Why the expression levels of some known drug target genes are not regulated by the drug? (expression: balance between transcription & degradation)

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## **Conflict of interest disclosure**

### None

**Committee of Scientific Affairs** 



# Anti-diabetic rosiglitazone (Rosi) is an agonist of PPARγ



Ahmadian et al. 2013 Nature Medicine

### Identifying PPARy target genes using ChIP-seq



# Gene expression levels (RNAseq) were not changed by rosi for Lipe



### Why?

Lipe is not a target of PPARγ. But log2FC after PPARγ KD: -3.2
 Lipe is upregulated but post-transcriptionally down regulated

# The balance between mRNA synthesis and decay determines mRNA expression levels



- Earliest changes can identity primary target genes rather than secondary targets
- Unbiased detection of novel transcripts, including non-coding RNAs
- detects enhancer RNAs

Core et al, Science, 2008

Wash / Elute

### **Up Regulation of Transcription Due to Rosi**



#### Up and Down Regulation of Transcription Due to Rosi



Gene expression levels (RNAseq) were not changed by rosi but the transcriptional levels were changed for Lipe



### **Nascent Transcription Changes Precedes mRNA Regulation**

Log2-FoldChange in Gene Body





GROseq Step et al. (2014) G&D

### **Nascent Transcription Changes Precedes mRNA Regulation**



GROseq Step et al. (2014) G&D

### **Nascent Transcription Change Precedes mRNA Regulation**



### Calculation of mRNA half-life using a dynamic model



- Groseq (2 replicates)
  - t: 0, 10, 30, 60, 180 min after rosi treatment
- RNAseq (2 replicates)
  - t: 0, 10, 30, 60, 180 min after rosi treatment

### Dynamic models for stability



### Modeling the signals in Fourier domain



Transfer function in the Fourier Domain

### Modeling the signals in Fourier domain



### Modeling the signals in Fourier domain



### Predicted mRNA half-lives



### Modeling variable time constant

$$\frac{dR(t)}{dt} + \frac{R(t)}{T} = KG(t) \Longrightarrow \frac{dR(t)}{dt} + \frac{R(t)}{T(t)} = K(t)G(t)$$



-3 -1 1 3

### Genes with reduced half lives are associated with metabolic process & many of them are targets of PPARy





### **Examples:** Lipe



Minutes



Lipe: hormone-sensitive lipase Pnpla2: adipose triglyceride lipase : related with lipolytic function

#### Majority of genes induced early by rosi are belong the C1

#### Log2-FoldChange in Gene Body



### Among the 567 genes, 320 are belong to C1

| GO Biolog             | gical Process   |
|-----------------------|---|
| ·                     | small molecule catabolic process (3e-14)<br>carboxylic acid catabolic process (2e-14)<br>fatty acid catabolic process (3e-12)     |
| Molecular             | function  |
|                       | coenzyme binding (9.6e-11)  |
|                       | NAD binding (0.00017), fatty-acyl-coA   |
| KEGG                  |   |
|                       | fatty acid degradation (4e-8)   |
|                       | fatty acid metabolism (1e-7)  |
|                       |   |
| GO Biological Process |   |
| -                     | positive regulation of glucose transport (0.007)<br>regulation of glucose import (0.02)<br>fatty acid biosynthetic process (0.09) |
| Molecular fu          | unction   |
|                       | protein kinase A catalytic subunit binding (0.2)  |
| KEGG                  |   |
|                       | PPAR signaling nathway (0.004)  |

PPAR signaling pathway (0.004) insulin signaling pathway (0.02)

### Degradation model I



miRNA mediated degradation

#### C1 is depleted with AU-rich elements (AREs) compared with C2 and C3







Hypothesize: Genes with abrupt induction will lose its stability. But HuR stabilizes its target genes.



# siHuR in the mature adipocytes block the increase of expression only in Tomm70a





Non-C1: increase in expression

### E2 treatment in breast cancer cell lines (MCF7)



GROseq: from Lee Kraus Lab. (Hah et al. 2011 Cell) RNAseq : from Rattray lab (Honkela et al. 2015, PNAS)

# Why?

RNA homeostasis Similar to protein homeostasis => The changes in expression levels do not guarantee the changes in the protein levels.

### Summary



• Not all PPAR $\gamma$  binding sites are responsive to its agonist

### Pancreatic tissue



#### **Pancreatic Tissue**

### Single cell RNAseq for mouse pancreatic epithelial cells at e13.5

Total 77 single cells.

At e13.5, the pancreatic progenitors express endocrine progenitors of each lineage



#### Kim and MacDonald (2

### Clustering scRNAseq from pancreas



0•

Stanuescu et al., (2017) Physiol. Genomics

# Clustering with RNAseq data identifies cells committed to α cells

α
 β
 Single cells



α, β cells from Benneret al. BMC Genomics2015

# Clustering with RNAseq data identifies cells committed to α cells

• α • β • Single cells



SIc38a5 correlates with Gcg and is highly observed in the  $\alpha$  cell committed cells



SLC38A5 (Snat5) is uniquely and highly expressed in the developing pancreas until late gestation



Stanescu et al.

### Loss of SLC38A5 impairs Gcg expression throughout midgestation





# Loss of SLC38A5 leads to decreased numbers of MafB progenitors Pdx1 MafB

e13.5







#### ARTICLE

#### Amino Acid Transporter Slc38a5 Controls Glucagon Receptor Inhibition-Induced Pancreatic a Cell Hyperplasia in Mice

Jinrang Kim, Haruka Okamoto, ZhiJiang Huang, Guillermo Anguiano, Shiuhwei Chen, Qing Liu, Katie Cavino, Yurong Xin, Erqian Na, Rachid Hamid, Joseph Lee, Brian Zambrowicz, Roger Unger, Andrew J. Murphy, Yan Xu, George D. Yancopoulos, Wen-hong Li<sup>4</sup>, Jesper Gromada<sup>4,5</sup>, Senior author

<sup>5</sup> Lead Contact



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# Bimodal distribution is observed when 2 cell types are mixed in scRNAseq



# scRNAseq data from pancreatic islet show bimodal patterns for key genes



### A new clustering algorithm using bimodal distribution



### A top-down clustering for human pancreatic islet data



### **Evaluation of the top-down clustering**

Pyramidal S1 Endothelial-Mural Oligodendrocytes Astrocyte-Ependymal Interneurons Microglia Pyramidal CA1 Alpha Delta Beta PP Duct Acinar Gold standard **Bimodal clustering** tSNE+DBSCAN **Hierarchical clustering** SC3 0.8 0.9 ARI 0 n Bimodal BS AN Hierarchical SC3 Bimodal Break DBS AN chical B درج Zeisel (n=3005) Wang et al (n=1083) Mouse cortex Human pancreas

## SHARP for Park (ARI)



## SHARP for Park (Time)



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