Cell type-specific Transcription Factor Co-associations:

Partner TF Importance & Combinatorial Associations

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Genome-wide Co-association Matrix



TF-centric regulatory modules

Find biclusters of a focus-TF's peaks (e.g. GATA1) that co-associate with distinct combinations of other TFs



GATA1-centric regulatory modules



Core Module

GATA1-centric regulatory modules



GATA1-centric regulatory modules

Focus TF: GATA1 in K562



Defining Proximity Regions



Classification/Regression models for associations





 Quantify importance of individual associated TFs and frequently occuring sets of TFs – by Random Forest/Boosted trees/RuleFit

Types of positive/negative sets

- Classification scenarios
 - True association matrix (+) vs Shuffled coassociation matrix that breaks associations (-)
 - True association matrix (+) vs association matrix for peaks present in other cell-lines but NOT in target cell-line (-)
 - TSS-Proximal (+) vs TSS-Distal sites (-)
 - Peaks near highly expressed genes (+) vs. Peaks near low expressed genes (-)
- Regression scenarios
 - True association matrix -> focus TF binding strength
 - True association matrix -> gene expression

Discriminative/Regression models for associations

Matrix of Rank-normalized signals for each TF (after removing focus TF signals)

Shuffle rows in each column (partner TF) independently – to break all TF association signals

Classification true matrix from shuffled one - Random Forest

Define new modelbased distance matrix for all focus-TF target locations for biclustering



Feature Importance (TF association importance)

K562 Gata1 Partner TF OOBError Plot



K562 Gata1 Partner TF OOB Feature Importance



Change in Order of Partner TF Importance in Different Bins

based on Feature Importance scores of Random Forest Bagging



- Core-partners get lower score in distant bins
- Long-range partners get higher score in distant bins

- **OA** Actual Window (Peak boundaries width of peak)
- **OO** 500nt flanking peak summit (1000nt wide)
- OT 1250nt flanking peak summit (2500nt wide)
- OF 2500nt flanking peak summit (5000nt wide)

Jund Partner TF Feature Importance – Effect of data quality



c-fos (Snyder) c-fos (Uchicago)

Why *fos* doesn't come up..

Partner	Cfos	Cfos	Fosl2	Cfos
	K562	HelaS3	HepG2	GM12878
OA	299	5444	8561	4
00	330	5703	9119	4
OT	354	6294	10179	7
OF	374	7035	11436	14
nJund_Peaks	759	26074	22441	1715
nPartner_Peaks	5810	6236	19714	1744

Results can be made more robust to data quality by using predicted binding sites (using integrative TF binding models)

Combinatorial Coassociation



- Does signal from X independently do better than that from X and Y?
- Does signal from X and Y do better than X and Z?

Frequent TF-set Mining

Gatal <- Egata2 Gata2 (70.3, 99.9) Gatal <- Ccnt2 Pol2 (70.5, 99.9) Gatal <- Ccnt2 Gata2 (70.6, 99.9) Gatal <- Tall Cmyc (70.4, 99.9) Gatal <- Tall Cmyc (70.4, 99.9) Gatal <- Nrsf Pol2 (60.7, 99.9) Gatal <- Nrsf Gata2 (60.8, 99.9) Gatal <- Nrsf Pol2 Gata2 (60.0, 99.9) Gatal <- Nrsf Pol2 Gata2 (60.0, 99.9) Gatal <- Ccnt2 Tall Cmyc (60.1, 99.9)

```
Gata1 <- Tal1 Cmyc (70.4, 99.9)
Gata1 <- Taf1 Nrsf Max Egata2 Tal1 Cmyc (20.2, 99.8)
Gata1 <- Efos Cjun P300f4 Hdac2 Nrsf Tal1 Cmyc Pol2 Gata2 (5.8, 100.0)</pre>
```

We are also using TF item sets from

- hierarchical clustering of columns of TF-specific association matrix
- Can also use significant global associations (post GSC)

Regulatory network construction



- Feature importance can determine edge strength for each TF
- This gives bidirectional asymmetric network
- Highlight asymmetric edges & see if that is because of quality or master regulator vs cofactor effect
- Superimpose protein interaction data/annotations
- We can then extract cliques with strong edges as modules