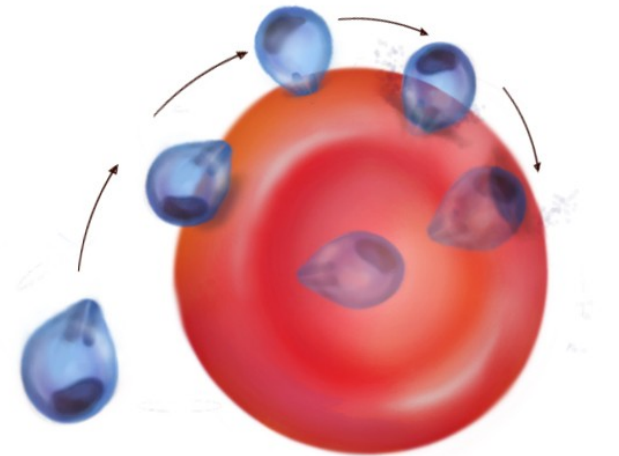
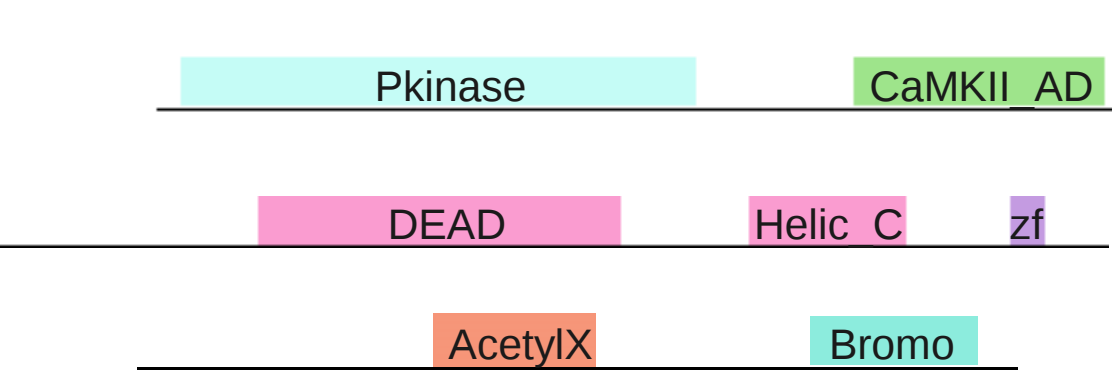


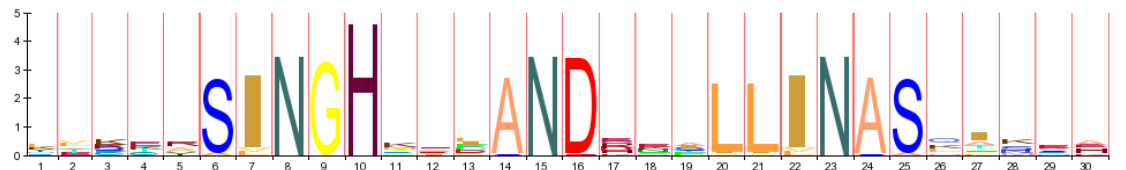
Protein domain prediction using context statistics, the false discovery rate, and comparative genomics, with application to *Plasmodium falciparum*



Alejandro Ochoa

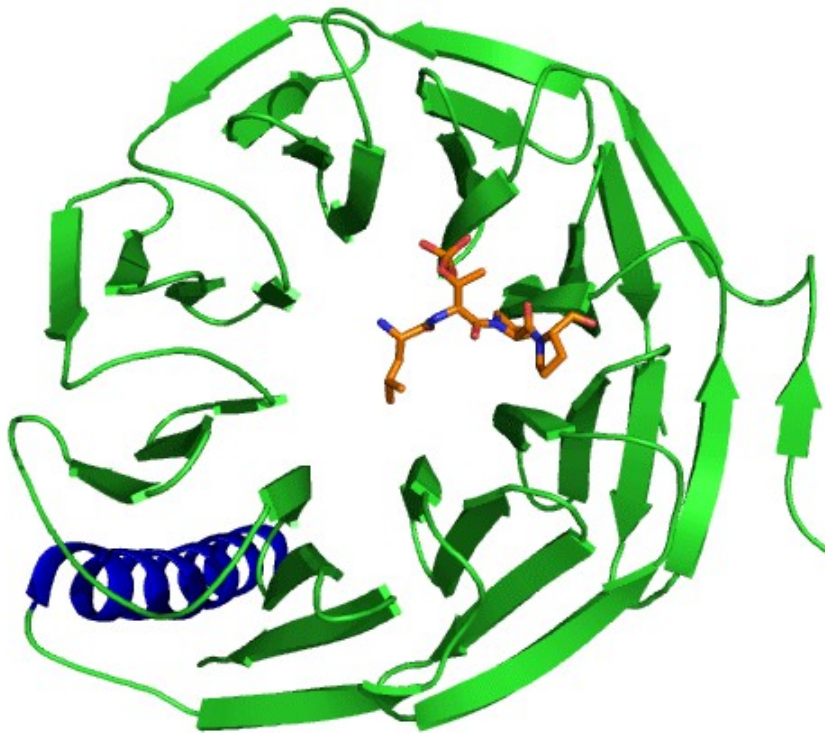
Molecular Biology, Princeton University

Final Public Oral, 2013-06-27



Labs

Protein domains

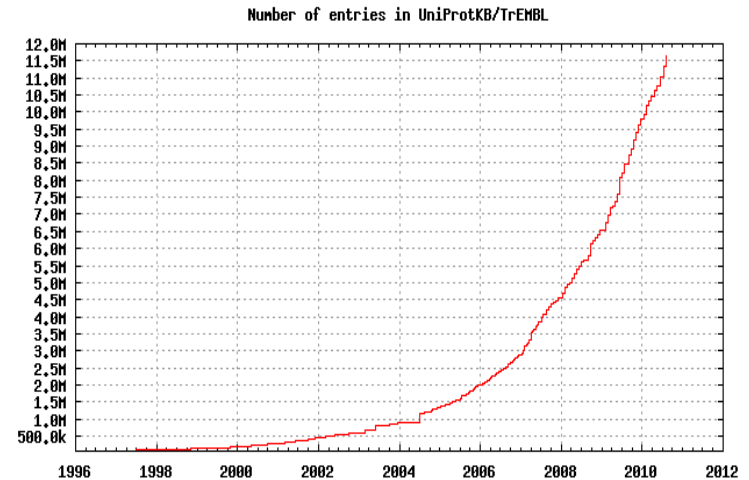
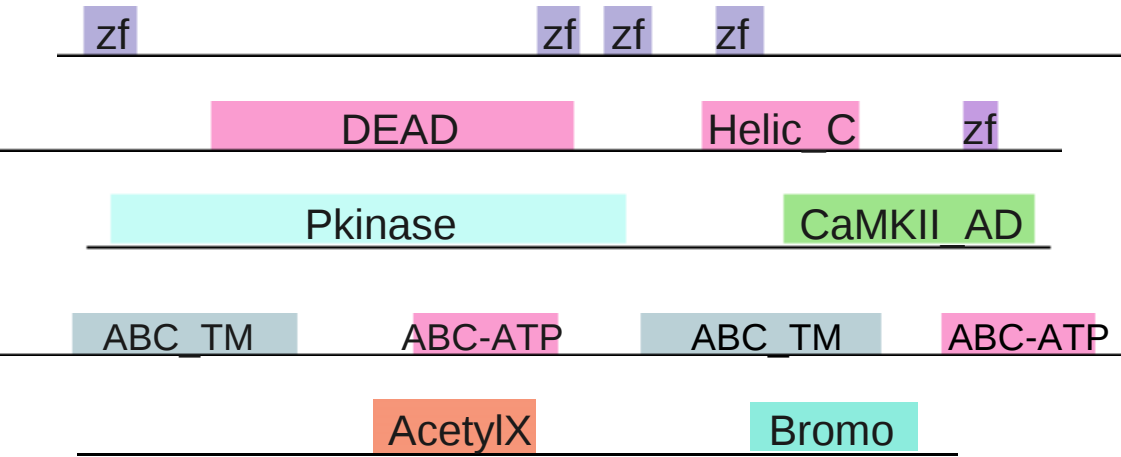


Structure
Evolution
Function

Sequence-based domain prediction:



Why predict domains?



For new sequences, before experiments start...

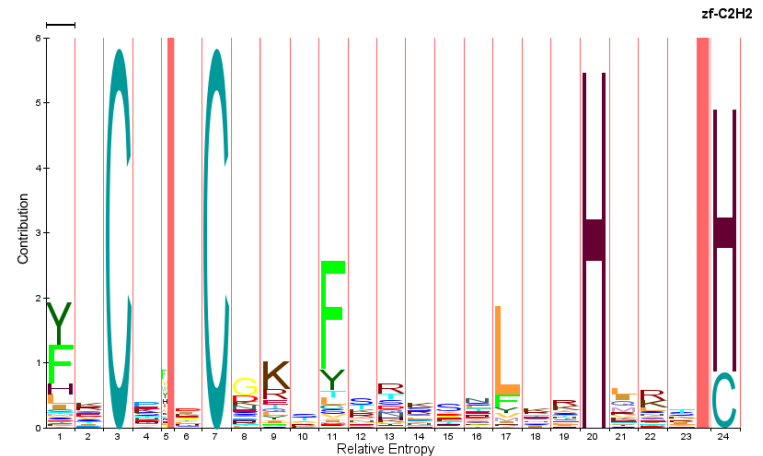
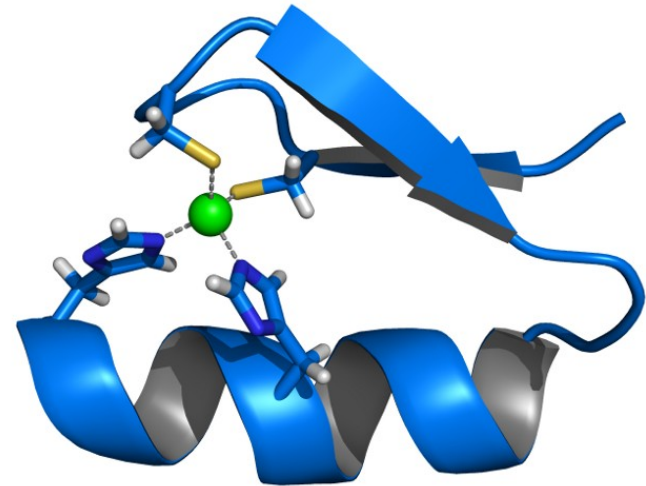
Domains may imply functions

Experimental alternatives are unfeasible as protein databases grow exponentially

Representing Domains

SNAI_DROME/362-385
 SNAI_XENLA/232-255
 SNAI_MOUSE/236-259
 ESCA_DROME/426-449
 SUHW_DROAN/221-243
 TERM_DROME/323-346
 Z020_XENLA/174-196
 EVI1_HUMAN/217-239
 Z02_XENLA/34-59
 EVI1_HUMAN/21-44
 ZNF10_HUMAN/517-539
 ZNF91_HUMAN/238-260
 ZFP58_MOUSE/120-142
 TRAI_CAEEL/306-331
 ZNF76_HUMAN/345-368
 ZN12_MICSA/106-129
 LOLA1_DROME/794-817
 ZNF17_HUMAN/435-457
 ZG32_XENLA/34-56
 TF3A_BUFAM/104-128
 ZG46_XENLA/146-168
 MZF1_HUMAN/412-434
 ZN239_MOUSE/6-28
 ZSC22_HUMAN/352-374
 EGR1_HUMAN/396-418
 SUHW_DROAN/349-373
 CF2_DROME/485-508
 CF2_DROME/401-423
 KRUP_DROME/306-328
 TTY1_HUMAN/383-407
 ZG52_XENLA/61-83
 TTKB_DROME/538-561
 ZNF76_HUMAN/285-309
 SDC1_CAEEL/145-168
 SRYC_DROME/358-380
 SDC1_CAEEL/270-292
 TRAI_CAEEL/276-300
 ESCA_DROME/370-392

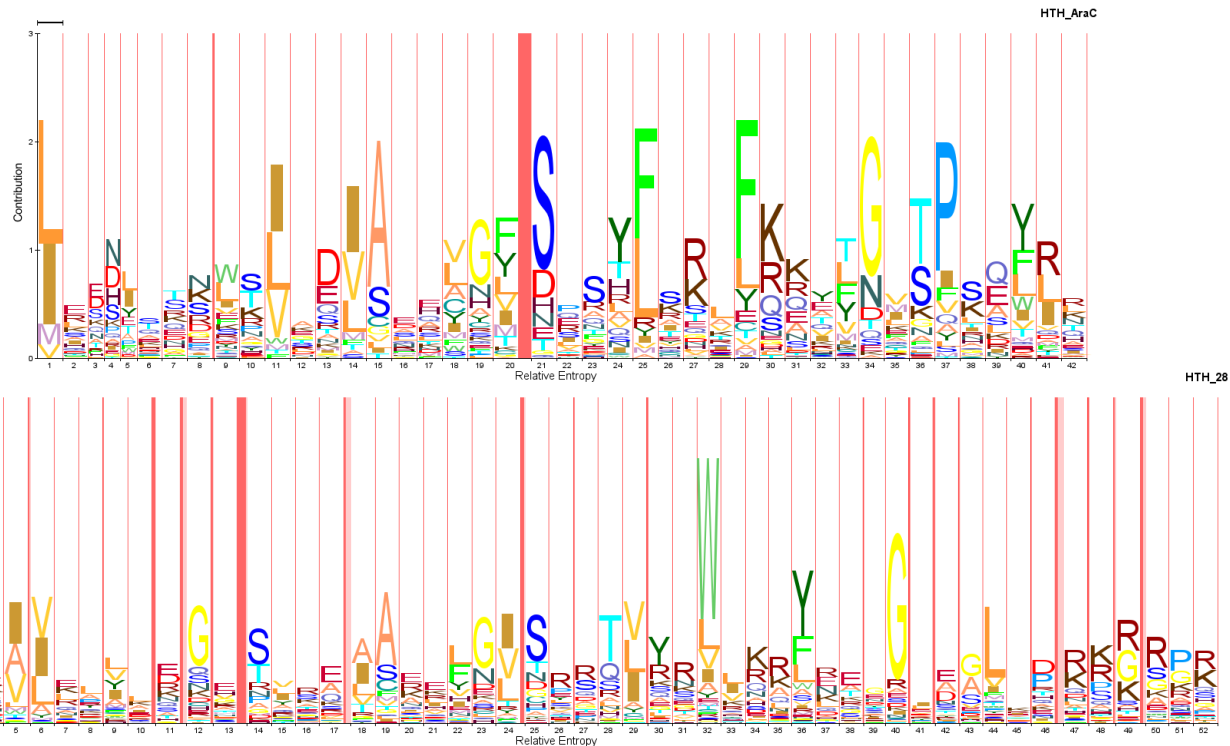
YACQ...VCH...KSFSRM...SLLNKHSSS...NC
 YQCK...SCS...RTFSRM...SLLHKHEET...GC
 YQCQ...ACA...RTFSRM...SLLHKHSES...GC
 YSCT...SCS...KTFSRM...SLLTKHSEG...GC
 HVCG...KCY...KTFRRL...MSLKKHLEF...C
 LHCR...RCR...TQFSRR...SKLHIHQKL...RC
 FMCA...DCG...RCFSVS...SSLKYHQRI...C
 IKCK...DCG...QMFSTT...SSLNKHRRF...C
 YSCA...DCG...KHFSEK...MYLQFHQKNPSEC
 YRCE...DCD...QLFESK...AELADHQKF...PC
 YKCN...QCG...IIFSQN...SPFIVHQIA...H
 YKCE...ECG...KAFKQL...STLTTHKII...C
 IKCE...ECG...KAFSTR...STYYRHQKN...H
 YKCEF.ADCE...KAFSNA...SDRAKHQNR...TH
 YTCS...TCG...KTYRQT...STLAMHKRS...AH
 YRCS...QCG...KAFRRT...SDLSSHRT...QC
 YECR...HCG...KKYRWK...STLRRHENV...EC
 YECN...KCG...KFFRYC...FTLNRHQRV...H
 FVCV...HCG...KGFRDM...YKLSLHLRI...H
 YVCYF.ADCG...QQFRKH...NQLKIHOYI...H
 YVCT...ECG...TSFRVR...POLRIHLRT...H
 FVCG...DCG...QGFRVS...ARLEEHRV...H
 YKCD...KCG...KGFRS...SSLLVHHSV...H
 YKCG...ECG...KTFSRS...THLTQHQRV...H
 FACD...ICG...RKFRS...DERKRHTKI...H
 YACK...ICG...KDFTRS...YHLKRHQKYS...SC
 YTCP...YCD...KRFTQR...SALTVHTTK...LH
 YTCS...YCG...KSFTQS...NTLKQHTRI...H
 YTCE...ICD...GKFSDS...NQLKSHMLV...H
 YVCPF.DGCN...KKFAQS...TNLKSHILT...H
 YTCT...QCN...KQFSHS...AQLRAHIST...H
 YPCP...FCF...KEFTRK...DMMTAHVKI...IH
 YTCPE.PHCG...RGFTSA...TMYKNHVRI...H
 YMCQ...VCL...TLFGHT...YNLFMHURT...SC
 YQCD...ICG...QKFVQK...INLTHHARI...H
 YFCH...ICG...TVFIEQ...DMLFKHWRL...H
 NKCEY.PCGG...KEYSRL...ENLKTHRRT...H
 CKCN...LCG...KAFSRP...WLLQGHIRT...H



Databases of Domain Families

This work uses Pfam and HMMER, but theory and results are general

Two members of Pfam “Clan” HTH



Plasmodium falciparum

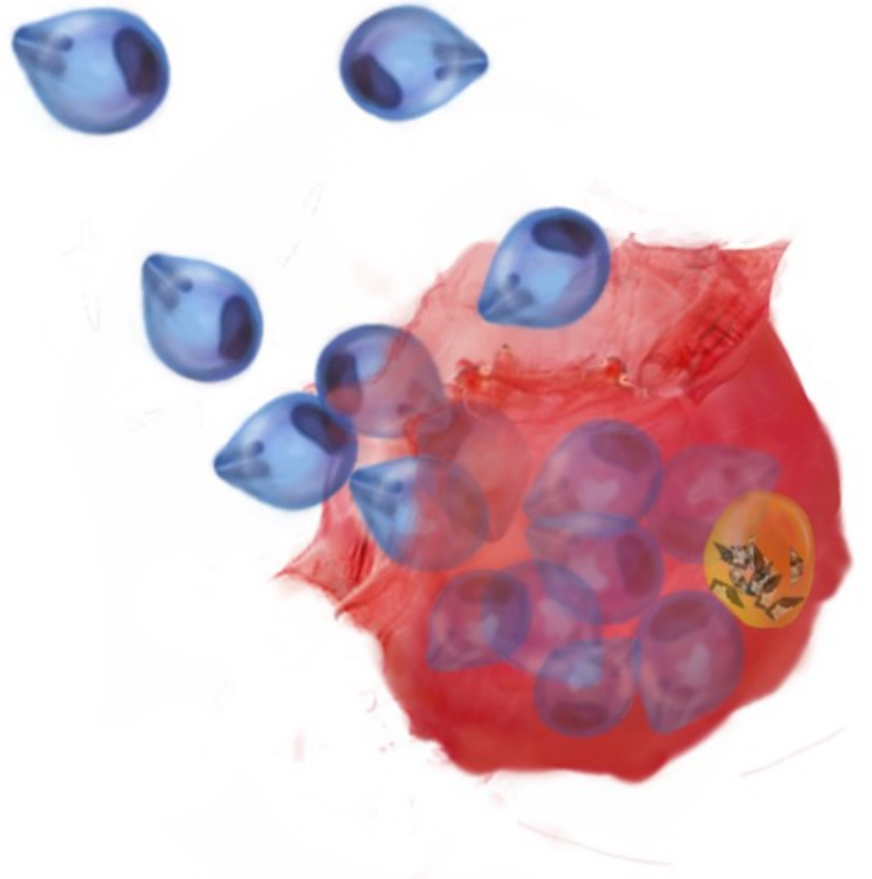
Malaria

Information challenges

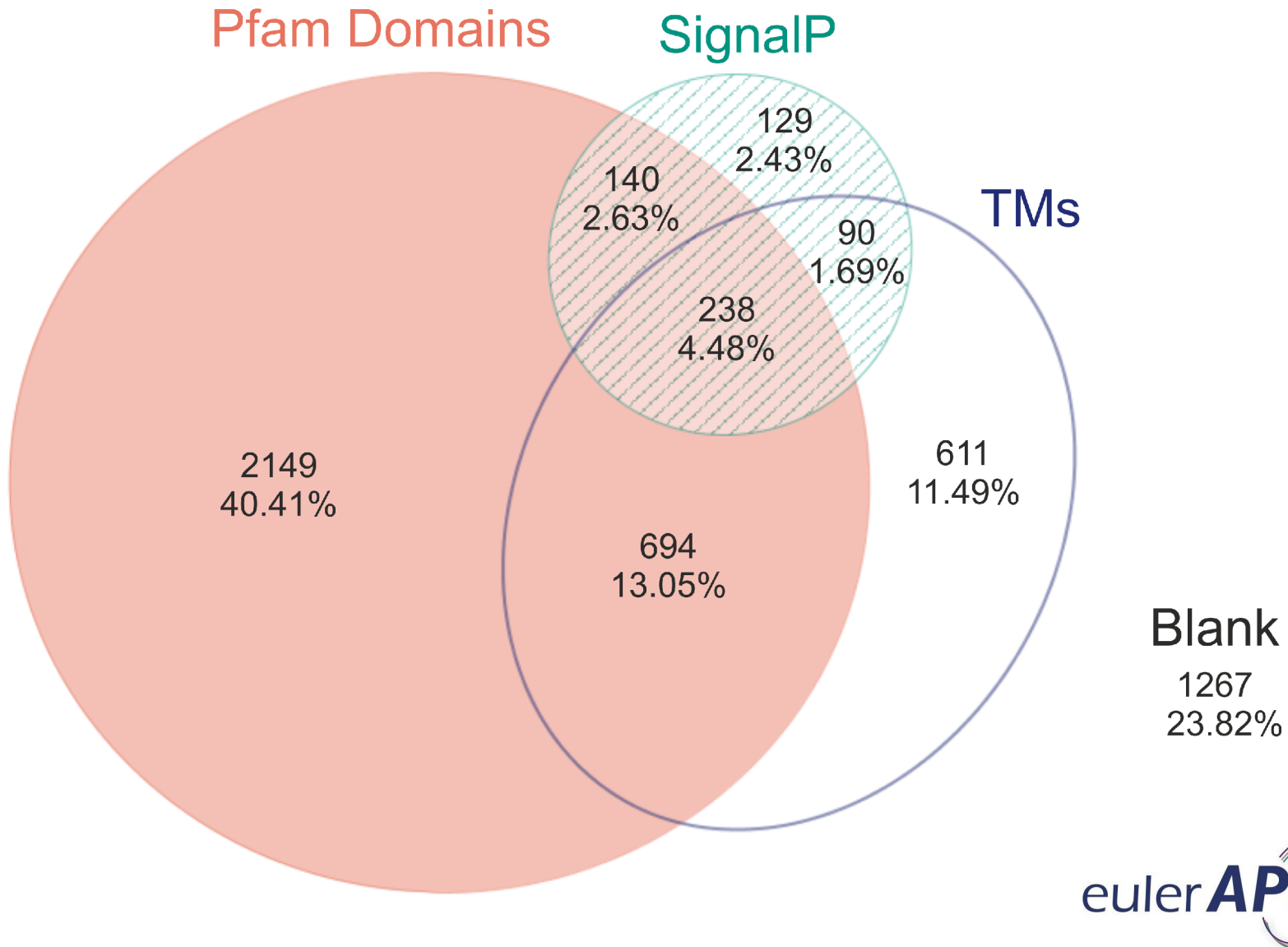
- Diverged eukaryote
- 80% AT-bias
- Low-complexity regions

Annotation

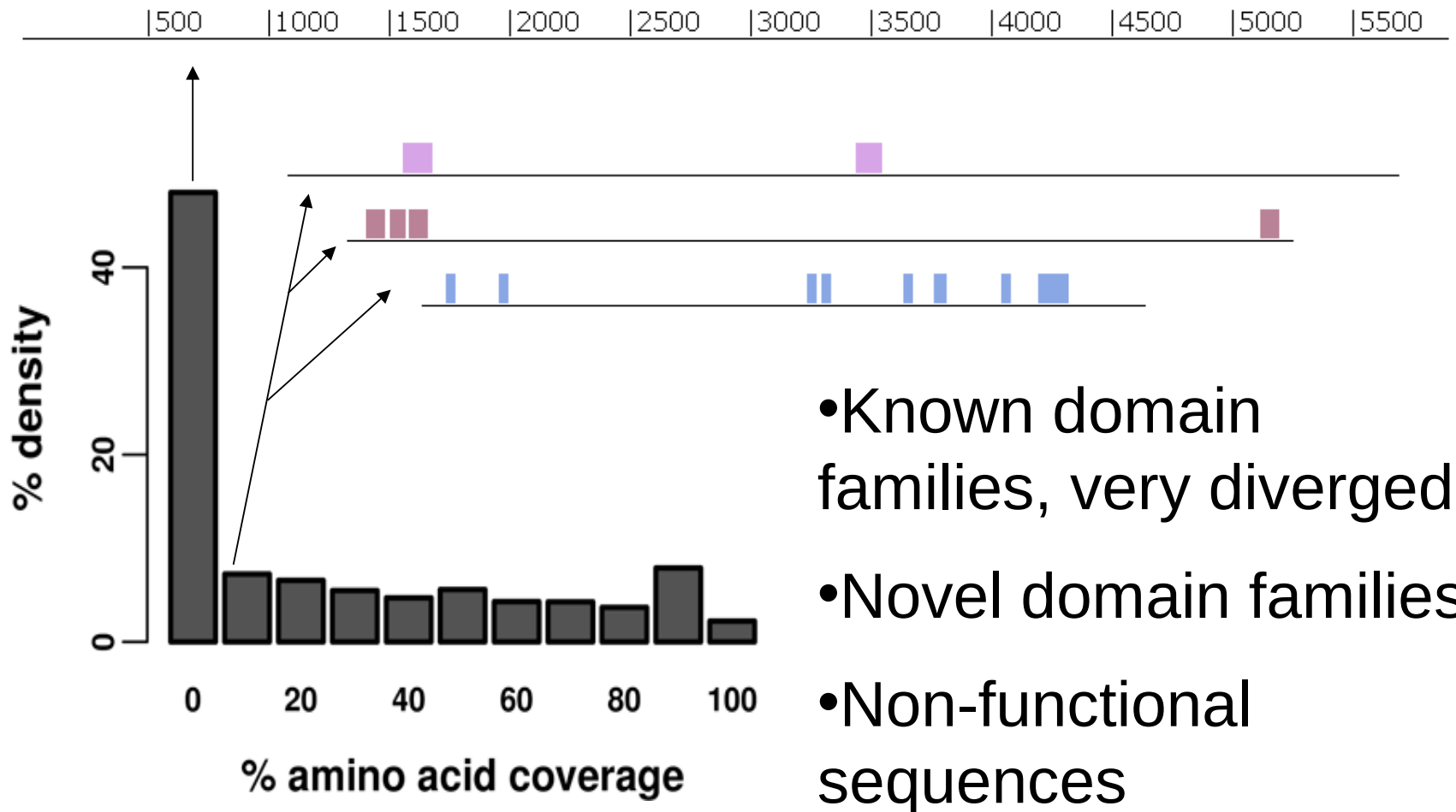
- 5.5K proteins
- 45% unknown function
 - 20% unknown in yeast
- 88% of annotations are bioinformatical



Sequence-based annotation of *Plasmodium falciparum* proteome



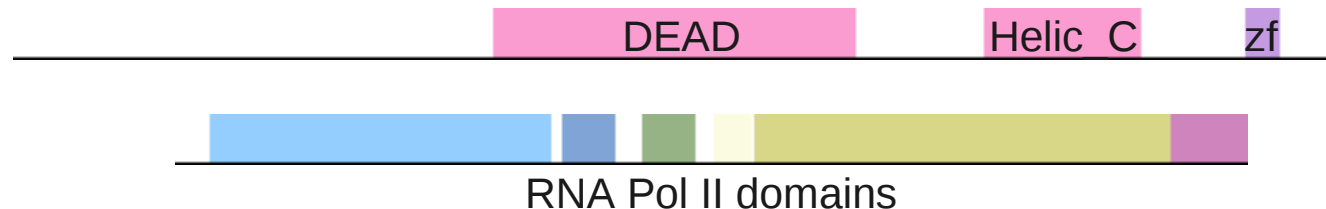
Poor domain coverage of *Plasmodium falciparum*



Outline of results

- Domain prediction using context
 - Application to malaria parasite
- Optimal *FDR* control for domains
 - Family-based q -values, adapted to domain problem
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 - Implications for sequence models
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Domain Prediction Using Context: dPUC



Background

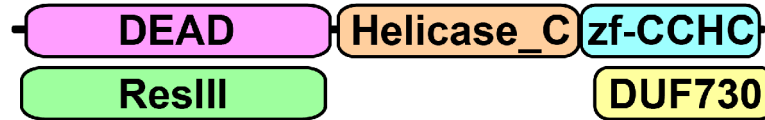
- Domains co-occur in limited combinations
- Domains are scored independently of each other

Idea

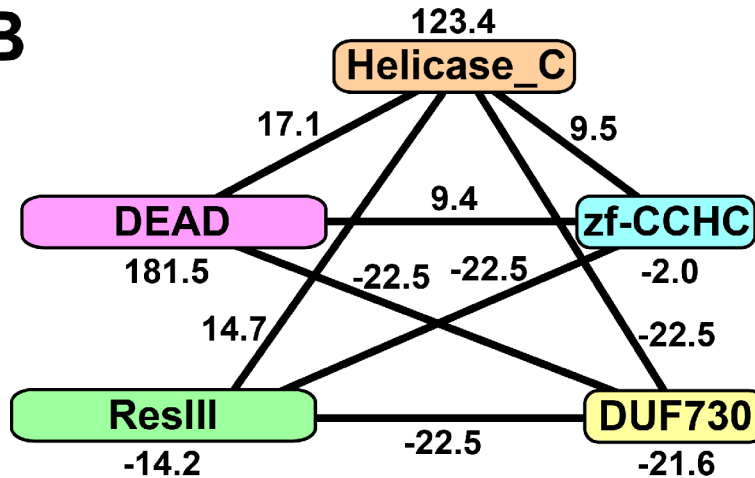
- Score domains in combination
- Context + Sequence evidence

The dPUC method

A



B



C

Standard Pfam:



dPUC Pfam:



<http://compbio.cs.princeton.edu/dpuc/>

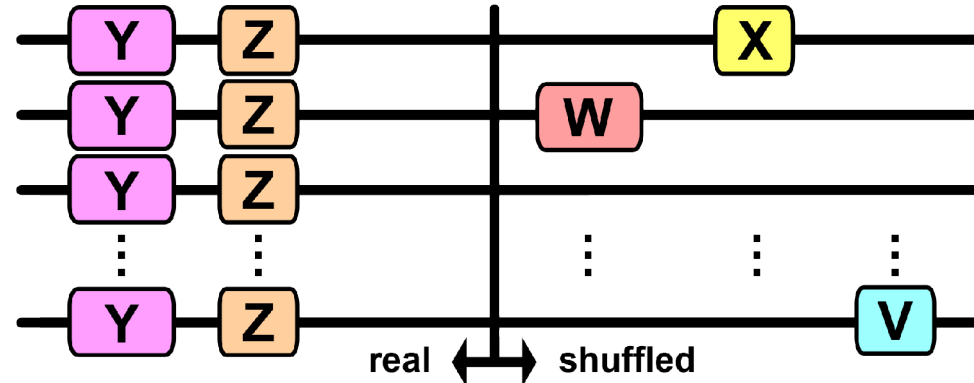
Ochoa, *et al.* BMC Bioinformatics 12, 90 (2011).

Improved signal to noise

Real protein

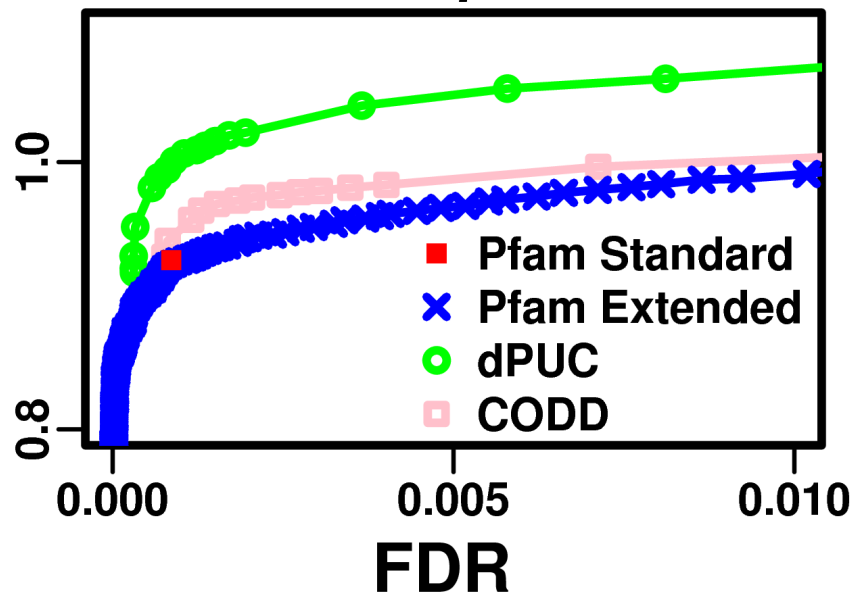


Real protein with shuffled sequences

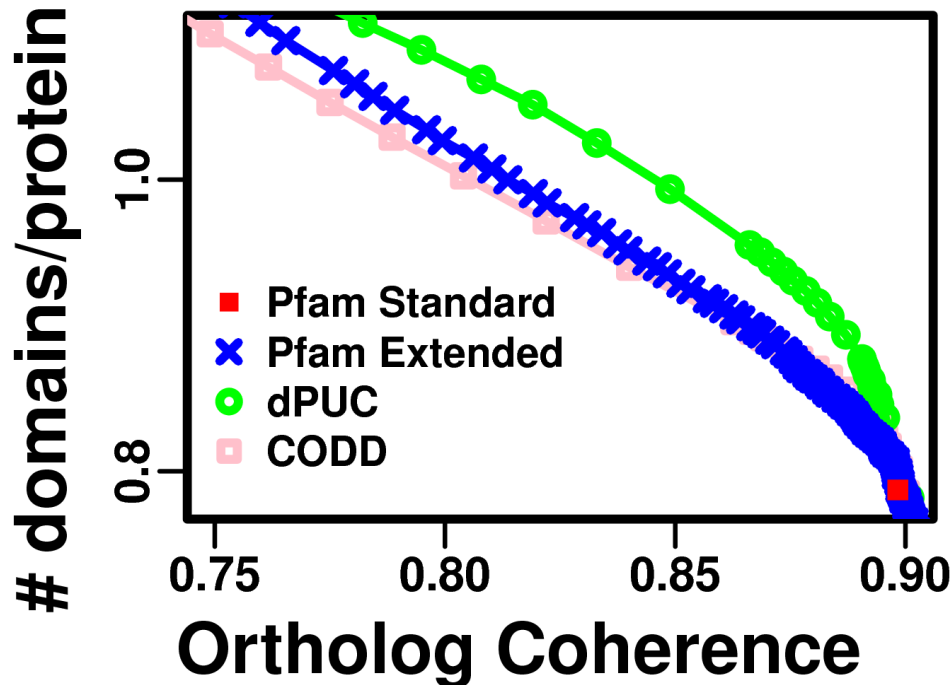
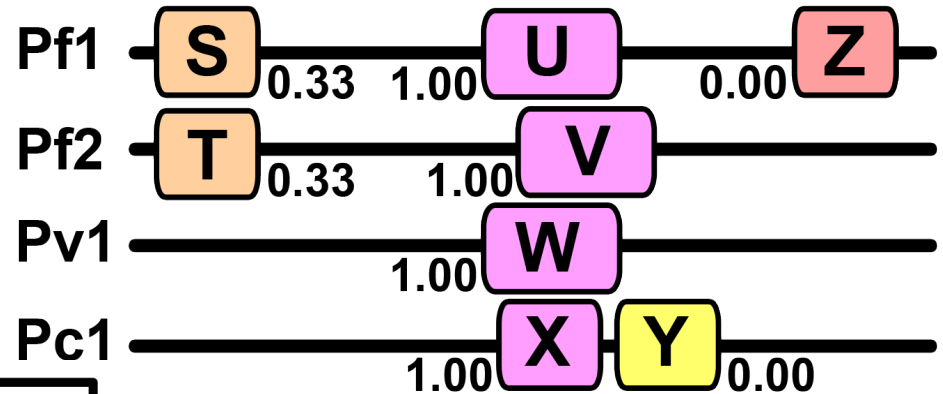


domains/protein

P. falciparum



Improved ortholog coherence on *Plasmodium* species



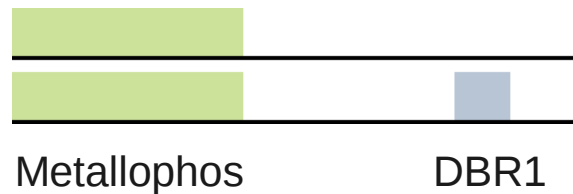
New predictions

Phosphatase -> RNA lariat debranching enzyme

P. falciparum

Std Pfam

dPUC



S. cerevisiae

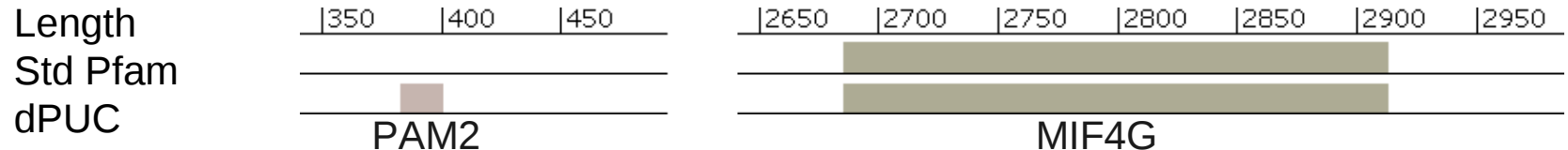
Std Pfam & dPUC



New predictions

MIF4G domain-containing protein ->
Poly-A binding protein-interacting protein 1

P. falciparum



H. sapiens



Interested malaria curator:
Hagai Ginsburg,
Hebrew U of Jerusalem

<http://compbio.cs.princeton.edu/dpuc/>
Ochoa, *et al.* BMC Bioinformatics 12, 90 (2011).

New predictions

RNA helicase -> mRNA sequestration

Description
Organism
Standard Pfam
dPUC Pfam

RNA helicase-1
P. falciparum



Description
Organism
Standard Pfam
dPUC Pfam

DDX41_DROME ATP-dependent RNA helicase abstrakt
D. melanogaster



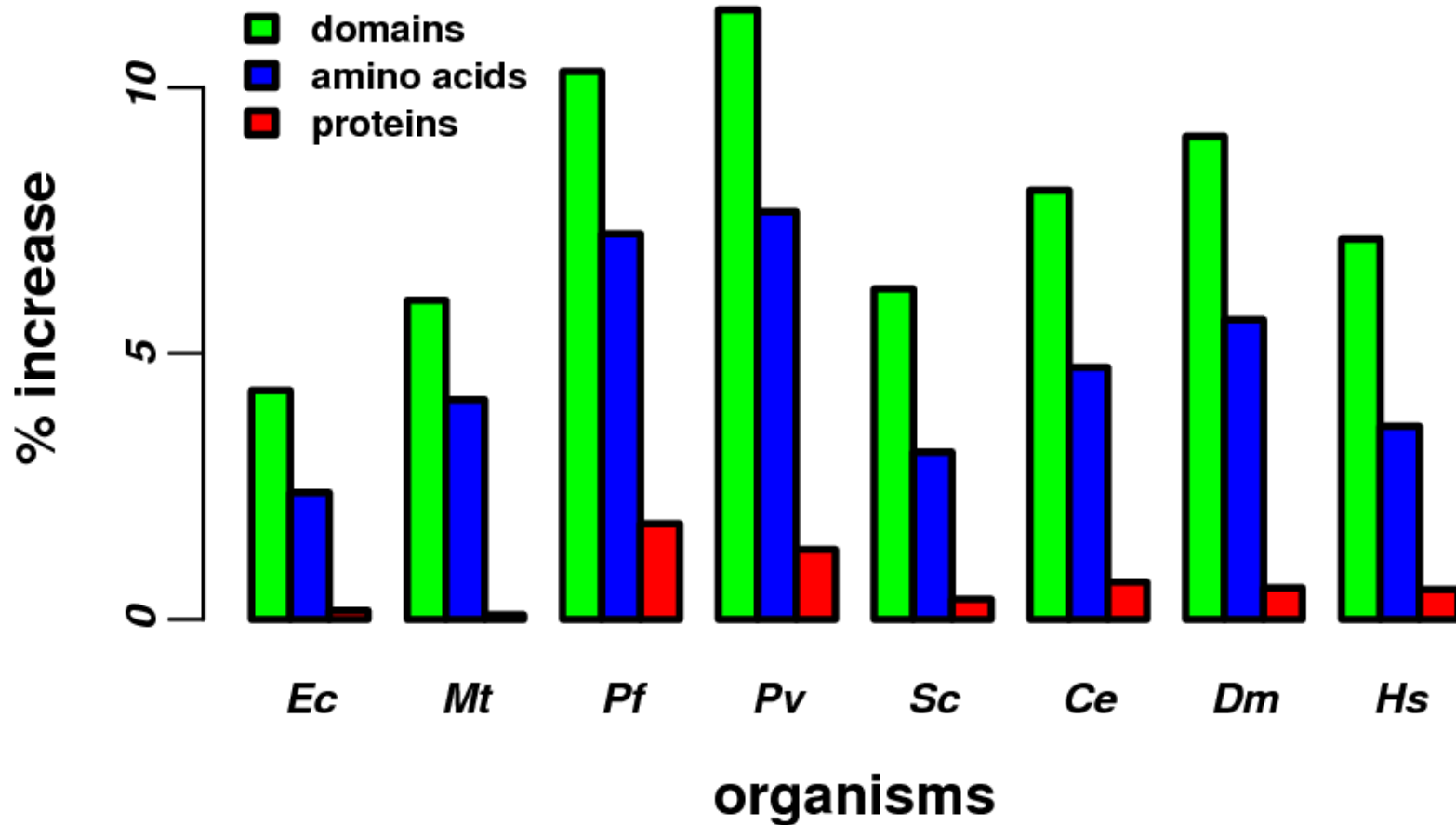
Description
Organism
Standard Pfam
dPUC Pfam

DDX41_HUMAN Probable ATP-dependent RNA helicase DDX41
H. sapiens

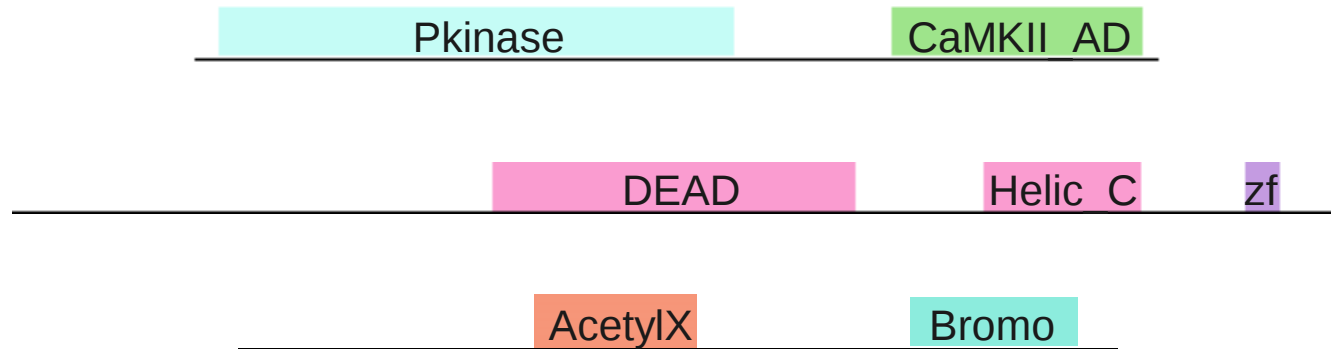


DEAD Helicase_C zf-CCHC

dPUC increases coverage

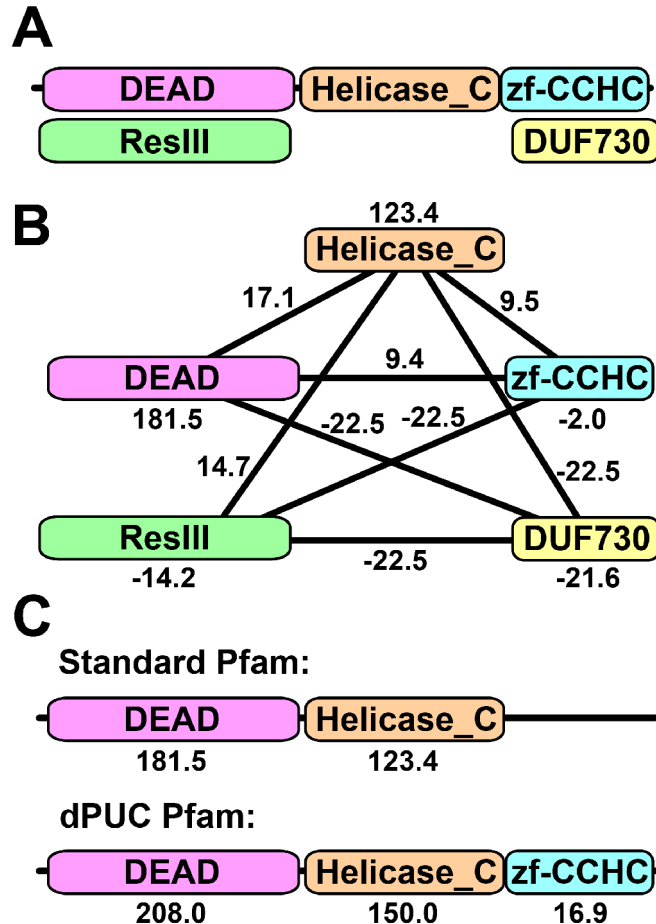


Domain context



Complements sequence evidence
Improves domain predictions
Works best on diverged organisms

The dPUC method: problems

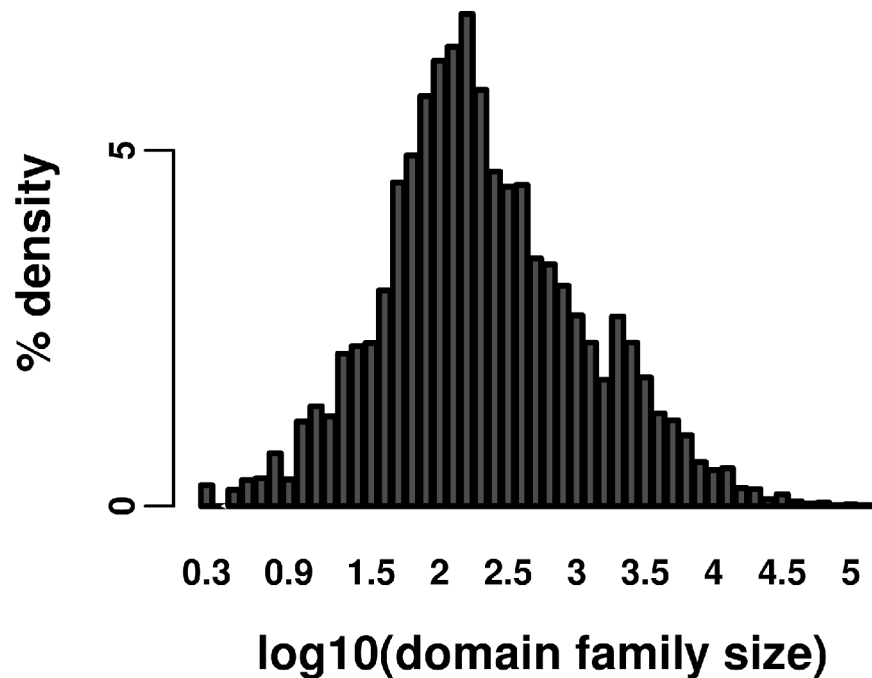


Outline of results

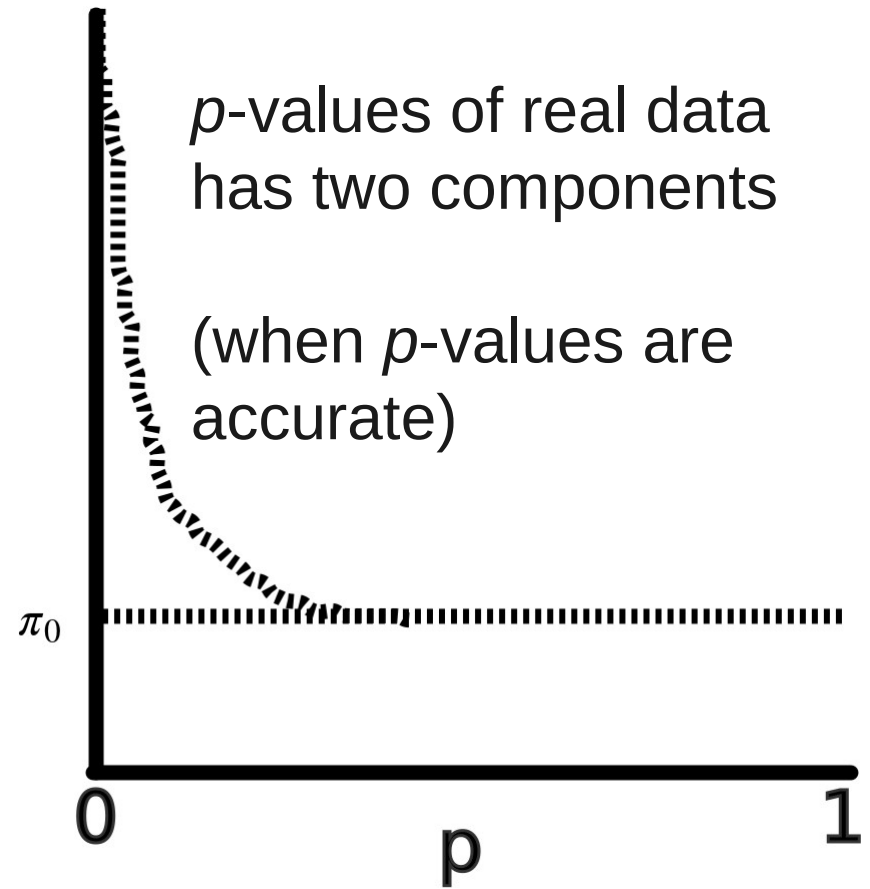
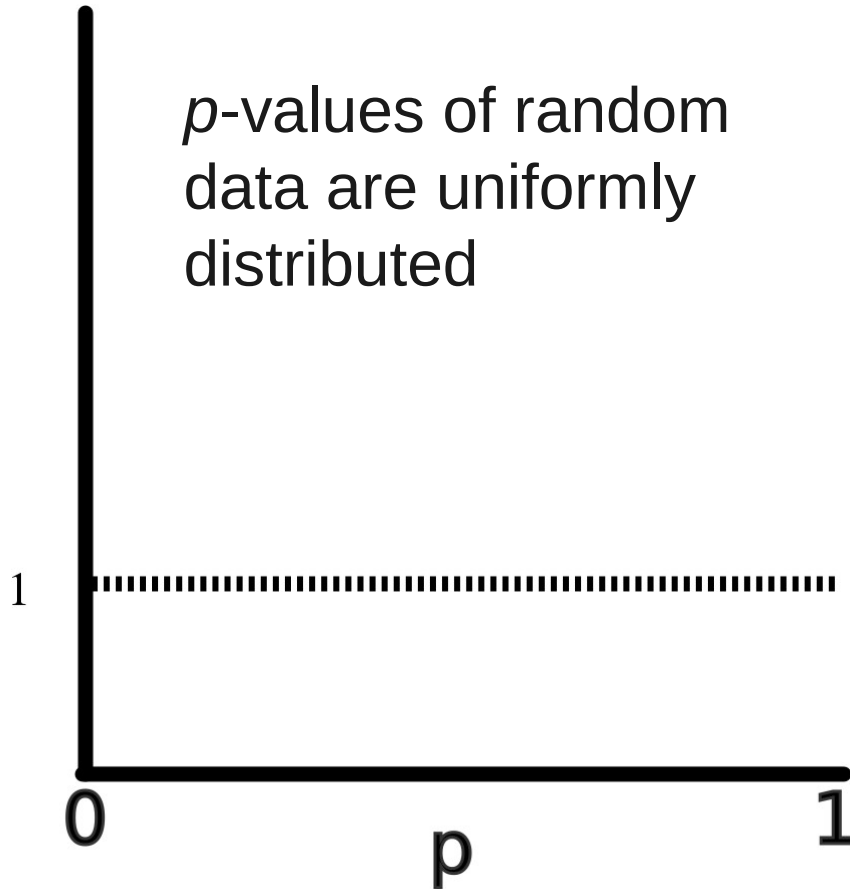
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$$FDR = E/n$$

FDR = average posterior error probability

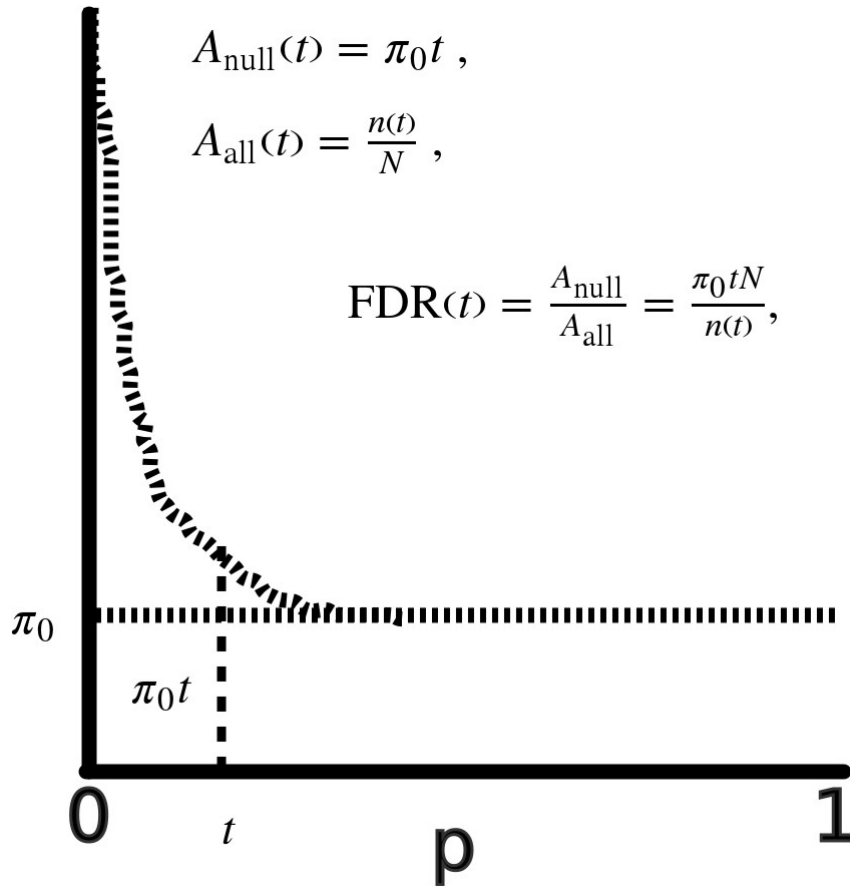


Computing q -values



Step 1: estimate π_0 (proportion of data that is false)

Computing q -values



Step 2: Directly estimate $FDR(t)$ for all thresholds t

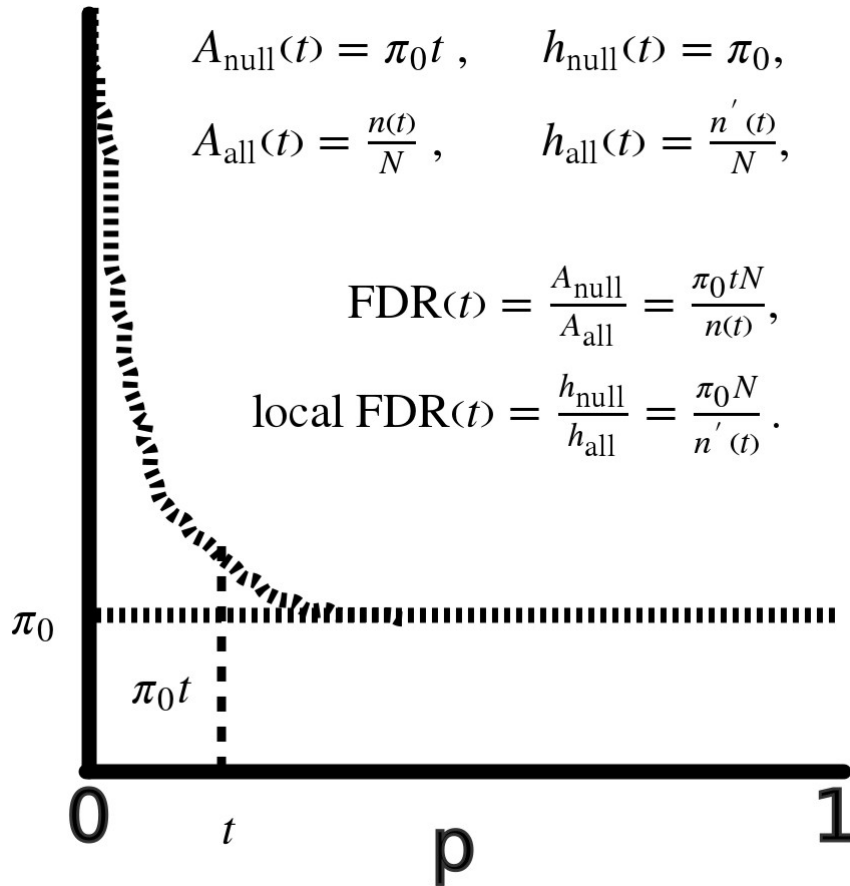
– $N = \#$ tests

– $n(t) = \#$ sig tests

Step 3: Ensure monotonicity

$$q(p) = \min_{t; p \leq t} FDR(t)$$

FDR and local FDR



Local FDR =
Posterior Error
Probability (*PEP*)

FDR = average *PEP*
of significant
predictions

Local FDRs optimize domain prediction

Find domain family thresholds t_i (for each family i) to maximize predictions

$$M = \sum n_i(t_i),$$

while constraining the combined *FDR* of all families to Q

$$Q \geq \frac{\sum \pi_{0,i} \cdot t_i \cdot N_i}{\sum n_i(t_i)} = \frac{\sum FDR_i(t_i) \cdot n_i(t_i)}{\sum n_i(t_i)}.$$

Equal family *local FDRs* solve this optimization!
Also holds when constraining combined *E-value*!

Empirical null models

$$FDR = \#FP / (\#FP + \#TP)$$

2nd order Markov Random Sequences

- FP if domain came from random sequence

Improved from Ochoa, *et al.* BMC Bioinformatics 12, 90 (2011).

Ortholog Set Coherence

- FP if orthologs don't predict any homologous domains

Improved from Ochoa, *et al.* BMC Bioinformatics 12, 90 (2011).

Clan Overlap

- FP if domain overlaps stronger non-homologous domain

Inspired/adapted from S. Eddy (p.c., 2012).

Context Coherence

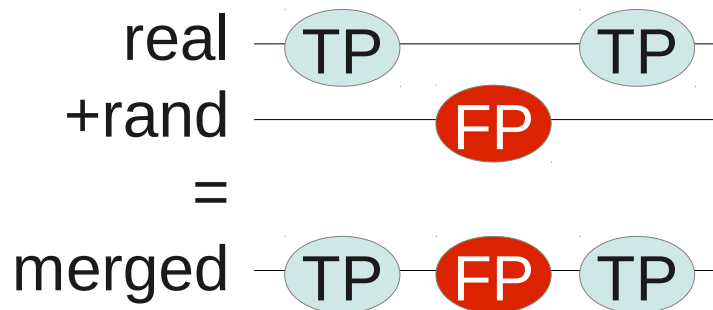
- FP if domain doesn't co-occur with any stronger domains

Inspired/adapted from Terrapon, *et al.* BMC Bioinformatics 13, 67 (2012).

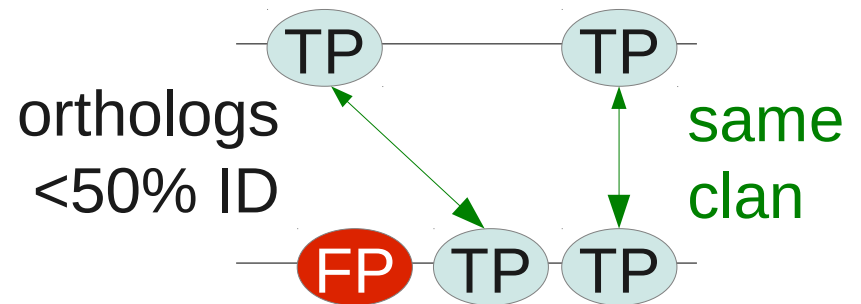
Empirical null models

$$FDR = \#FP / (\#FP + \#TP)$$

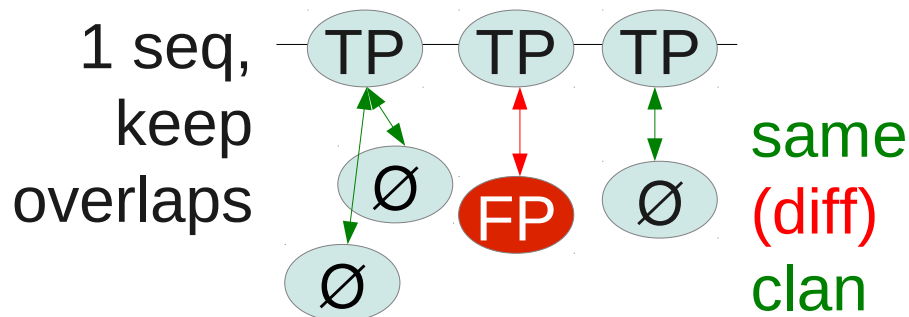
2nd order Markov Random



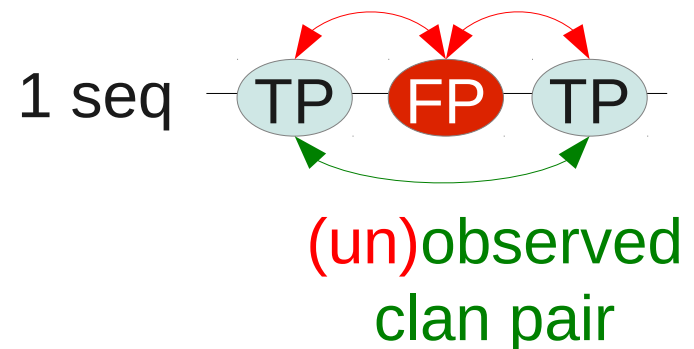
Ortholog Set Coherence



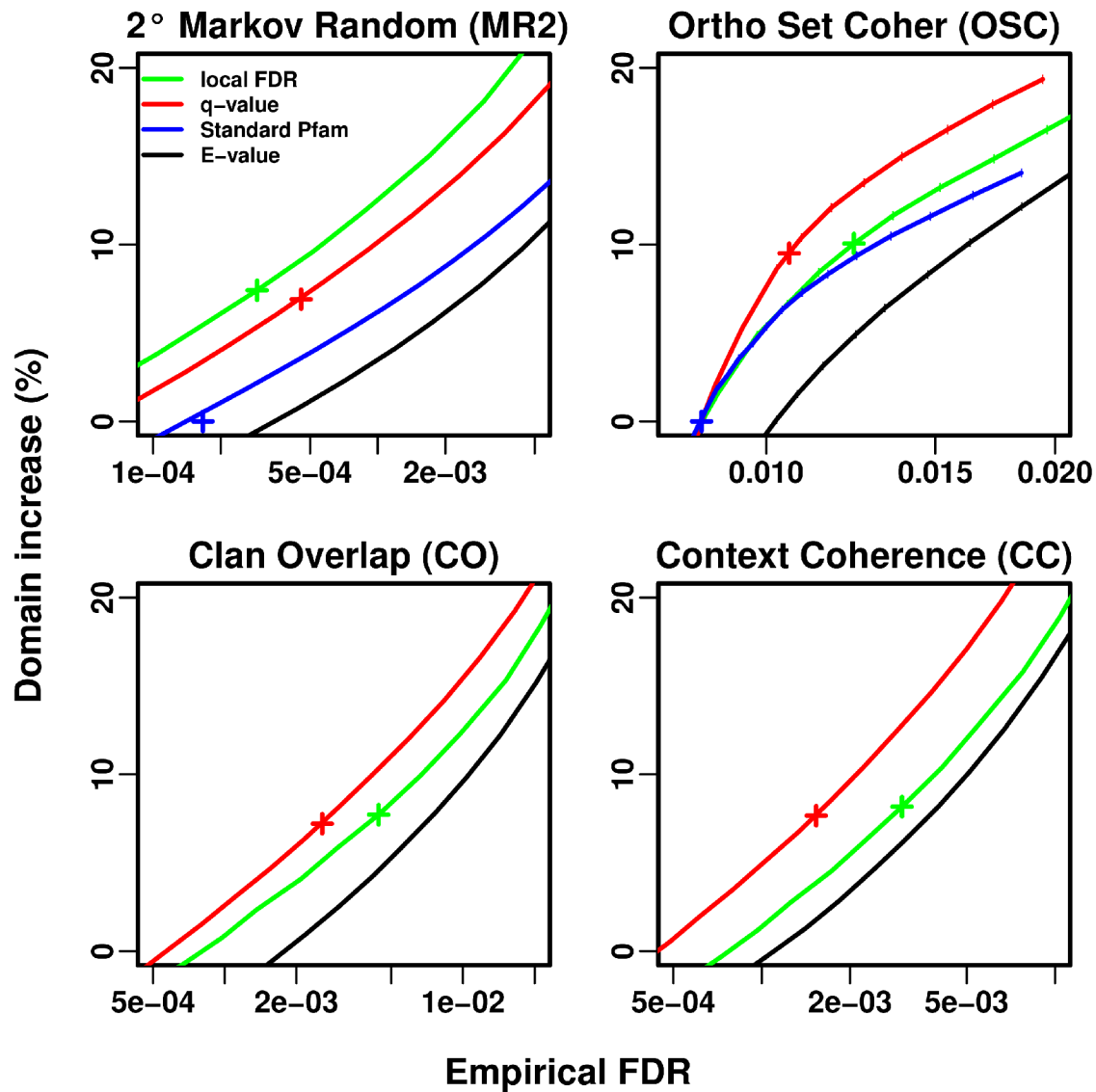
Clan Overlap



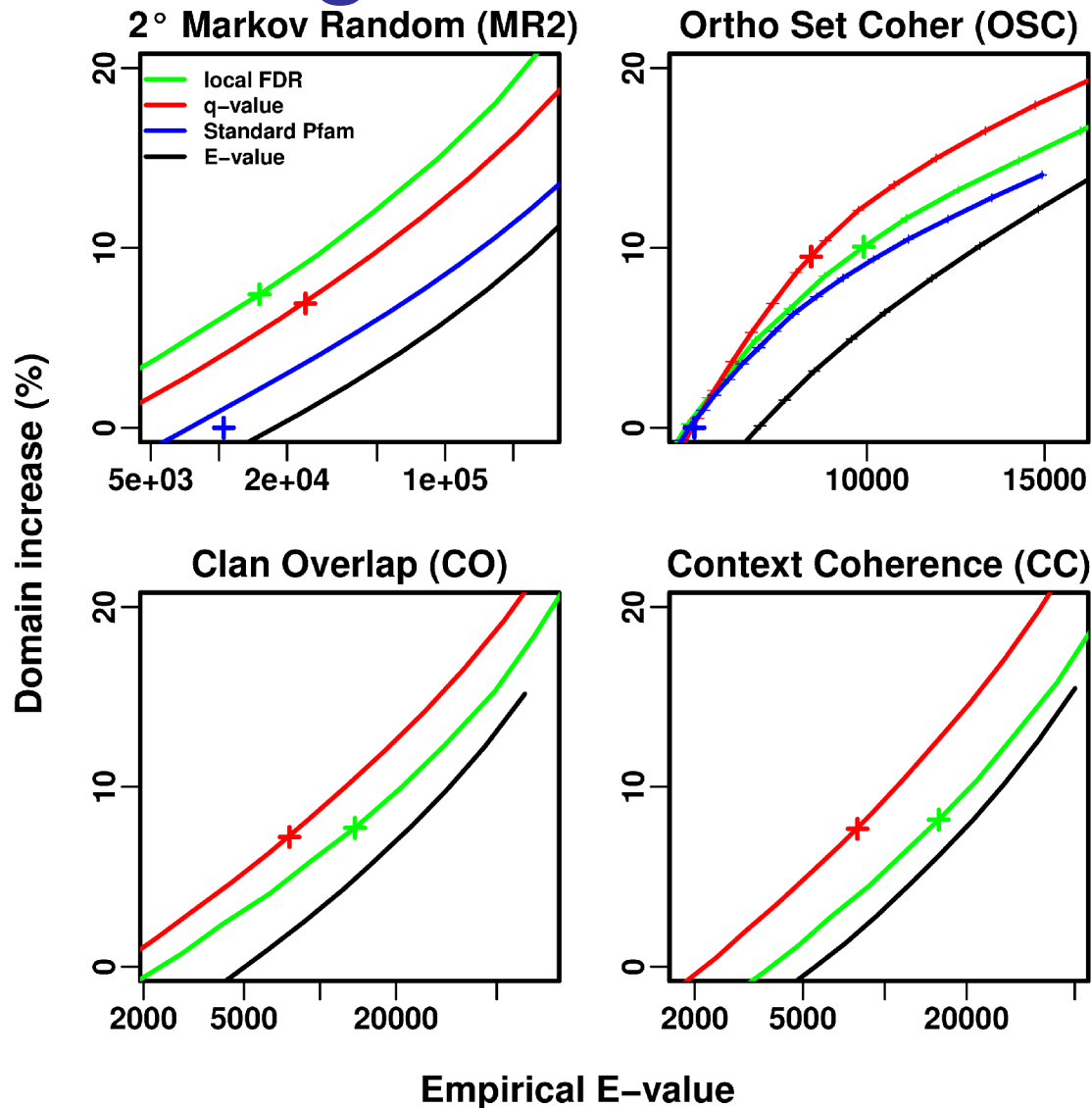
Context Coherence



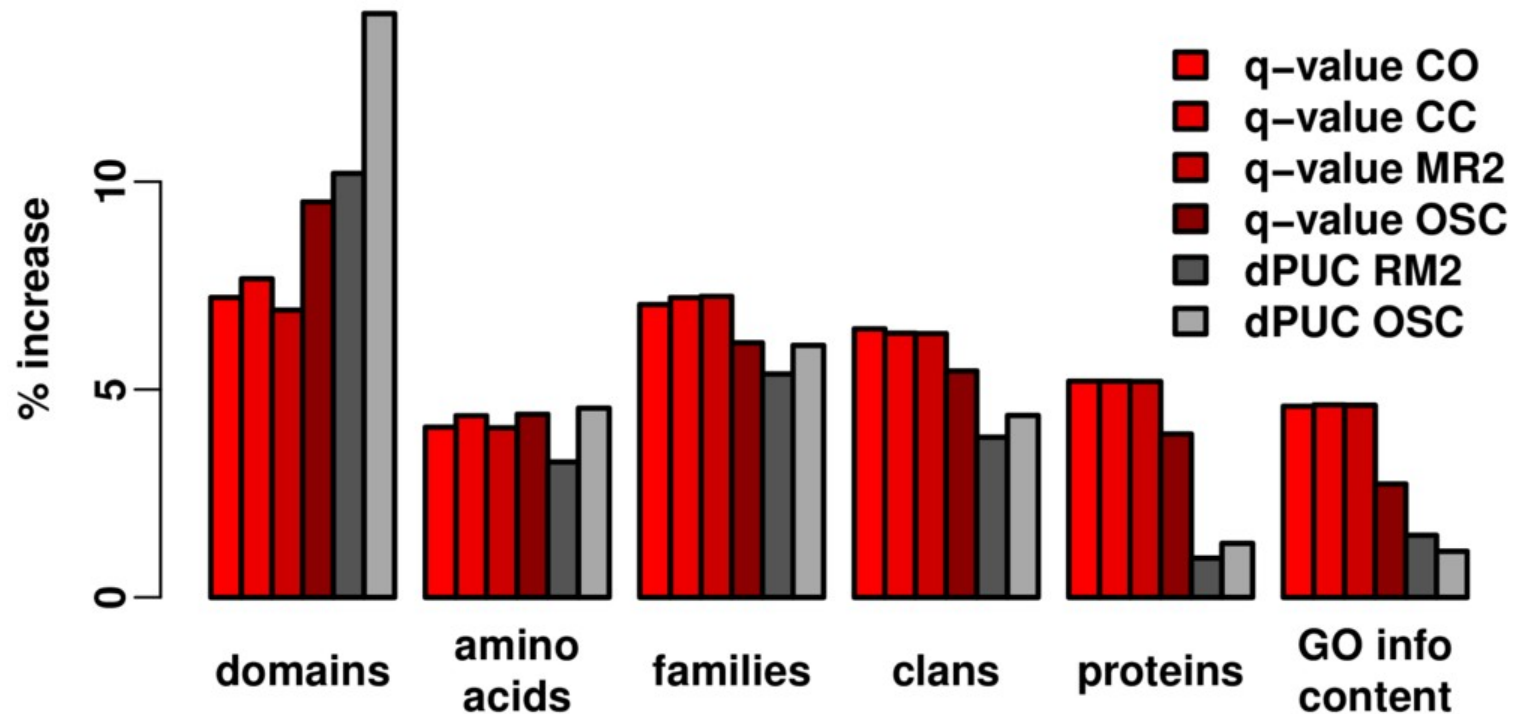
$$q > \text{local FDR}$$



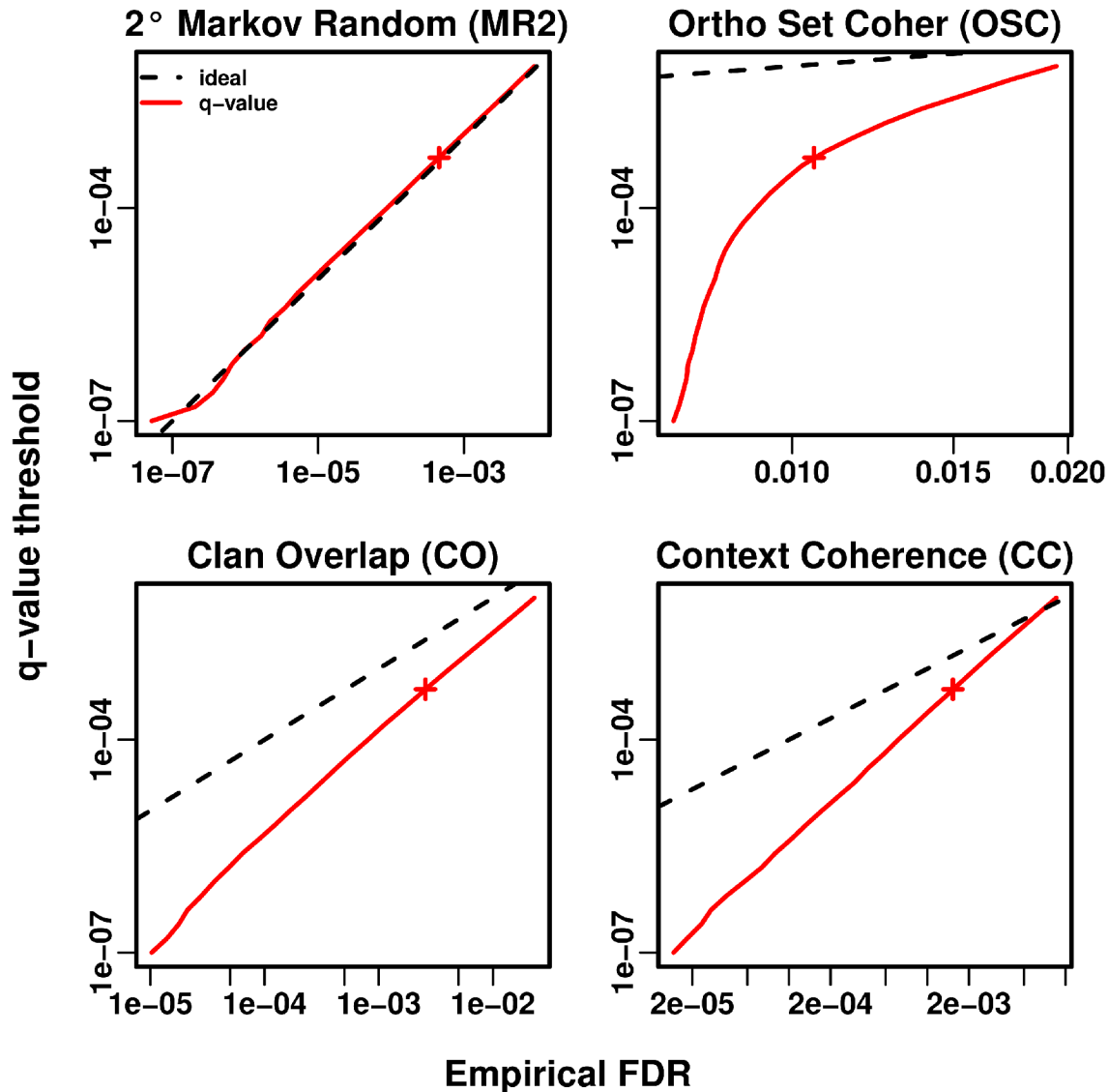
$q > \text{local FDR}$, E worst at
constraining combined E -value!



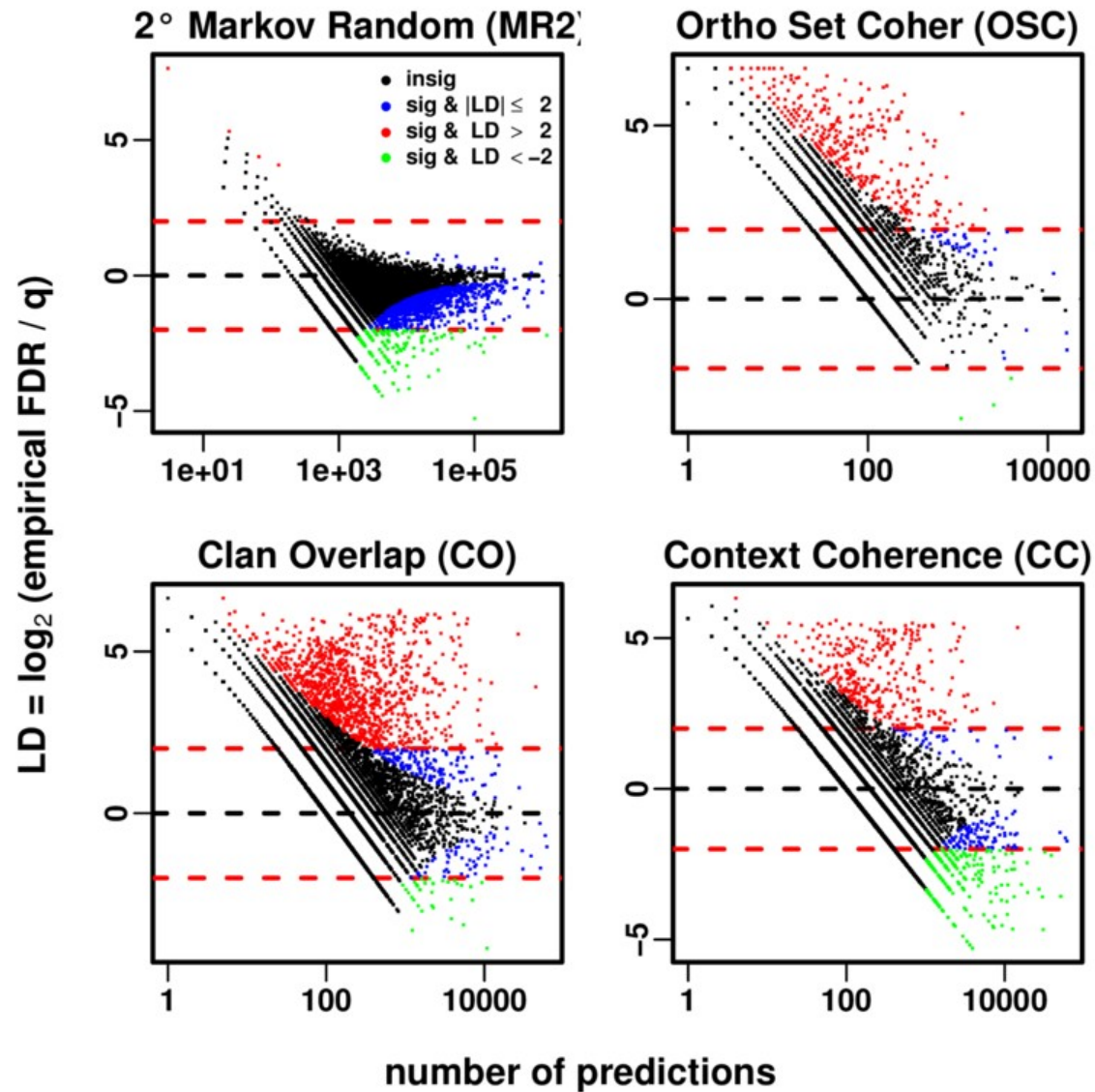
Improving more than just domain counts



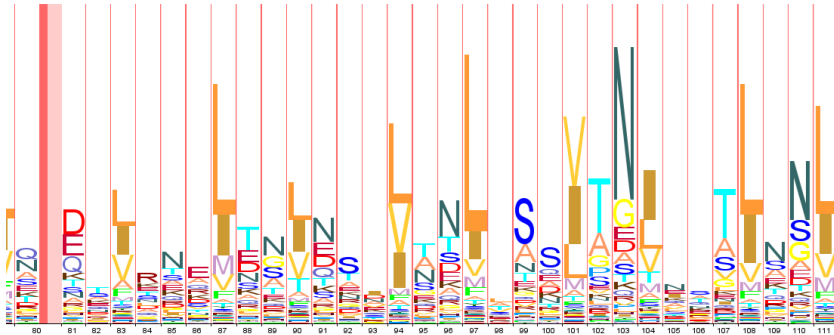
The q -values underestimate empirical $FDRs$



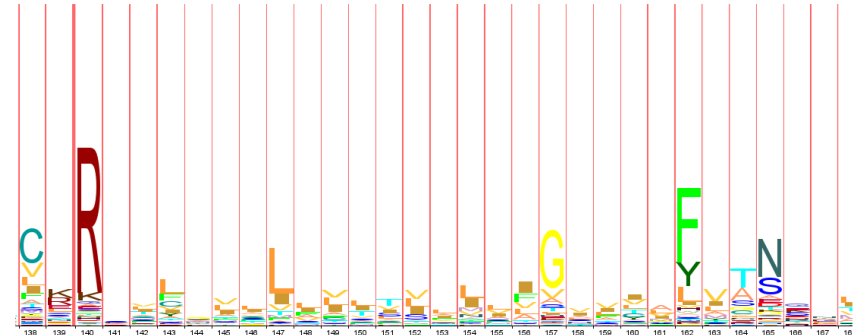
Per family analysis of noise



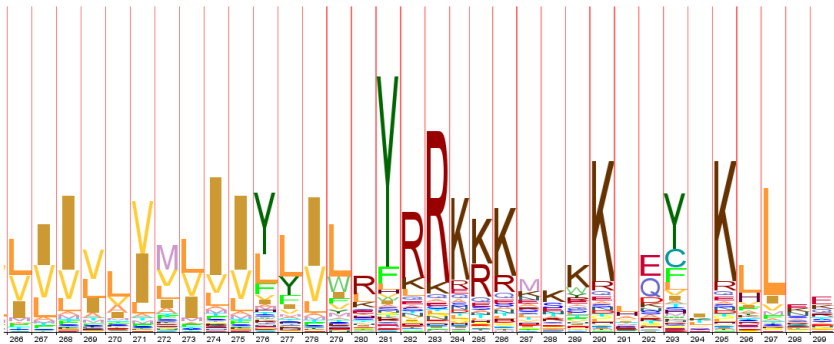
Classes of noisy domains



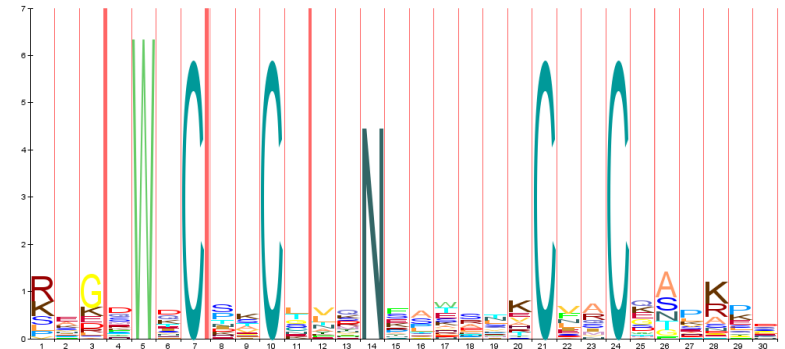
Coiled coils: 403 (3%)



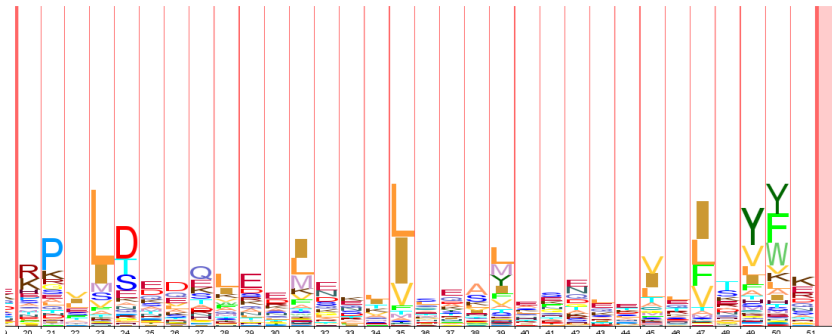
Transmembranes: 1466 (12%)



Low complexity: 1987 (16%)



Cysteine-rich: 587 (5%)

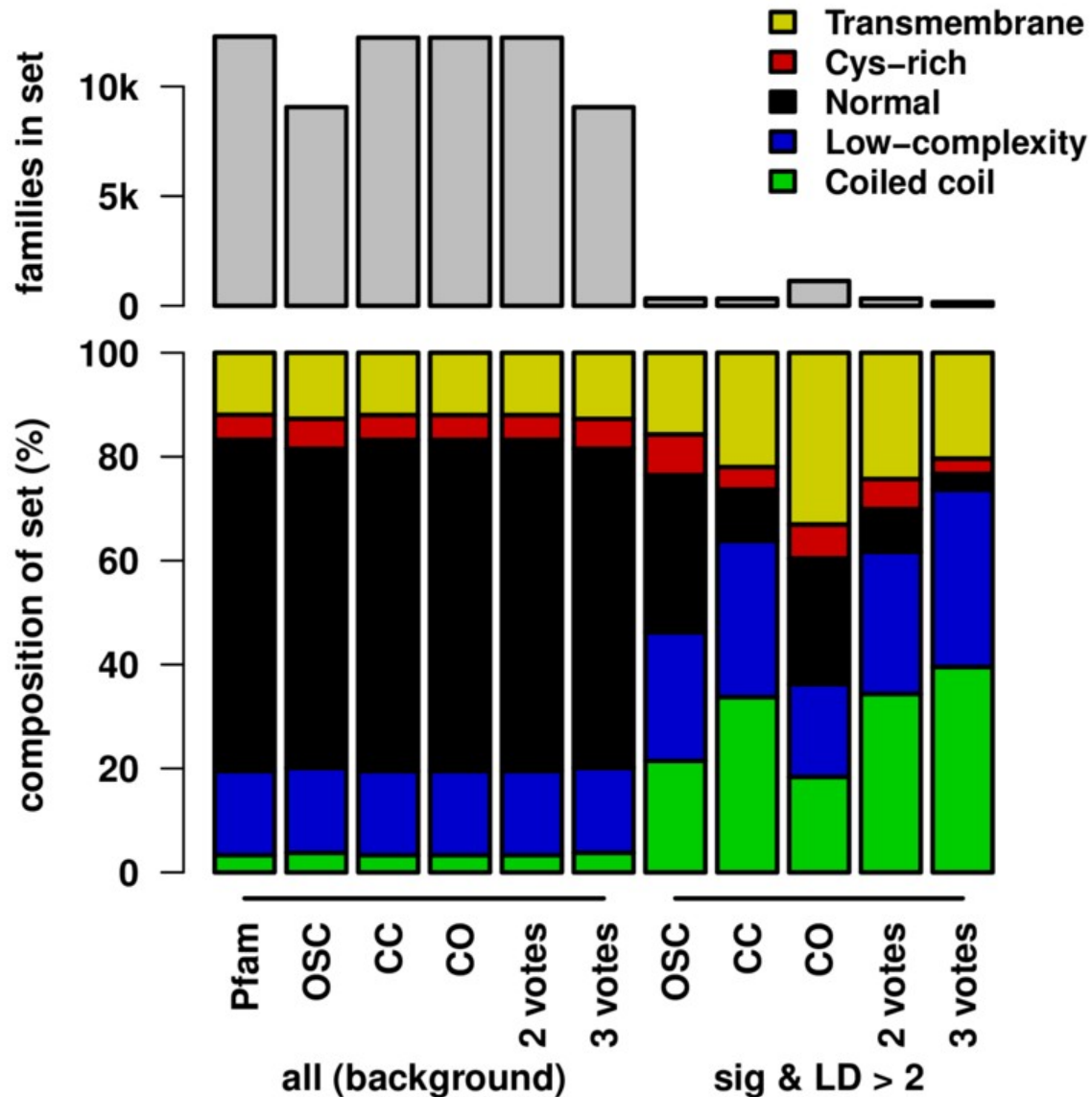


**Normal:
7830 (64%)**

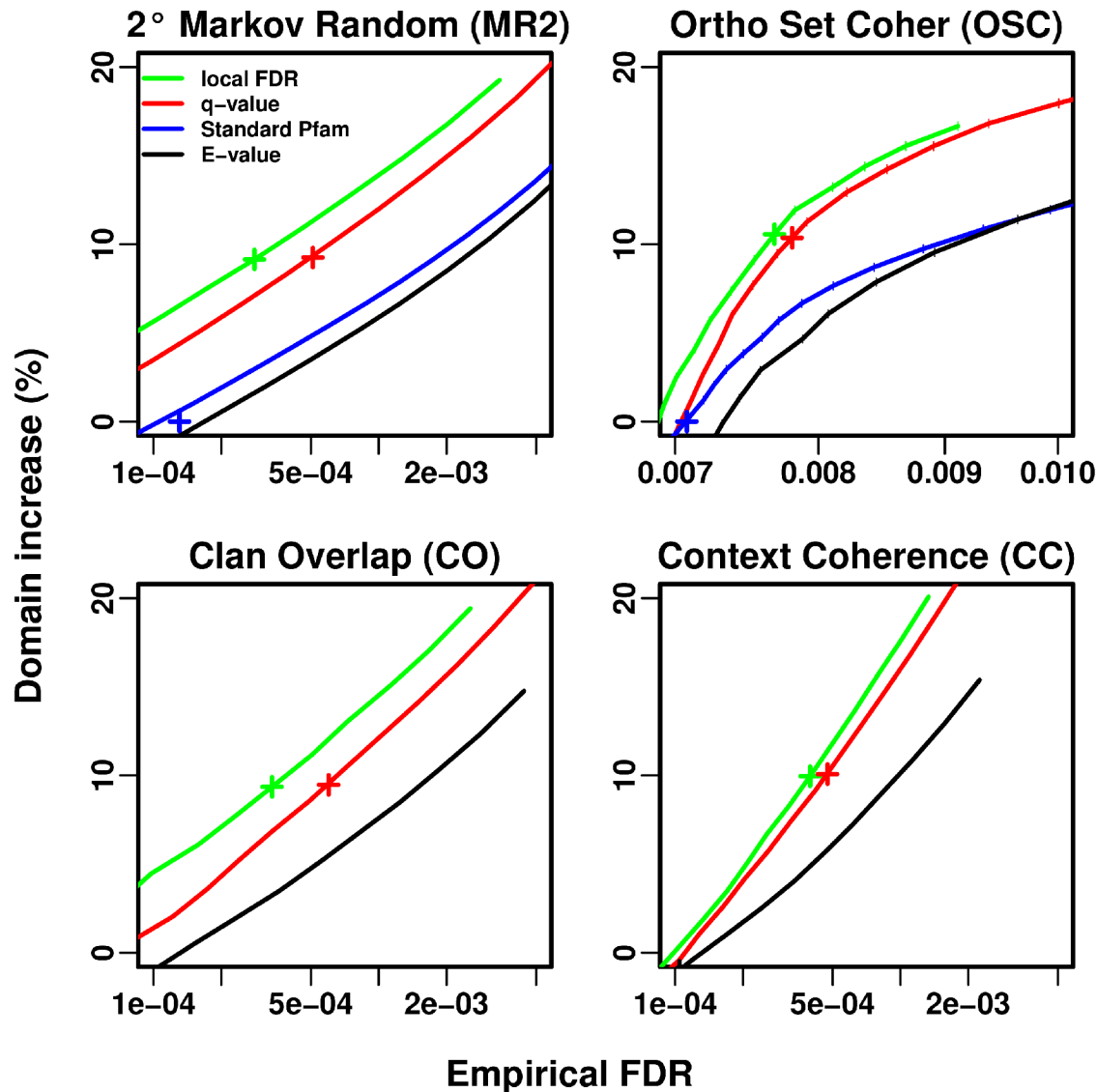
Similar but non-homologous families
(different clans)

Rackham *et al.* JMB 403, 480–493 (2010).
Wong, *et al.* PLoS Comput Biol 6, e1000867 (2010).

Classes enriched in noisy domains



Local FDR > q in families with correct stats



Conclusions

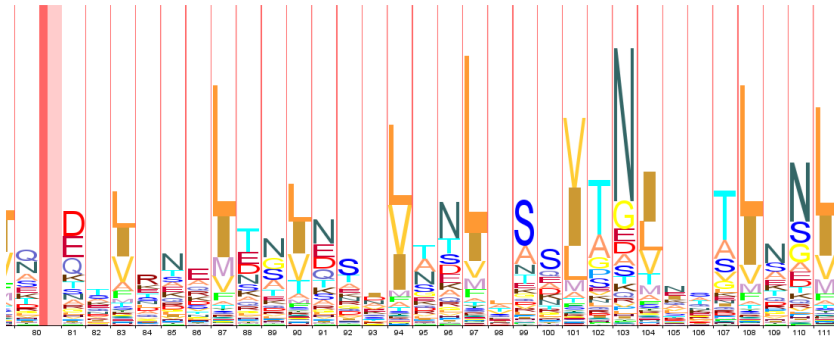
The q -value and *local FDR* are better for domains

- And likely better for regular sequence database searches, iterated searches, orthology prediction
- E -values do not control posterior error probability

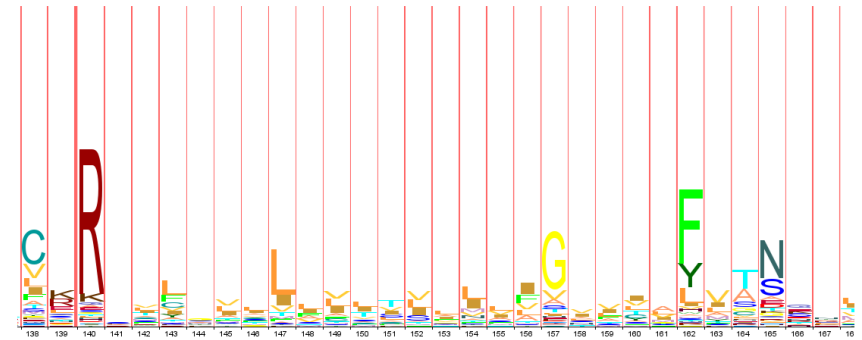
Presented novel empirical null models

- Needed to verify theory is correct
- Uses common-sense biological information and real, full protein sequences
- Structural benchmarks (i.e. SCOP) are limited to well-studied, single domains from model organisms, and exclude coiled coils and transmembrane domains

Noisy domains



Coiled coils



Transmembranes

These remain large problems in all sequence analysis.

Solutions?

- Cannot prevent or ignore these queries
- Masking removes too much information
- Benchmarks not powerful enough to give better thresholds
- Can we properly handle these common, correlated patterns that do not imply homology?

Rackham *et al.* JMB 403, 480–493 (2010).

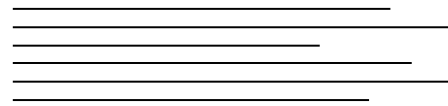
Wong *et al.* PLoS Comput Biol 6, e1000867 (2010).

Outline of results

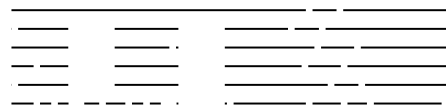
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Identification of novel domain families in *Plasmodium falciparum*

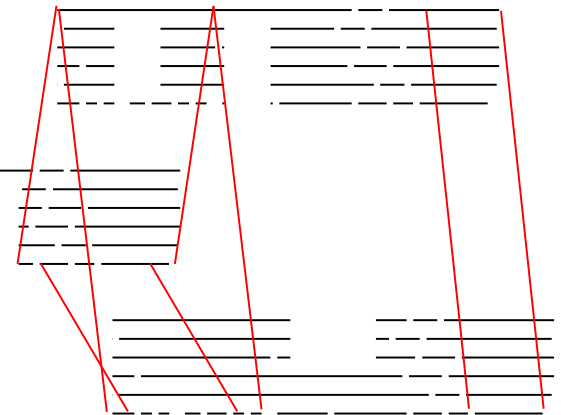
15 Apicomplexan proteomes



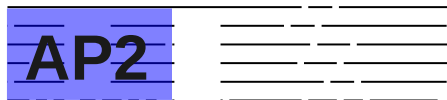
Orthogroup alignments and HMMs



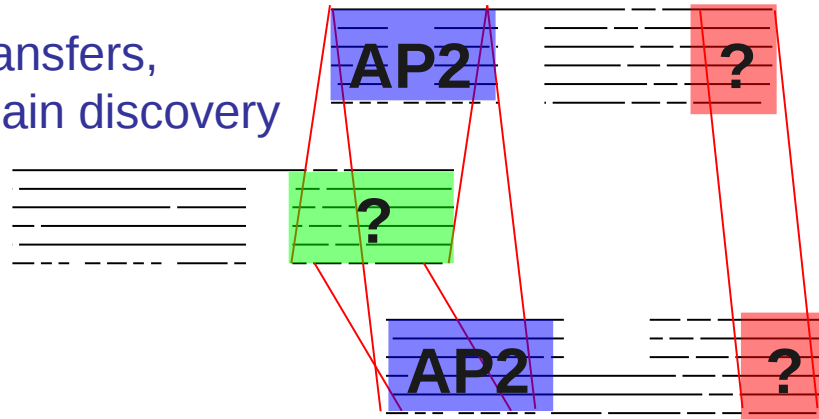
HMM-HMM comparison between orthogroup HMMs



Fancy orthogroup Pfam domain assignments







Domain transfers, novel domain discovery



Pfam domains:
+33% doms, +14% prots
Pfam domain transfer
New non-Pfam domains:
+7.2% doms, +8.7% prots

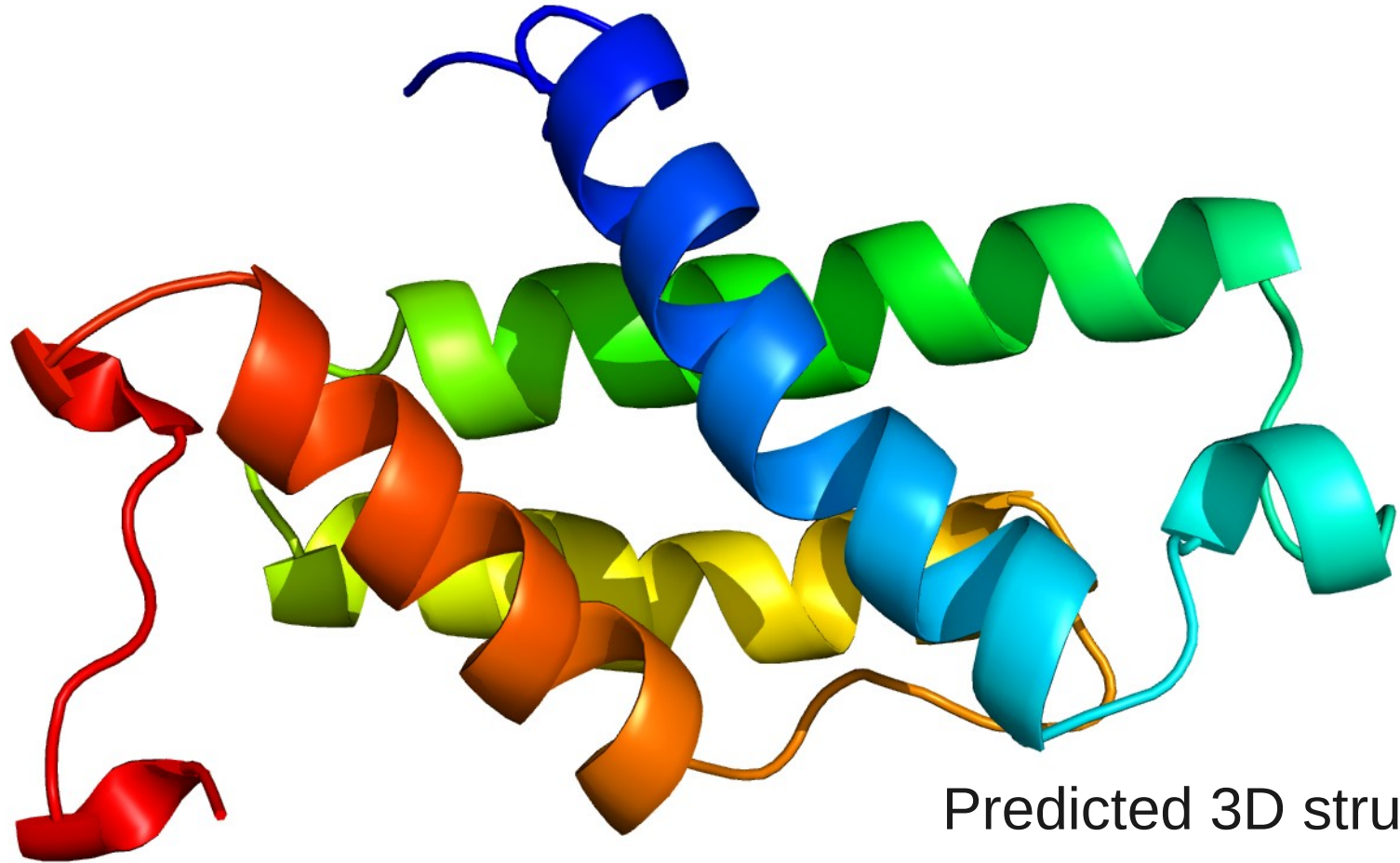
Prediction and experimental validation of novel AP2 domains

Protein	Domain	<i>q</i> -value	BILD	Context	Significant motifs
PF13_0026	Dx1	0.0031	*	*	NA
PF13_0114 [§]	Dx1	0.29		*	NA
	Dx2	0.049	*		NA
MAL7P1.167 [§]	Dx1	0.31	*		NA
	Dx2	0.4		*	
	Dx1+Dx2	-	-	-	
	Dx3	<i>q</i> >0.61	*		NA
PF11_0404	Dx1	0.53		*	NA
	Dx2	0.47		*	NA
	Dx3	<i>q</i> >0.61	*		
PF13_0267	Dx1	0.5	*	*	
PF07_0126	Dx1	<i>q</i> >0.61	*		NA
PF13_0235	Dx1	<i>q</i> >0.61	*		NA

[§] = Proteins without strong Pfam predictions

With Ariel Schieler & Sebastian Nasamu

Computational analysis of a novel domain that co-occurs with AP2s



Predicted 3D structure

Follow up experiments by Joana Santos,
Mehak Mumtaz & Marcos Lanio



PRINCETON
UNIVERSITY



Mona Singh, Computer Science

- Eric Banks
- Tony Capra
- Jesse Farnham
- Dario Gherzi
- Peng Jiang
- Zia Khan
- Elena Nabieva
- Shilpa Nadimpalli
- Anton Persikov
- Yuri Pritykin
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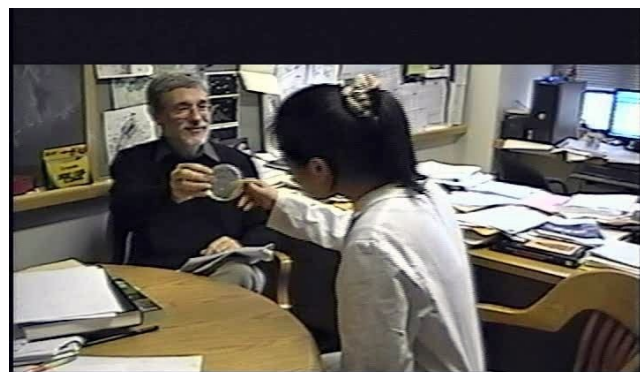
Singh lab, 2009



Llinás lab, 2011



Molbio skits!





Big Dumb Rock Show

