

# Molecular grammars of intrinsically disordered regions that span the human proteome

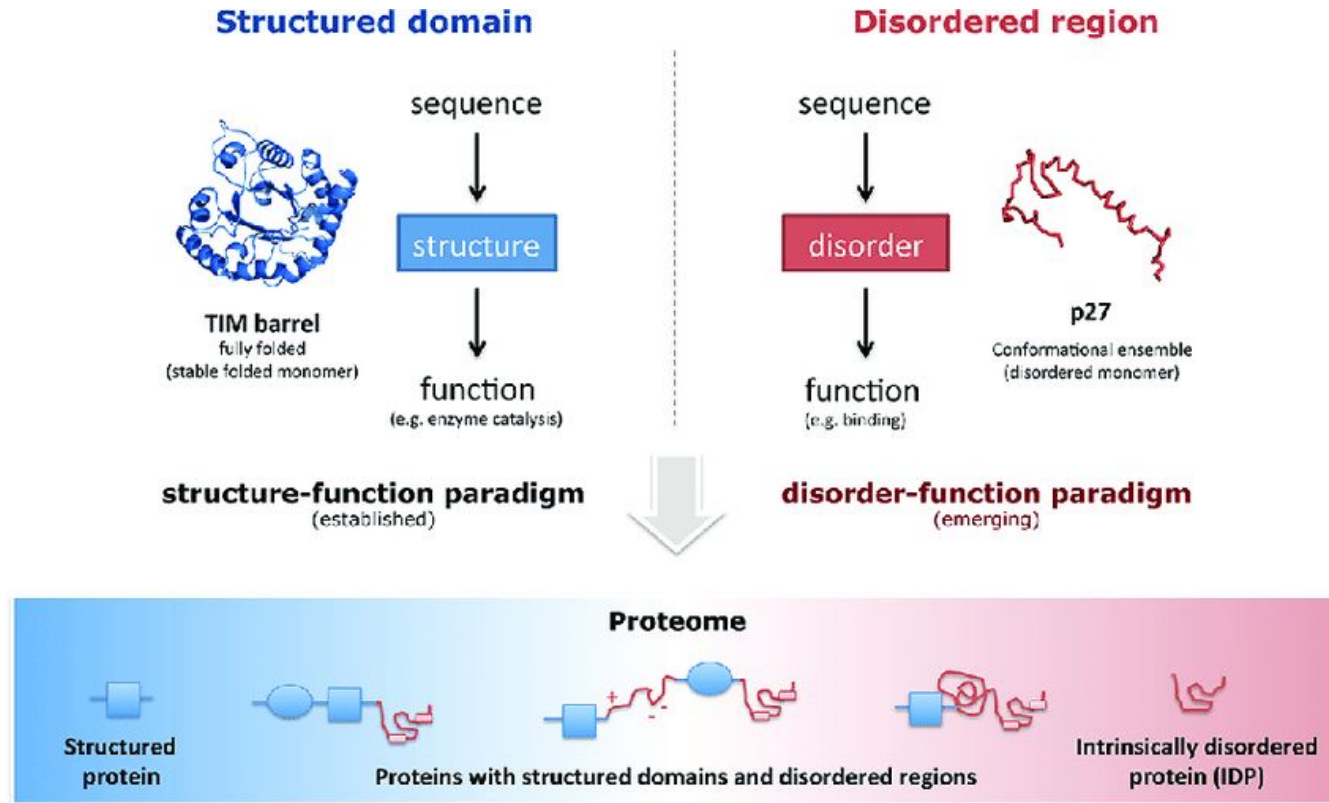
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LifeLU reading group

presented by Özdeniz Dolu

10.04.2025

# Intrinsically Disordered Region



# Intrinsically Disordered Region

- In human proteome, more than 50% of proteins contain at least one IDR
- IDRs show conformational heterogeneity and poor sequence conservation

Authors attempt to associate molecular grammars with IDRs:

- “IDR-specific molecular grammars are defined jointly by the non-random amino acid composition and the non-random patterning of distinct pairs of amino acid types with respect to one another.”

# Proposed Approach

**NARDINI+ Algorithm:** Combination of NARDINI algorithm and compositional analyses inspired by other work in the literature.

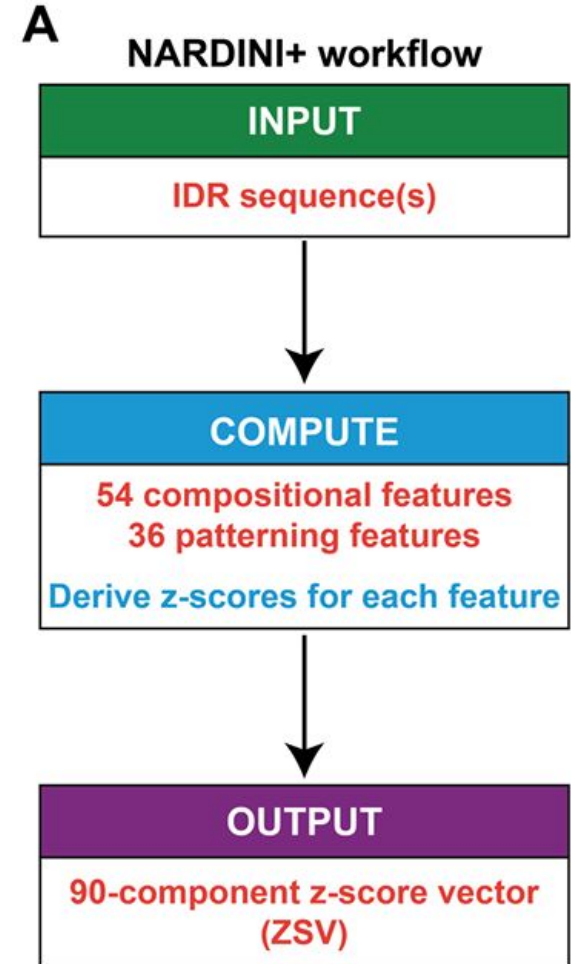
Through unsupervised learning methods and clustering, characterize IDRs in human proteome based on molecular grammars.

GIN: Grammars Inferred using NARDINI+

IDRome-spanning basis set, 30 clusters, unique fingerprint for each cluster

# NARDINI+

- 36 non-random binary patterns computed by NARDINI algorithm.
- Statistics on alphabet symbols used in the IDR yield 54 compositional features.
- Combined into a z-score vector.



# NARDINI+ (Compositional Features)

- The fraction of each of the 20 naturally occurring residues (20 features)
- Fractions of polar, aliphatic, aromatic, (Lys + Arg), (Asp + Glu), charged residue, residues that promote chain expansion, disorder promoting residues (8 features)
- “20 values that quantify the presence of specific residue or RG patches”
- Arg/Lys ratio, Glu/Asp ratio
- Net charge per residue (NCPR) (4 features)
- Apparent isoelectric point (pI), Kyte-Doolittle hydrophobicity

## B

CHD7 IDR8: VGSSEKKAADKAEGGPFKDGETLEGSDAEEESLDKTAESSLLEDEIAQGEELDSLGGDEIENNENDE

DDX10 IDR2: QKGGKRLEGT EHRQDNDTGNEEQEEEEDEEEMEEKLAKAKGSQAPSLPNTSEAKIKIEVPTQFLDRDEEEEDAD

# NARDINI+ (Compositional Features)

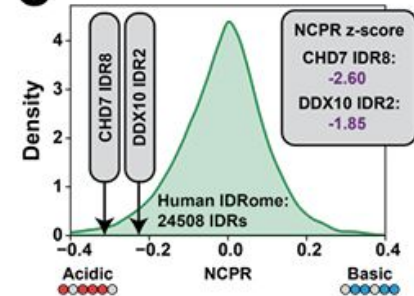
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**B**

CHD7 IDR8: VGSSEKAAADKAEGGPFDGETLEGSDAEEESLDKTAESSLLEDEIAQGEELDSLGGDEIENNENDE

DDX10 IDR2: QKGGKRLGTEHRQDNDTGNEEQEEEEDEEEMEEKLAKAKGSQAPSLPNTSEAQKIKEVPTQFLDRDEEEEDAD

**C**



# NARDINI+ (Binary Patterning Features)

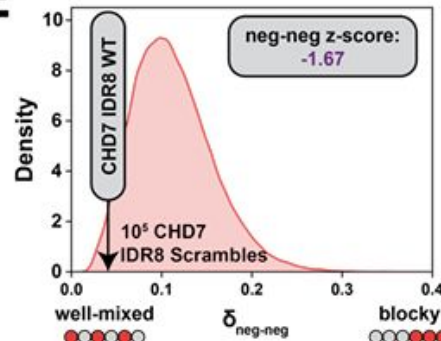
- 36 binary features computed by NARDINI algorithm.
- Grouping of residues into 8 groups: polar, hydrophobic, positively charged, negatively charged, aromatic, alanine, glycine, and proline
- “For each unique pair of residue types U and X, the NARDINI algorithm computes a parameter  $\delta_{UX}$  that quantifies the extent to which U and X are mixed or segregated along the linear sequence.”

## Patterning of residues of type U vs. X

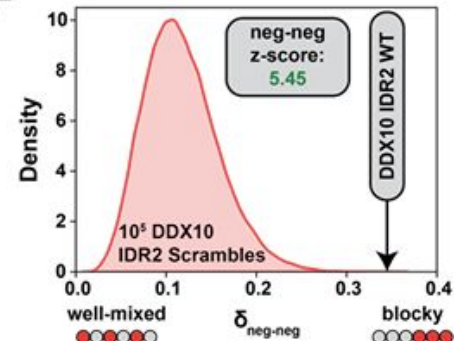
- **pol**≡**Polar**: S, T, N, Q, C, H
- **hyd**≡**Hydrophobic**: I, L, V, M
- **pos**≡**Positive**: K, R
- **neg**≡**Negative**: D, E
- **aro**≡**Aromatic**: F, W, Y
- **ala**≡**Alanine**: A
- **pro**≡**Proline**: P
- **gly**≡**Glycine**: G



**E**



**F**





# NARDINI+ (An Example)

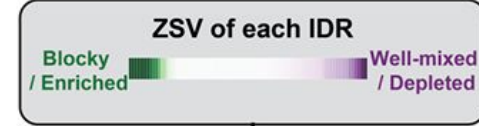
Calculated features for 2 IDRs. Zero features are omitted.



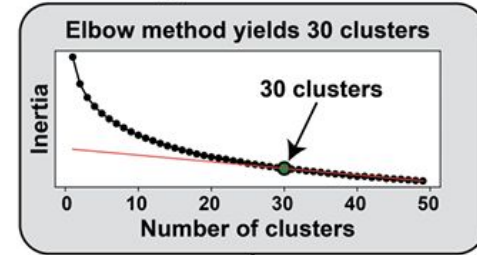
# Grammars Inferred using NARDINI+ (GIN)

- 4,529 IDRs in preferred length range (out of 24508) used to cluster.
- K-means clustering with  $K = 30$
- **All human IDRs** then mapped to clusters.
- 30 clusters are referred as **GIN clusters**.

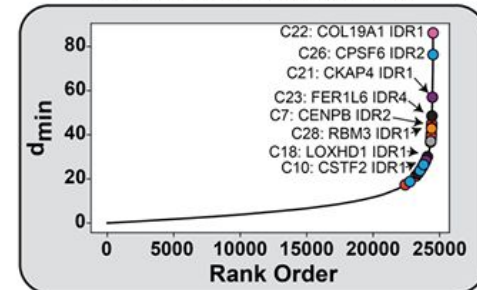
NARDINI+ applied to  
cluster all 24508 human IDRs



K-means clustering of IDRs  
of length  $100 \leq n \leq 300$



Map all human IDRs  
onto 30 GIN clusters



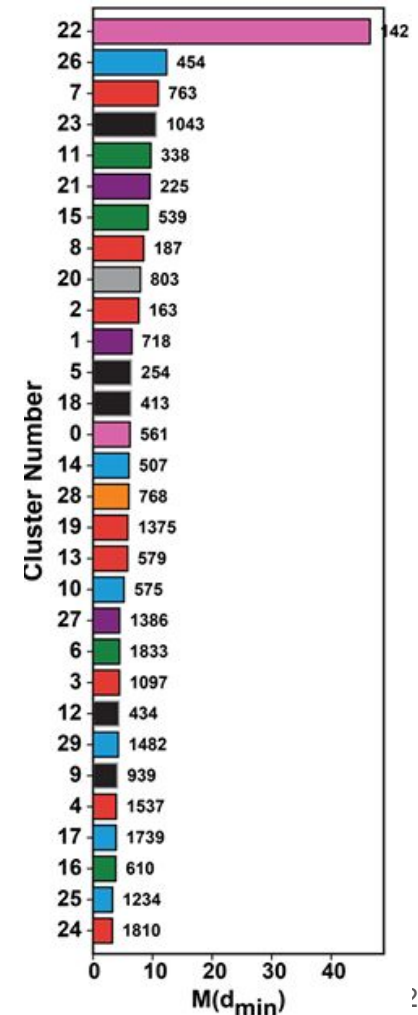


# Grammars Inferred using NARDINI+ (GIN)

Higher  $M(d_{\min})$  (Median of minimum inter-cluster distance) values imply stronger mapping to clusters.

Higher number of IDRs in a cluster should result in a weaker mapping. However there are some exceptions.

For example, cluster 22 is very strongly mapped. 22 is uniquely defined by uniform distribution of Pro and Gly. Defining grammar for elastomers and collagens.



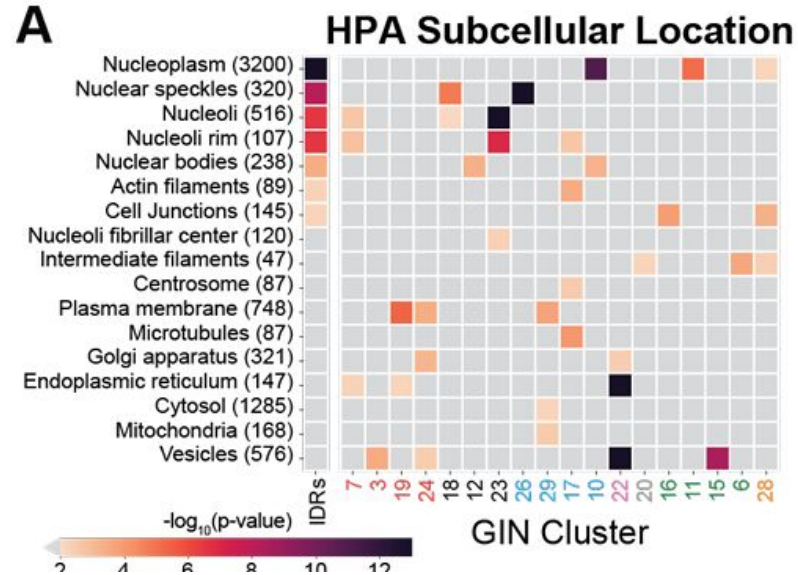
## Results: GIN as a resource

- Authors provide 2 Google Colab notebooks.
- First notebook takes gene id or uniprot accessions and calculates GIN annotations and ZSV vectors.
- Second notebook takes IDRs as fasta or a list of sequences and make the calculations from scratch.

[https://github.com/kierstenruff/RUFF\\_KING\\_Grammars\\_of\\_IDRs\\_using\\_NARDINI-](https://github.com/kierstenruff/RUFF_KING_Grammars_of_IDRs_using_NARDINI-)

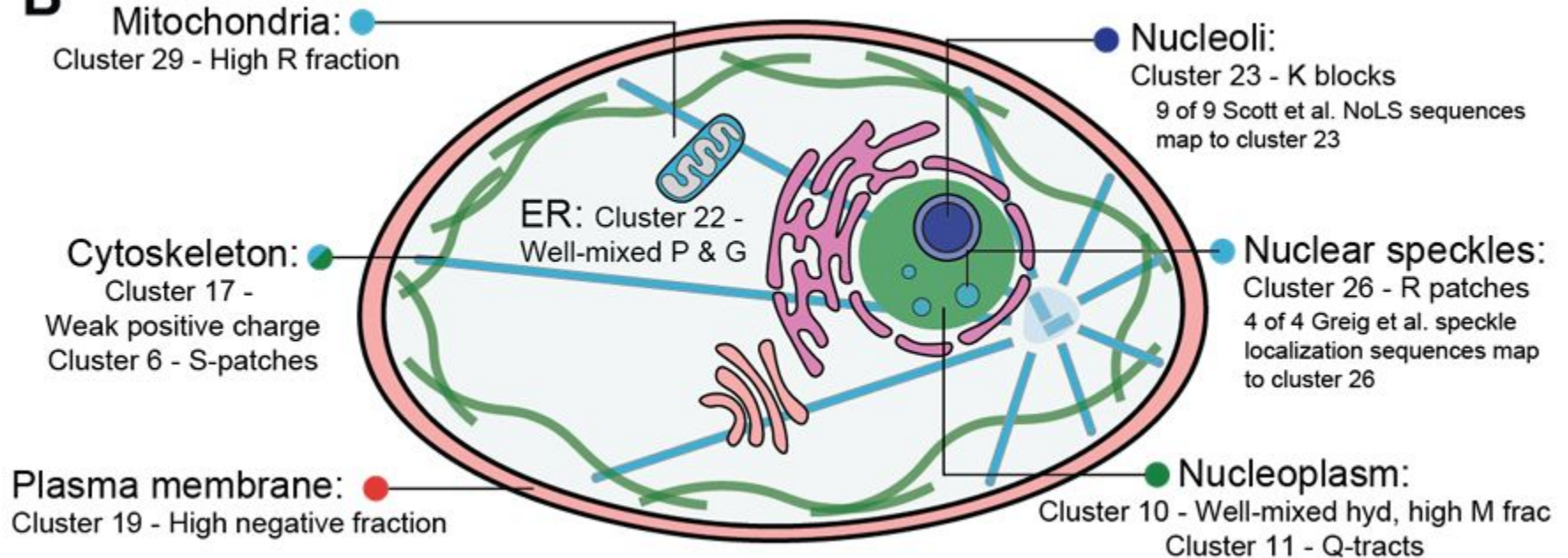
# Results: GIN clusters -> Sub-nuclear Localization

- Annotations of subcellular locations of proteins with IDRs from the Human Protein Atlas (HPA)
- Focus on IDRs of length  $\geq 70$  and non-linker IDRs of length  $\geq 50$  and high dmin values.
- Cluster 26 strongly associated with Nuclear speckles
- Cluster 23 strongly associated with Nucleoli rim



# Results: GIN clusters -> Sub-nuclear Localization

**B**



# Results: GIN clusters -> Sub-nuclear Localization

For further testing the case for cluster 23 and 26:

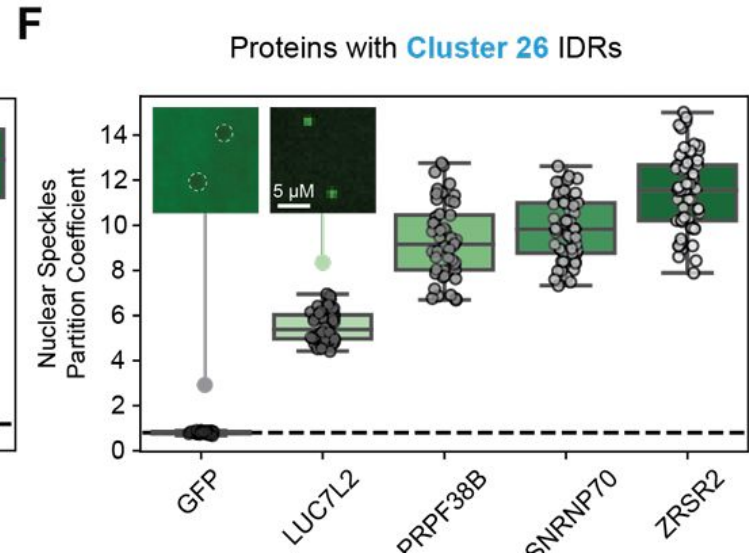
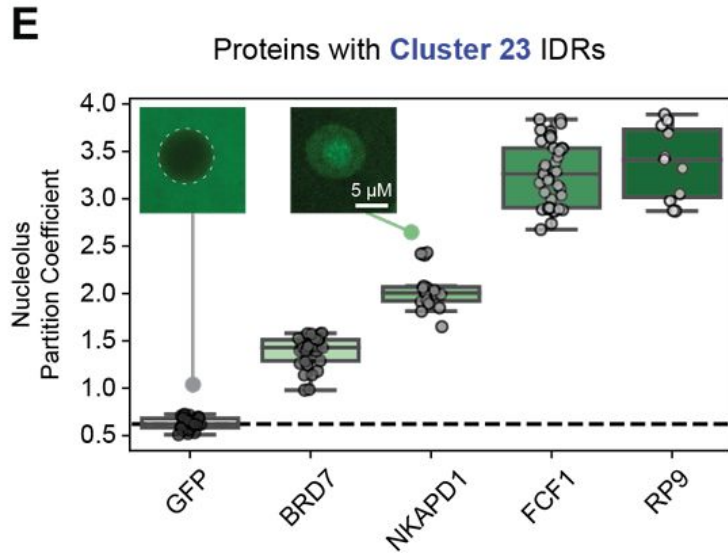
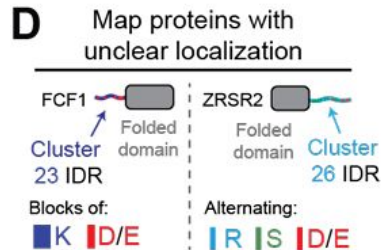
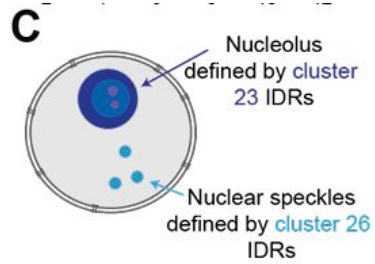
- 5 monomeric proteins from HPA dataset whose IDRs are clustered into either 23 or 26 are selected. These proteins have ambiguous annotations as to sub-nuclear localizations.
- Experiments are made on germinal vesicles (GVs) from live *Xenopus laevis* oocytes.
- Germinal Vesicle: Nucleus of an oocyte
- Oocyte: A developing egg





# Results: GIN clusters -> Sub-nuclear Localization

- Except for GFP, proteins with cluster 23 are partitioned into Nucleolus and proteins with cluster 26 are partitioned into Nuclear Speckles.
- The **partition coefficient**, abbreviated **P**, is defined as a particular ratio of the **concentrations** of a **solute** between the two solvents (a biphasic of liquid phases) (Wikipedia)

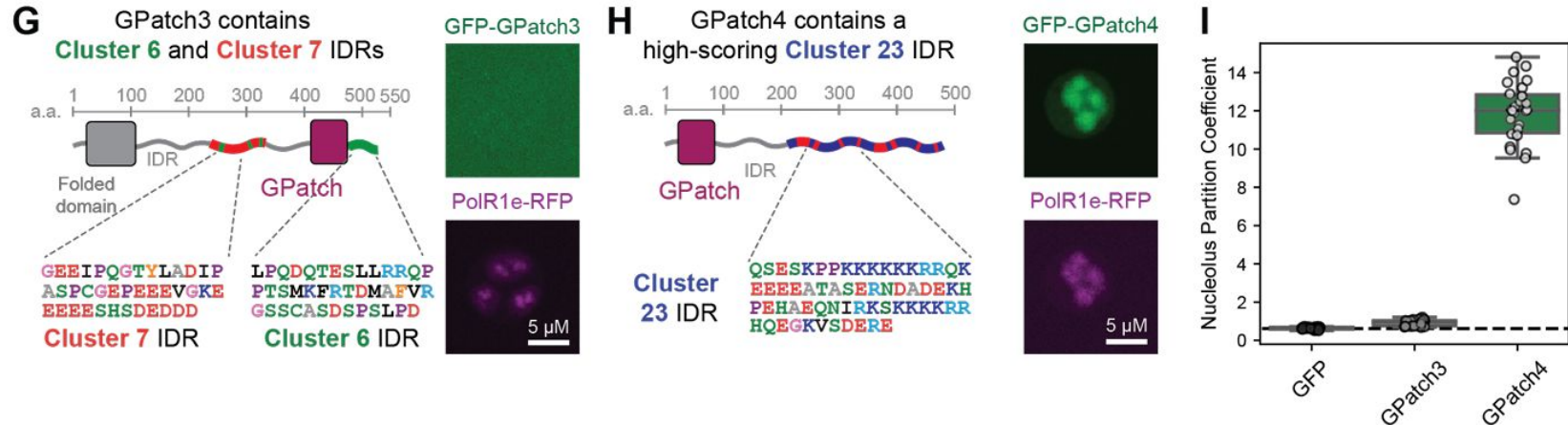


# Results: GIN clusters -> Sub-nuclear Localization

Two naturally occurring proteins GPatch3 and GPatch4 are investigated. Both contain GPatch domain. However, their localization preference are different.

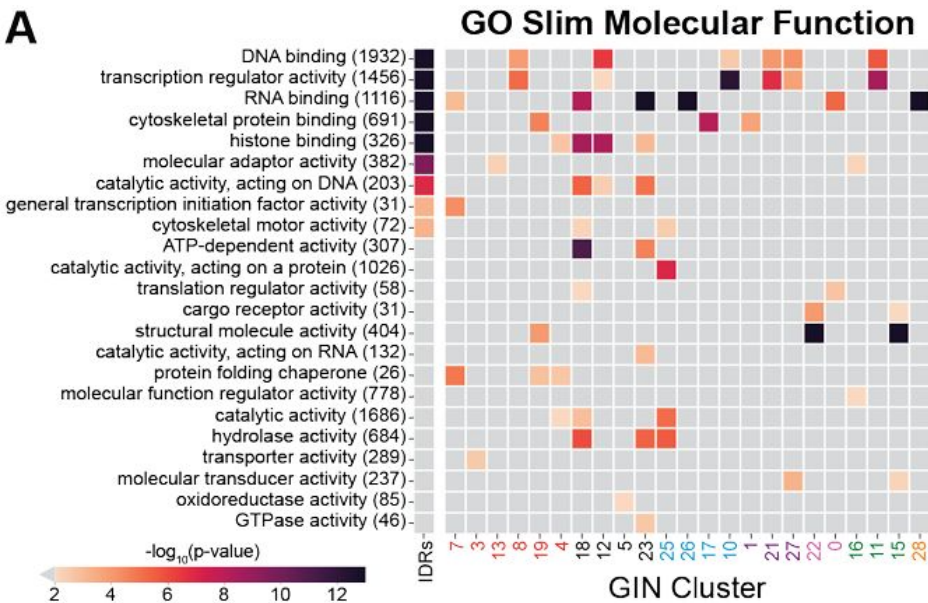
Localization preferences of proteins same domains can be determined by IDRs.

Authors note that, in some cases the vice versa is true.

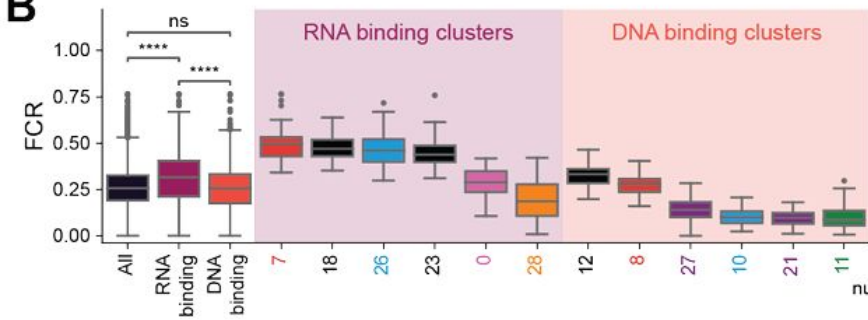


# Results: GIN clusters -> MF and BP

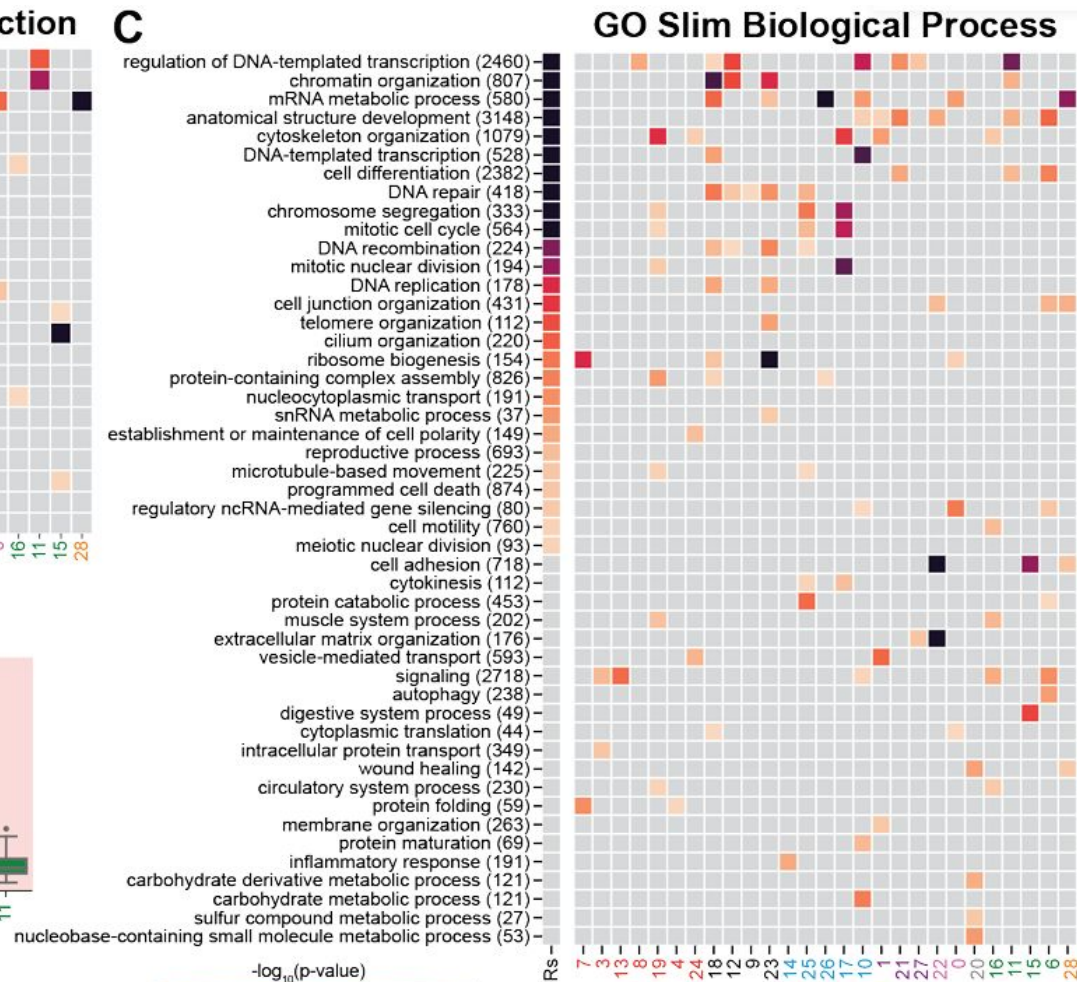
**A**



**B**



**C**



# Results: GIN clusters -> MF and BP

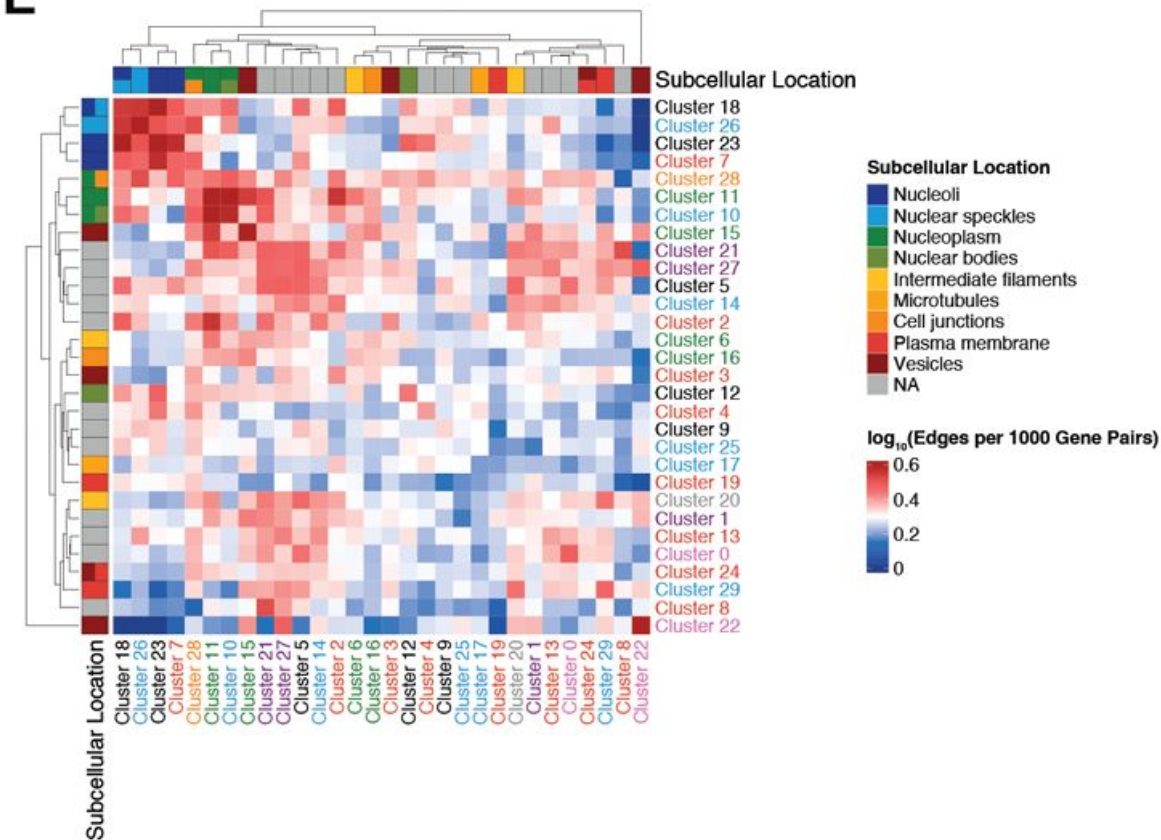
## DepMap24Q4 Dataset

- “Genome-wide CRISPR knockout screening in over 1000 cancer cell lines”
- “Two genes (proteins) might be functionally linked if their fitness effects upon knockout across the cell lines are correlated”
- How proteins with IDR GIN cluster annotations are associated?

# Results: GIN clusters -> MF and BP

Heatmap quantifying functional relationships based on **DepMap24Q4** dataset.

Higher value corresponds to more relationship.





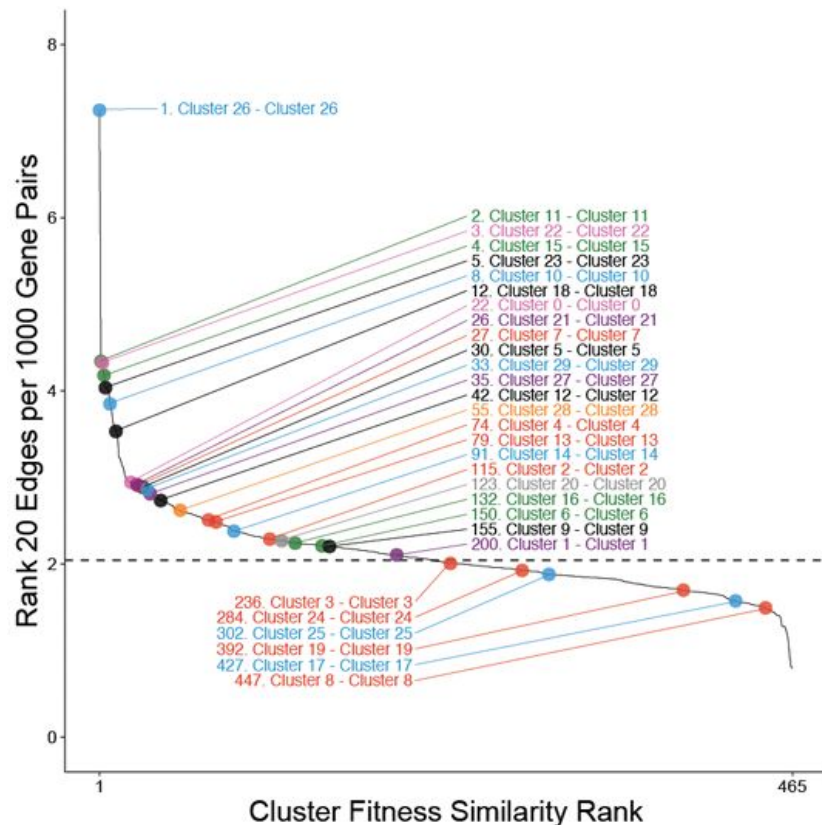
# Results: GIN clusters -> MF and BP

Inter and intra cluster fitness correlations based on **DepMap24Q4**.

Only the intra-cluster data shown on right.

An interesting note:

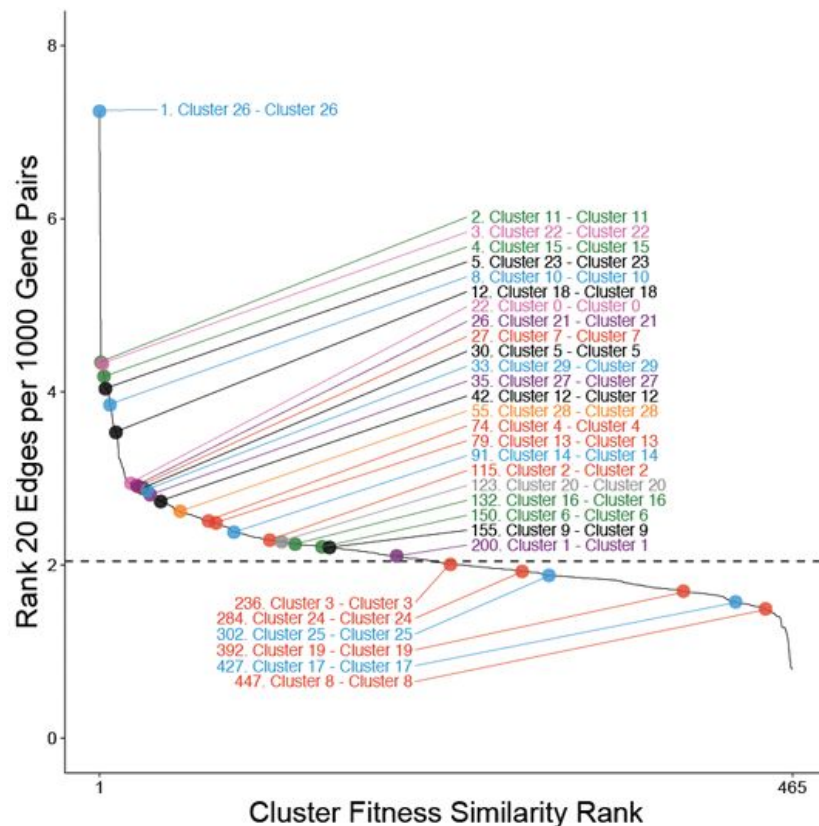
Top ranking clusters 26, 11, 22, 15, 23, and 10 are also highly enriched for specific subcellular locations. Suggesting that subcellular location might be an important signal for fitness data.



# Results: GIN clusters -> MF and BP

6 of 10 top-ranked gene (protein) fitness correlations are (within GIN clusters)

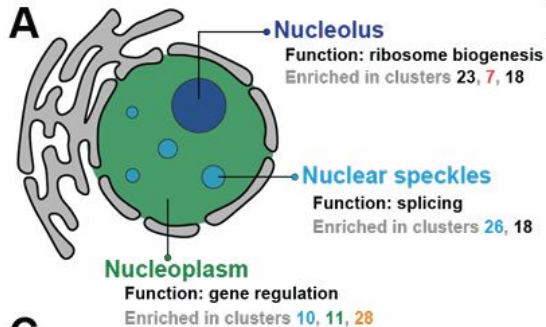
Genes/proteins that are associated some GIN clusters seem to have functional relationships.



# Results: GIN clusters -> Discrete Processes

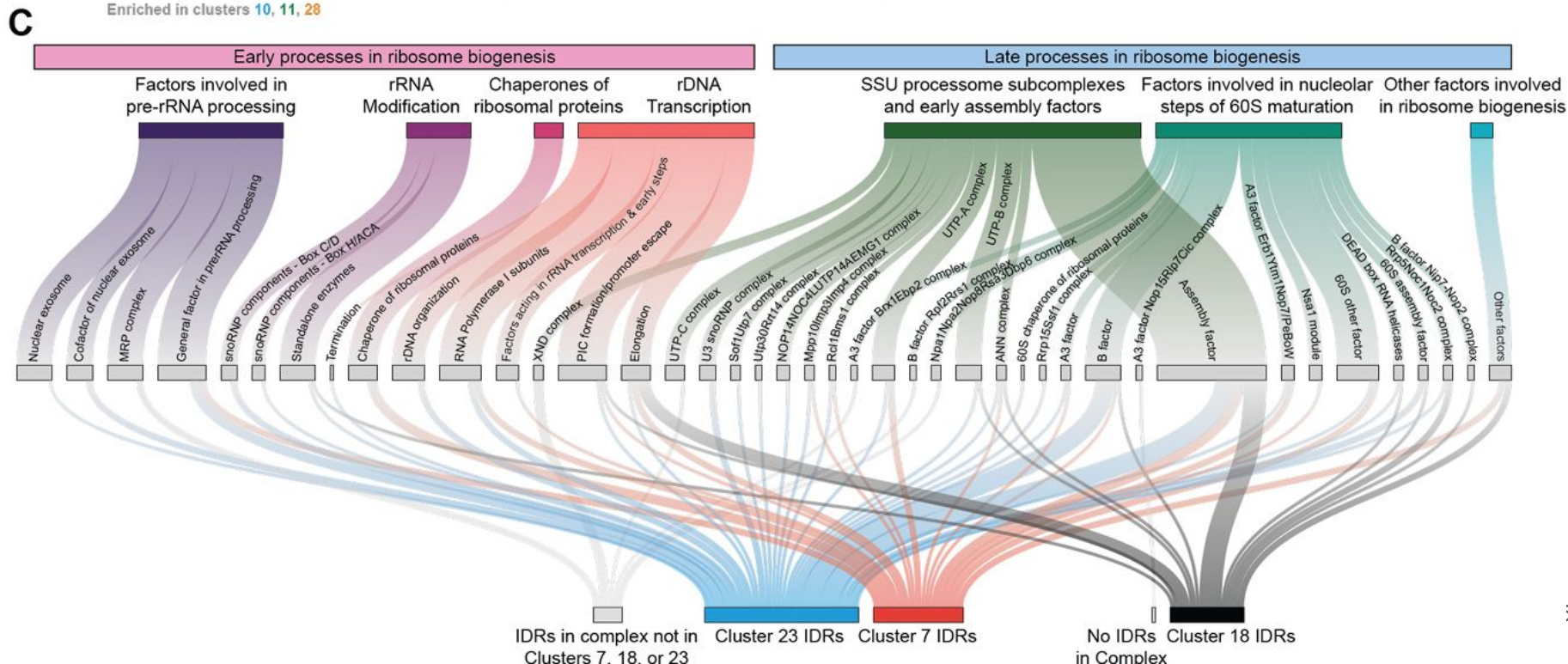
Closer look into by combining known functions of subcellular locations and GIN cluster enrichment.

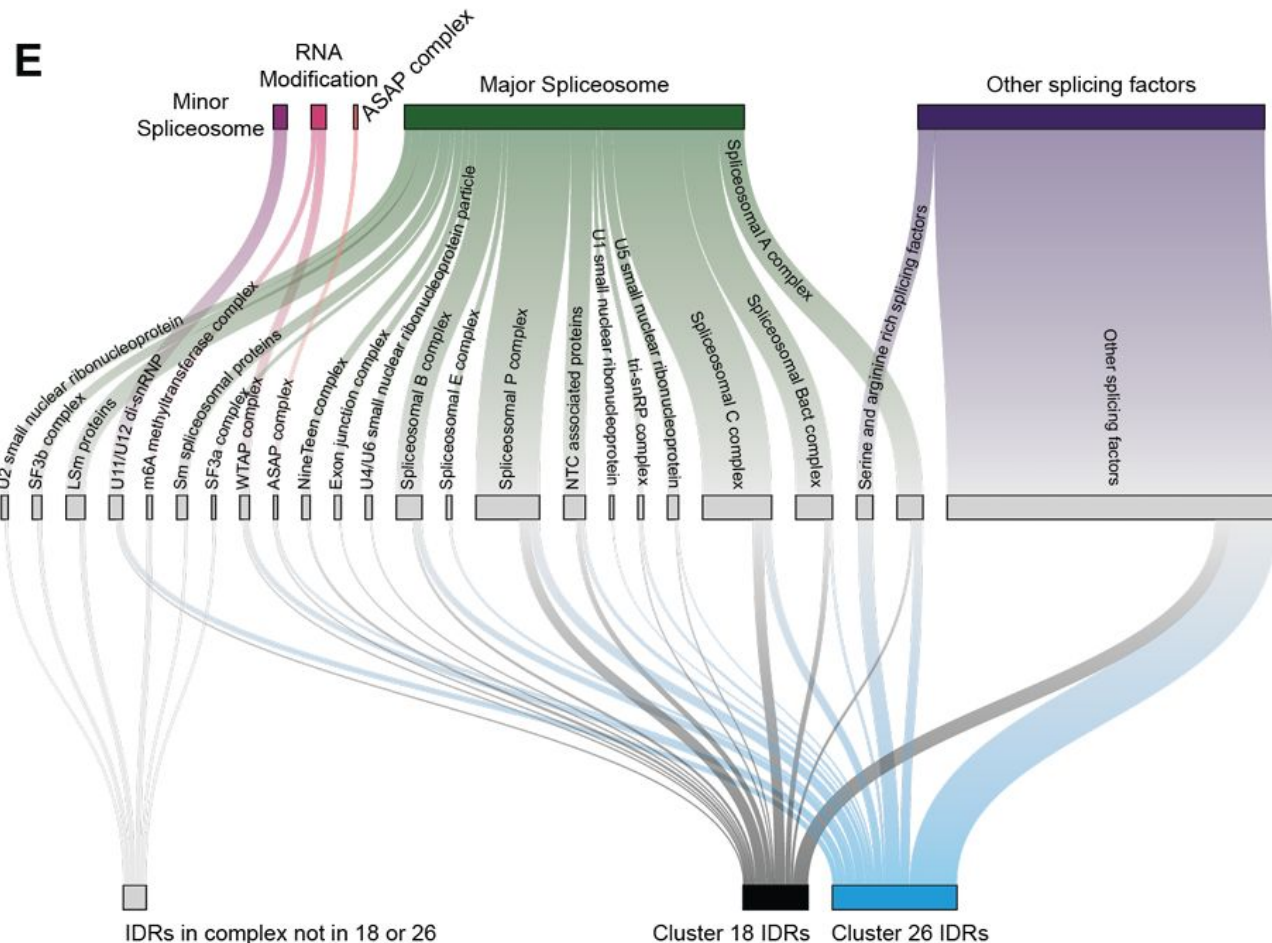
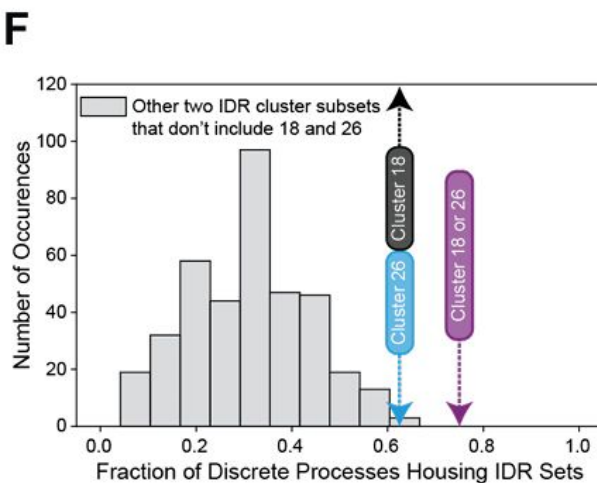
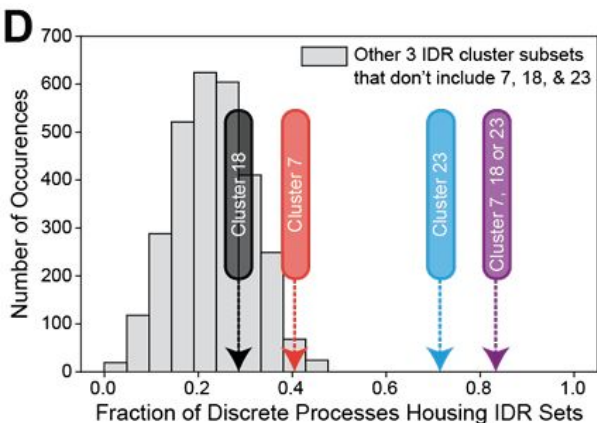




**B**

	GIN	IDR Grammar	Example IDR	Example Sequence
Nucleoli	23	K blocks	POLR1F IDR1	...KKKKKKKKHQEVDQDPVFQGSDDSGYQSDHKKKKKKRKHSEAEFTPL...
	7	D/E-tracts	UBTF IDR4	...EEDDEDEDEDEDEDEEEDDENGDSSEDDGSSSESSSEDESEDGDENEED...
	18	Large negative blocks with positive blocks	NCL IDR1	...DDDDDEEDDSEEAEMETTPAKGKKAQVVPVKANVAEDEDEDEDEDED...
NS	26	R patches	SRRM2 IDR2	...RGRSRSRTPARRGRSRSRTPARRRSRSTPTRRRSRSTPARRGRSRST...
	18	Large negative blocks with positive blocks	CD2BP2 IDR1	...DEEDEDEIIVPKKKLVDPVAGSGGGPSRFGKHSLSDEEEDDDGGSSK...
NP	10	Well-mixed hydrophobics, enriched in M	ARID1B IDR4	...MSMPDVMGRMPYEPNKKDPFGGMRKVPGSSPEPFMTQGGMPNSSMQDMYNQS...
	11	Q-tracts	PAXIP1 IDR2	...QQFHQLQQHQLQQQLLAQLQQHSLLLQQQQQQQIQQQQLQRMHQQQQQQQ...
	28	High aromatic fraction, specifically Ys	POLR2A IDR1	...YSPTSPSYSPSTSPSYSPSTSPSYSPSTSPSYSPSTSPSYSPSTSPSY...





# Results: Exceptional Grammars

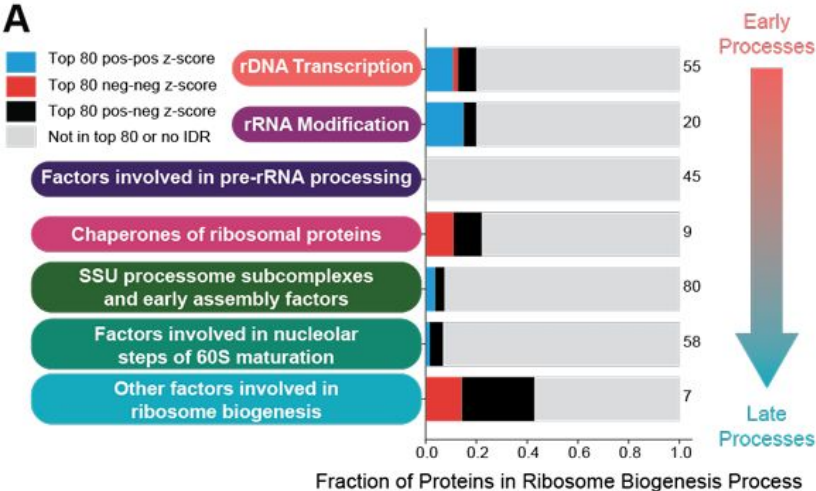
All 24508 IDRs in human proteome is sorted by a “given” GIN cluster feature.

IDRs in 99th percentile are examined.

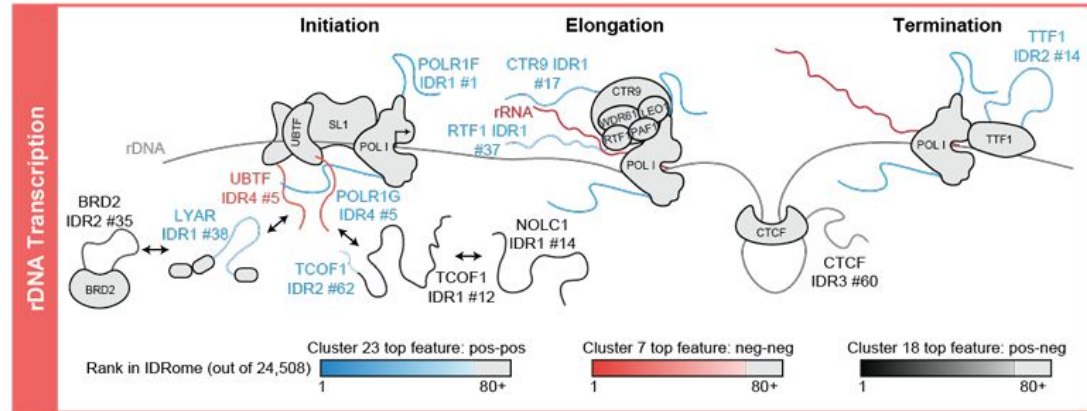
# Results: Exceptional Grammars

The earliest ribosome biogenesis processes are enriched in top scoring K blocks IDRs in the human IDRome (24,508 IDRs)

**A**



**B**

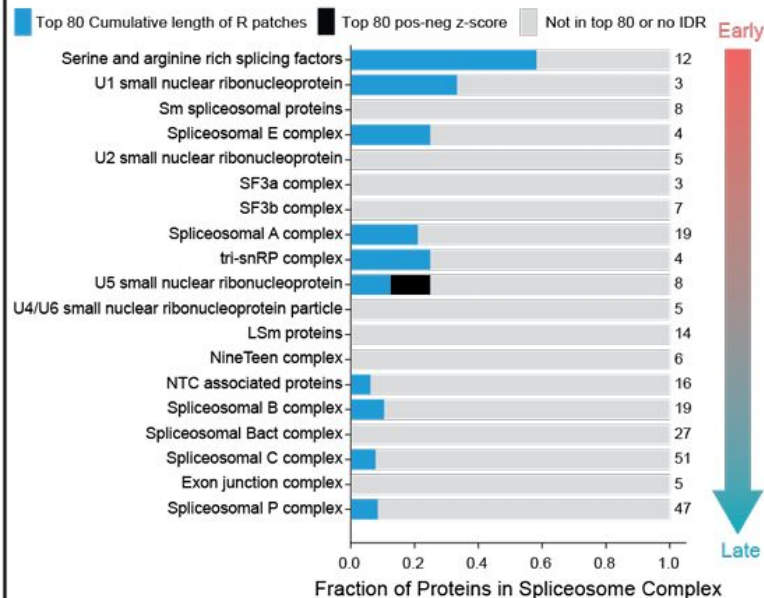




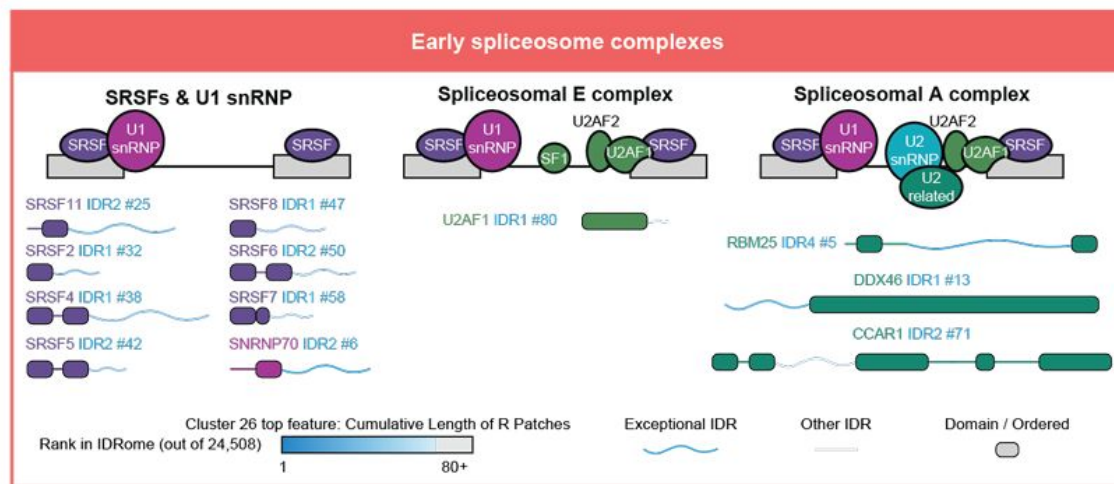
# Results: Exceptional Grammars

The earliest spliceosome processes are enriched in top scoring R patch IDRs in the human IDRome

C



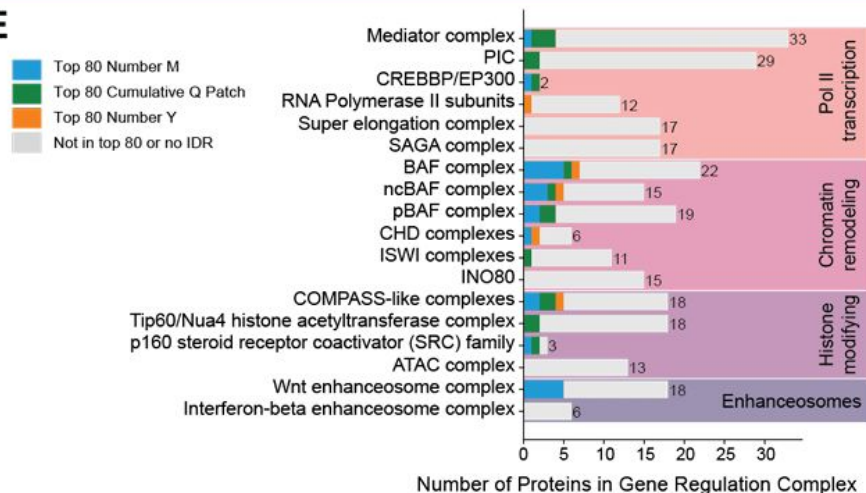
D



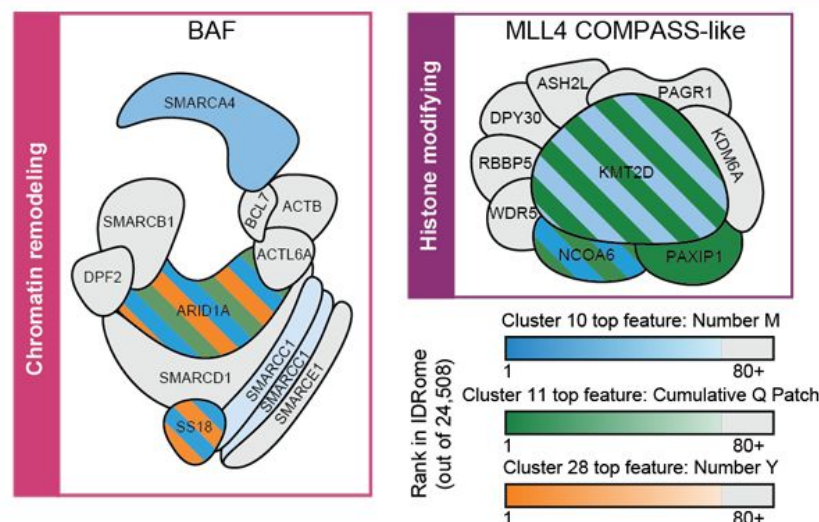
# Results: Exceptional Grammars

Core enhancer activation complexes are enriched in top scoring nucleoplasmic features in the human IDRome

E



F



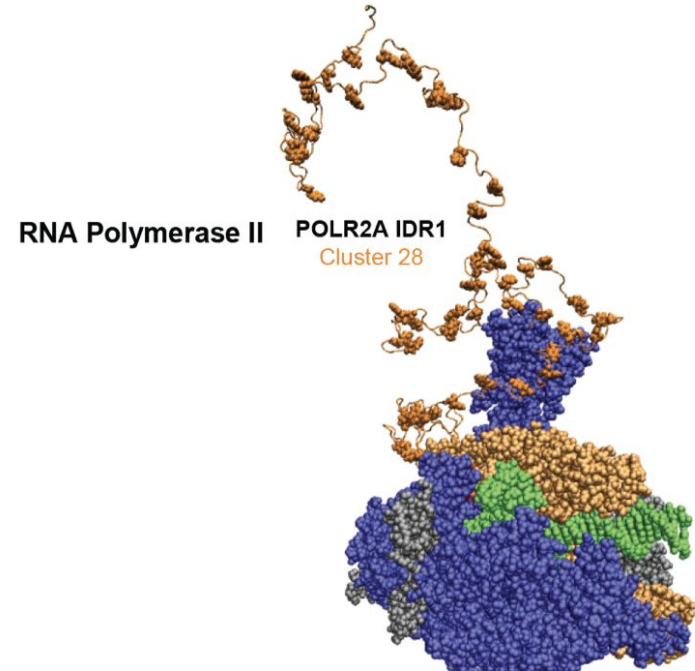
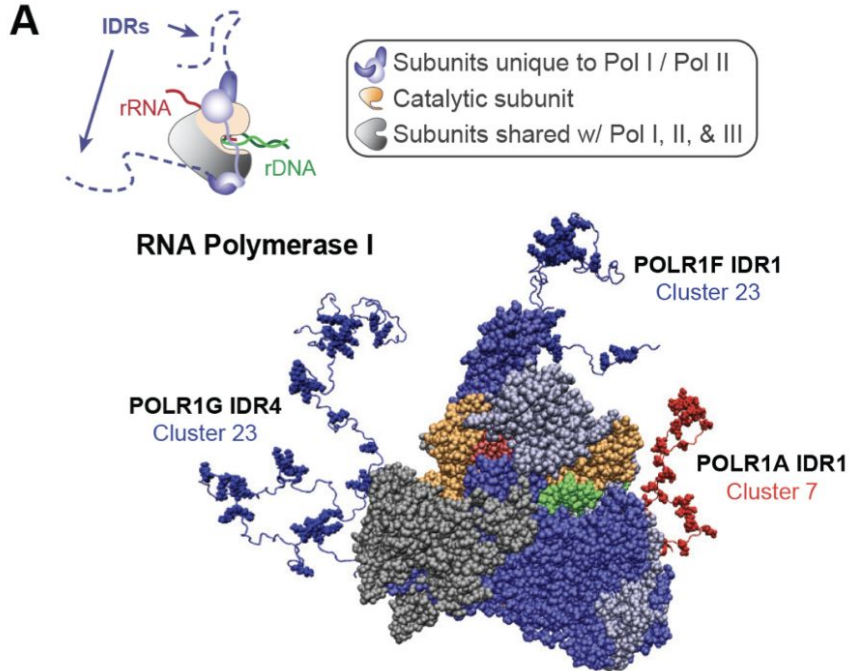
# Results: RNA Pol I & II and evolutionarily conserved exceptional grammars

RNA polymerases play an essential role in transcription.

The subunit-specific IDRs of Pol I and Pol II have grammars with exceptional features that define clusters 23 and 28, respectively.

Orthologs of these IDRs throughout IDRomes of 8 species is investigated.

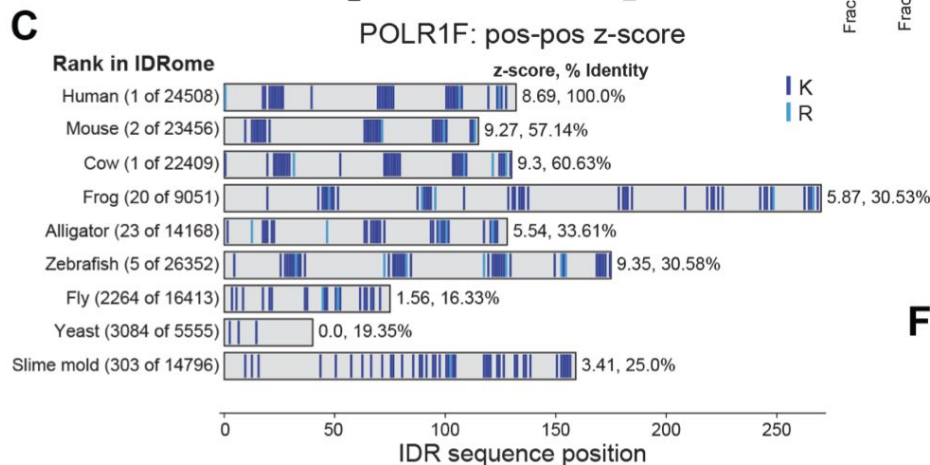
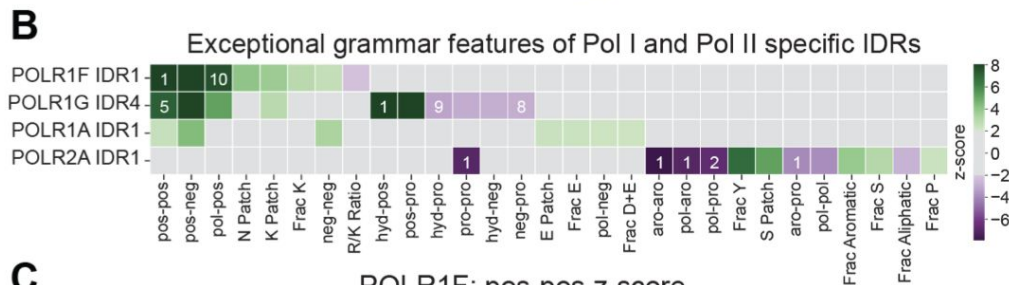
# Results: RNA Pol I & II and evolutionarily conserved exceptional grammars



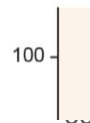


# Results: RNA Pol I & II and evolutionarily conserved exceptional grammars

Although sequences differ throughout,, “exceptionality” of POLR1F is conserved in many species.



**F**



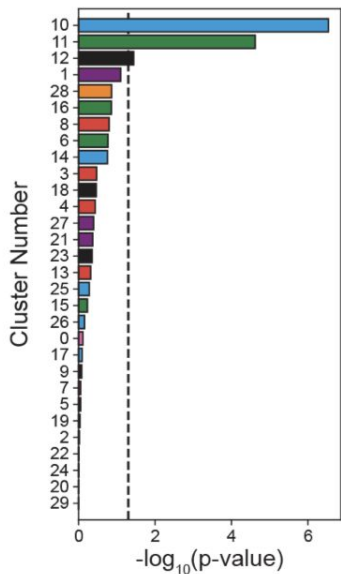
# Results: RNA Pol I & II and evolutionarily conserved exceptional grammars

Figures and analyses containing further investigation and atomic simulations are also included in Figure 6. However, omitted from this presentation.

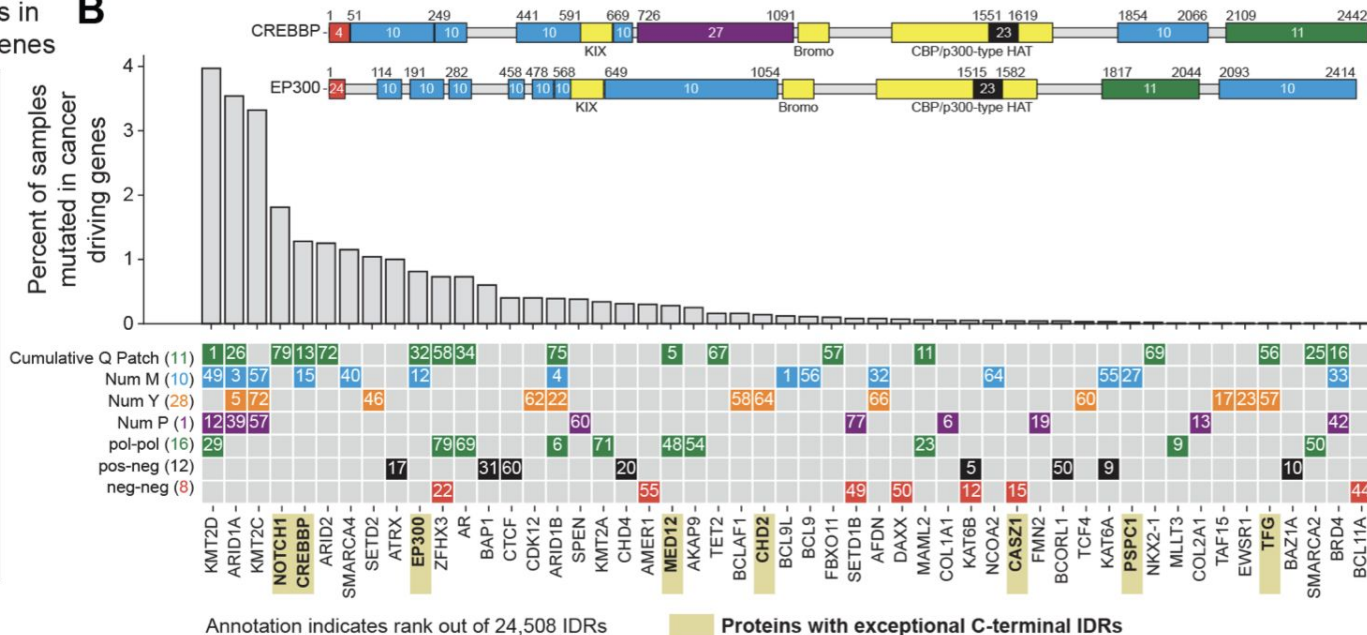
# Results: Cancer mutations disrupt exceptional grammars

Cancer driver genes are investigated. Proteins associated with them are seen to be enriched in clusters 10, 11, 12.

**A** Enrichment of IDR clusters in proteins of cancer driving genes



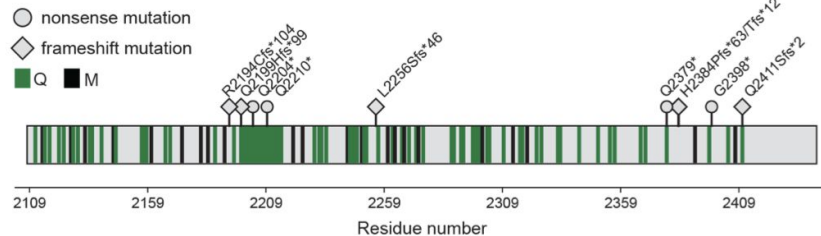
**B**



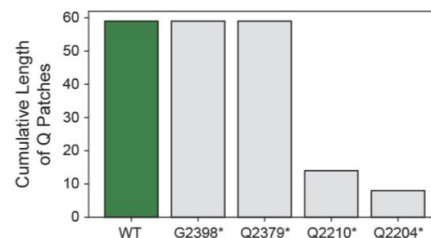
# Results: Cancer mutations disrupt exceptional grammars

Cancer driving mutations of C-terminus IDRs of CREBBP and EP300 can lead to “exceptionality loss” for grammars.

**C** CREBBP Cluster 11 C-terminal IDR



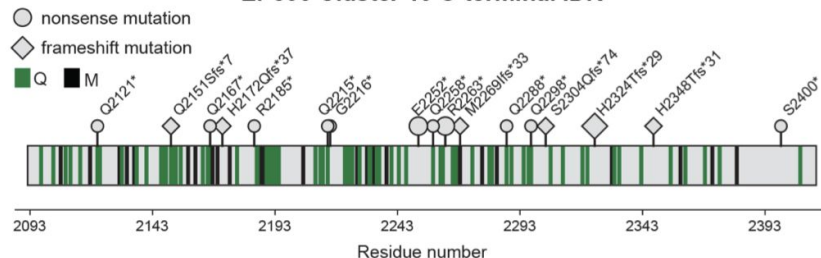
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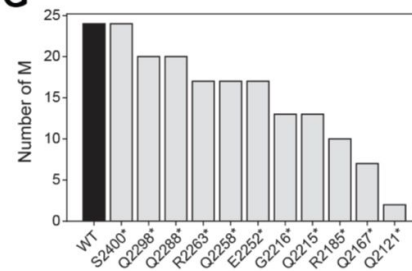
**E**

IDR	Q Patch
WT	13th, 59
G2398*	13th, 59
Q2379*	13th, 59
Q2210*	210th, 14
Q2204*	467th, 8
Total: 24,508	

**F** EP300 Cluster 10 C-terminal IDR



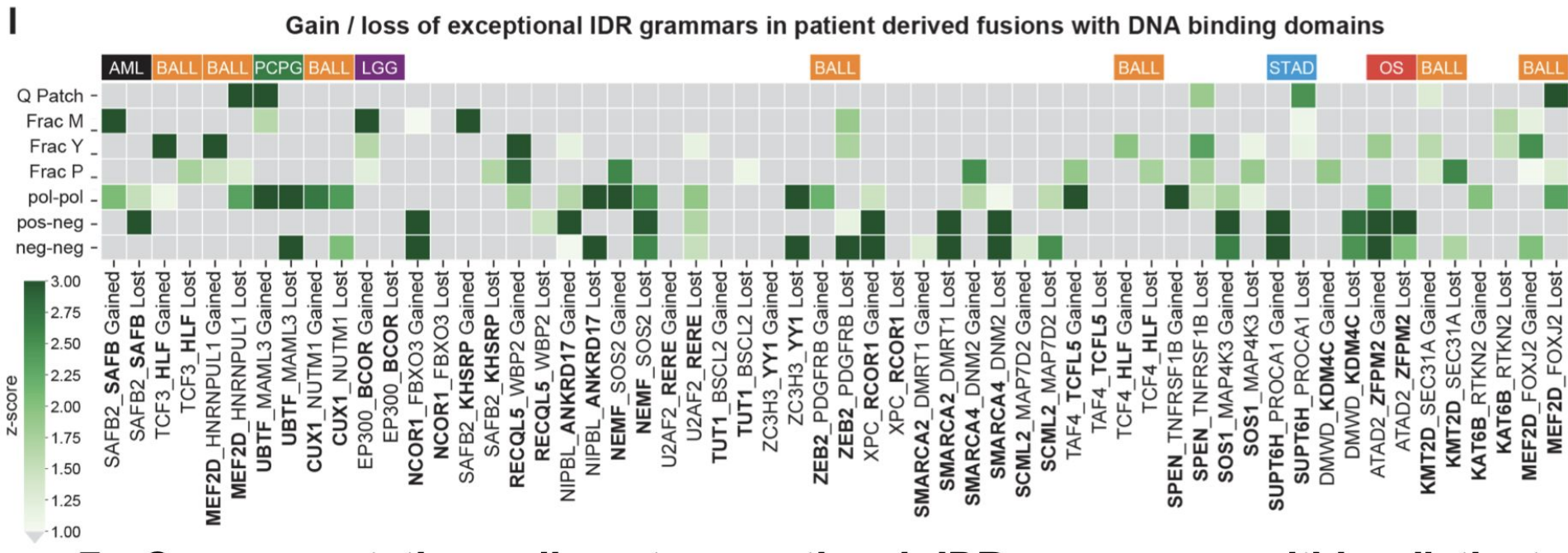
**G**

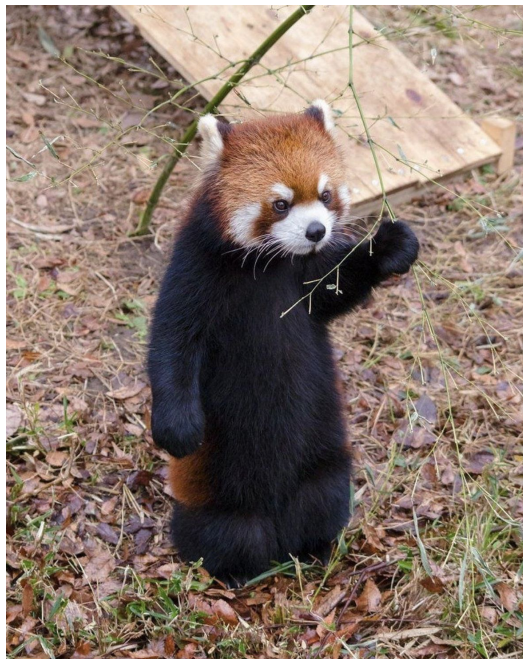


**H**

IDR	Num M
WT	12th, 24
E2252*	29th, 17
Q2215*	63rd, 13
Q2167*	263rd, 7
Q2121*	3157th, 2
Total: 24,508	

# Results: Cancer mutations disrupt exceptional grammars





Thanks for listening