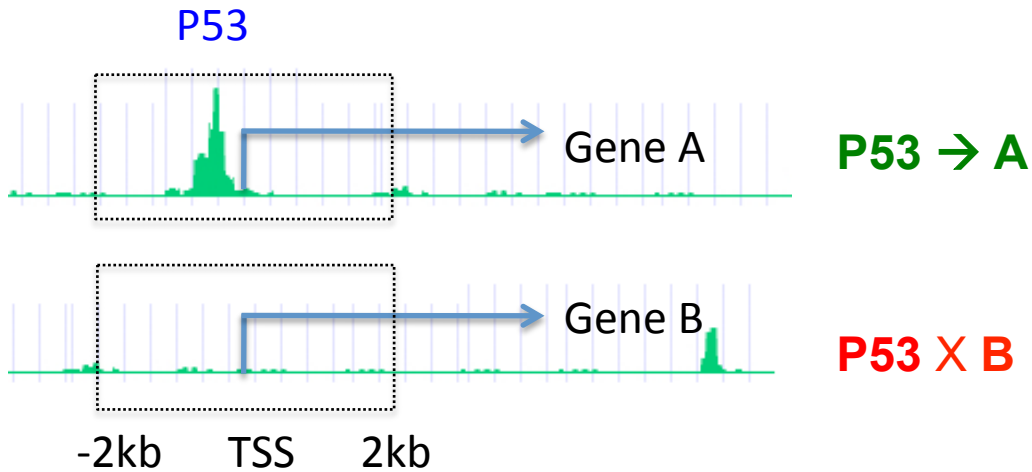


# **TIP: Transcription Factor Target Identification with a Probabilistic Model**

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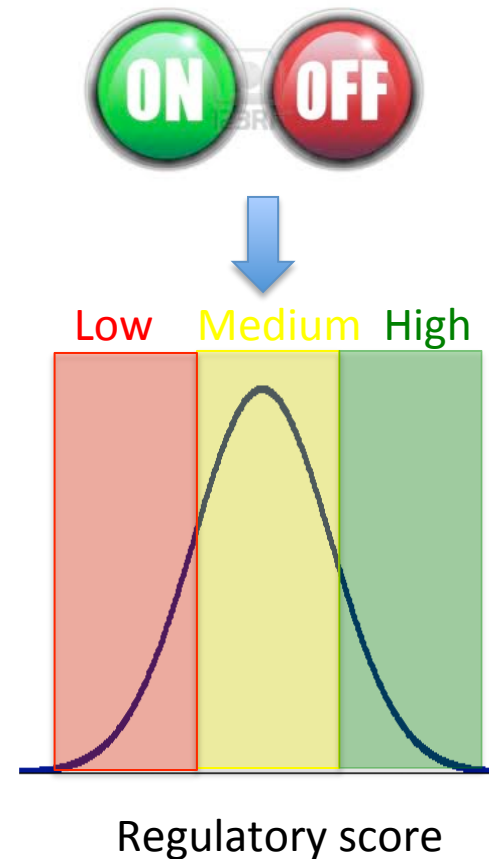
# Conventional way to identify TF target genes



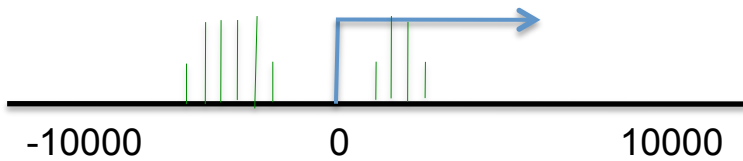
## Limitations of peak-based method:

- sensitive to # binding peaks
- sensitive to cut-off value (1, 2 or 5 kb of TSS?)
- no significance estimation

TF binding and TF → gene regulation is NOT binary but quantitative



# Target Identification with a Probabilistic model (TIP)

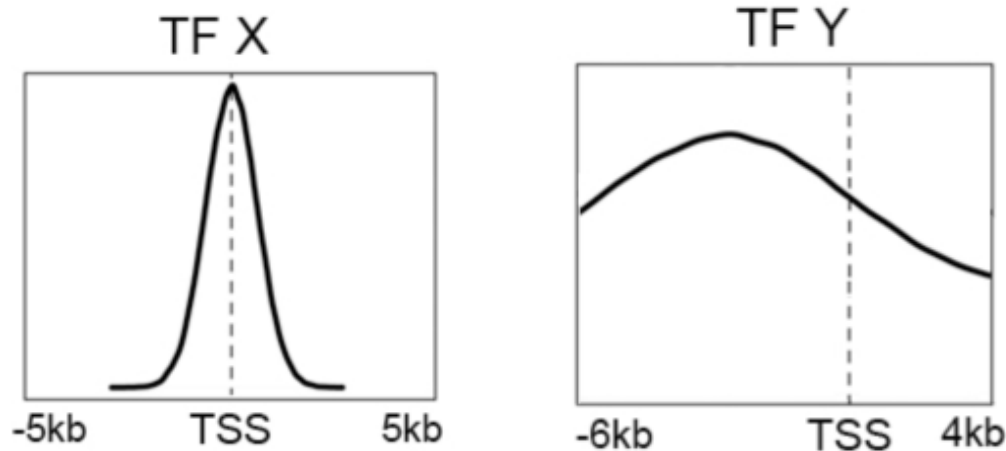


$S_i(\mathbf{g})$ : the signal of a TF at nucleotide  $i$  of gene  $\mathbf{g}$

-- TF binding signal at different position contribute differently

(1) Distance from TSS – binding signal closer to TSS contribute more

(2) Binding preference of a TF – TF specific



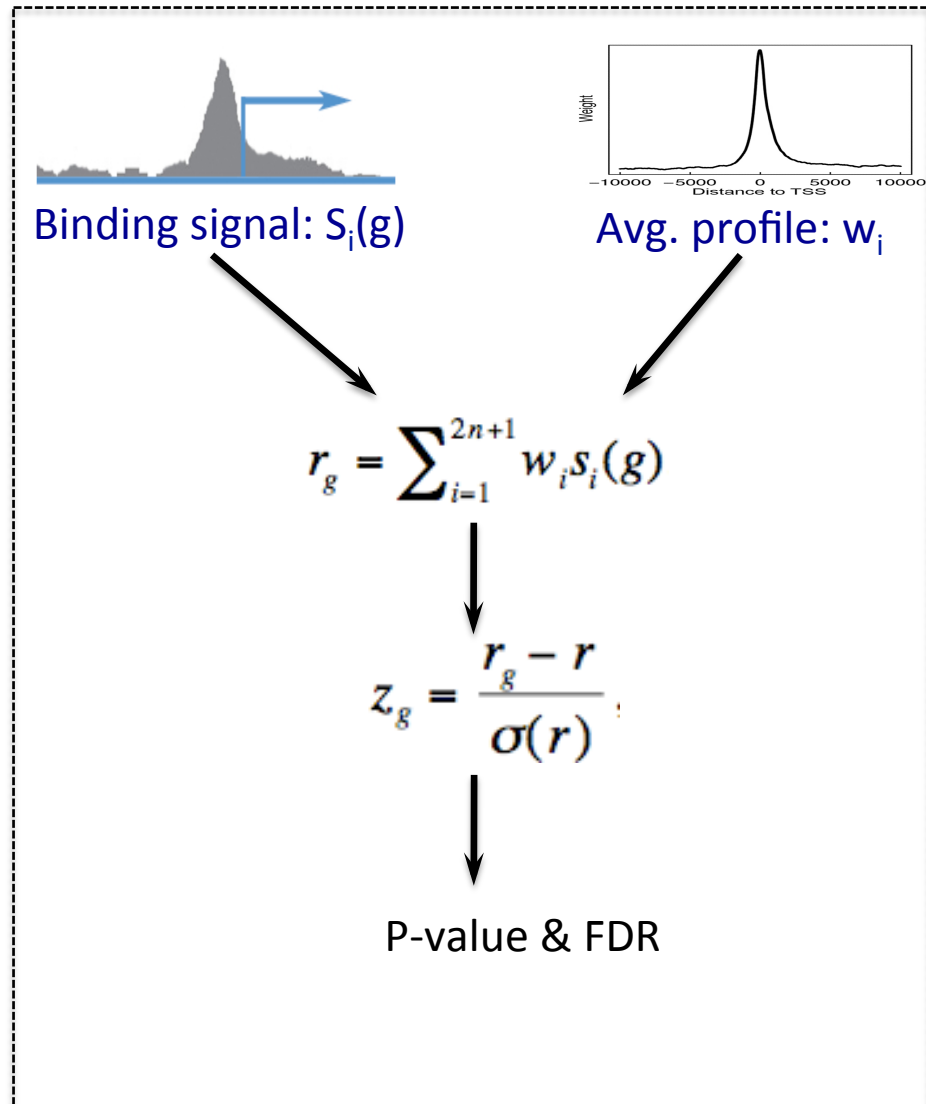
**Avg. binding signal across all genes**  
-- characteristic binding profile of a TF

# Three Steps of TIP

Step 1: Given a ChIP-seq wiggle file, estimate the characteristic binding profile of the TF – the aggregation signal in the 20kb DNA region centering at TSS

Step 2: For each gene, calculate the weighted sum of binding signal of the TF in its 20kb TSS region. – denote it as regulatory score.

Step 3: Normalize the regulatory scores of genes into z-scores, can estimate the P-value and FDR. – result in a ranked target gene list.



# Example 1: STAT4 targets identified by TIP are more down-regulated in STAT4 KO

## Data

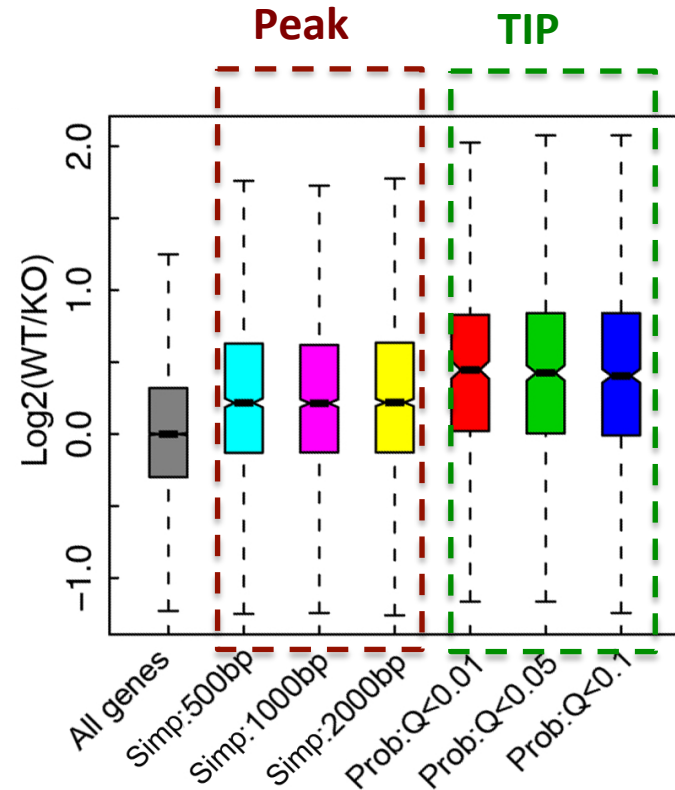
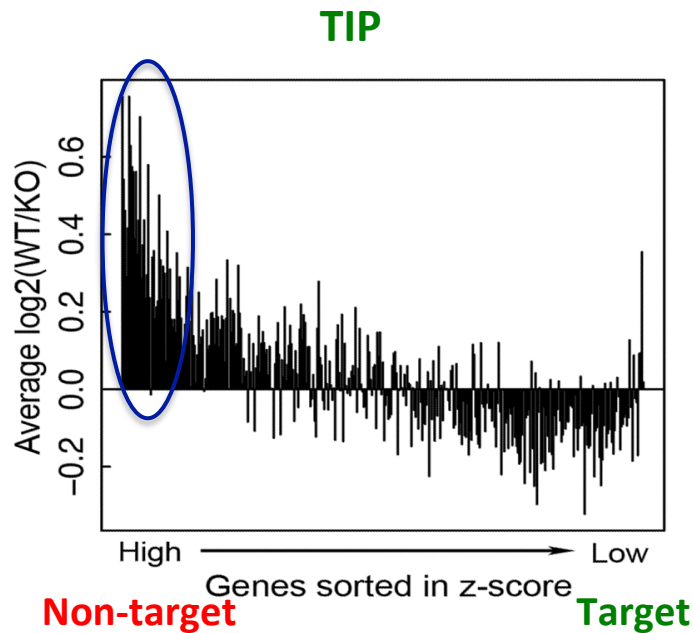
ChIP-seq data:

Stat4 binding in Th1 cell

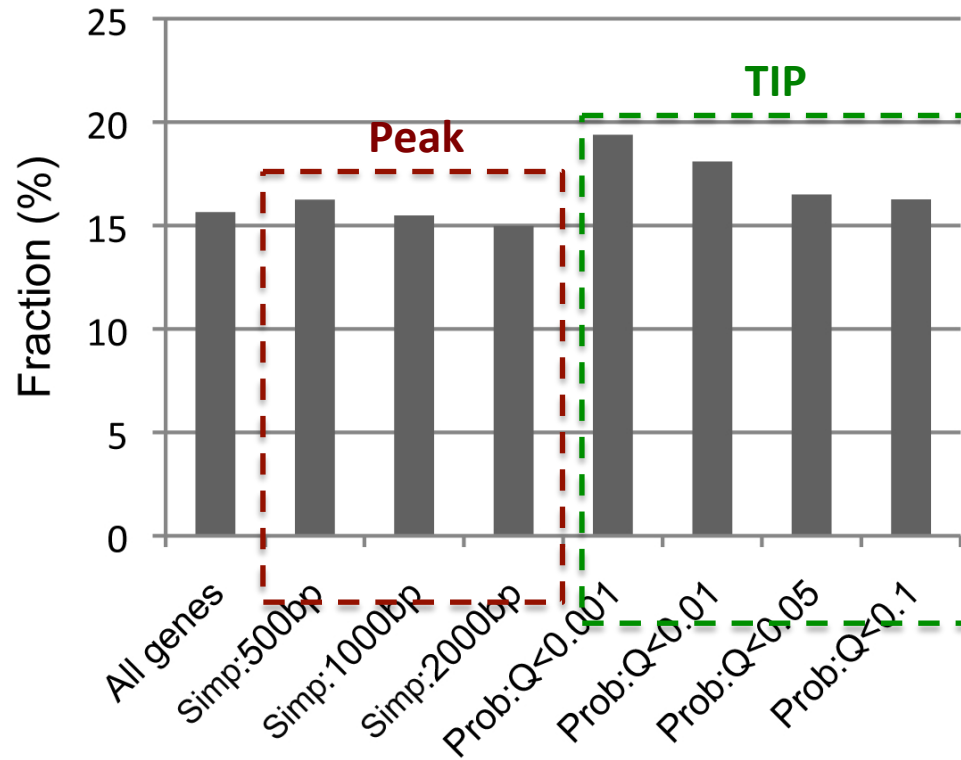
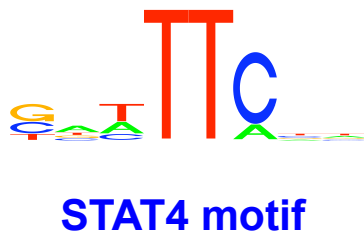
Gene expression:

STAT4KO + WT mice

*Wei et al., 2010*



# Example 1: STAT4 binding motif are more enriched in targets identified by TIP



# Example 2: ER $\alpha$ targets identified by TIP are more responsive to estrogen treatment

## Data

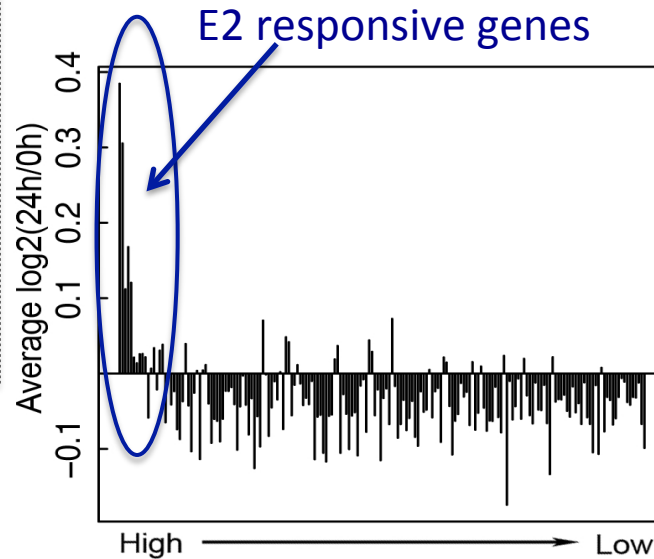
### ChIP-seq data:

ER $\alpha$  binding in MDA-MB-231 cell line

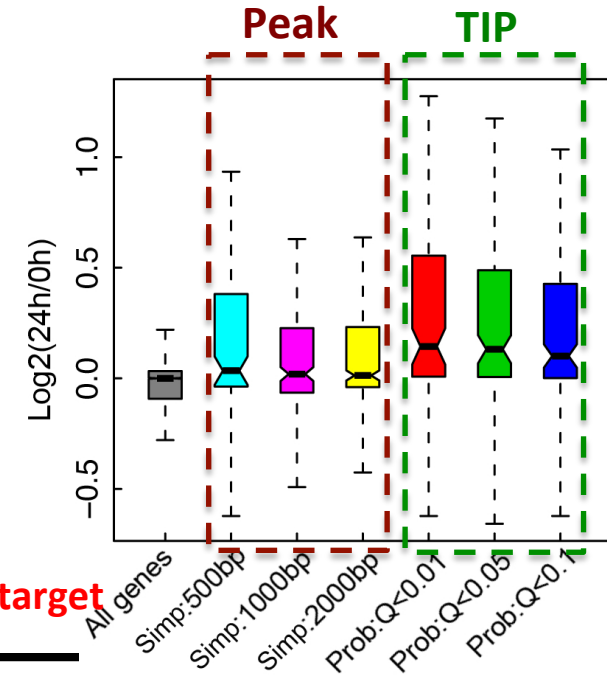
### Gene expression:

before and after E2 treatment

*Stender et al., 2010*



Target Non-target



# target genes

	Cut-off	ER $\alpha$
Peak-based	[-500bp, 500] bp	349
	[-1000, 1000] bp	651
	[-1500, 1500] bp	883
	[-2000, 2000] bp	1,091
TIP	Q<0.001	244
	Q<0.01	312
	Q<0.05	406
	Q<0.1	492

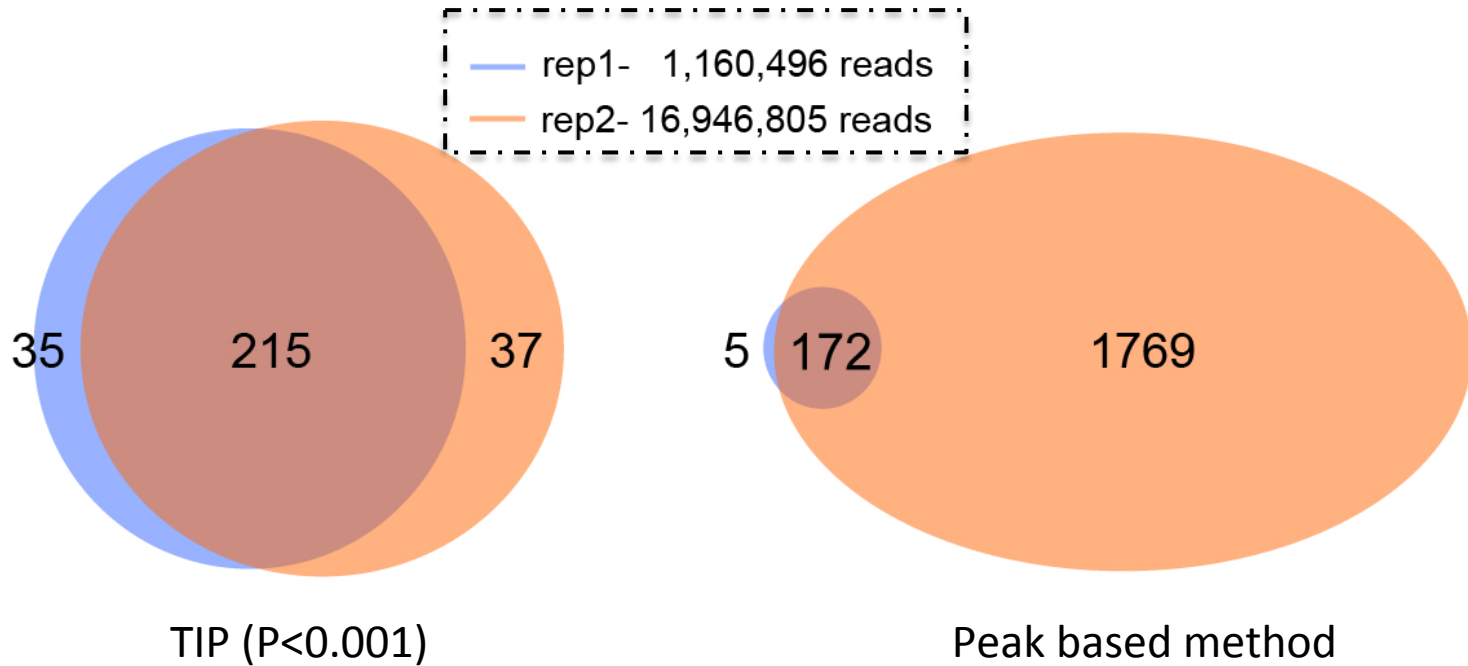
# # TIP target

# target genes

	Cut-off	ER $\alpha$	
Naïve	[-500bp, 500] bp	349	
method	[-1000, 1000] bp	651	
	[-1500, 1500] bp	883	
	[-2000, 2000] bp	1,091	
Probabilistic	Q<0.001	244	
	model	Q<0.01	312
		Q<0.05	406
		Q<0.1	492



# TIP is insensitive to sequencing depth



TCF4: Mokry et al., 2010

**TIP is cost-effective-- do not require high read-depth for target gene identification!**

# Apply TIP to ENCODE data

Table: TFs with both CHIP-seq and knock-down expression data

TF	Efficiency of siRNA (%)	#DE genes	#Up-regulated	#Down-regulated	#Targets (all cell lines)	#Targets that are TF (all cell lines)	#Targets (Peak-based)	#Targets (TIP)	#Distal targets	Level
CTCF	68.1	411	247	164	1406	84	5093	340	0	T
GATA1	73.8	1024	747	277	223	15	204	33	0	M
GATA2	45.3	111	54	57	403	32	703	178	114	M
JUN	30.9	90	27	63	284	21	1056	143	118	T
JUND	48.4	918	788	130	468	35	1956	24	288	M
NFE2	47.1	99	44	55	135	5	234	80	10	B
RAD21	91.7	2015	1100	915	762	61	4439	232	0	T
SMARCB1	63.2	802	403	399	221	13	40	176	32	T
TAL1	54.1	383	308	75	257	16	1307	274	142	M
USF2	57.2	455	382	73	268	10	620	79	165	B
YY1	50.1	437	282	155	450	29	9974	326	228	T

# Target expression in TF knockdown cells

## -- ENCODE data

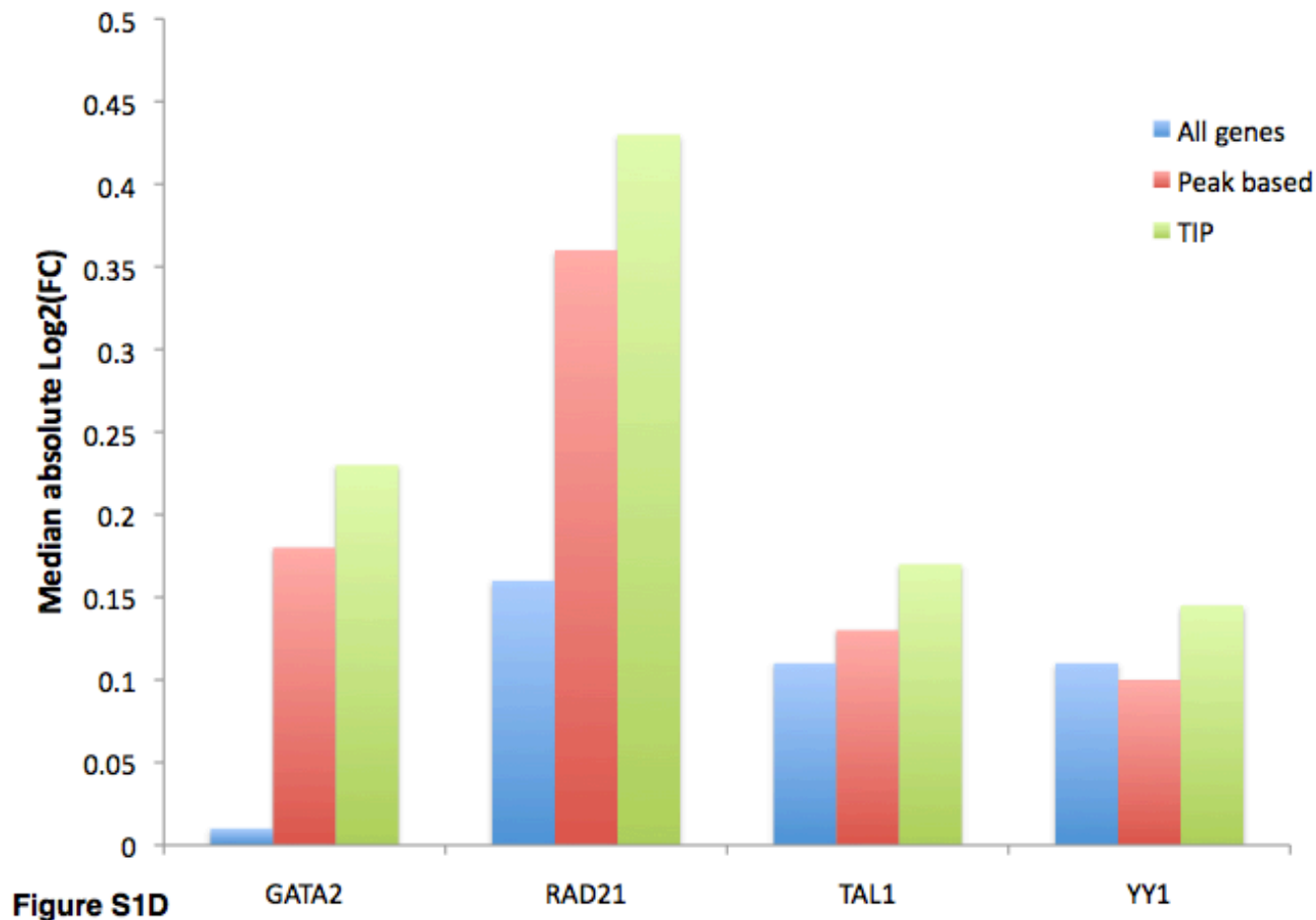
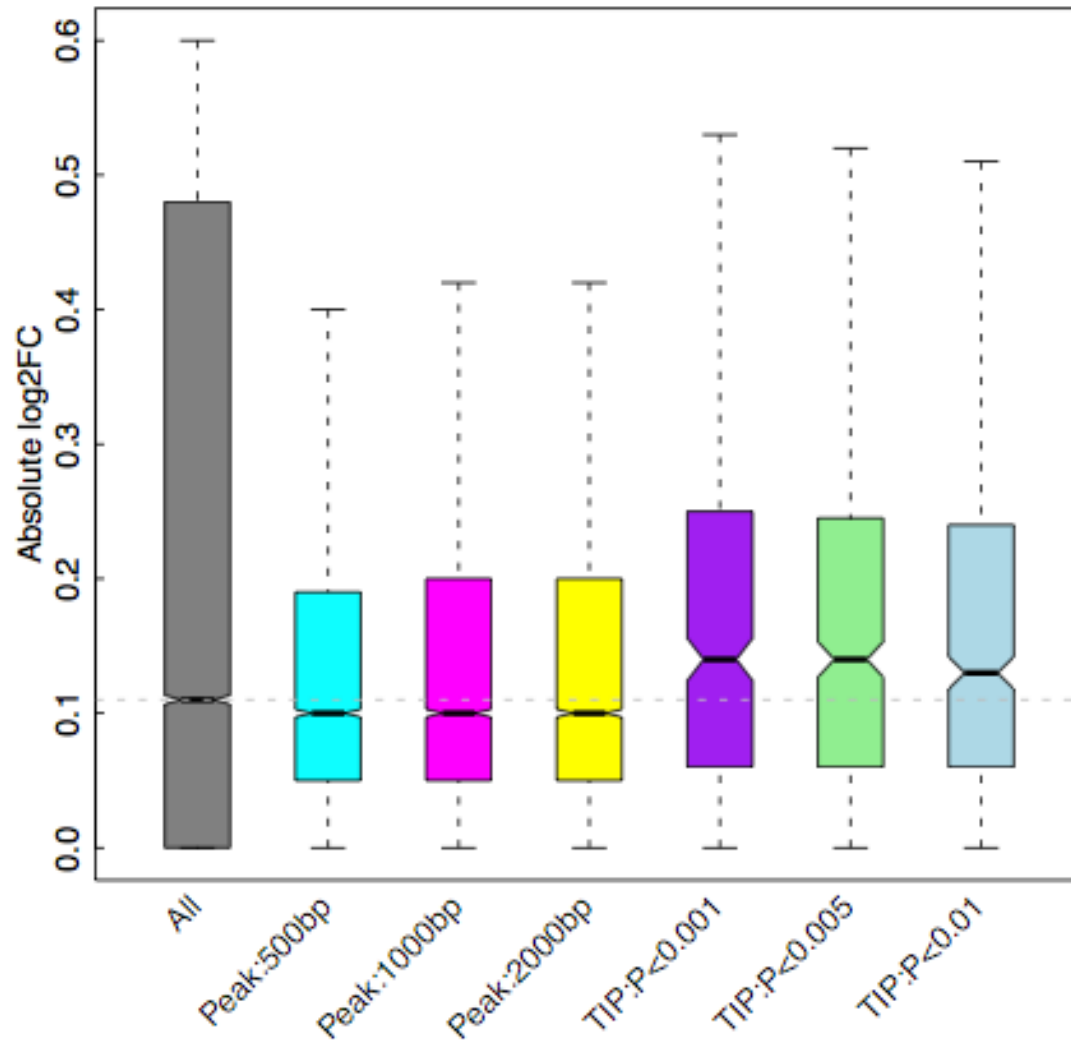


Figure S1D

# Expression changes of GATA1 targets in its knockdown K562 cells



# Extended version of TIP

## **(1) Combine TIP with peak based method**

- for each target gene, list all peaks nearby its TSS
- calculate the relative contribution of each peak to the regulatory score of a target gene

## **(2) Improve P-value calculation.**

- current version is too conservative in estimate significance.
- using 2-component mixture model (non-target and target genes)

## **(3) Build a webserver to implement TIP on line.**