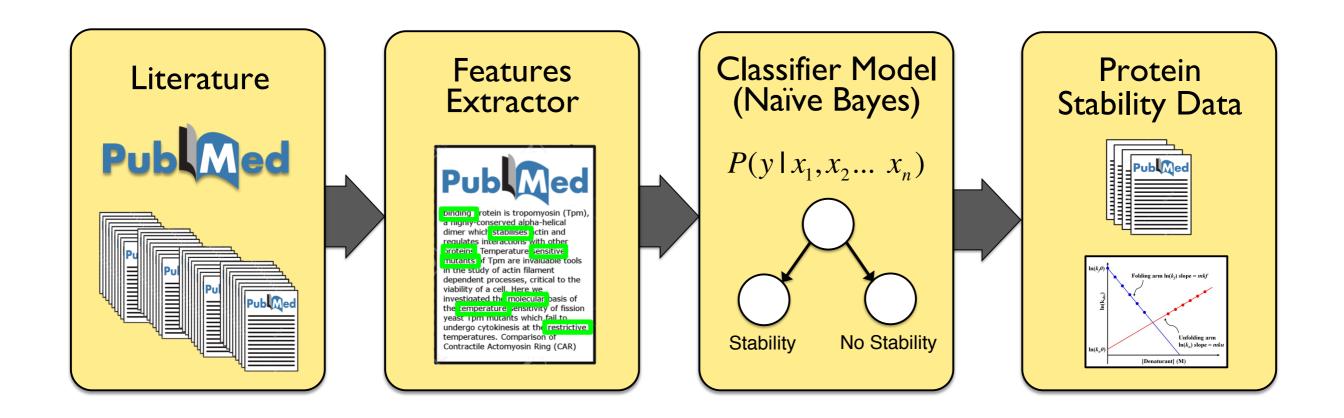
Literature dataset

We collected a set of articles from ProTherm database and use them for training a machine learning method. Negative set are selected from PubMed.



Data manipulation

We extracted the words from the full text version of the manuscript and build a Naive-Bayes classifier based on the occurrence of the different words.

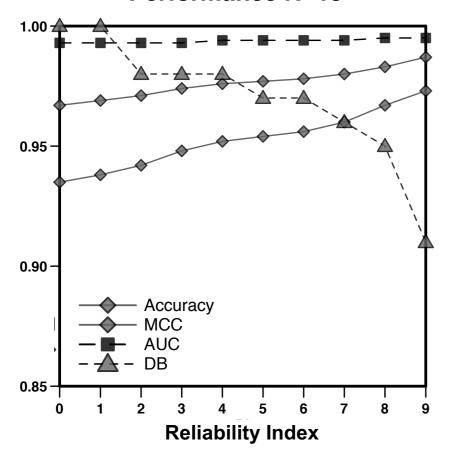


Method performance

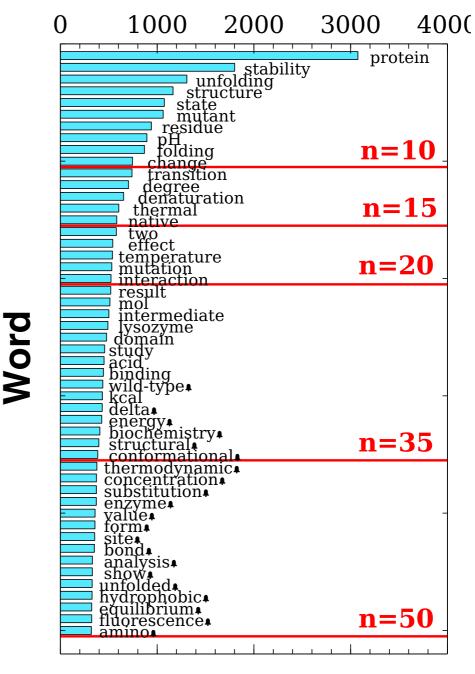
We performed different tests using different number of words as prediction features

Features	Accuracy	MCC	AUC
10	0.95	0.90	0.99
15	0.97	0.94	0.99
20	0.97	0.94	0.99
35	0.97	0.94	1.00
50	0.97	0.94	1.00

Performance N=15

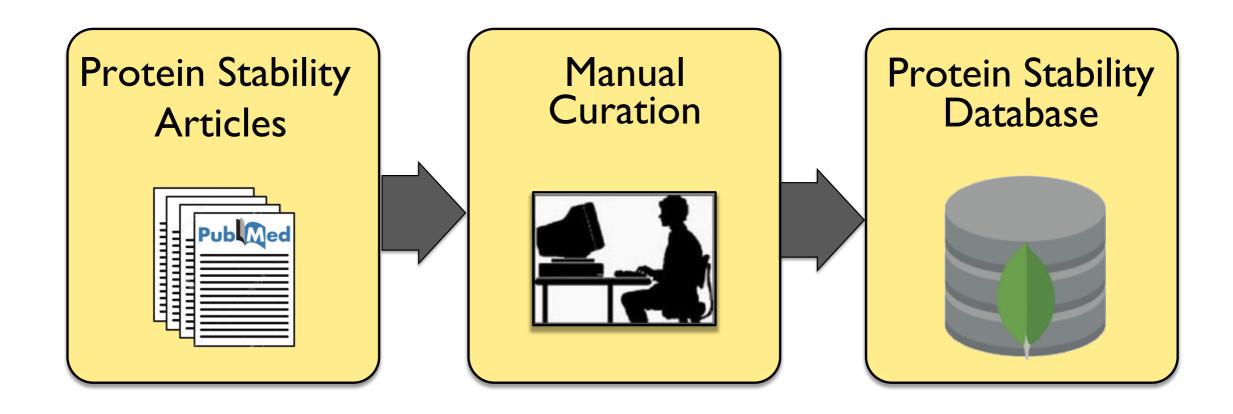






Data retrieval

The implementation of the previous machine learning method will be useful to simplify the manual curation process preselecting a set of manuscripts with high probability of including stability data.



Meta prediction

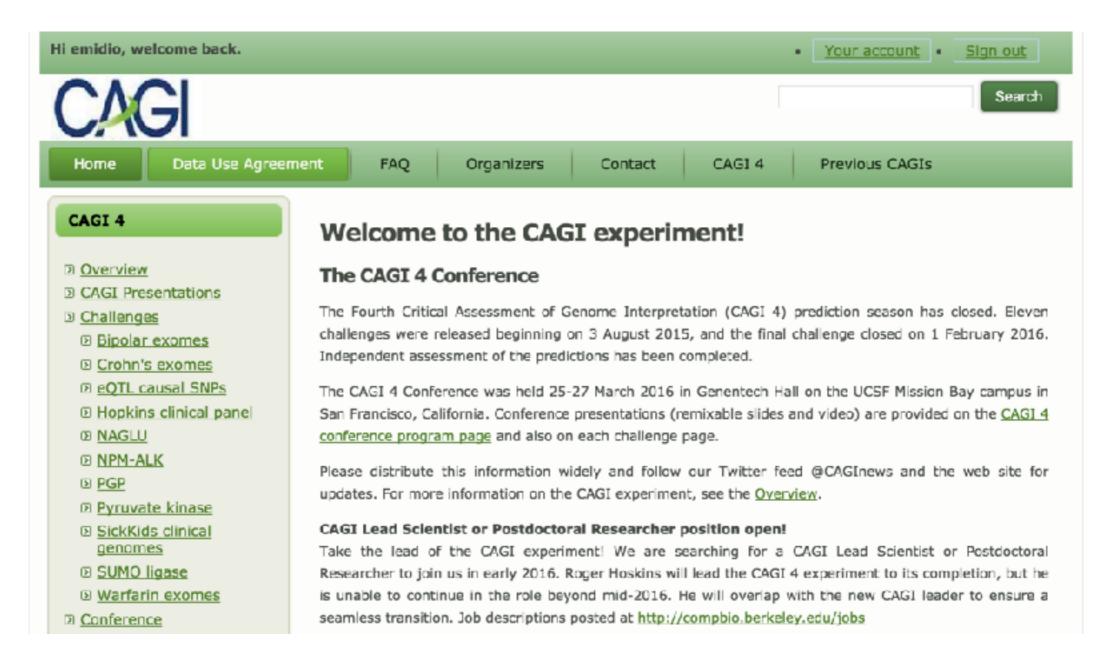
One approach that we previously tested for the prediction of functionally deleterious variants is the development of a meta prediction approach.

- We can integrate different prediction methods for protein stability change
- We can train methods on different type of variant datasets

The final goal consists in selecting highly-reliable predictions

CAGI experiments

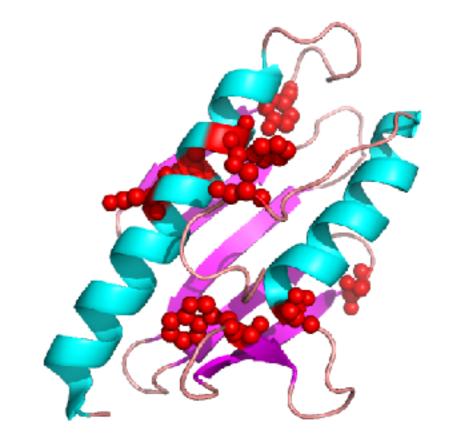
The Critical Assessment of Genome Interpretation is a community experiment to objectively assess computational methods for predicting the phenotypic impacts of genomic variation.

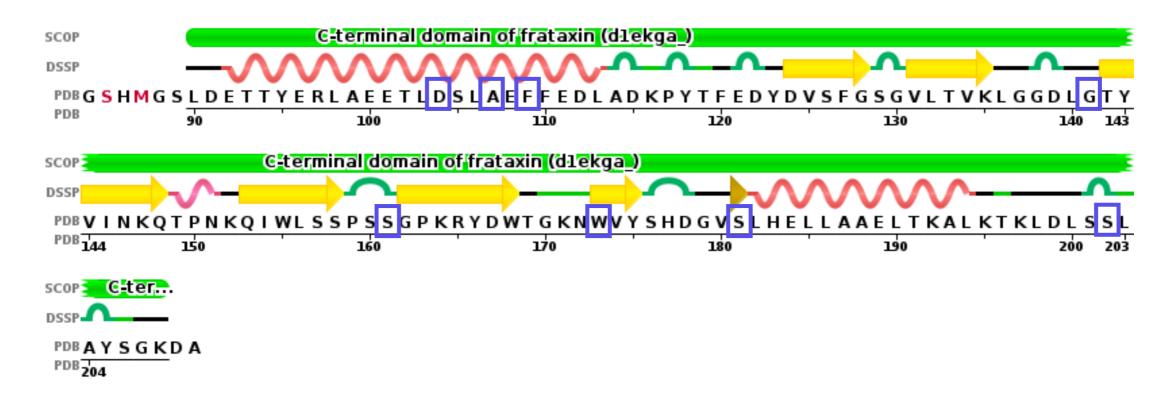


https://genomeinterpretation.org/

Frataxin challenge at CAGI

- Participants were asked to submit predictions of the variation of the unfolding free energy change upon mutation at concentration 0 of denaturant ($\Delta\Delta G_{H20}$).
- Data providers at the University of Roma experimentally determined the unfolding free energy of the wild-type and 8 mutants using CD and Fluorescence. The average value between the two ΔΔG_{H20} is considered for the assessment

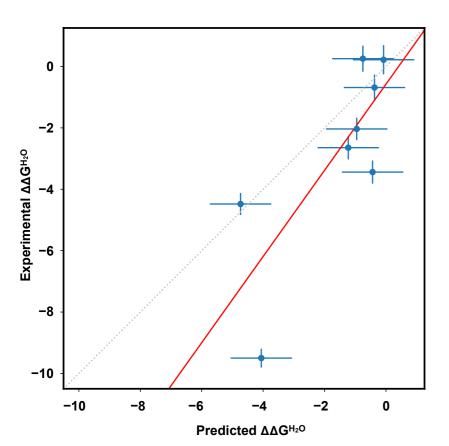




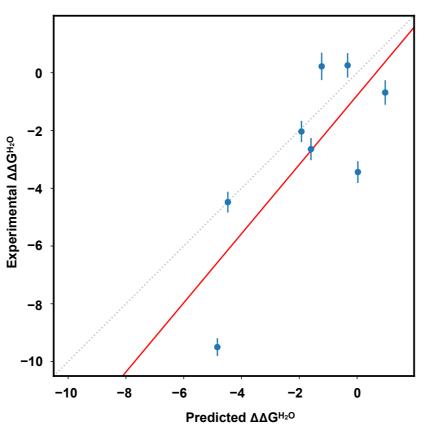
Fold-X and I-Mutant

The performance of Fold-X is comparable with the negative of predictions from Group 6

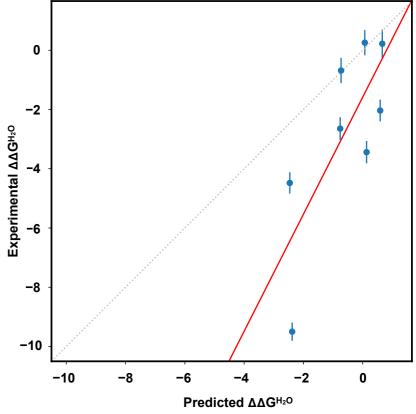
Reverse Group_6
PCC=0.78 - SPC=0.71 - KTC=0.57
RMSE=2.32 - MAE=1.60



Fold-X PCC=0.77 - SPC=0.62 - KTC=0.50 RMSE=2.24 - MAE=1.62



I-Mutant
PCC=0.76 - SPC=0.64 - KTC=0.50
RMSE=3.13 - MAE=2.24

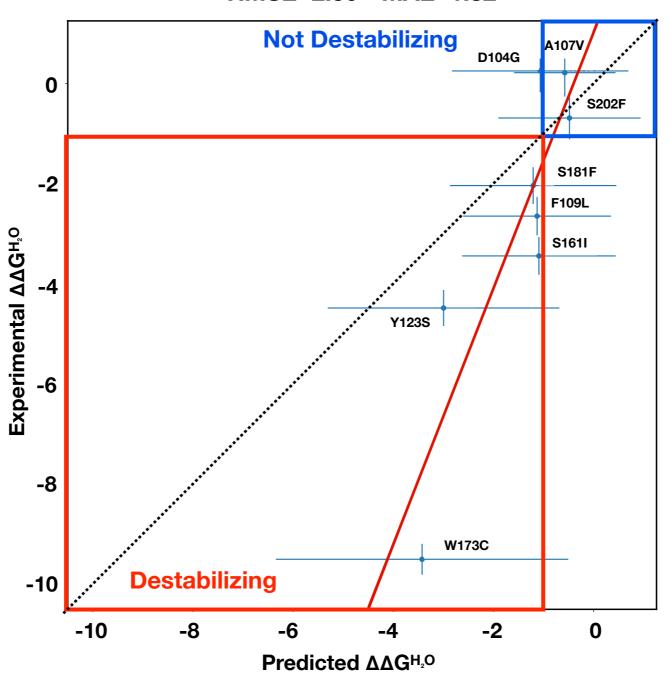


Averaged predictions

For W173C we found the highest difference between predicted and experimental $\Delta\Delta$ Gs

Average Predictions

PCC=0.88 - SPC=0.81 - KTC=0.57 RMSE=2.50 - MAE=1.82



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Gustavo Parisi, Univesidad de Quilmes
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Joost Schymkowitz, KU Leuven

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Marco Polo Research Project

BIOSAPIENS Network of Excellence

SPINNER Consortium

Biomolecules, Folding and Disease



http://biofold.org/