### **Supplemental Materials**

for

# Prioritization of regulatory variants with tissue-specific function in the non-coding regions of human genome

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## Supplemental Tables

Supplemental Table S1. Number of allele specific TF binding (ASB) training SNVs in 6 cell lines.

Cell type name	# of ASB SNVs	# of control SNVs
GM12878	1,848	12,382
A549	215	1,357
H1hESC	1,464	6,525
HepG2	2,717	24,479
K562	767	12,635
MCF7	653	8,034

#### Supplemental Table S2. Feature list in random forest models of TURF.

Generic features	
TF binding sites from ChIP-seq	Binary variable
DNase I hypersensitive sites from DNase-seq	Binary variable
DNase footprints	Binary variable
DNase footprints with matched TF ChIP-seq peaks	Binary variable
TF motifs from PWM matching	Binary variable
TF motifs from PWM matching with matched TF ChIP-seq peaks	Binary variable
Information content change of two alleles in PWM matching	Numerical variable
Information content change of two alleles in PWM matching with matched TF ChIP-seq peaks	Numerical variable
eQTLs	Binary variable
Quantiles (25%,50%,75% and 100%) and variance of ChIP-seq signals across all available ChIP-seq experiments from ENCODE	Numerical variables
Functional significance score from DeepSEA	Numerical variable
Tissue-specific features (in final ensemble model)	
H3K4me1 peaks from ChIP-seq	Binary variable
H3K4me3 peaks from ChIP-seq	Binary variable
H3K27ac peaks from ChIP-seq	Binary variable
H3K36me3 peaks from ChIP-seq	Binary variable
H3K27me3 peaks from ChIP-seq	Binary variable
DNase I hypersensitive sites from DNase-seq	Binary variable
DNase footprints	Binary variable

Supplemental Table S3. Feature importance of TURF generic scores. The importance is calculated based on mean decrease of impurity from the random forest model.

Feature name	Importance	
CHIP	0.124607	
DNASE	0.033935	
PWM	0.018686	
FOOTPRINT	0.008609	
EQTL_2	0.004167	
PWM_matched	0.005055	
FOOTPRINT_matched	0.001454	
IC_change	0.050415	
IC_matched_change	0.021396	
funsig	0.122206	
ChIP_var	0.149313	
ChIP_max	0.167037	
ChIP_quantile1	0.097849	
ChIP_quantile2	0.090261	
ChIP_quantile3	0.105011	

Supplemental Table S4. Comparison of performance on cell type-specific predictions for MPRA variants and ASB SNVs.

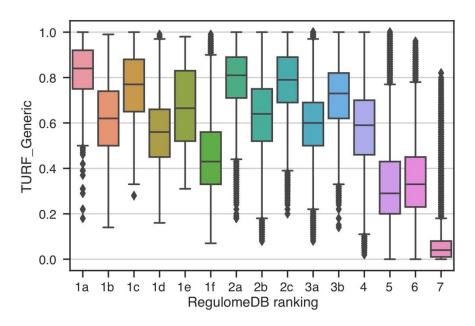
Method	AUPR	AUROC	Correlation	
Test on MPRA variants in E116 (GM12878)				
TURF_Tissue-Specific	0.548	0.731	0.450	
TURF_Generic	0.536	0.740	0.447	
GenoNet	0.536	0.723	0.442	
DeepSEA	0.451	0.671	0.115	
GenoSkyline	0.394	0.673	0.352	
CADD	0.240	0.566	0.082	
TURF_Ensemble (trained from ASB SNVs)	0.484	0.724	0.393	
Test on MPRA variants in E118 (HepG2)				
TURF_Tissue-Specific	0.571	0.765	0.423	
TURF_Generic	0.502	0.733	0.372	
GenoNet	0.572	0.757	0.429	
DeepSEA	0.401	0.667	0.166	
GenoSkyline	0.326	0.628	0.225	
CADD	0.332	0.607	0.152	
TURF_Ensemble (trained from ASB SNVs)	0.479	0.728	0.357	
Test on MPRA variants in E123 (K562)				
TURF_Tissue-Specific	0.441	0.681	0.344	
TURF_Generic	0.400	0.664	0.294	
GenoNet	0.420	0.706	0.340	
DeepSEA	0.290	0.607	0.099	

GenoSkyline	0.284	0.659	0.274		
CADD	0.212	0.539	0.046		
TURF_Ensemble (trained from ASB SNVs)	0.329	0.676	0.253		
Test on ASB SNVs in GM12	Test on ASB SNVs in GM12878				
TURF_Ensemble	0.511	0.894	0.538		
TURF_Tissue-Specific	0.510	0.882	0.537		
TURF_Generic	0.284	0.822	0.390		
GenoSkyline	0.246	0.743	0.340		
DeepSEA	0.222	0.688	0.190		
GenoNet	0.129	0.685	0.158		
Test on ASB SNVs in HepG	2				
TURF_Ensemble	0.355	0.857	0.406		
TURF_Tissue-Specific	0.353	0.831	0.401		
TURF_Generic	0.221	0.797	0.323		
DeepSEA	0.226	0.611	0.138		
GenoSkyline	0.212	0.582	0.133		
GenoNet	0.161	0.654	0.134		
Test on ASB SNVs in K562					
TURF_Ensemble	0.316	0.847	0.376		
TURF_Tissue-Specific	0.313	0.815	0.366		
TURF_Generic	0.193	0.782	0.301		
GenoNet	0.140	0.644	0.113		
GenoSkyline	0.090	0.628	0.144		
DeepSEA	0.062	0.572	0.054		

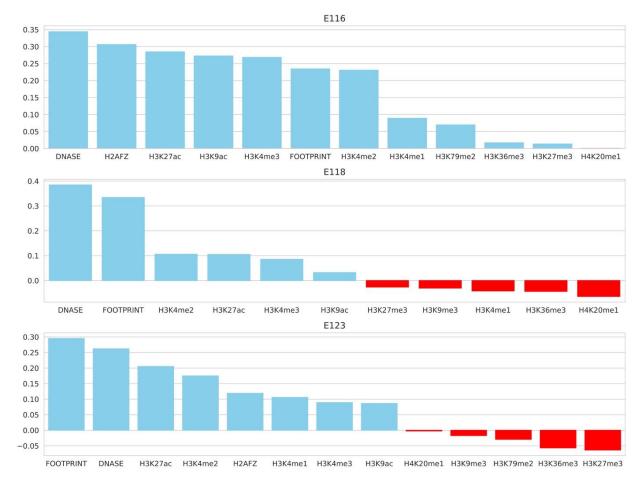
#### Test on ASB SNVs in A549

TURF_Ensemble	0.313	0.872	0.369
TURF_Tissue-Specific	0.315	0.833	0.369
TURF_Generic	0.158	0.796	0.256
DeepSEA	0.052	0.760	0.119
GenoNet	0.030	0.690	0.094
GenoSkyline	0.022	0.632	0.049

#### **Supplemental Figures**



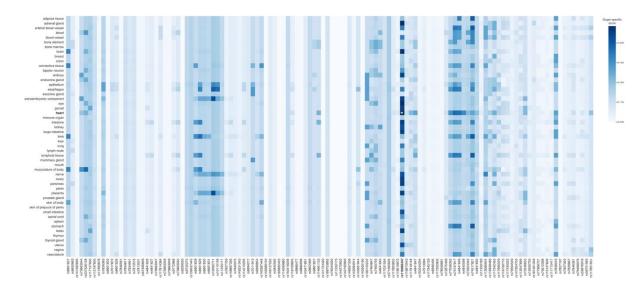
Supplemental Figure S1. Boxplot of TURF generic scores VS RegulomeDB ranking scores on 10,422,004 common SNVs from dbSNP153. X axis represents the original ranking scores from RegulomeDB v1.1, y axis represents the TURF generic scores from random forest model.



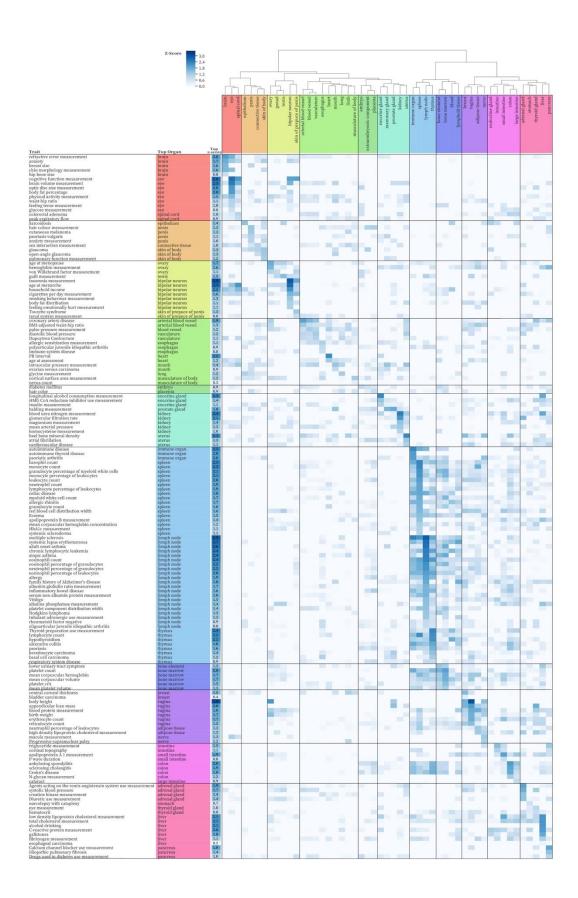
Supplemental Figure S2. Pearson correlation of labels and tissue-specific features in three MPRA datasets (E116: GM12878; E118: HepG2; E123: K562). Blue bars represent positive correlations while red bars represent negative correlations.



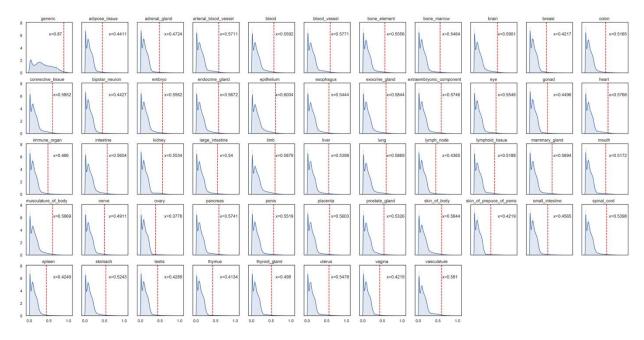
Supplementary Figure 3. Pearson correlation of labels and tissue-specific features in 6 ASB datasets. Blue bars represent positive correlations while red bars represent negative correlations.



Supplemental Figure S4. Organ-specific scores of variants in the *GATA4* locus. rs118065347 has the top organ-specific score in heart compared to other candidate SNPs found through genome-wide association scan on bicuspid aortic valve cases. rs118065347 was shown to have hear-specific regulatory functionality from a previous study Yang et al. 2017.



Supplemental Figure S5. Enrichment of regulatory variants with high organ-specific scores over variants associated with diverse traits. Only organ-trait pairs with z-scores higher than 0 and passing multiple test correction (FDR threshold of 5%) are shown.



Supplemental Figure S6. Distribution of TURF generic and organ-specific scores on 3,102,261 GM12878 SNVs. The red vertical dashed line represents the 99<sup>th</sup> percentile score in each subplot.