

Table S1. Training dataset composition

	<i>Toggle</i>	<i>Neutral</i>	<i>Rheostat</i>	SNP-Possible	unknown	filtered	detailed
raw	-	-	-	-	822	-	
X-residues					820	2	undetermined amino acid X
SNP-possible	-	-	-	593	227	-	
variant effect-labeling	66	153	-	374	227	-	k-means clustering of experimental scores
manual refinement	60	147	-	374	227	12	removed: T (6), N (6)
ntModel labels	94	181	151	62	113	-	removed: T (74), N (109), R (47)
manual refinement	20	72	104	-	-	230	
<i>funtrpTraining</i>	80	219	104	62	113	244	Final <i>funtrpTraining</i> = 403

Detailed overview of the of training dataset composition for training of both random forest-based models in the *funtrp* pipeline. Show are the total numbers of instances remaining after the variant effect-labeling, filtering and prediction steps.

Table S2. Set of sequence-based features used to train *funtrp* random forest-based models

id	Feature	Source	Description	Parameters	Notes
1	Solvent Accessibility	PROF (*)	predicted solvent accessibility (PACC)	default	
2	Secondary Structure	PROF (*)	predicted helix (pH), strand (pE) or loop (pL)	default	
3	Residue Flexibility	PROFbval (*)	predicted residue flexibility (PROFbval)	default	
4	Protein Disorder	MD (*)	predicted protein disorder (MDraw)	default	
5	Amino Acid	-	amino acids encoded as a vector of length 20	NA	
6	Residue Size	-	basic amino acid property (small or large)	NA	
7	Residue Charge	-	basic amino acid property (uncharged / + / -)	NA	
8	SNP possible	-	number of possible nsSNPs (all codons)	NA	
9	Conservation	ConSurf (*)	predicted conservation	default	MSAs: Big80 + PSI-Blast + MUSCLE
10	MSA Ratio	-	fractions of residue amino acid per MSA column	NA	MSAs: Big80 + PSI-Blast + MAFFT

(*) tools are applied via the PredictProtein pipeline (Yachdav et al., 2014) using a redundancy reduced sequence database. (Big80) which combines UniProt (SwissProt+TrEMBL) and PDB. Features were ranked by importance towards *funtrp* position type labels in Swiss-Prot using ReliefF; weights were rounded off (Kononenko, RobnikSikonja, & Pompe, 1996). If applicable, parameters used in feature computation are specified.

Table S3. Protein subsets for model training

Identifier	Source	Proteins \diamond <i>funtrp</i>	w/ E.C. annotation	w/o E.C. annotation
Swiss-Prot	UniProtKB/SwissProt	20,410 \diamond 19,501	4,273 \diamond 4,241	16,137 \diamond 15,260
TrEMBL	UniProtKB/TrEMBL		9,668 \diamond 9,554	144,277 \diamond 5,254
EXPV	UniProtKB/SwissProt	1,250 \diamond 1,239	1,250 \diamond 1,239	x
PMD	PMD & (Bromberg, Kahn, & Rost, 2013)	1.224 \diamond 1,220	x	x

Extracted datasets used in analysis. EXPV is a subset of experimentally verified enzymes in Swiss-Prot (Mahlich, Steinegger, Rost, & Bromberg, 2018). Literature based annotations of effect (PMD database) were taken from (Bromberg et al., 2013).

Table S4. Confusion matrices of position type predictions for (A) *ntModel* und (B) *funtrpModel*

(A)			(B)			
<i>Neutral</i>	<i>Toggle</i>	Observed \downarrow	<i>Neutral</i>	<i>Toggle</i>	<i>Rheostat</i>	Observed \downarrow
140	7	<i>Neutral</i>	199	4	16	<i>Neutral</i>
9	51	<i>Toggle</i>	2	64	14	<i>Toggle</i>
			19	5	80	<i>Rheostat</i>

Predictions for both models are based on LOO-CV results.

Table S5. Performance of predicting position types for a Random Forest (RF) based classifier model using evolutionary conservation alone

Position Type	Sensitivity	Specificity	PPV	NPV	Precision	Recall	F1	Prevalence	Detection Rate	Detection Prevalence	Balanced Accuracy
<i>Neutral</i>	0.66	0.87	0.75	0.81	0.75	0.66	0.70	0.38	0.25	0.33	0.76
<i>Rheostat</i>	0.46	0.70	0.29	0.83	0.29	0.46	0.35	0.21	0.10	0.33	0.58
<i>Toggle</i>	0.66	0.89	0.81	0.79	0.81	0.66	0.72	0.41	0.27	0.33	0.77

Shown are the averaged performances per class over 100 resample runs. For each run, 3000 residue positions from Swiss-Prot were resampled randomly (without replacement), selecting 1000 instances of each position type respectively. The same was repeated for the test set and a total of 300 residue positions. Position type labels were based on *funtrp* predictions. PPV = positive predictive value; NPV = negative predictive value.

Table S6. Performance of VarCards Ensemble prediction for PMD effect annotation dataset

Position Type	VarCards prediction	neutral		non-neutral		total per effect	total per position type
		(neutral)	(mild/moderate)	(severe)			
<i>Neutral</i>	effect	273	445	652	1,370	2,784	
	no-effect	535	465	414	1,414		
<i>Rheostat</i>	effect	245	624	1,412	2,281	2,836	
	no-effect	191	198	166	555		
<i>Toggle</i>	effect	217	941	1,813	2,971	3,180	
	no-effect	63	94	52	209		
total Neutral		<i>808</i>	<i>910</i>	<i>1,066</i>			
total Rheostat		<i>436</i>	<i>822</i>	<i>1,578</i>			
total Toggle		<i>280</i>	<i>1,035</i>	<i>1,865</i>			
total per PMD effect		1,524	2,767	4,509			

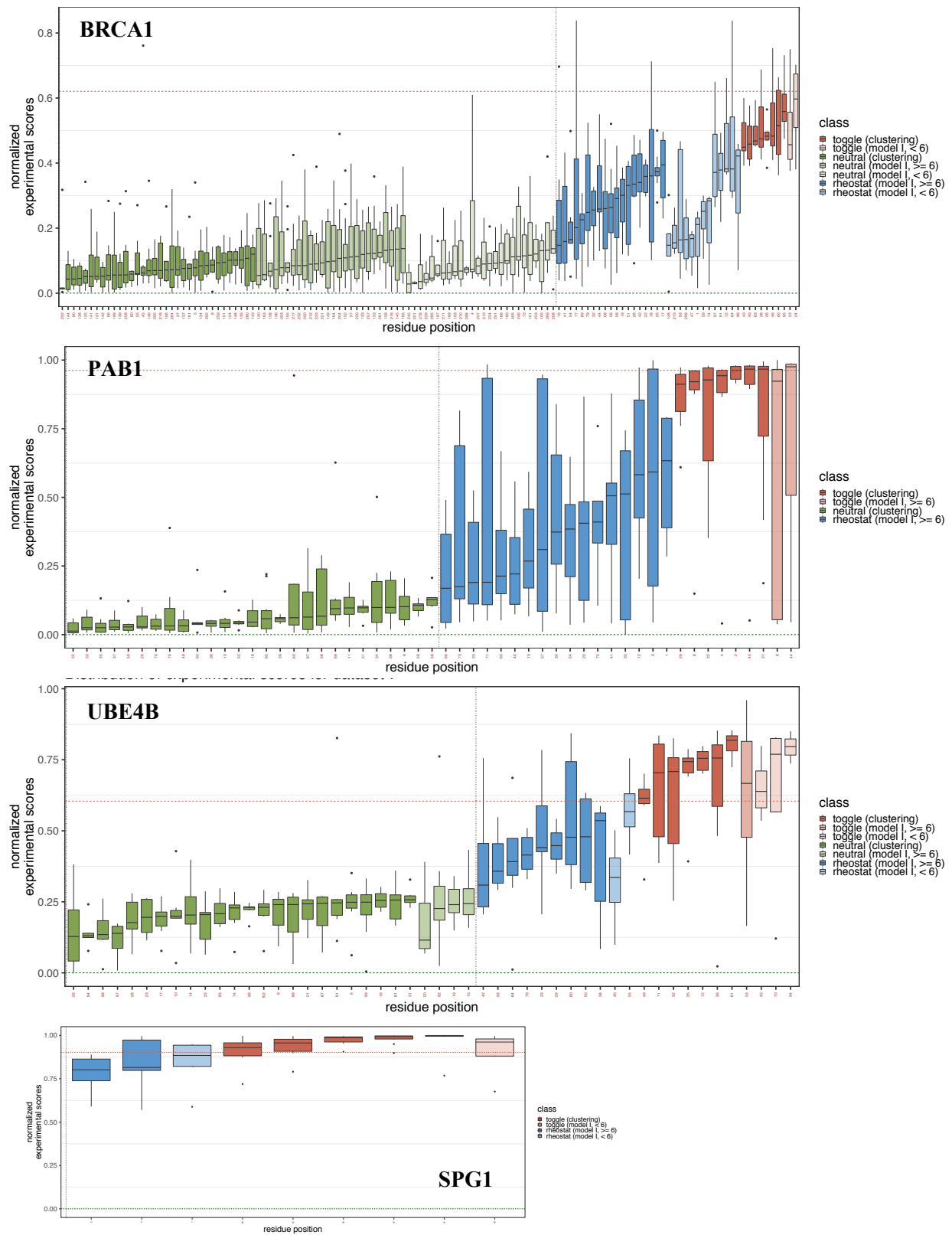


Figure S1. Distribution of experimental (DMS) variant effect scores for training datasets. Measured experimental scores extracted from DMS datasets were normalized to [0,1]. Residue positions on the x-Axis are grouped by (i) position types, (ii) way of labeling and (iii) within these groupings ordered based on increasing distribution medians. The labeling types are: **variant effect-labeled**, predicted with more than six experimental scores available and predicted with less than six experimental scores available.

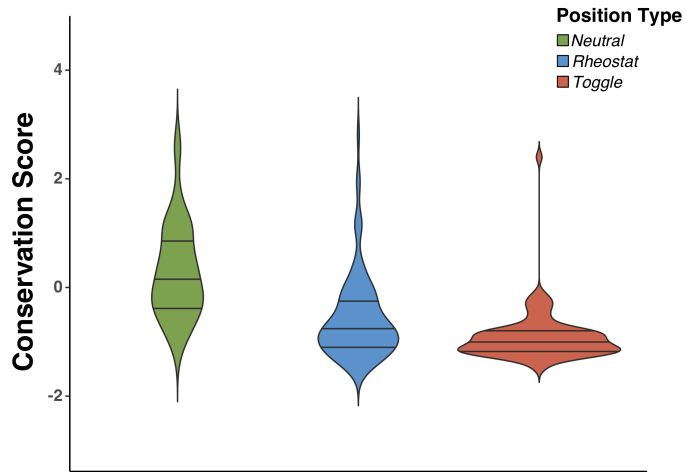


Figure S2. Distribution of ConSurf conservation scores for *funtrp* training dataset. Density distributions of evolutionary conservation (ConSurf) compared between position types for the *funtrp* model training dataset. ConSurf predictions scores are by default normalized such as 0 depicts the average score over the entire protein and standard deviation is 1. Colors are according to position type (green =Neutral, blue =Rheostat, red =Toggle).

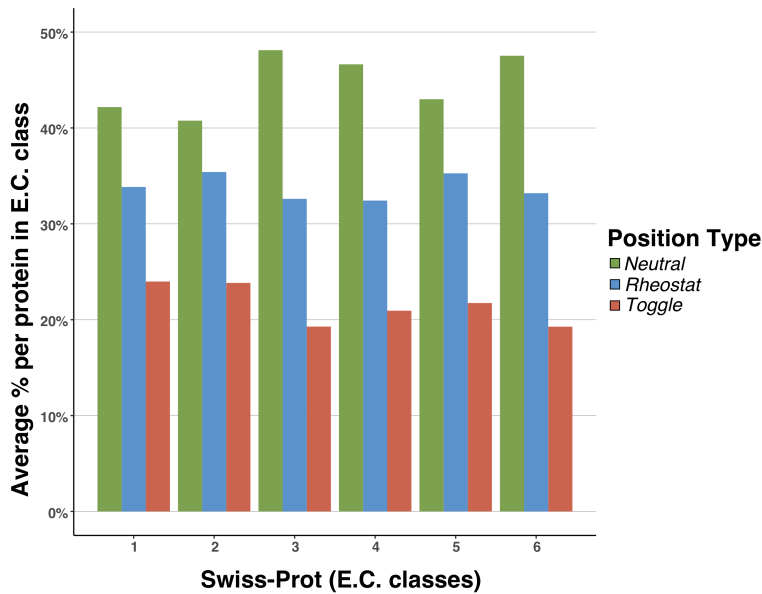


Figure S3. Average fraction of position types on per-protein basis for main E.C. classes in the entire Swiss-Prot dataset. Colors are according to position type (green =Neutral, blue =Rheostat, red =Toggle). Mean fractions of position types differ significantly among enzyme classes based on the standard error of the mean: 1 (N= 6.0E-04, R=6.4E-04, T=5.8E-04), 2 (N=3.7E-04, R=4.1E-04, T=2.4E-04), 3 (N=5.2E-04, R=4.2E-04, T=3.6E-04), 4 (N=9.2E-04, R=1.0E-03, T=7.6E-04), 5 (N=1.5E-03, R=1.1E-03, T=1.2E-03), 6 (N=8.9E-04, R=9.5E-04, T=9.1E-04).

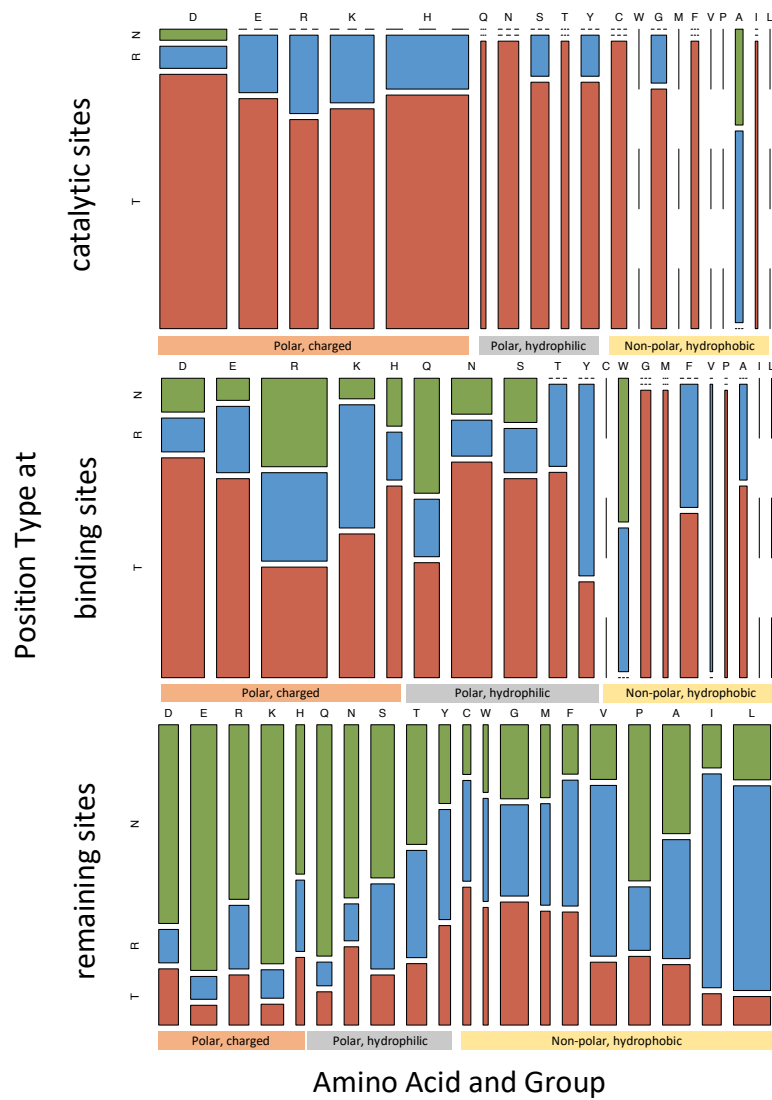
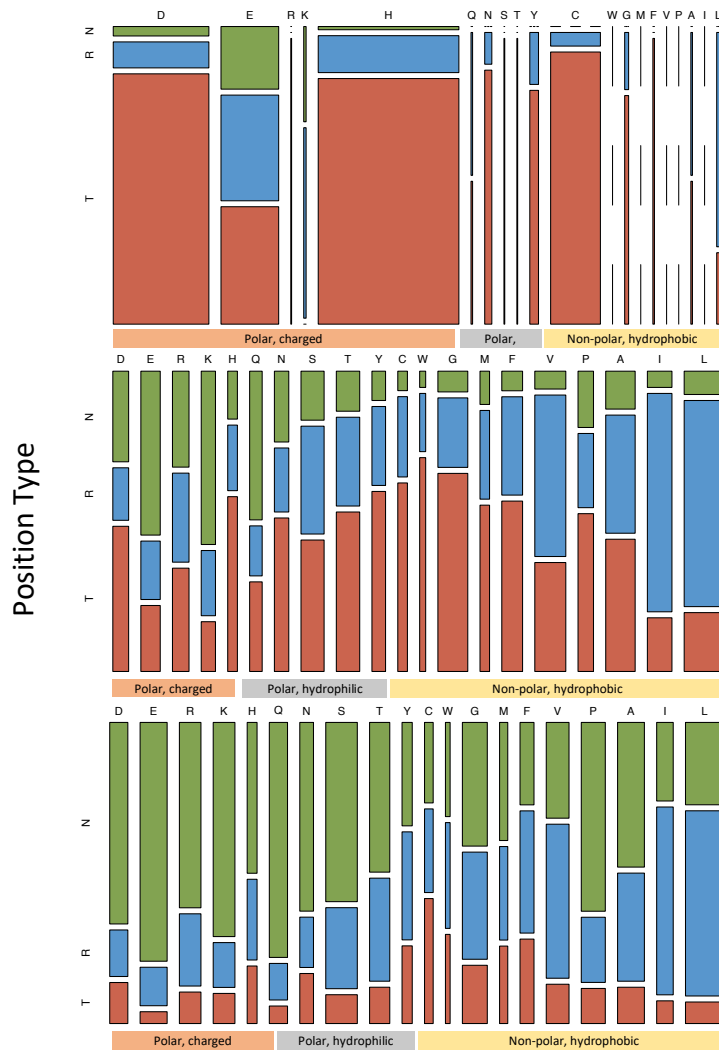


Figure S4. Fractions of position types per amino acid compared by site characteristic. Comparison of fractions at catalytic sites and binding sites against the remaining residues of the respective Swiss-Prot enzymes. Colors are according to position type (green = *Neutral*, blue = *Rheostat*, red = *Toggle*).



Amino Acid and Group

Figure S5. Fractions of position types per amino acid for metal binding sites and spheres. Comparison of SaHLe spheres and residues annotated as metal binding sites within spheres vs remaining residues of the respective Swiss-Prot enzymes. Colors are according to position type (green =Neutral, blue =Rheostat, red =Toggle).

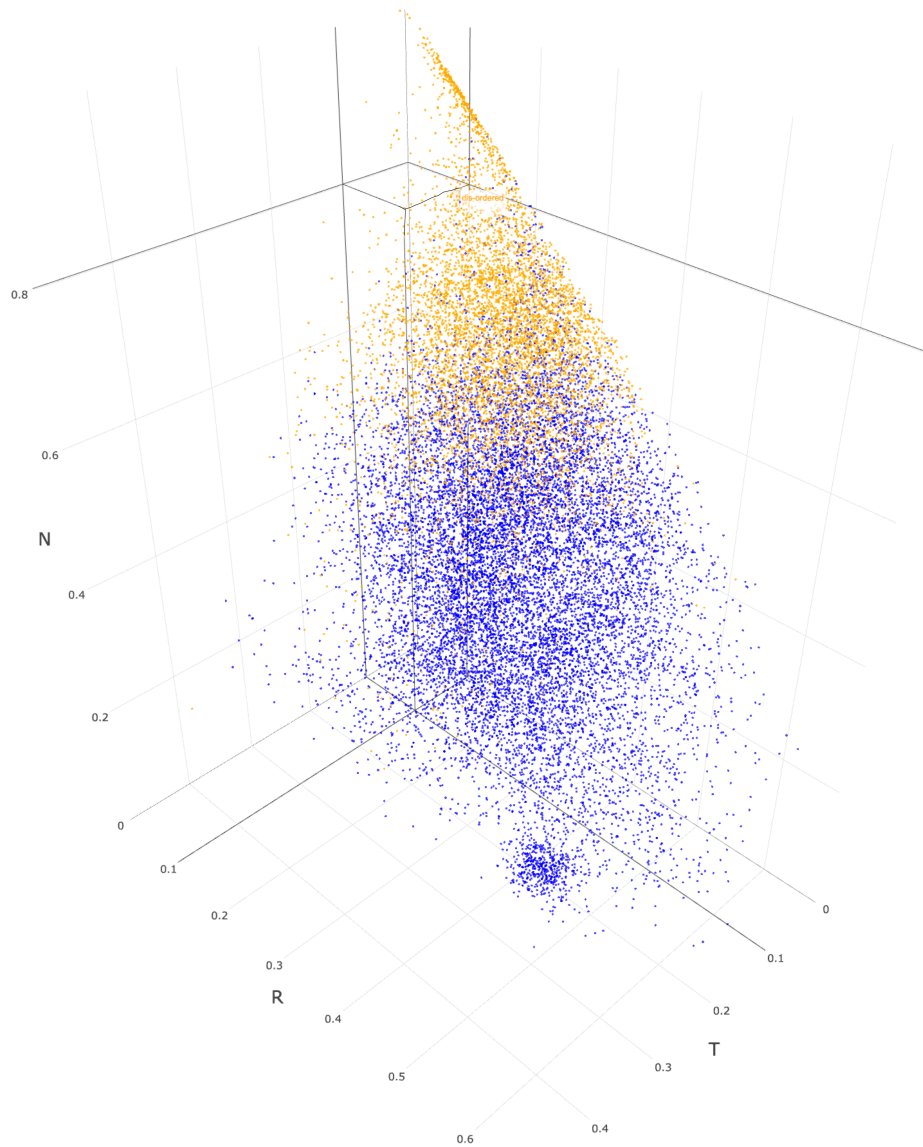


Figure S6. *funtrp* prediction scores for disordered Proteins compared within position types. Proteins in Swiss-Prot were labeled as either ordered or disordered based on MetaDisorder predictions (Methods). Residues located in disordered proteins are highlighted in yellow, those found in ordered proteins are shown in blue.

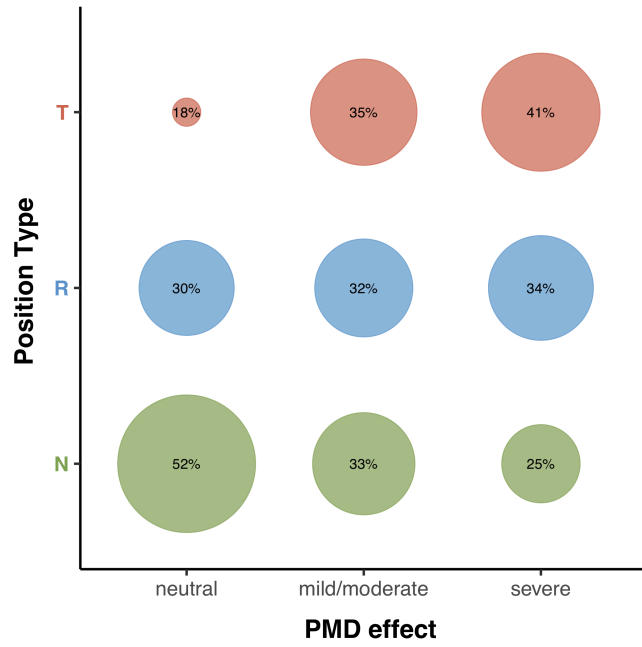


Figure S7. Distribution of position types for PMD effect annotations. PMD *mild* and *moderate* effects annotations were grouped into *mild/moderate*. Percentages are rounded; colors are according to position type (green = *Neutral*, blue = *Rheostat*, red = *Toggle*).