Single cell analysis of midbrain dopamine neurons - Thomas Perlmann lab (LICR/KI)

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Midbrain dopaminergic neurons

Less than 1% of the neurons in a brain Important for:

- motivation & reward systems
- motor behavior
- working memory

Parkinsons Disease – death of DA

neurons in Substantia Nigra Successful grafting with ES derived DA neurons in 80-90s Animal trials with iPS derived DA neurons ongoing





(Björklund et al. 2014)





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Pitx3 eGFP knock in mice labels mature DA neurons.

FACS sorted eGFP positive cells from midbrains of two mouse strains.

All libraries prepared in the Perlmann lab

- eGFP Pitx3 heterozygote mice at:
 - Embryo E13.5, E15.5, E18.5
 - Juvenile P1, P7
 - Adult P90
- eGFP Pitx3 homozygote (Pitx3 double KO) mice at:
 - E13.5
 - P1
- Total **1395** SmartSeq2 libraries after quality control











Quality Control



Quality Control – removal of non Pitx3 cells









Graph based clustering to remove non-DA neurons





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Develop. Stage	Number of Cells	
E13.5	135	
E15.5	140	
E18.5	181	
P1	258	
P7	80	
P90	312	
E13.5H	141	
P1H	148	



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Analysis of Pitx3 Heterozygote cells

- Main aims
 - Find possible subgroups of cells
 - Find marker genes for the groups
 - Understand development of the lineage







Selection of variable genes

- Using ERCC spike-ins and Brennecke method
 - 453 variable genes using all ages
 - 300 800 variable genes using individual ages



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Initial PCA – mainly time separation



Clear separation by 3 maturation stages: early embryo (E13/E15), perinatal (E18-P7) and adult (P90)







Genes along pseudotime



PCA



Main problem: All cells are very similar – same celltype. Hard to identify subgroups since the signal is weak relative to developmental time, noise and other sources of variation.

Top loadings of PC2, PC3 seems to define our lineages: Slc6a3 (Dat), Nxph4, Gad2 and Vip

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t-distributed stochastic neighbor embedding (tSNE) and igraph



Variable genes at each stage should reflect lineage differences more than developmental differences







Classifying cells into subgroups









Classifying cells into subgroups



Clusters defined by infomap community detection.

Each cluster classified into a subgroup based on proportion of cells exclusively expressing one of our markers.

Manual definition of Gad2+ and Nxph4+ lineage into Th high/low.

Subgroups		Nxph4 Dat ^{low}	N-Dat ^{low}
		Nxph4 Th Dat ^{low}	NT-Dat ^{low}
	•	Gad2 Dat ^{low}	G-Dat ^{low}
		Gad2 Th Dat ^{łow}	GT-Dat ^{low}
	•	Th Dat ^{high}	T-Dat ^{high}
	•	Aldh1a1 Th Dat ^{high}	AT-Dat ^{high}
		Vip Th Dat ^{high}	VT-Dat ^{high}
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Marker gene discovery

Differential expression between the clusters both across all stages and at one developomental stage at a time was done using SAMseq.



Validations with immunohistochemistry



Pitx3 double KO cells



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Padlock probe - in situ sequencing

- Method developed at Mats Nilsson lab (SU)
- Selected 49 genes among our differential expression data:
 - Fairly high expression
 - Markers for subgroups
 - Some general markers to exclude other celltypes
- Cells defined as expanded area around nuclei (from Dapi staining)





Filter in situ data

E_Tissue1 – 709 cells



G_Tissue2 – 755 cells



F_Tissue8 – 828 cells



H_Tissue5 – 49 cells



- Filtering of *in situ* cells for DA neurons:
 - Cells in selected regions
 - Require epression of Pitx3, EGFP or Th.
- Keep only cells with at least 3 subgroup marker genes.
- Out of around 60K cells per section only 2141 cells kept.

Predict subgroup using Random Forest

- Convert both scRNAseq data and in situ data to rank based matrices.
- Train random forest with scRNA-seq data using cluster membership
- Predict subgroup for *in situ* cells
- Several rounds of training prediction only keep consistent predictions.







Conclusions

- Most extensive classification of subgroups of midbrain DA neuron subtypes to date.
- Several verification experiments with antibody staining (also with human tissue), in situ sequencing, retrograde labeling of innervation.
- Main issue in the data was that time separation was much stronger than separation of the subgroups.









Shiny apps

- http://shiny.rstudio.com/
- Interactive R applications used to present the data
- <u>http://rshiny.nbis.se/shiny-server-apps/shiny-apps-scrnaseq/</u> <u>perlmannlab_mouseDA</u>







