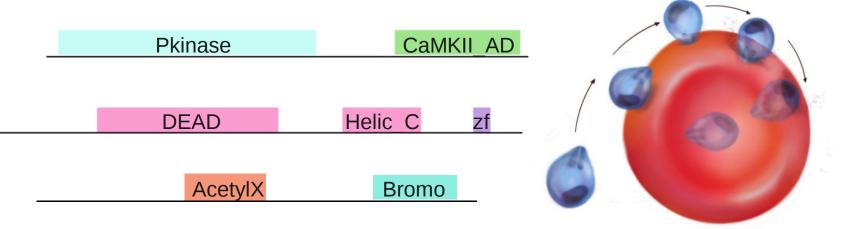
Forget the E-value: family-based q-values for protein domain prediction, and empirical error detection

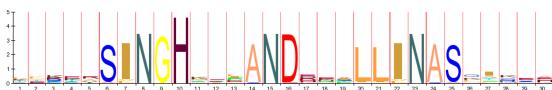


Alejandro Ochoa

Molecular Biology, Princeton University

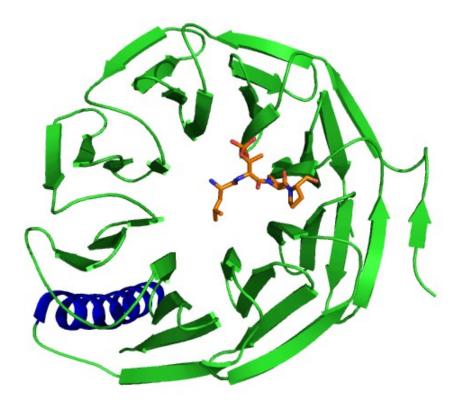
Yun Song's group, UC Berkeley, 2013-04-17





Labs

Protein domains



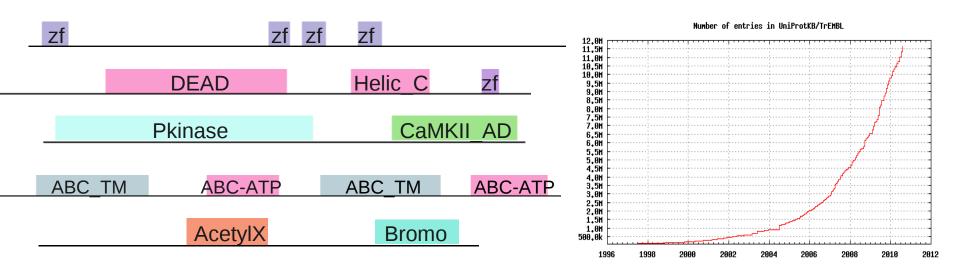
Structure Evolution Function

Sequence-based domain prediction:

F-box

WD4(WD40 WD40 WD40 WD40 WD40

Why predict domains?



For new sequences, before experiments start...

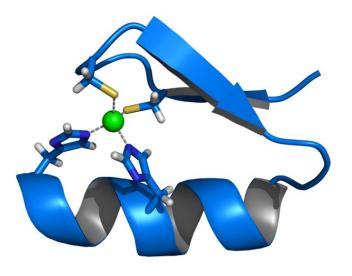
Domains may imply functions

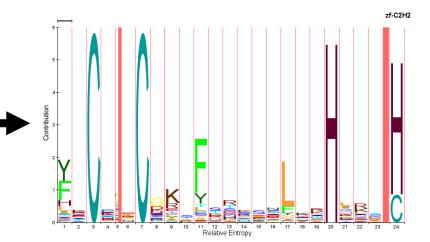
Experimental alternatives are unfeasible as protein databases grow exponentially

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 TRA1 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 TYY1 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 TRA1 CAEEL/276-300 ESCA DROME/370-392

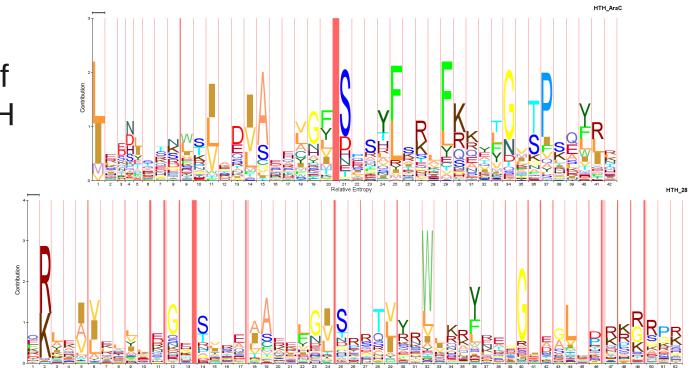
YACQVCHKSFSRMSLLNKHSSSNC
YQCKSCSRTFSRMSLLHKHEETGC
YQCQACARTFSRMSLLHKHQESGC
YSCTSCSKTFSRMSLLTKHSEGGC
HVCGKCYKTFRRLMSLKKHLEFC
LHCRRCRTQFSRRSKLHIHQKLRC
FMCADCGRCFSVSSSLKYHQRIC
IKCKDCGQMFSTTSSLNKHRRFC
YSCADCGKHFSEKMYLQFHQKNPSEC
YRCEDCDQLFESKAELADHQKFPC
YKCNQCGIIFSQNSPFIVHQIAH
YKCEECGKAFKQLSTLTTHKIIC
IKCEECGKAFSTRSTYYRHQKNH
YKCEF.ADCEKAFSNASDRAKHQNRTH
YTCSTCGKTYRQTSTLAMHKRSAH
YRCSQCGKAFRRTSDLSSHRRTQC
YECRHCGKKYRWKSTLRRHENVEC
YECNKCGKFFRYCFTLNRHQRVH
FVCVHCGKGFRDNYKLSLHLRIH
YVCYF.ADCGQQFRKHNQLKIHQYIH
YVCTECGTSFRVRPQLRIHLRTH
FVCGDCGQGFVRSARLEEHRRVH
YKCDKCGKGFTRSSSLLVHHSVH
YKCGECGKTFSRSTHLTQHQRVH
FACDICGRKFARSDERKRHTKIH
YACKICGKDFTRSYHLKRHOKYS.SC
YTCPYCDKRFTQRSALTVHTTKLH
YTCSYCGKSFTQSNTLKQHTRIH
YTCEICDGKFSDSNQLKSHMLVH
YVCPF.DGCNKKFAQSTNLKSHILTH
YTCTQCNKQFSHSAQLRAHISTH
YPCPFCFKEFTRKDNMTAHVKIIH
YTCPE.PHCGRGFTSATNYKNHVRIH
YMCQVCLTLFGHTYNLFMHWRTSC
YQCDICGQKFVQKINLTHHARIH
YFCHICGTVFIEQDNLFKHWRLH
NKCEY.PGCGKEYSRLENLKTHRRTH
CKCNLCGKAFSRPWLLQGHIRTH





Databases of Domain Families

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



Two members of Pfam "Clan" HTH

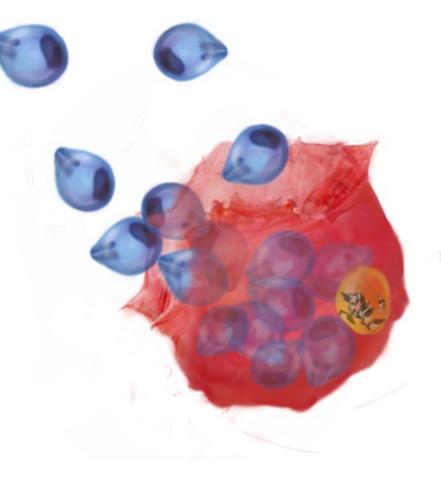
Plasmodium falciparum

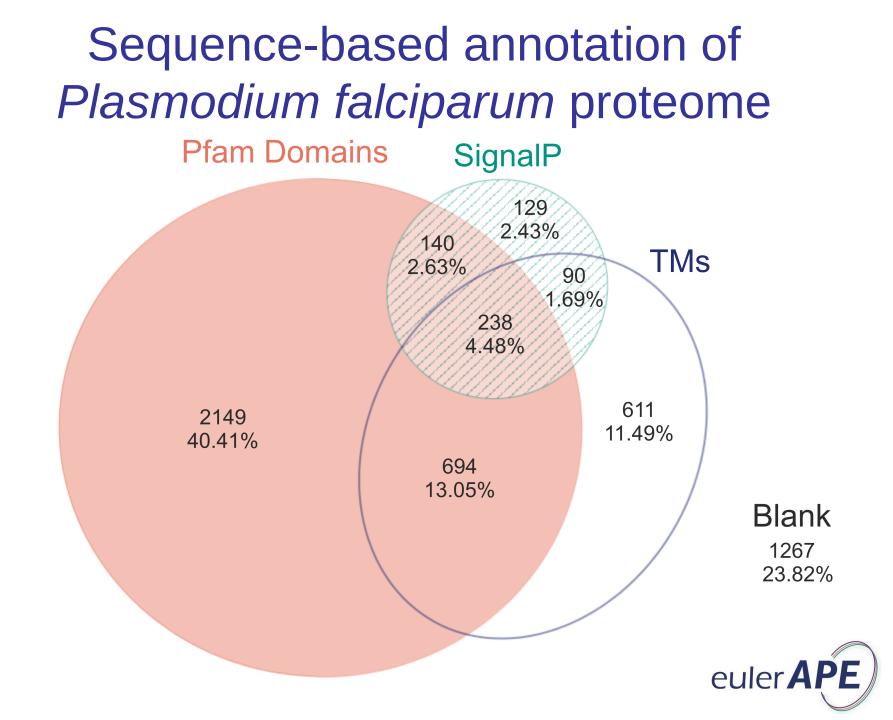
Malaria Information challenges

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

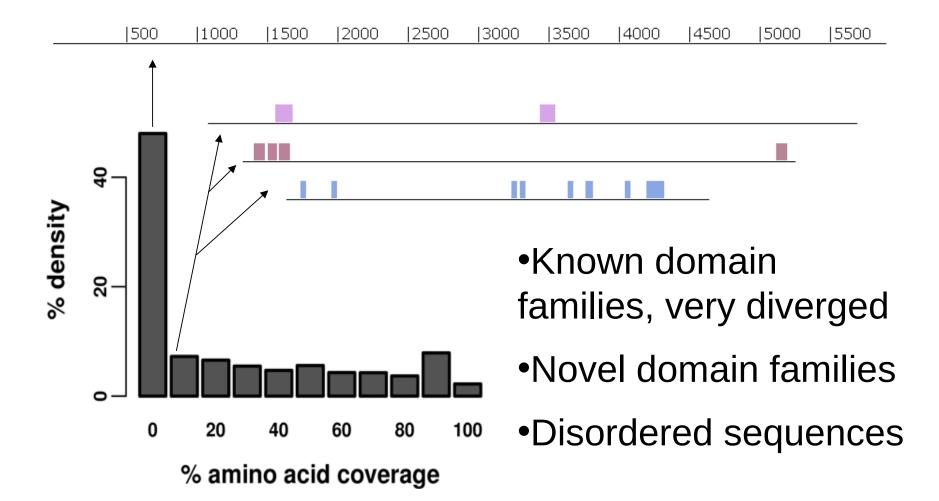
Annotation

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





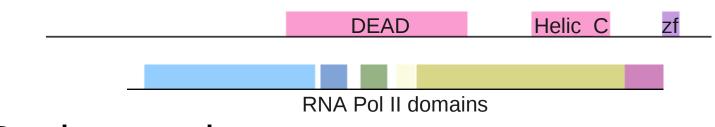
Poor domain coverage of *Plasmodium falciparum*



Outline of results

- Domain prediction using context
 - Application to malaria parasite
- Optimal *FDR* control for domains
 - Family-based q-values, adapted to domain problem
 - Local FDR optimizes problem, but q-values are more robust
 - Large problems with coiled coils, transmembrane domains
 - Implications for sequence models

Domain Prediction Using Context: dPUC

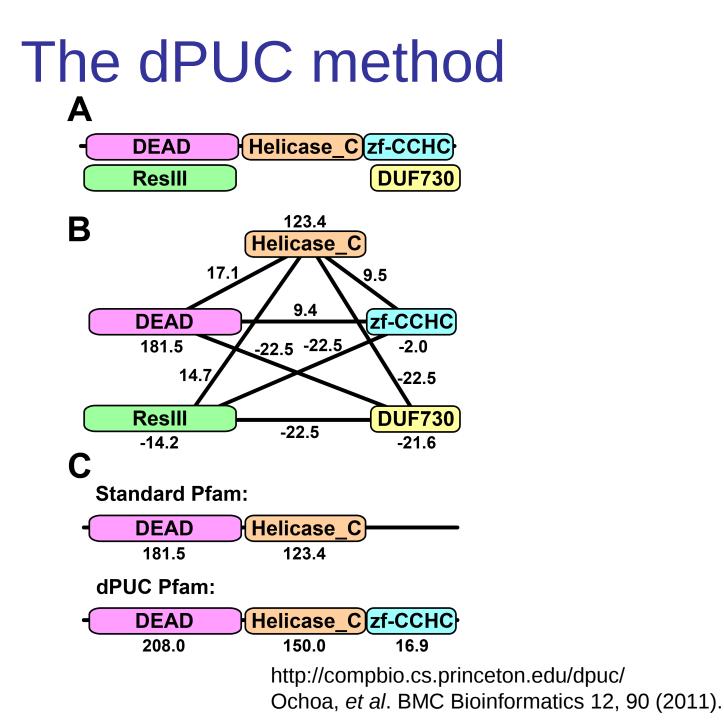


Background

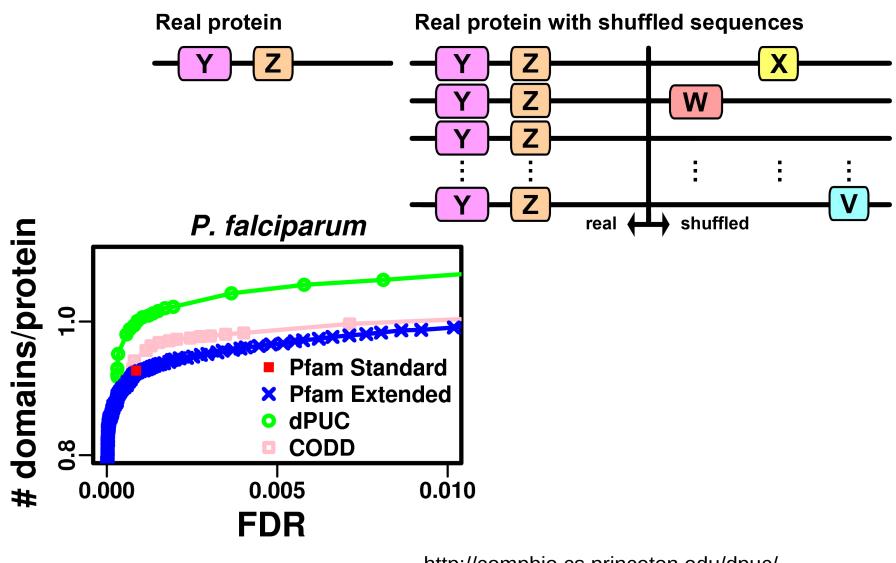
- Domains co-occur in limited combinations
- Domains are scored independently of each other

Idea

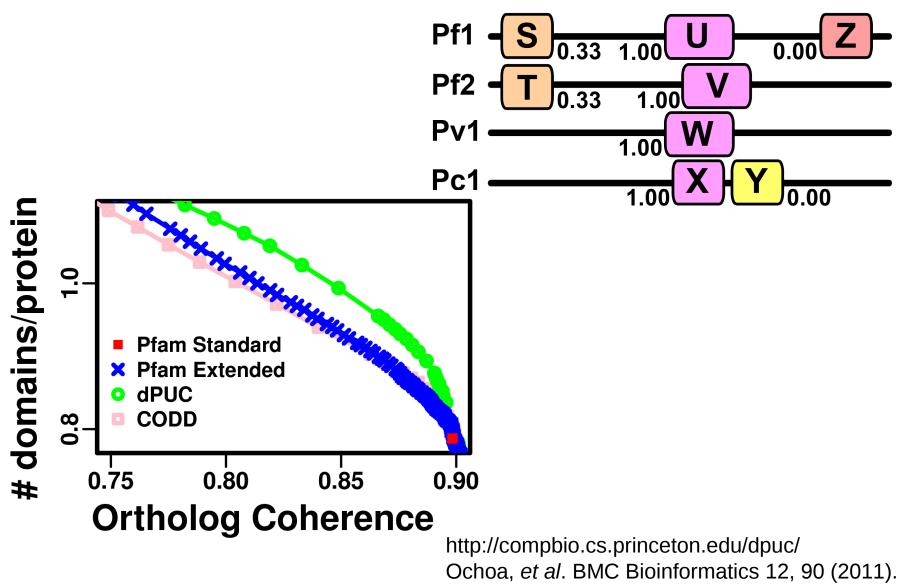
- Score domains in combination
- Context + Sequence evidence



Improved signal to noise



Improved ortholog coherence on *Plasmodium* species



New predictions

Phosphatase -> RNA lariat debranching enzyme

P. falciparum

Std Pfam dPUC



S. cerevisiae

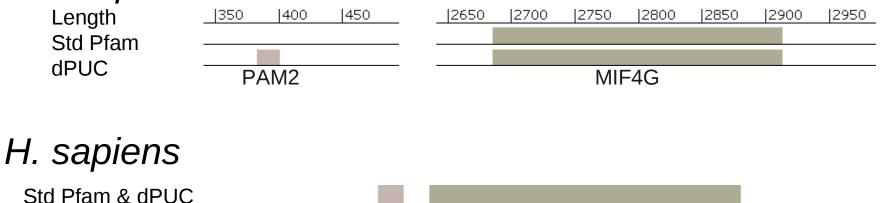
Std Pfam & dPUC



New predictions

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

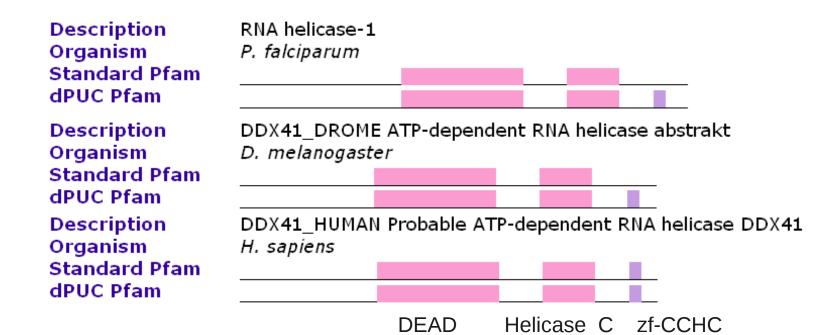
P. falciparum



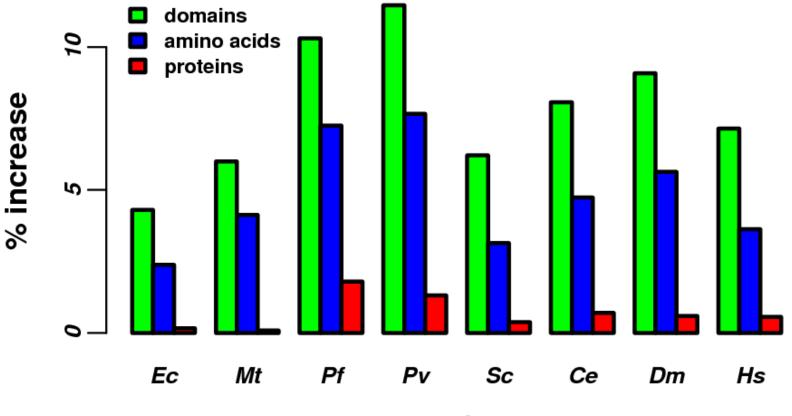
Interested malaria curator: Hagai Ginsburg, Hebrew U of Jerusalem

New predictions

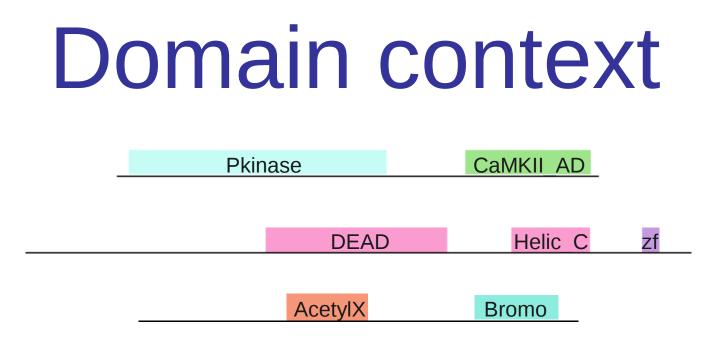
RNA helicase -> mRNA sequestration



dPUC increases coverage

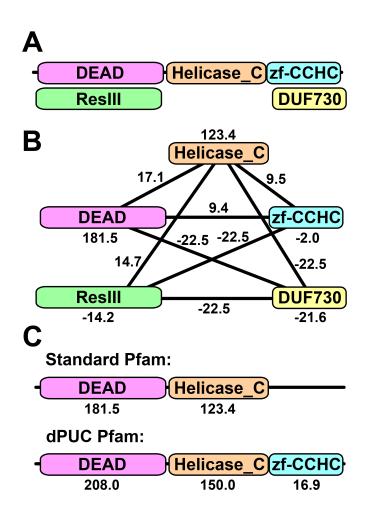


organisms



Complements sequence evidence Improves domain predictions Works best on diverged organisms

The dPUC method: problems



Domain scores are normalized by curated thresholds

Why do *E*-values perform worse?

Outline of results

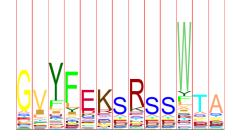
- Domain prediction using context
 - Application to malaria parasite
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The *E*-value: the statistic of sequence analysis

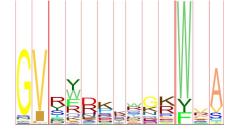
GVRQRKNSNKWVS PSI-BLAST (E 1997)

Query Type	Database Type	Software
GVYFEKSRSSWTA	GVRQRKNSNKWVS	FASTA (<i>Z</i> 1985) BLAST (<i>E</i> 1990)

First genomes 1976: virus 1995: bacteria 1996: yeast 1998: worm 2000: human



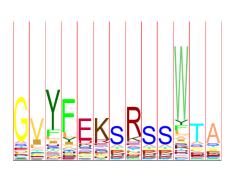
GVYFEKSRSSWTA

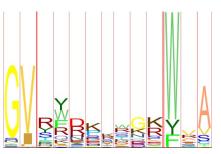


SAM (*Z* 1994, ~*E* 2005) HMMer (bit 1995, ~*E* 1998, *E* 2008) Pfam DB (curated 1997) IMPALA, RPS-BLAST (*E* 1999)

PRC (bit 2002, ~*E* 2004) HHsearch (~*E*,*P* 2004) CORAL (*E* 2009)

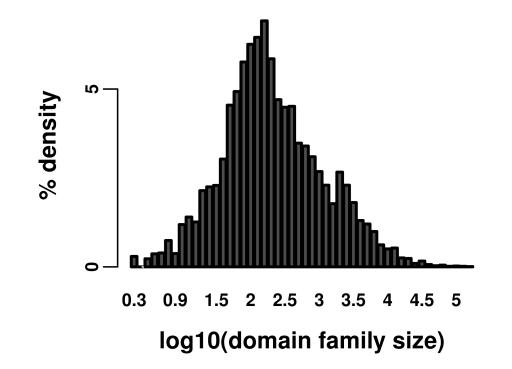
FDR: BH 1995, *q* 2002



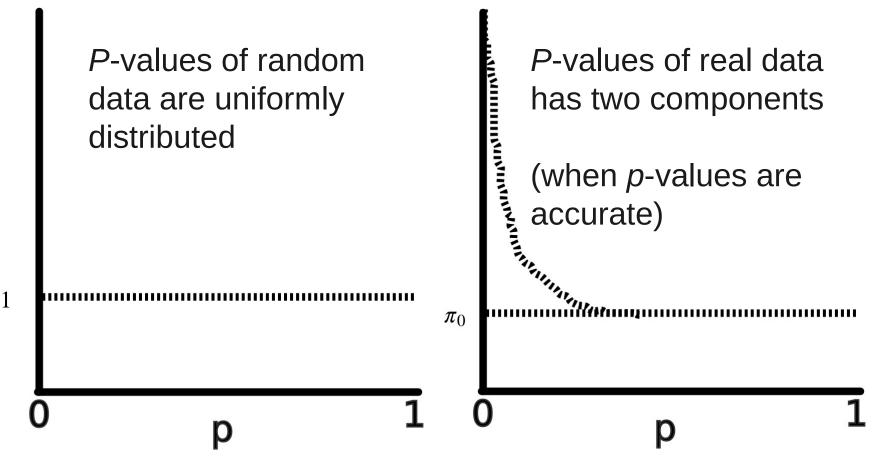


FDR = E/n

FDR = average posterior error probability

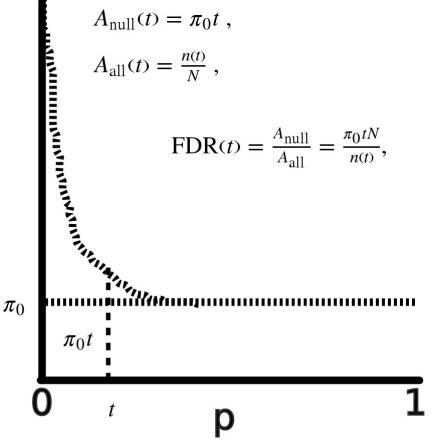


Computing *q*-values



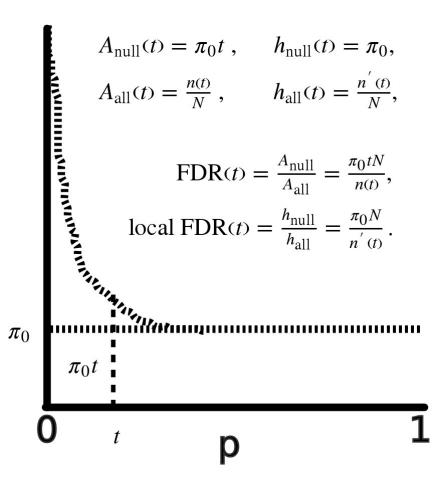
Step 1: estimate π_0 (proportion of data that is false)

Computing *q*-values



Step 2: Directly estimate *FDR(t*) for all thresholds t -N = # tests -n(t) = # sig tests Step 3: Ensure monotonicity $q(p) = \min_{t:p \le t} FDR(t)$

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

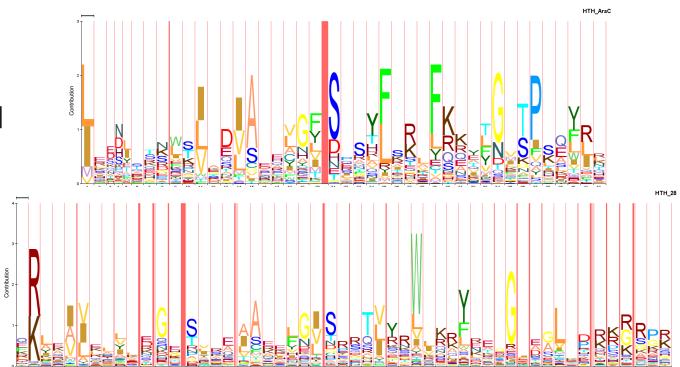
$$Q \ge \frac{\sum \pi_{0,i} \cdot t_i \cdot N_i}{\sum n_i(t_i)} = \frac{\sum FDR_i(t_i) \cdot n_i(t_i)}{\sum n_i(t_i)} .$$

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

Q-values for domains

Introduced domain-specific features

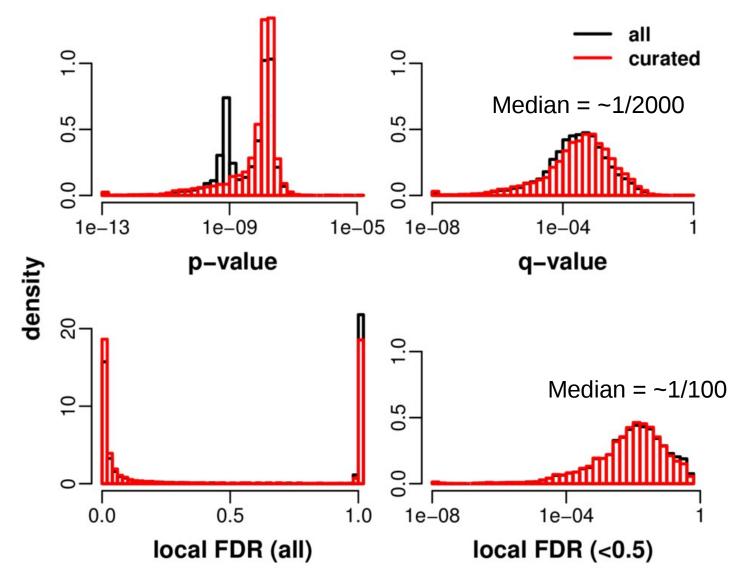
- Incomplete list of *p*-values (HMMER3, BLAST and descendants)
- Correction for domain overlaps (heuristic)



Two members of Pfam "Clan" HTH

FDR = E/n

Pfam curated thresholds



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

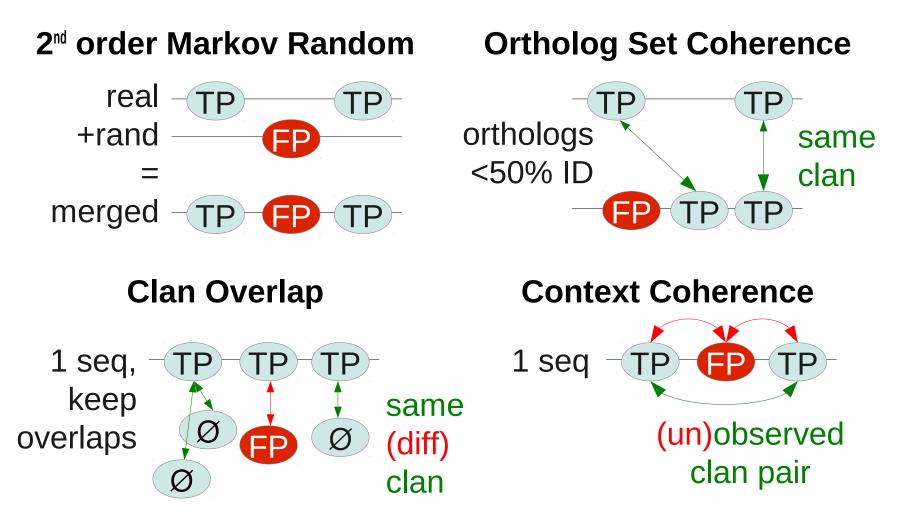
Inspired/adapted from S. Eddy (p.c., 2012).

Context Coherence

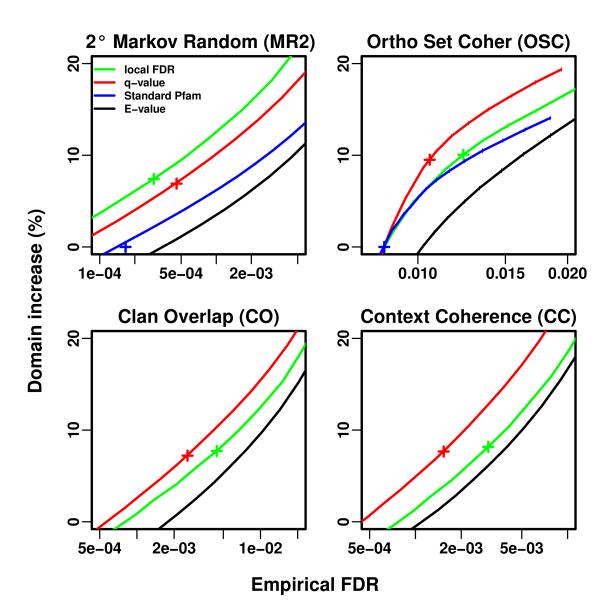
FP if domain doesn't co-occur with any stronger domains

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

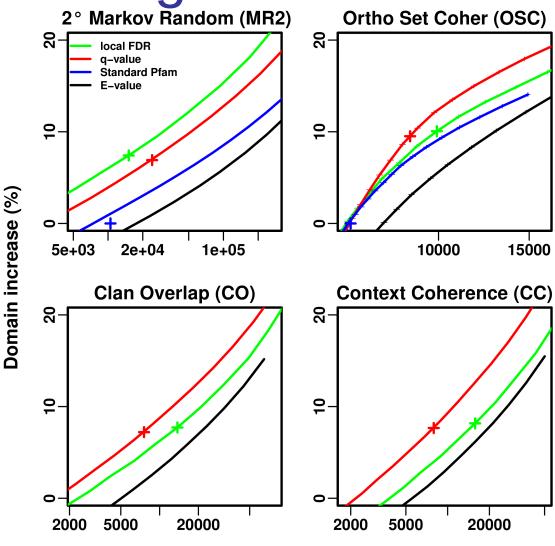
Empirical null models FDR = #FP / (#FP + #TP)



q > local FDR

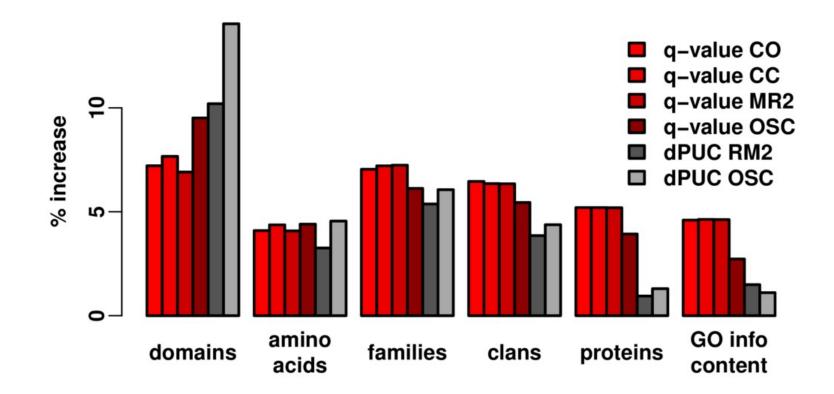


q > *local FDR, E* worst at constraining combined *E*-value!



Empirical E-value

Improving more than just domains



The *q*-values underestimate empirical FDRs 2° Markov Random (MR2) **Ortho Set Coher (OSC)** ideal q-value 1e-04 1e-04 q-value threshold e-07 1e-07 1e-05 1e-03 0.015 1e-07 0.010 0.020 Clan Overlap (CO) **Context Coherence (CC)** 1e-04 1e-04 6-**↓**-1e-05 1e-07

Empirical FDR

2e-05

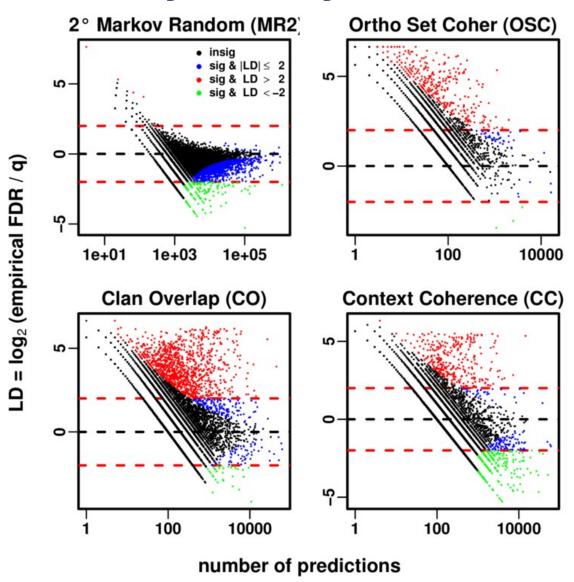
2e-04

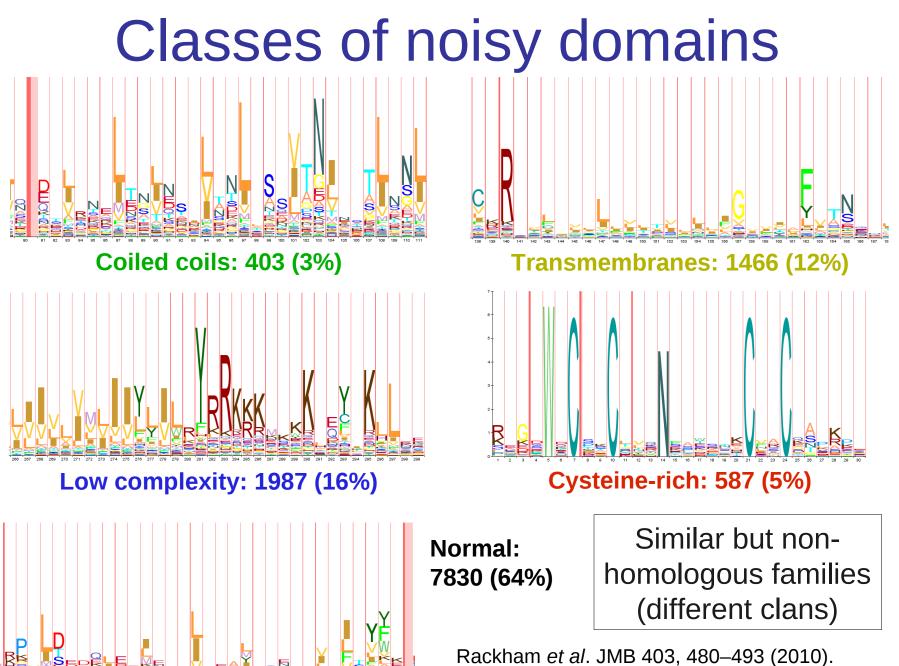
2e-03

1e-02

1e-04 1e-03

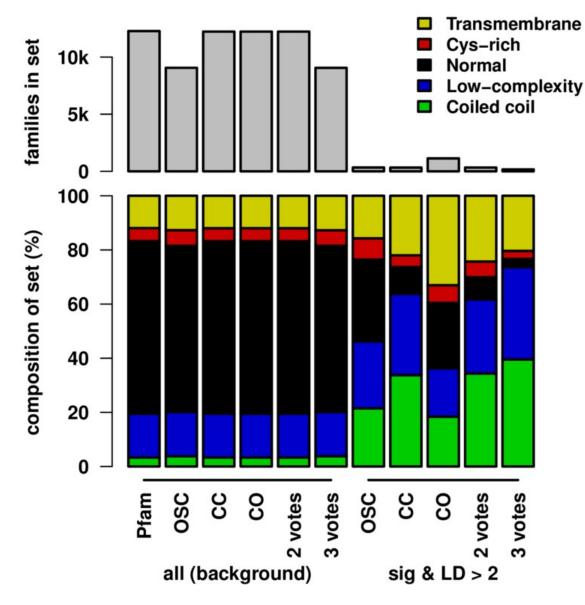
Per family analysis of noise



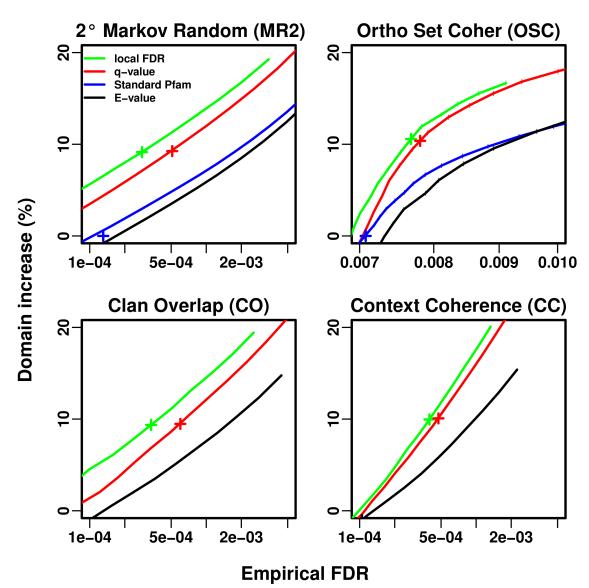


Wong, et al. PLoS Comput Biol 6, e1000867 (2010).

Classes enriched in noisy domains



Local FDR > q in families with correct stats

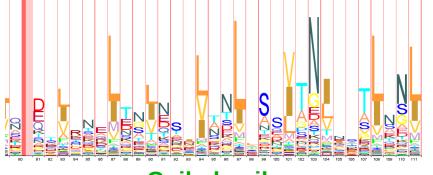


Conclusions

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 wellstudied, 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).





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- Peng Jiang
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