

A Proteomic Survey of the Postnatal Human Brain

Becky Carlyle, Associate Research Scientist

Nairn Lab

NIDA proteomics 06/02/17



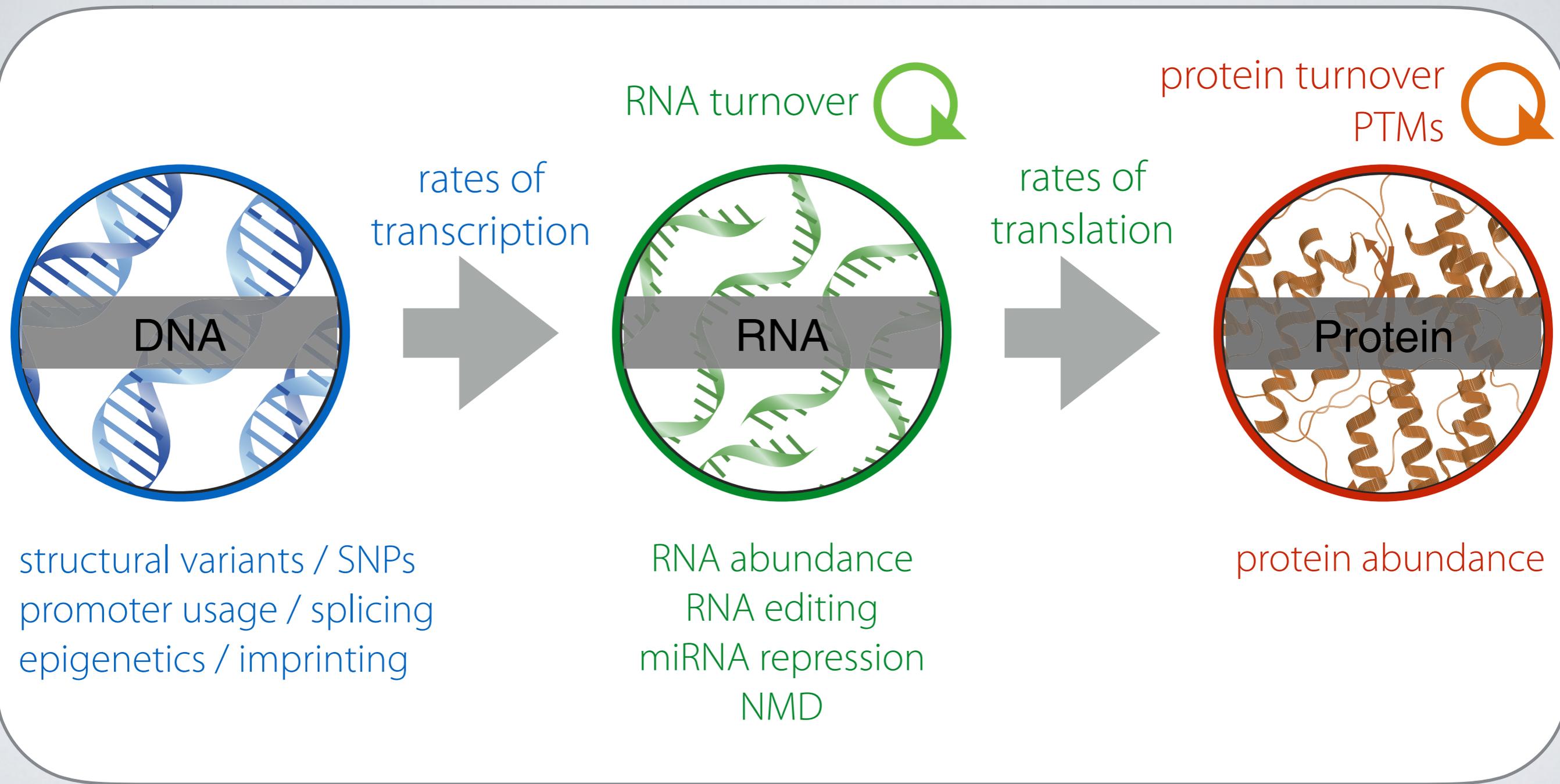
Talk outline

- Why proteomics?
- BrainSpan/psychENCODE project samples
- Proteomic methods
- Results
- Conclusions

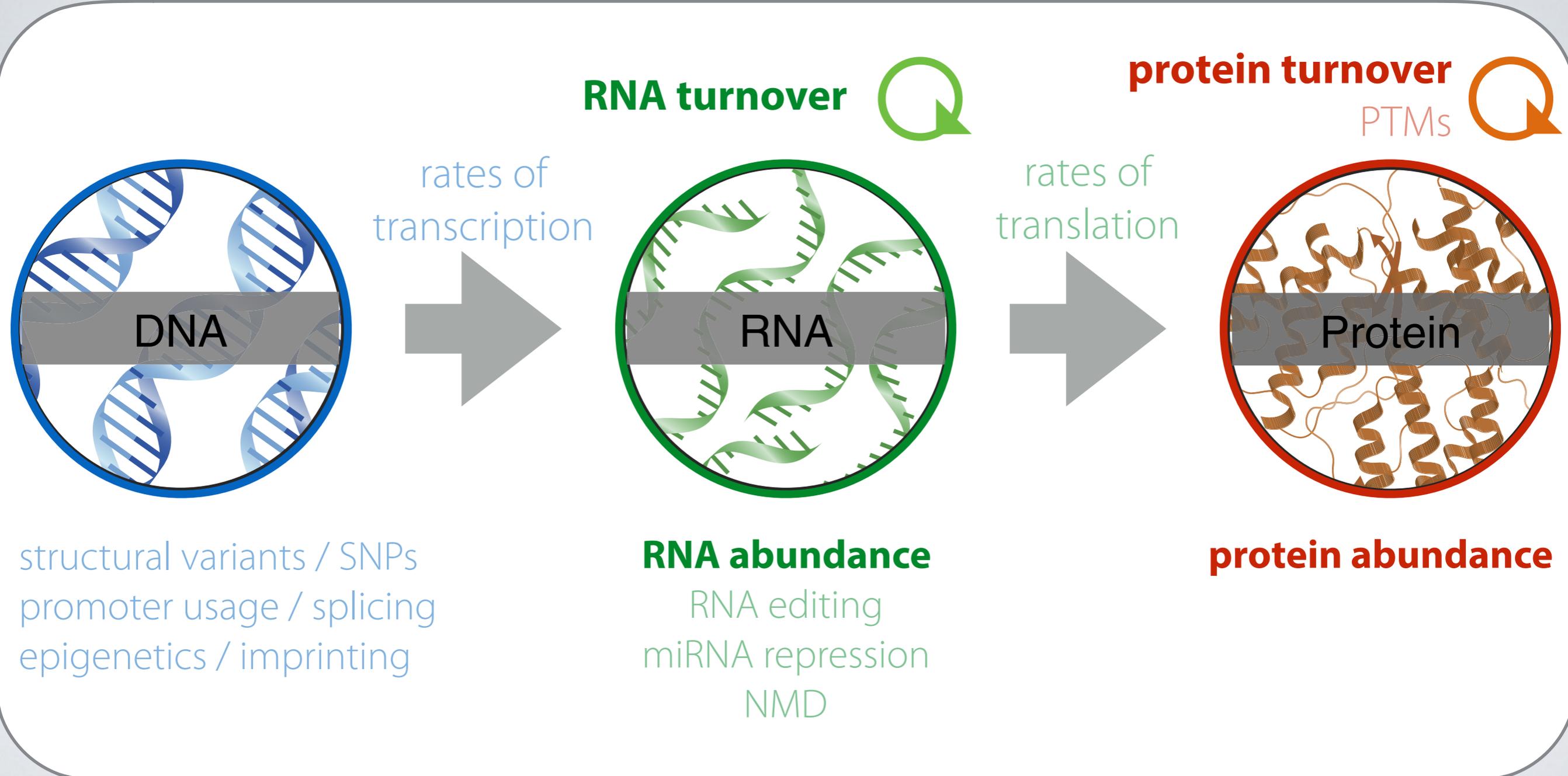
Talk outline

- **Why proteomics?**
- BrainSpan/psychENCODE project samples
- Proteomic methods
- Results
- Conclusions

Protein abundance is the final output of the central dogma

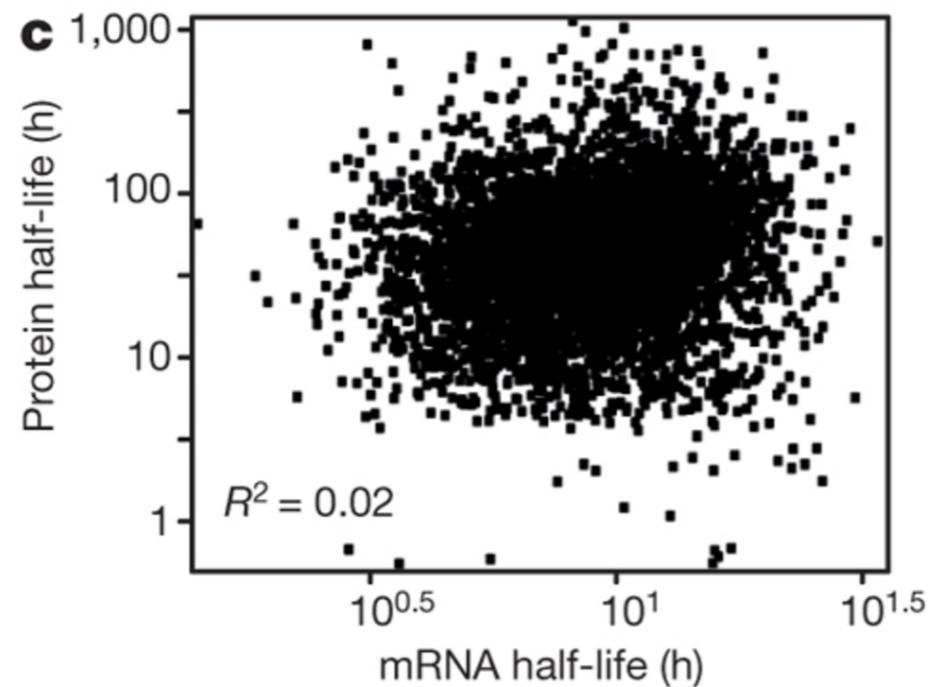
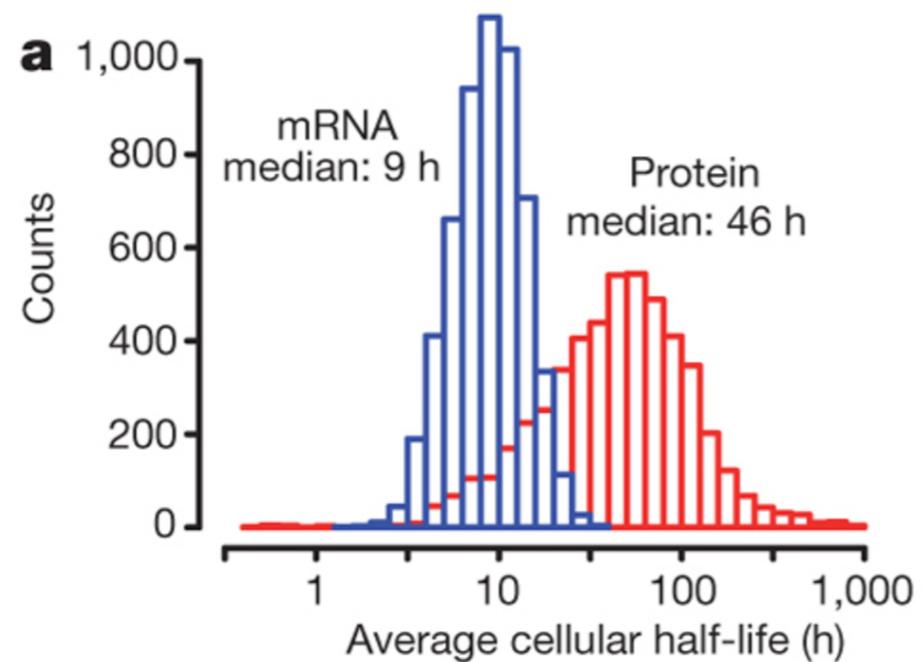


Protein abundance is the final output of the central dogma

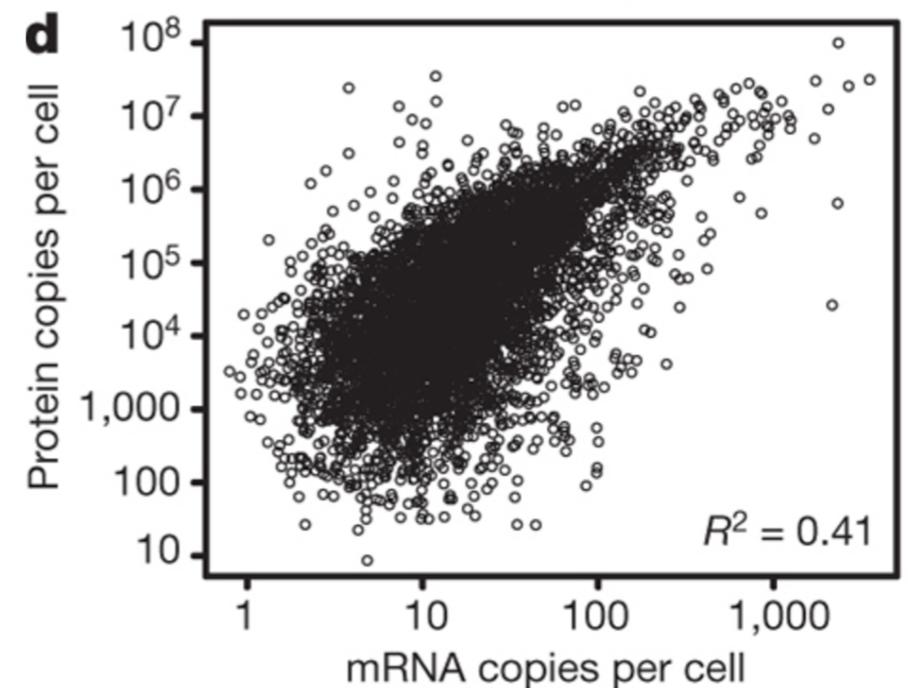
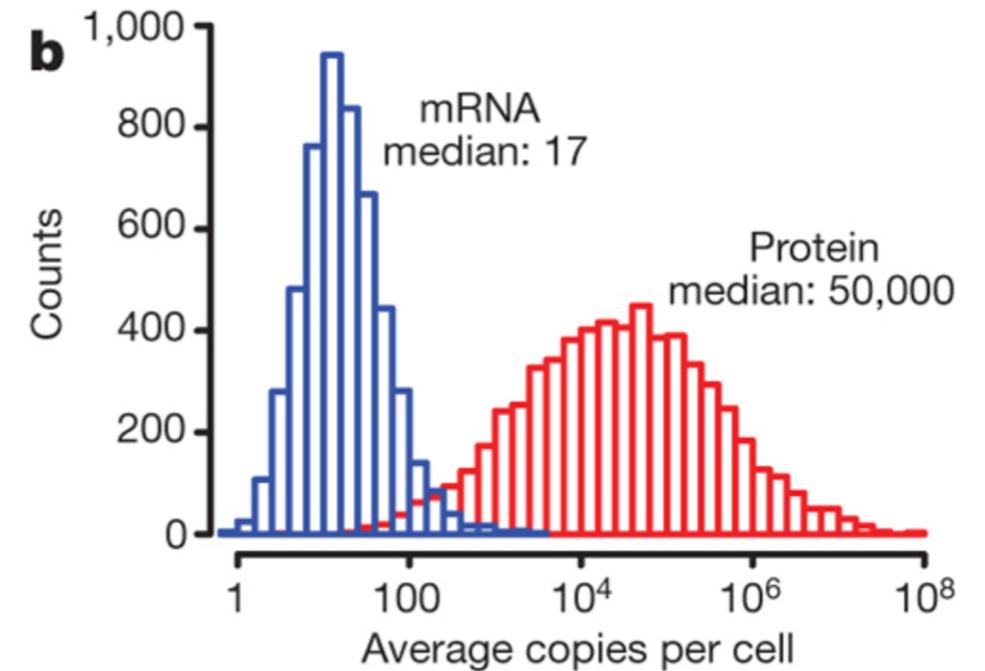


Protein and mRNA may correlate poorly

In both *turnover* rates...



... and *abundance*



Human brain - mRNA abundance

a NCX, HIP, AMY, STR, MD, CBC

15,132 (86.1%) surveyed genes expressed

90.8%

135
(0.9%) 10,594
(70%) 3,018
(19.9%)

■ Spatial DEX ■ Temporal DEX

NCX areas

14,375 (81.8%) surveyed genes expressed

85.5%

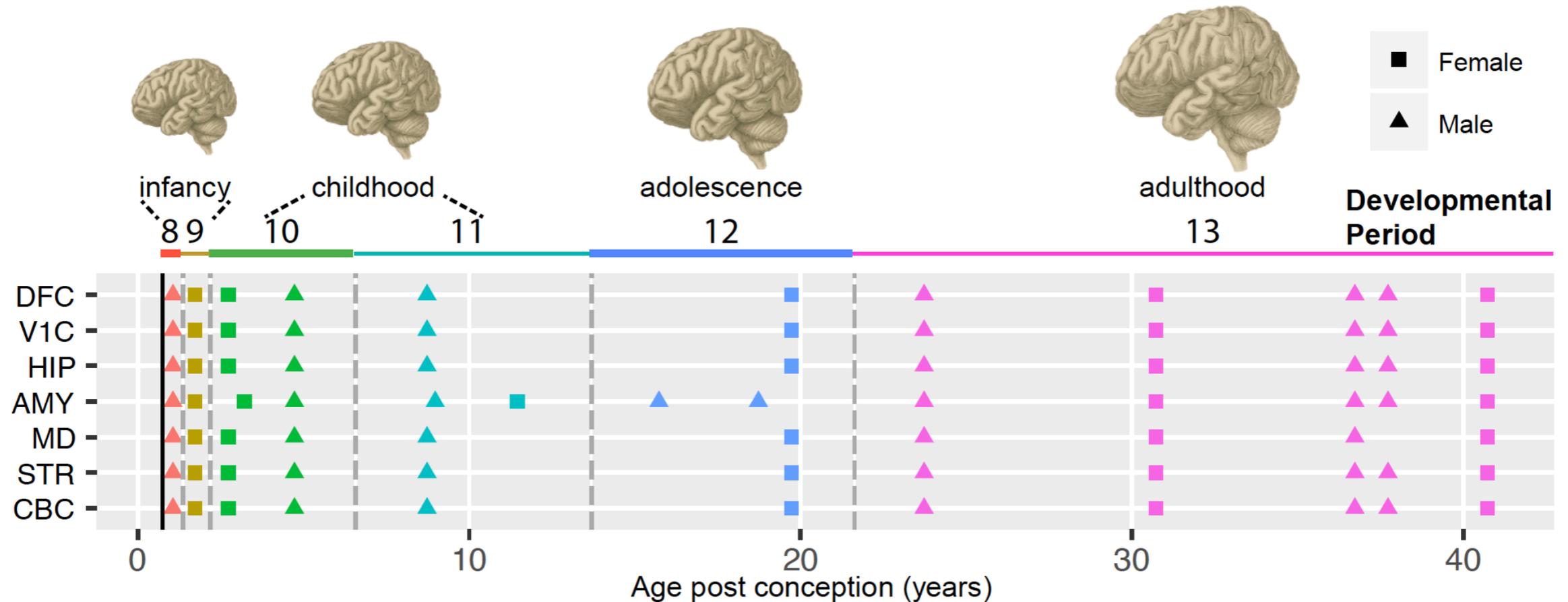
23
(0.2%) 3,445
(23.9%) 8,823
(61.4%)

- The vast majority of mRNA variation between regions comes during development
- Fewer genes differ in abundance between cortical regions in the adult
- Given the wider range in protein half lives and abundance, there may be differences in protein between these regions that are not reflected in mRNA

Talk outline

- Why proteomics?
- **BrainSpan/psychENCODE project samples**
- Proteomic methods
- Results
- Conclusions

BrainSpan/psychENCODE project samples



Same subjects used in BrainSpan for RNA-seq

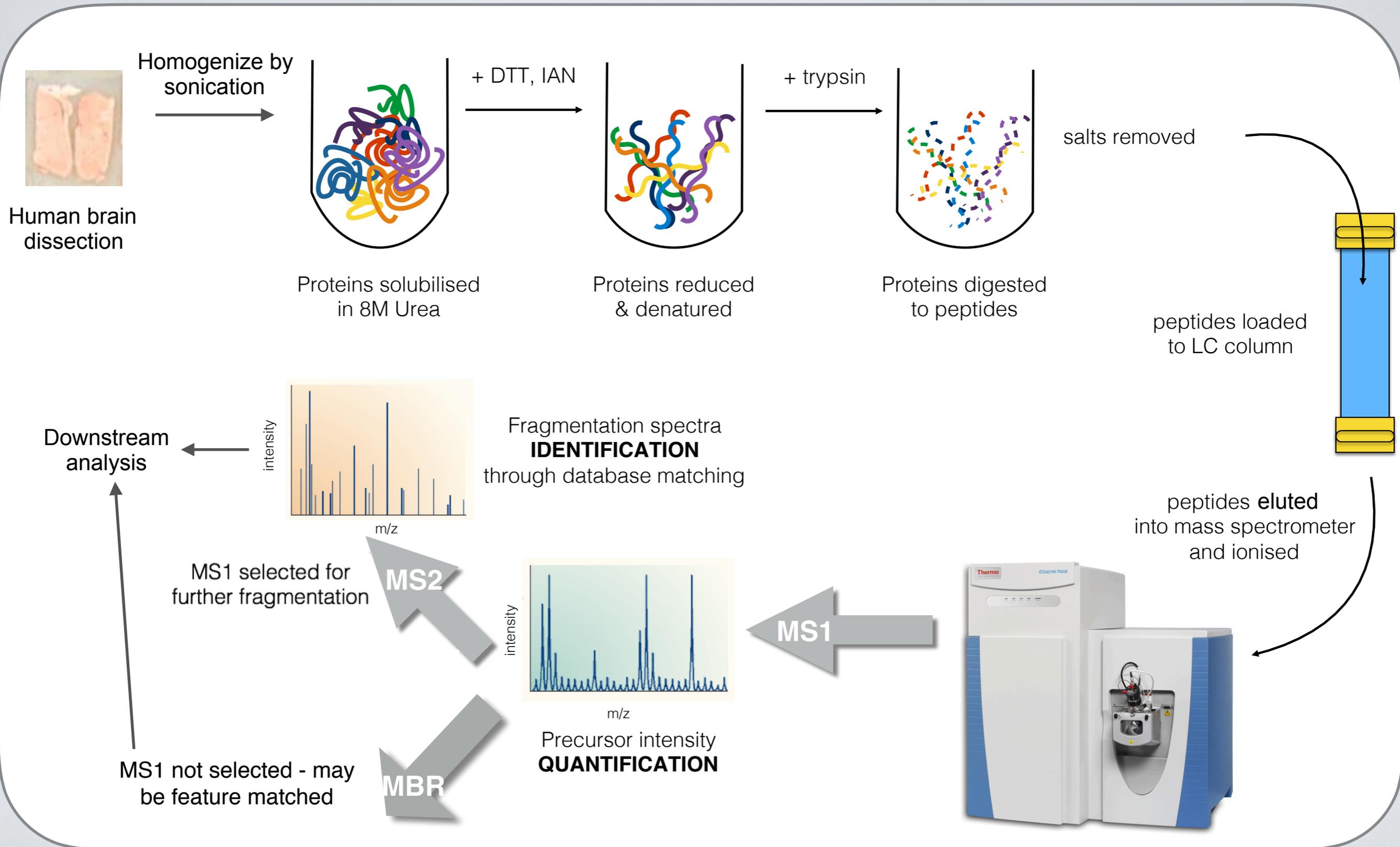
~6 subjects spanning postnatal development

5 adult subjects, 7 brain regions

Talk outline

- Why proteomics?
- BrainSpan/psychENCODE project samples
- **Proteomic methods**
- Results
- Conclusions

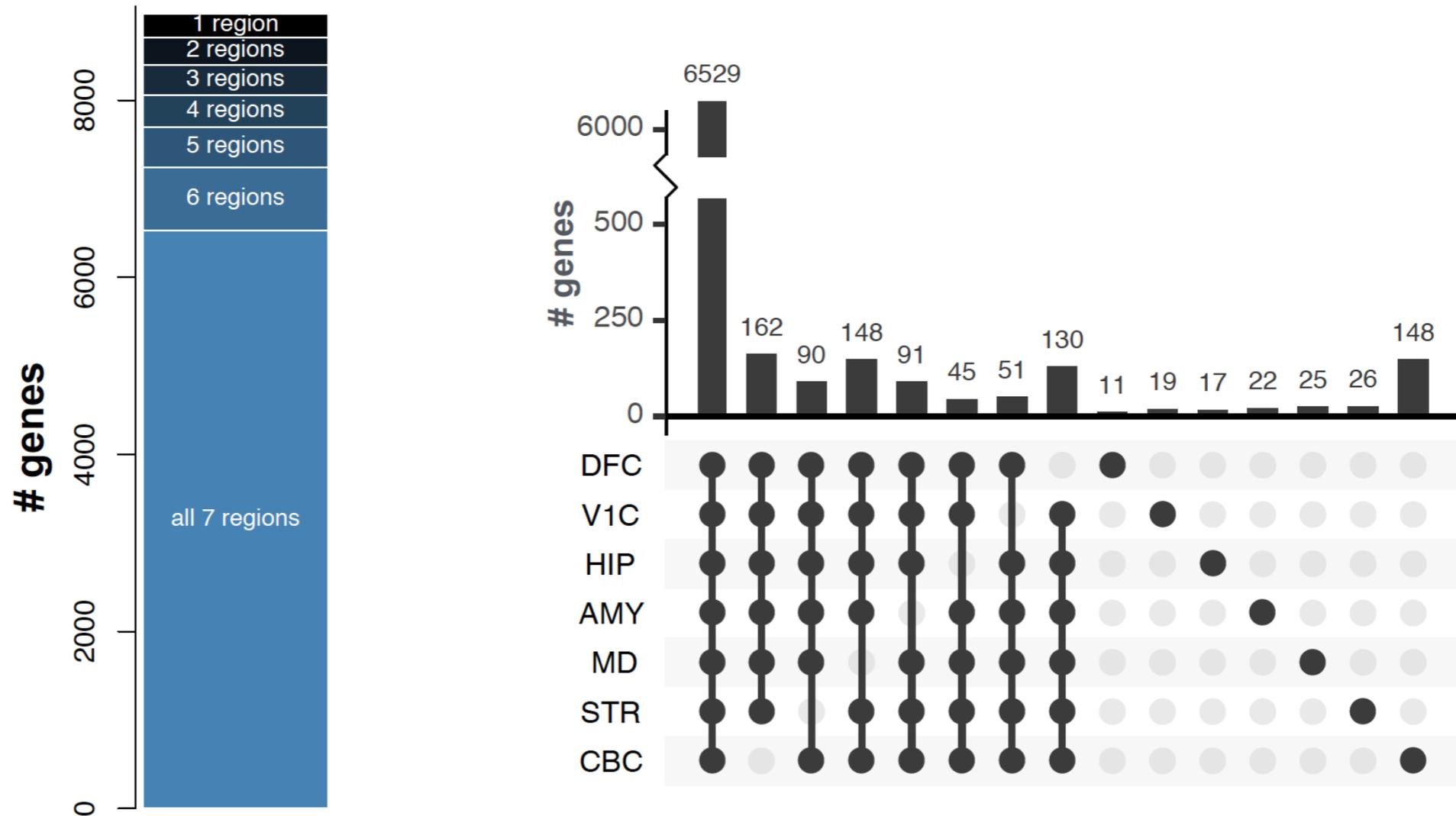
Simple proteomic workflow



Talk outline

- Why proteomics?
- BrainSpan/psychENCODE project samples
- Proteomic methods
- **Results**
- Conclusions

Fractionated regions

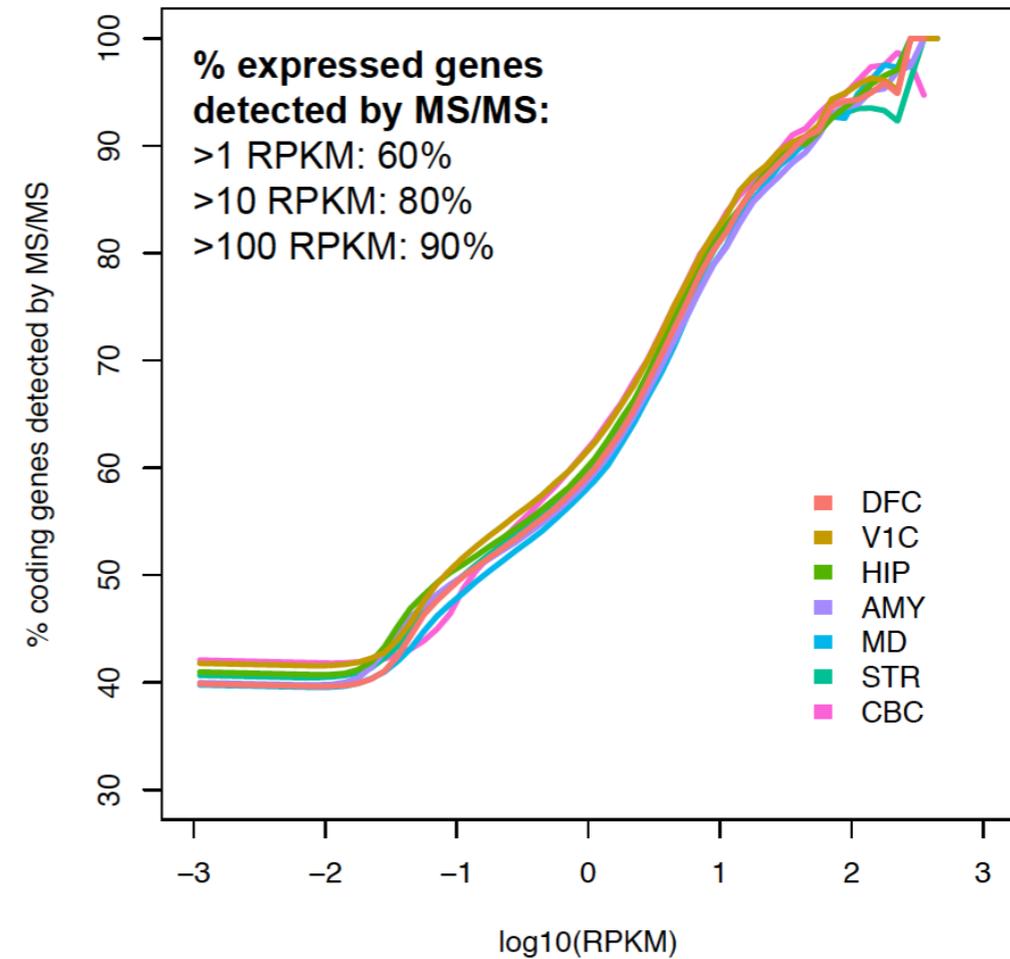
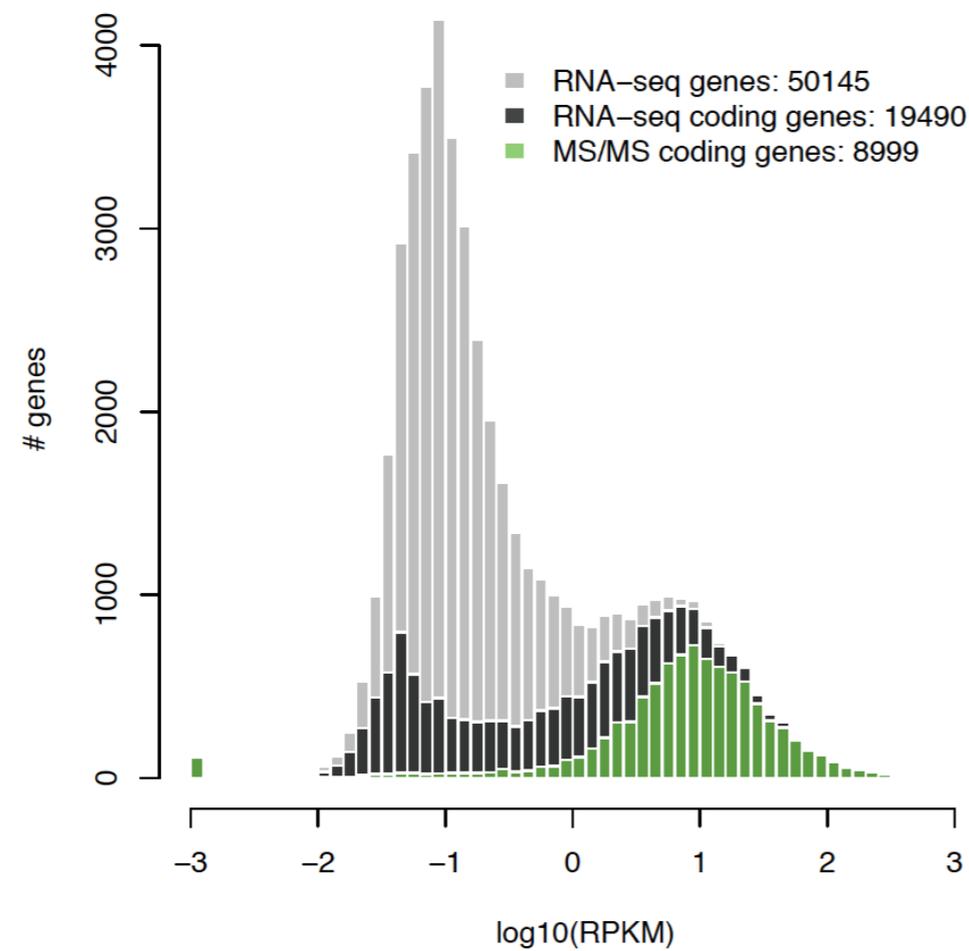


Total protein numbers are comparable to other studies showing ~11,000 proteins (in relatively simpler mixtures)

Most proteins are common to all regions

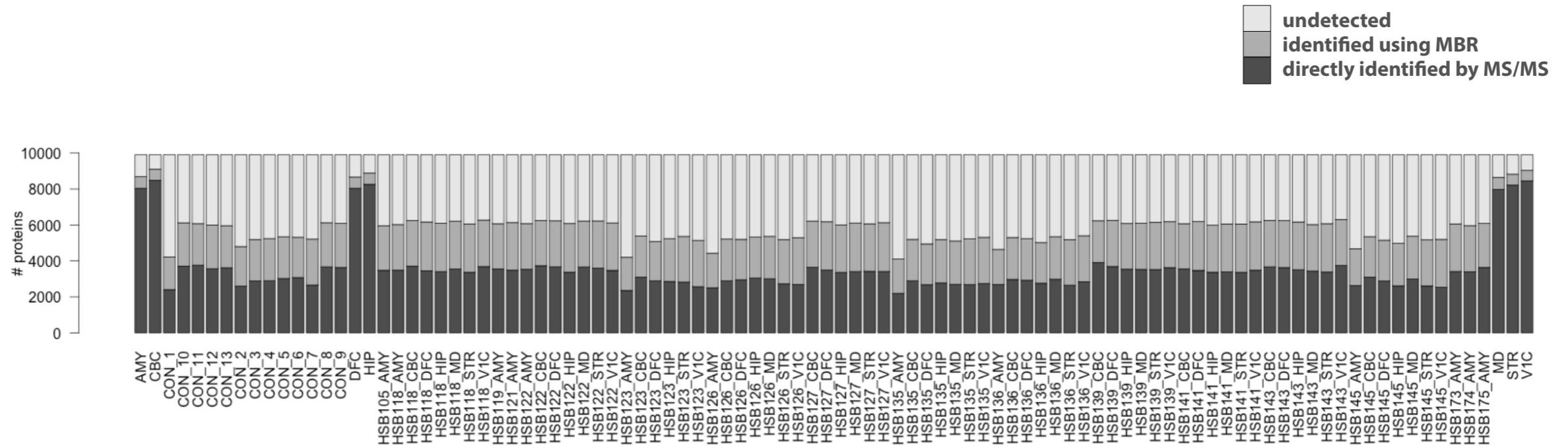
The cerebellum is a clear outlier

Fractionated regions - Comparison with RNA-seq



Unsurprisingly, coverage improves the more abundant a gene is

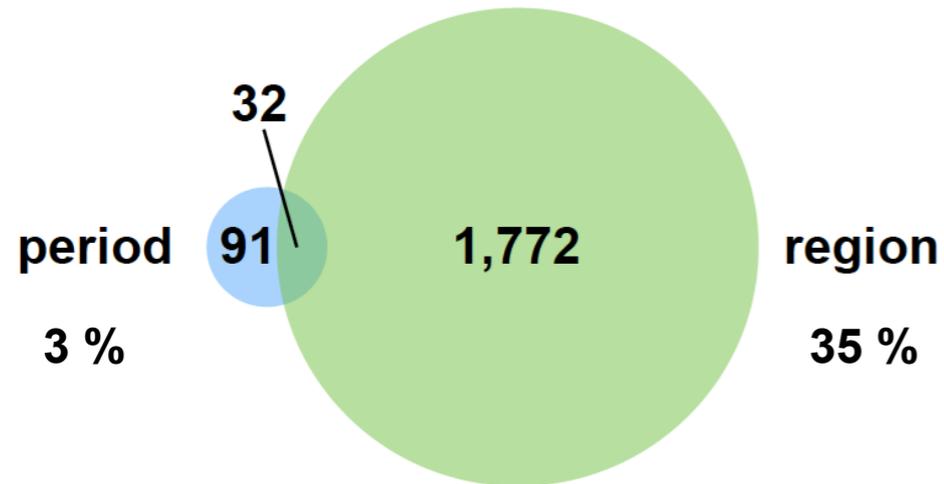
Single shot samples



On average, without matching between runs, we detect 3612 proteins per sample

Using match between runs significantly increases the number of protein IDs in the single shot samples by up to 50%

Single shot data: DEX genes - regions

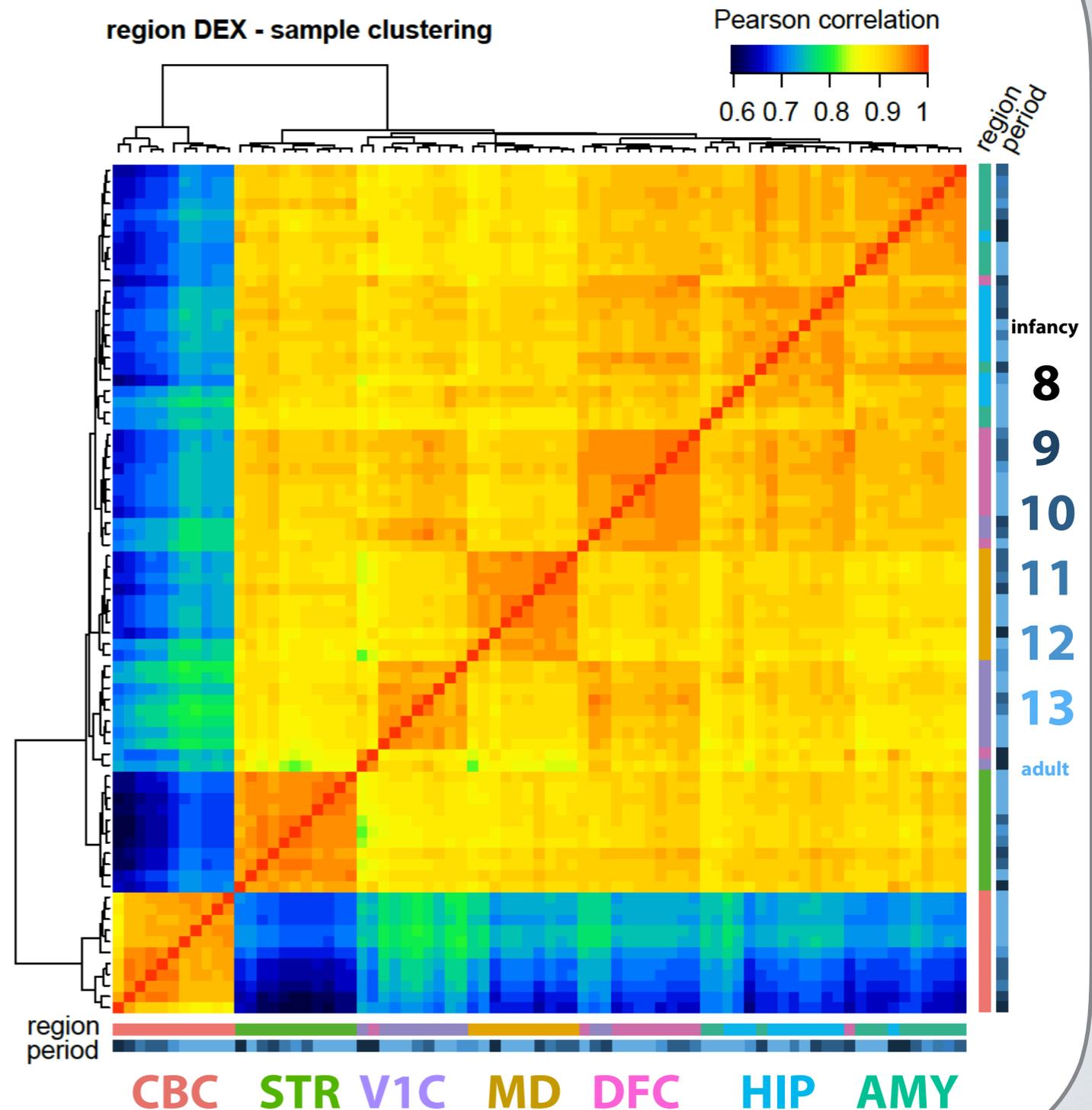


5151 proteins were reliably quantified

Samples were clustered on the basis of DEX genes

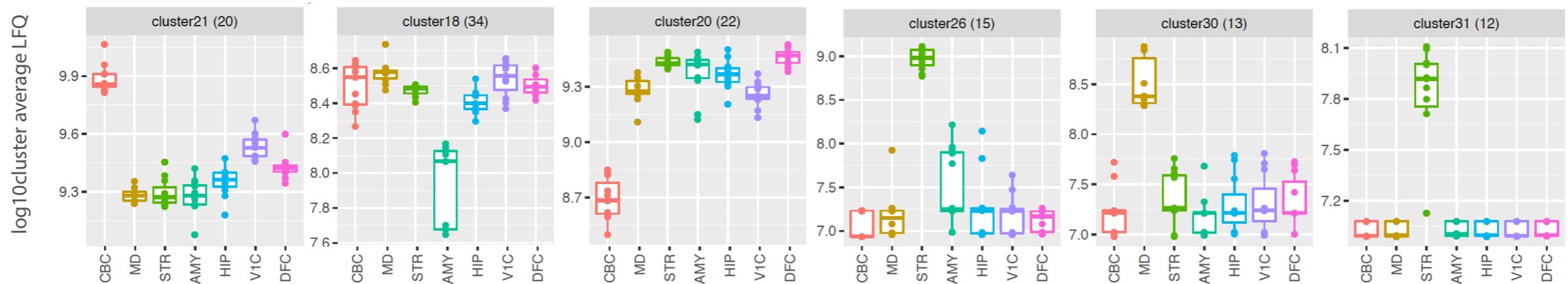
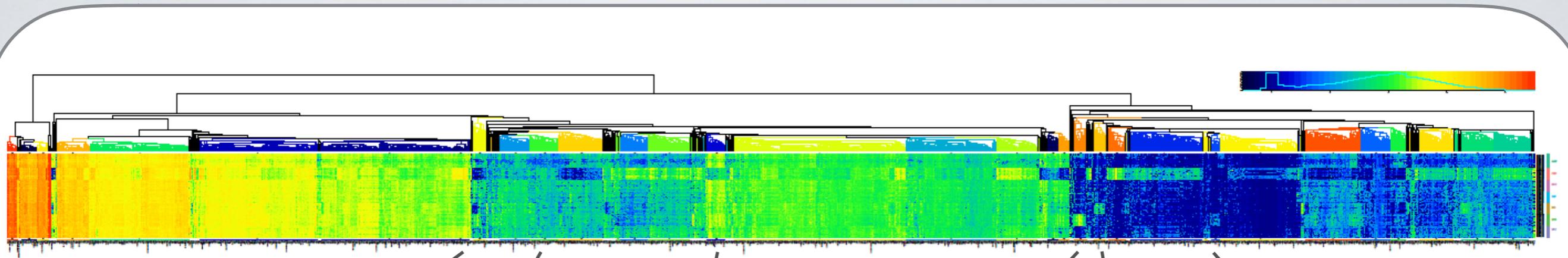
Cerebellum and striatum are clearly defined by their proteome in both development and adulthood

These two regions are markedly more homogenous with regards to cell type than the other 5 regions



Single shot data: DEX - genes

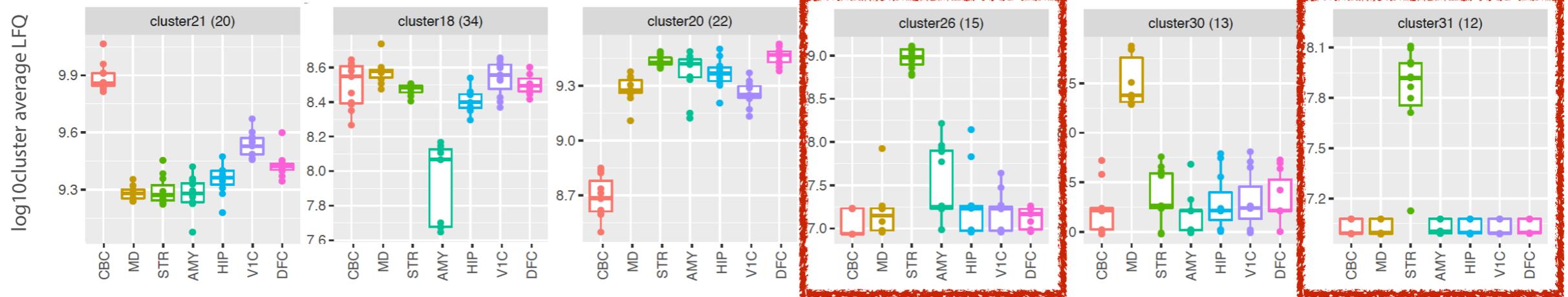
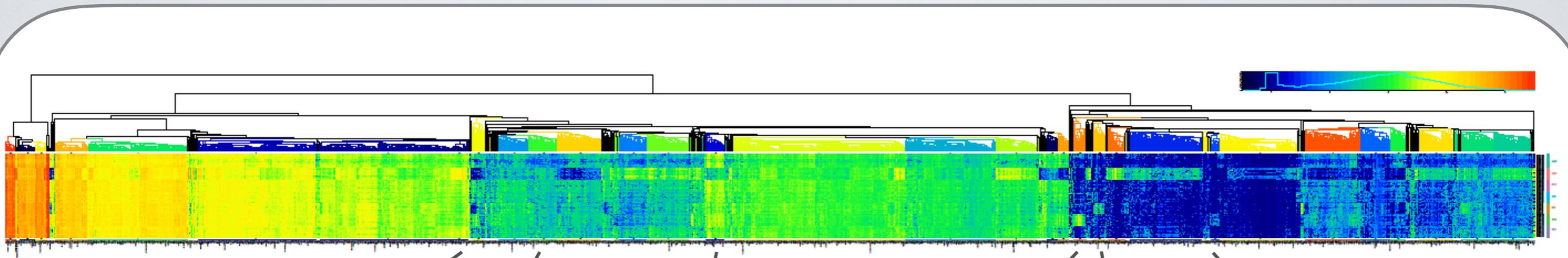
32 region DEX clusters



Single shot data: DEX - genes

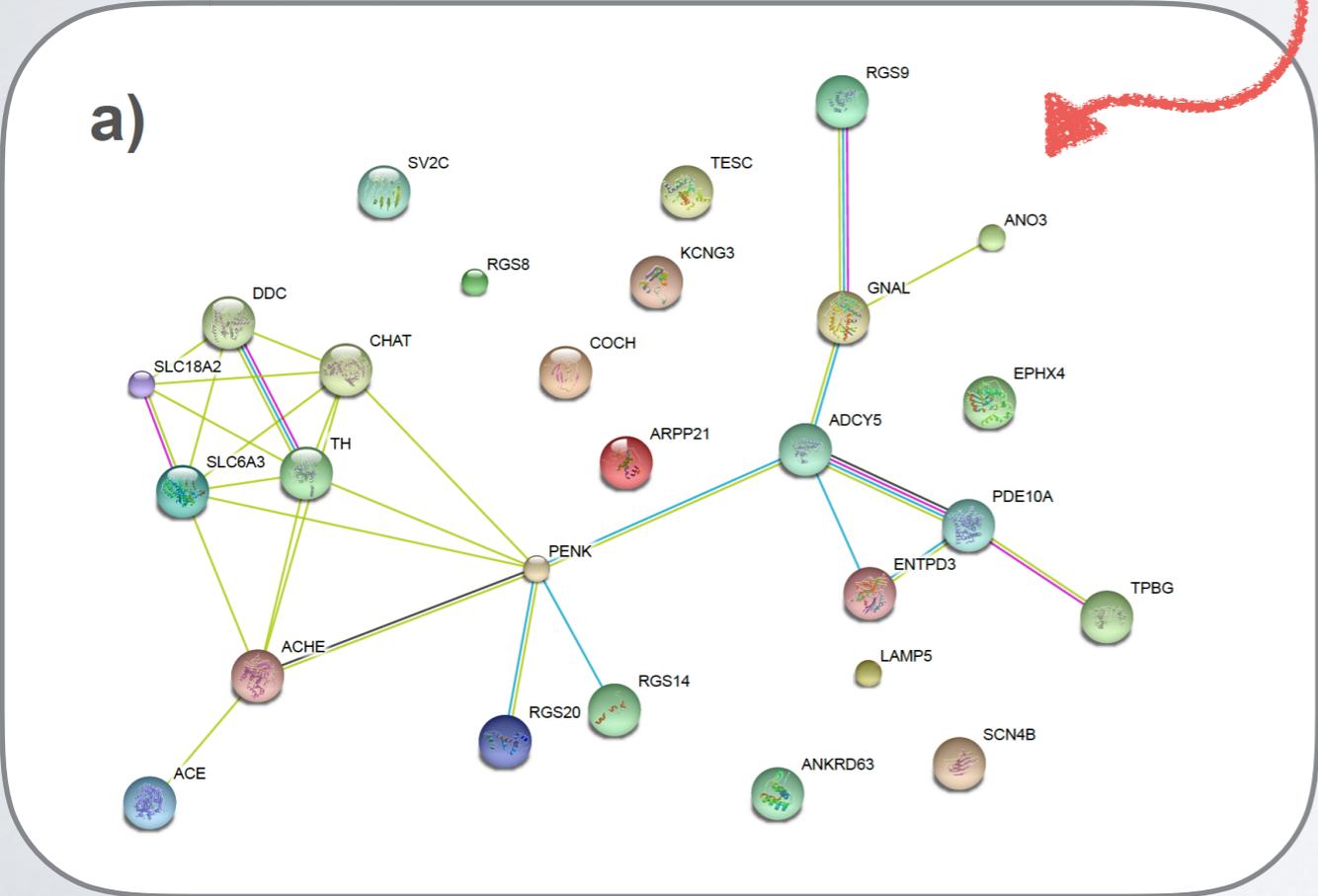
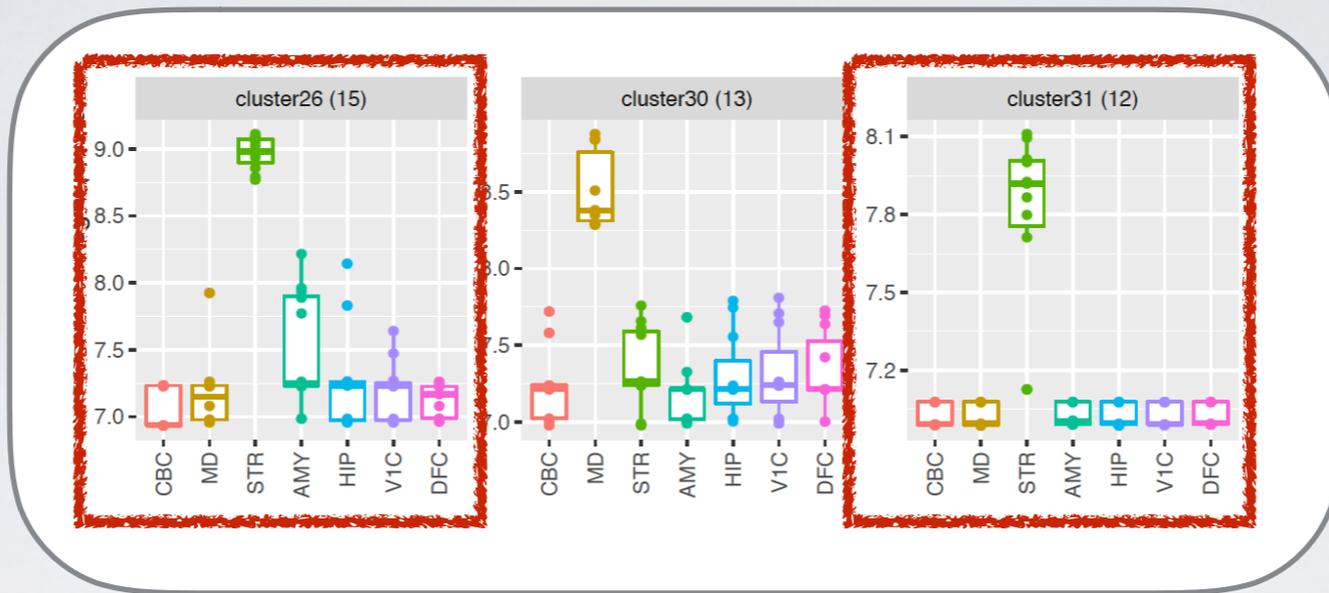
32 region DEX clusters

9 period DEX clusters



Striatally enriched clusters

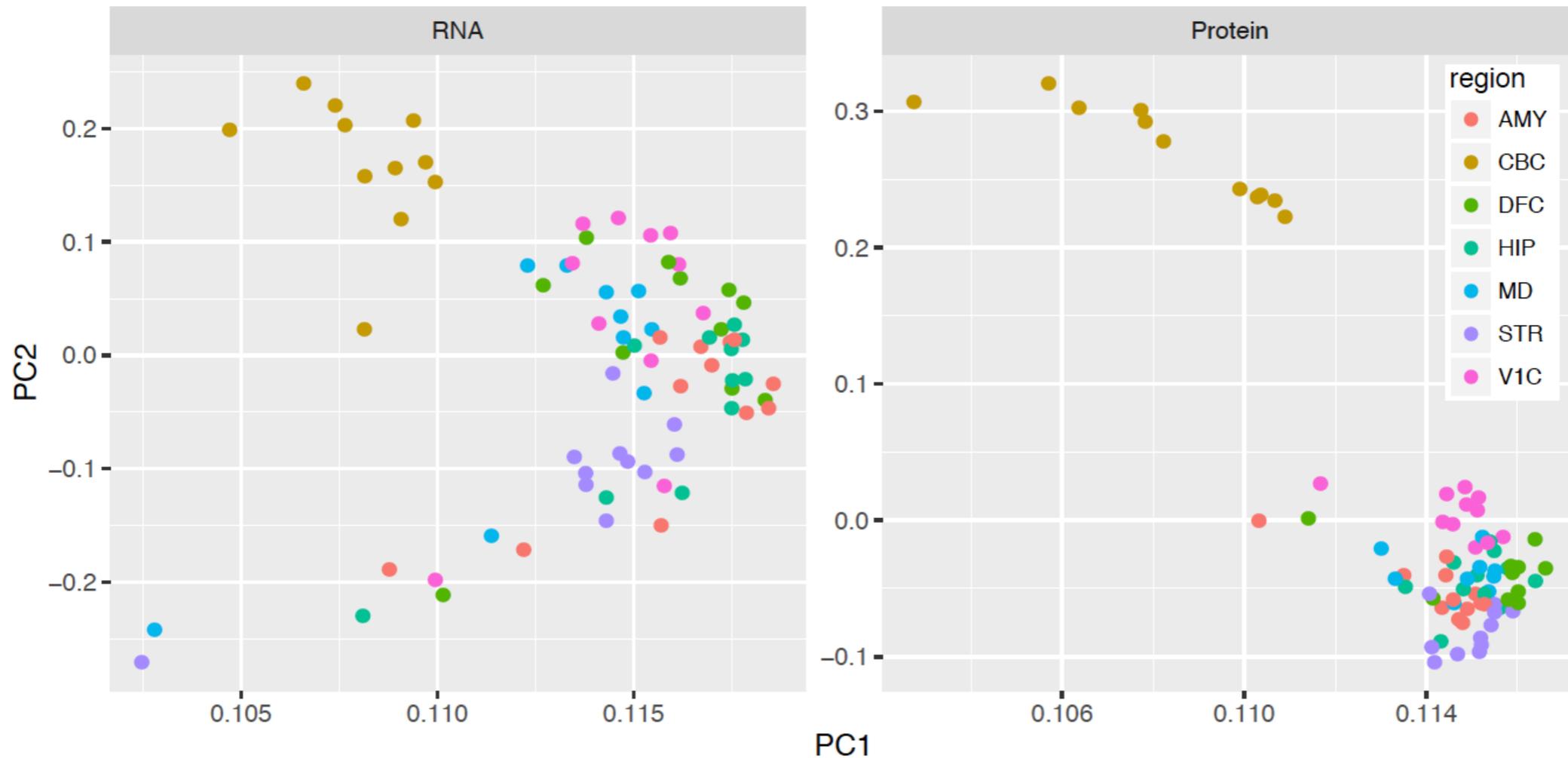
DEX genes - clustering



KEGG pathway	pAdj	proteins
Cocaine addiction	2.55E-06	ADCY5, DDC, RGS9, SLC18A2, SLC6A3, TH
Amphetamine addiction	0.0006	ADCY5, DDC, SLC18A2, SLC6A3, TH
Dopaminergic synapse	0.001	ADCY5, DDC, GNAL, SLC18A2, SLC6A3, TH
Parkinson's disease	0.0147	ADCY5, GNAL, SLC18A2, SLC6A3, TH

Proteins from striatal enriched clusters are functionally related, and enriched for appropriate KEGG pathways

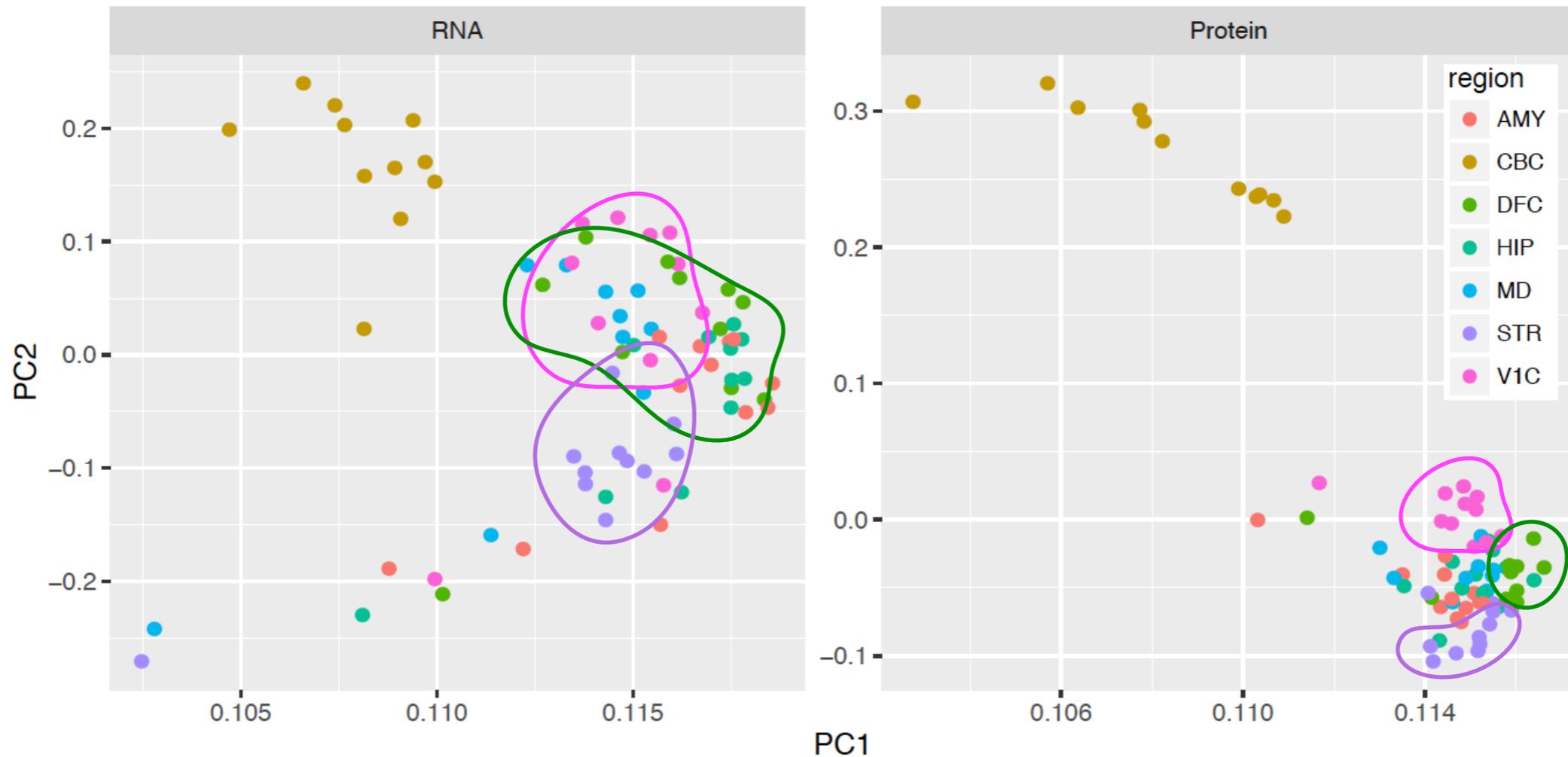
Comparison to RNA-seq



Cerebellum is more clearly separated from the other regions by protein

The other regions are easier to define by protein

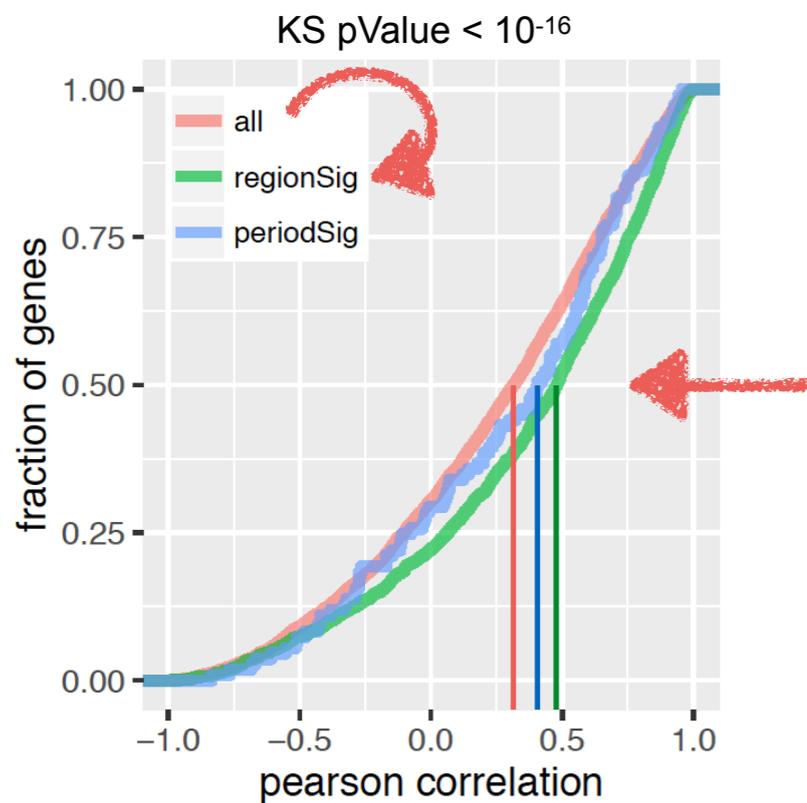
Comparison to RNA-seq



Cerebellum is more clearly separated from the other regions by protein

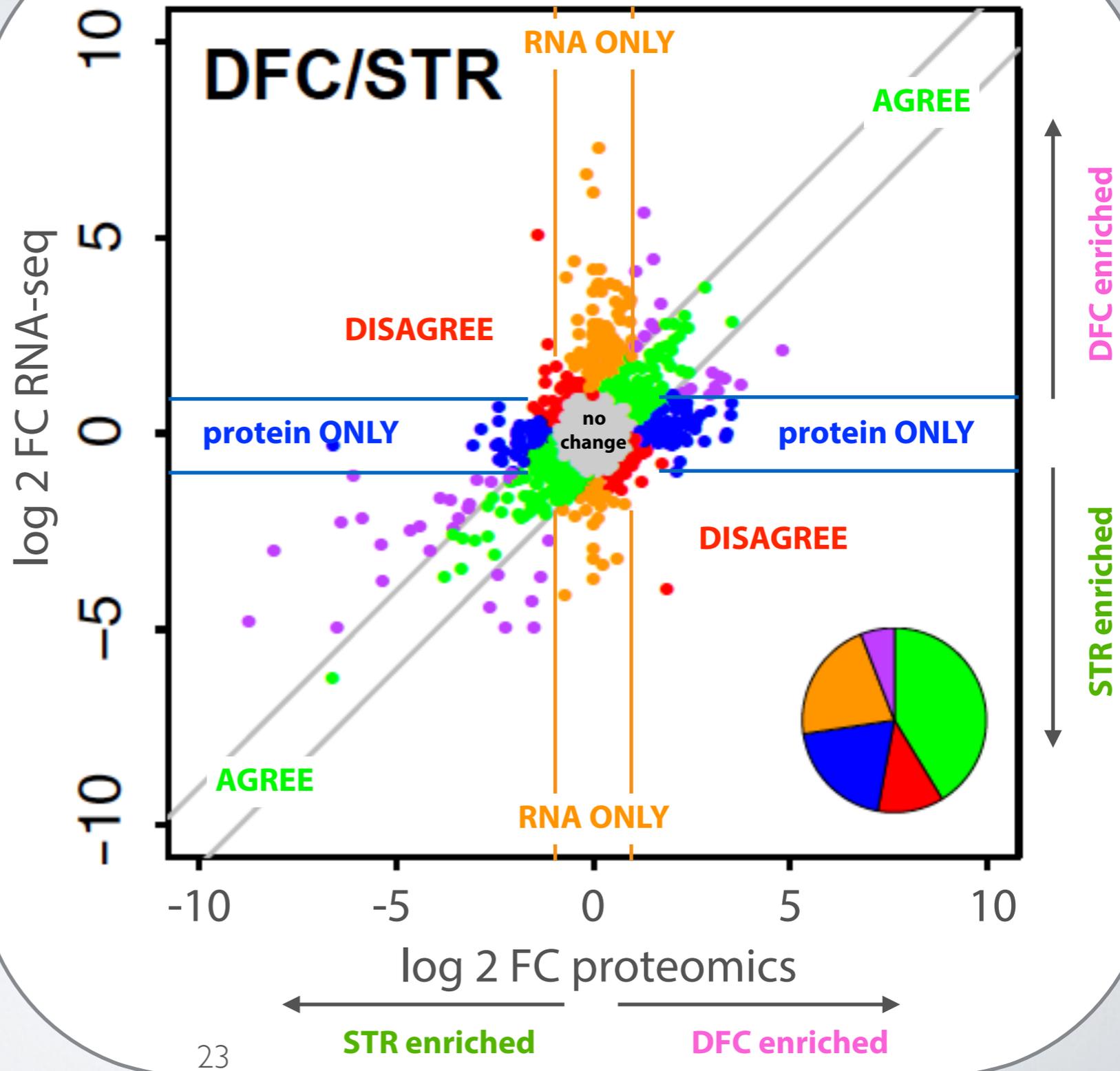
The other regions are easier to define by protein

RNA - protein comparison



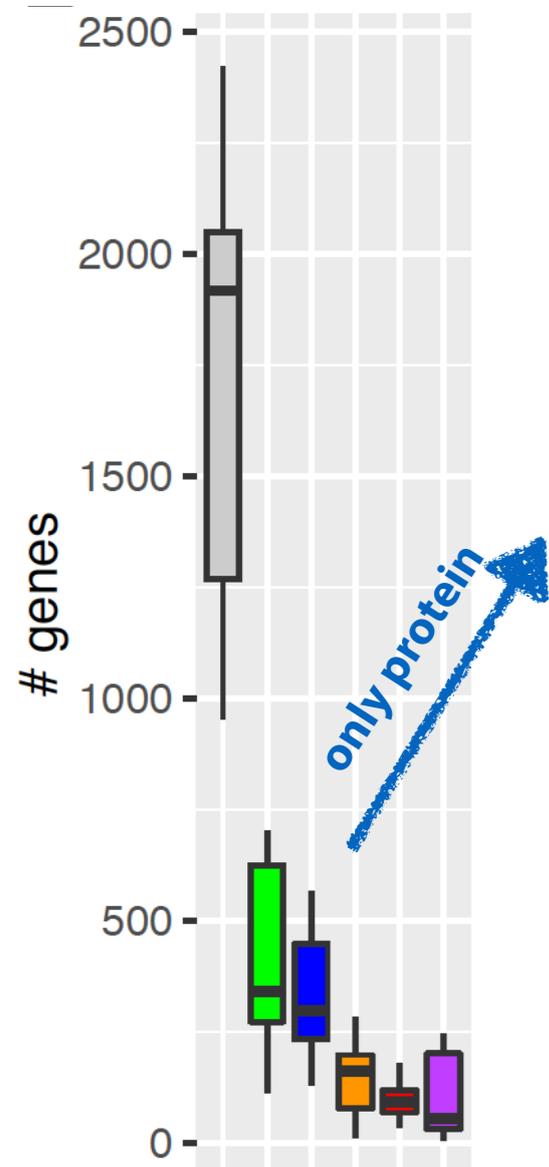
Region DEX genes significantly more correlated between RNA and protein

Trend towards period expression (too few genes)



Ontological analysis

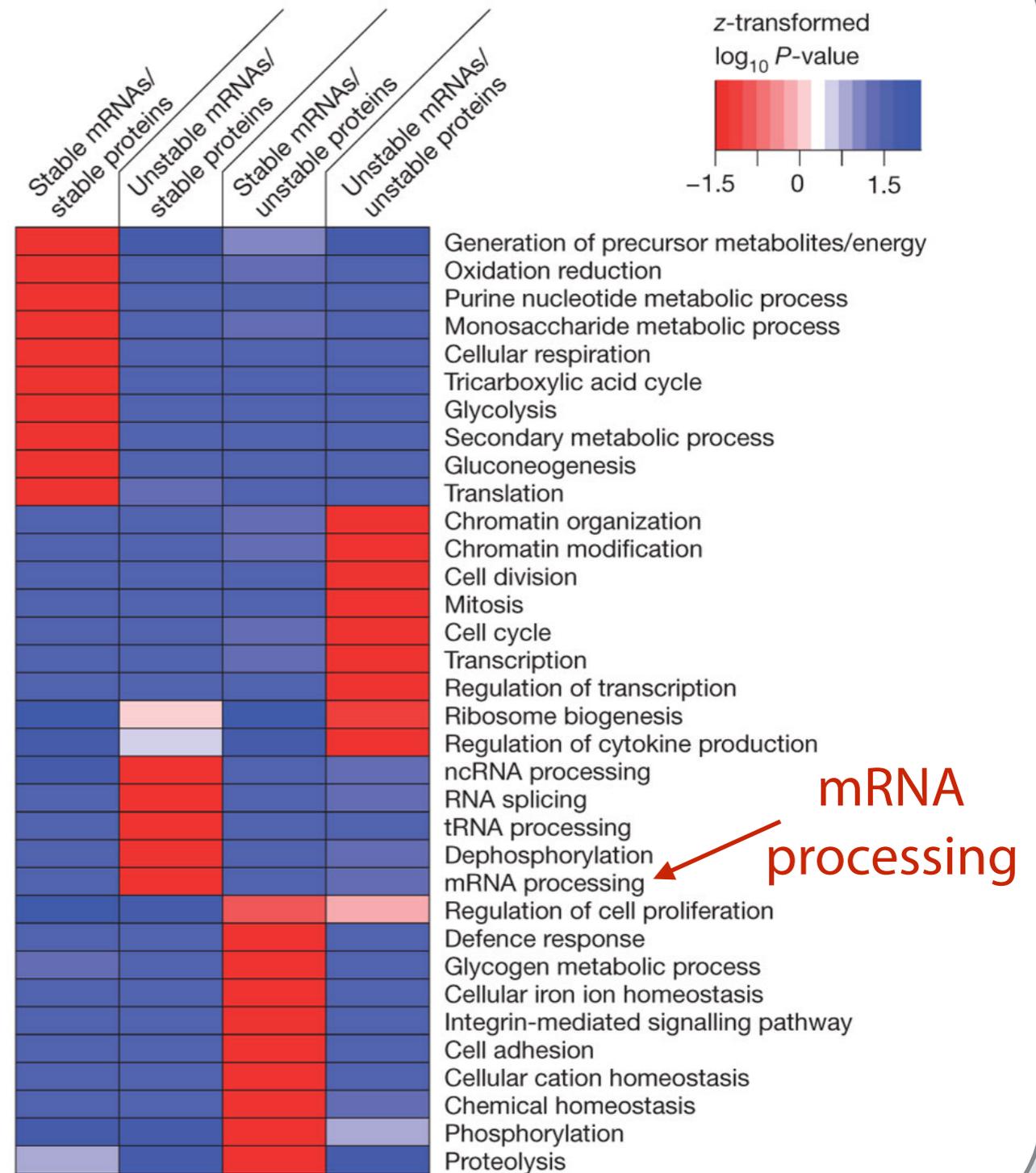
Do these groups of genes relate to ontology in any way?



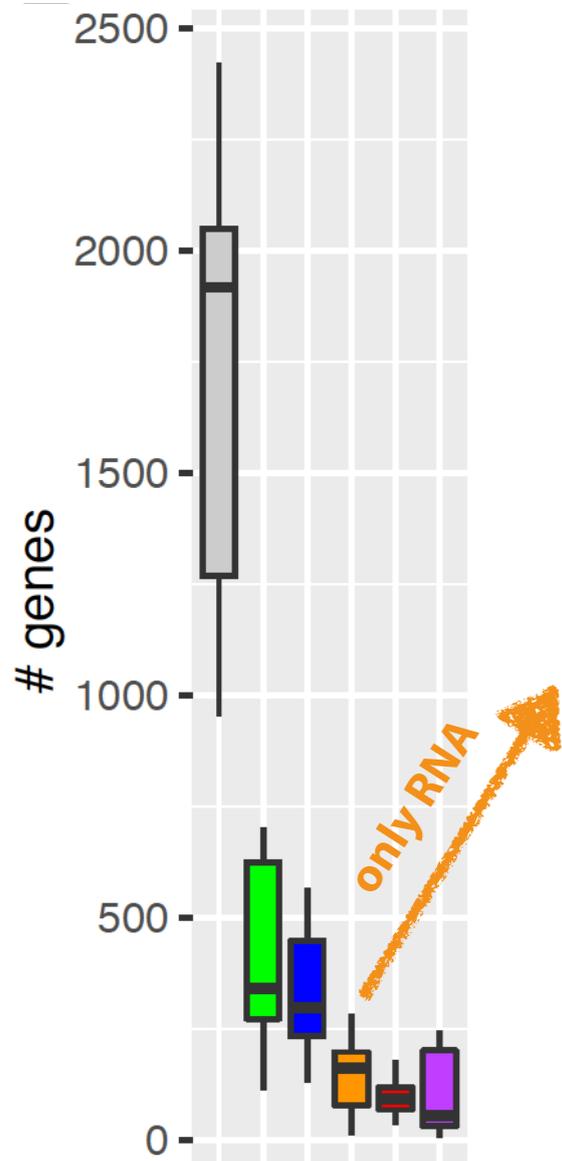
Term	pAdj	Enrichment
RNA processing	3.31E-28	3.23
mRNA processing	3.31E-28	3.58
RNA splicing	3.31E-28	3.57
RNA splicing, via transesterification	3.31E-28	3.7
mRNA splicing, via spliceosome	3.31E-28	3.69
RNA metabolic process	3.31E-28	1.95
nucleic acid metabolic process	3.31E-28	1.88
gene expression	3.31E-28	1.78
nucleobase-containing compound	3.31E-28	1.65
nucleoplasm	8.58E-29	2.19
nuclear part	8.58E-29	1.93
nuclear lumen	8.58E-29	2.01
nucleus	8.58E-29	1.57

Protein only changes

- Overall enrichment for nuclear proteins driven by cell body density
- RNA processing terms more significant than other nuclear terms
 - may be a reflection of relative stability
- Other interesting proteins appear in region comparisons with more similar cytoarchitecture



RNA only changes

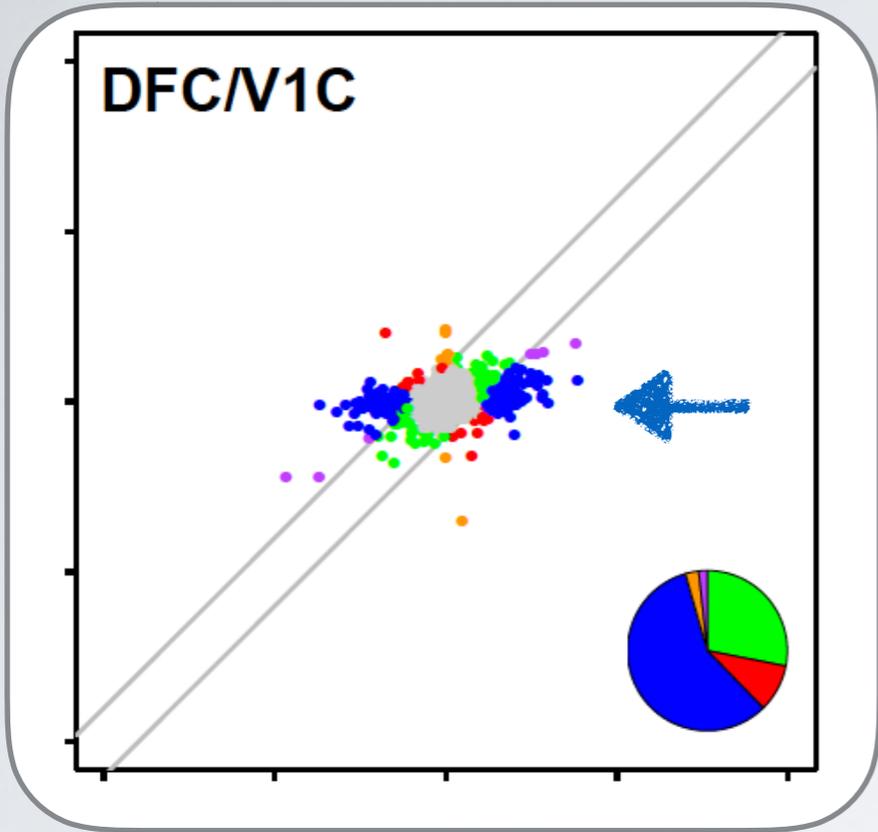


Term	pAdj	Enrichment
signaling	0.00071	1.39
cell communication	0.0008	1.38
metal ion transport	0.0016	2.33
ion transmembrane transport	0.0017	2.09
ion transport	0.0027	1.84
neurological system process	0.0027	2.23
transmembrane transporter complex	0.039	2.29
plasma membrane region	0.04	1.73
transporter complex	0.041	2.24
ion channel complex	0.042	2.36
dense core granule	0.05	6.78

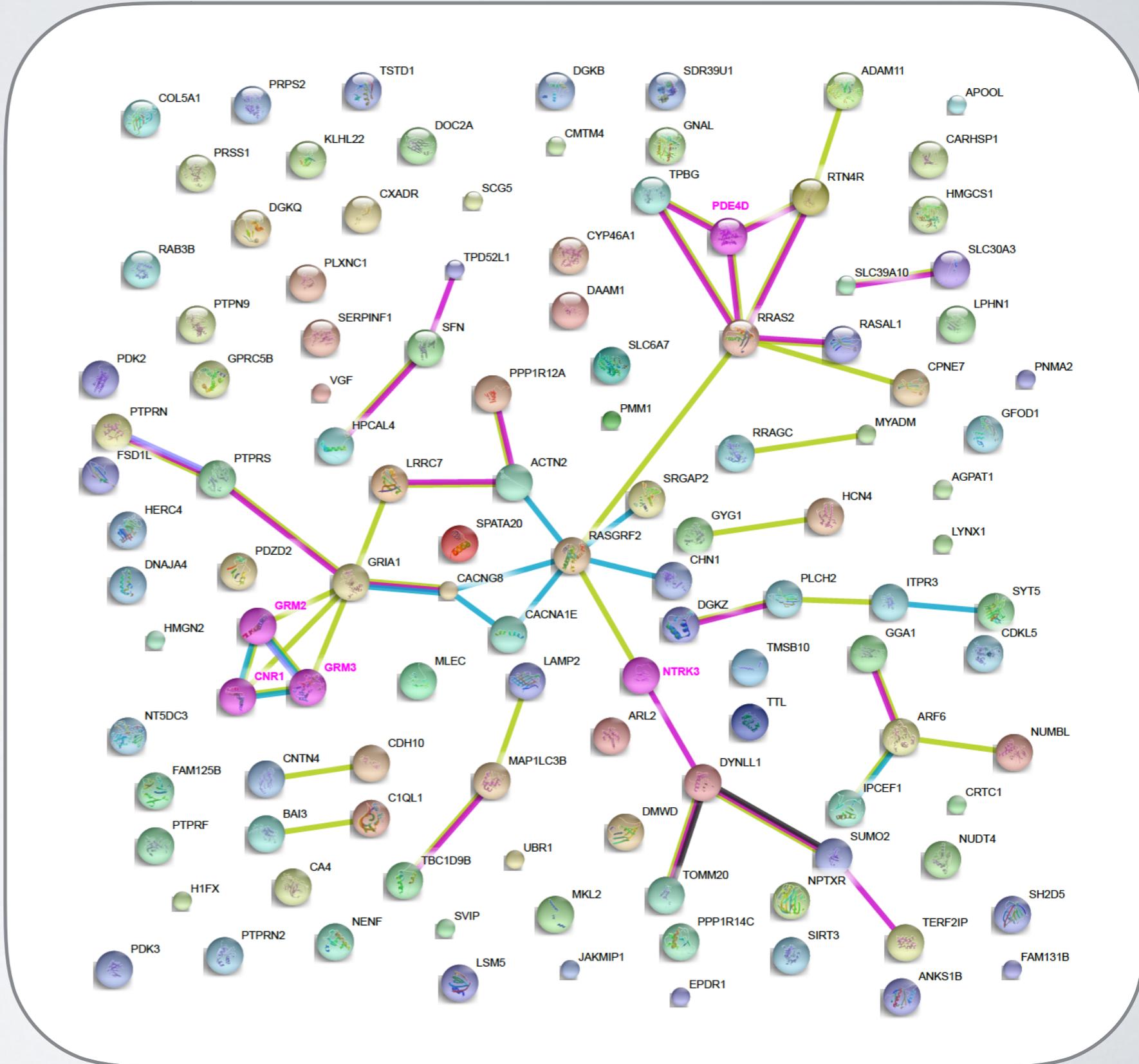
RNA only changes

- Tend to be membrane based, signalling proteins
- These proteins may be synthesized in one region, but transported to another
- Could have important implications for targeting knockdowns, for example

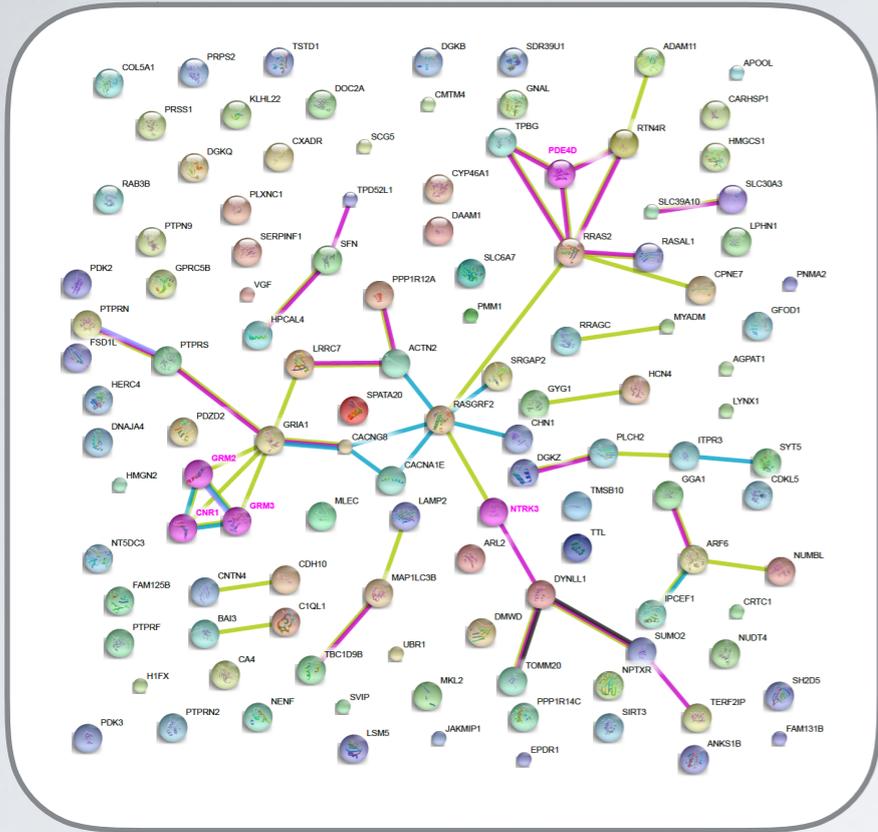
Comparison of areas with similar cytoarchitecture



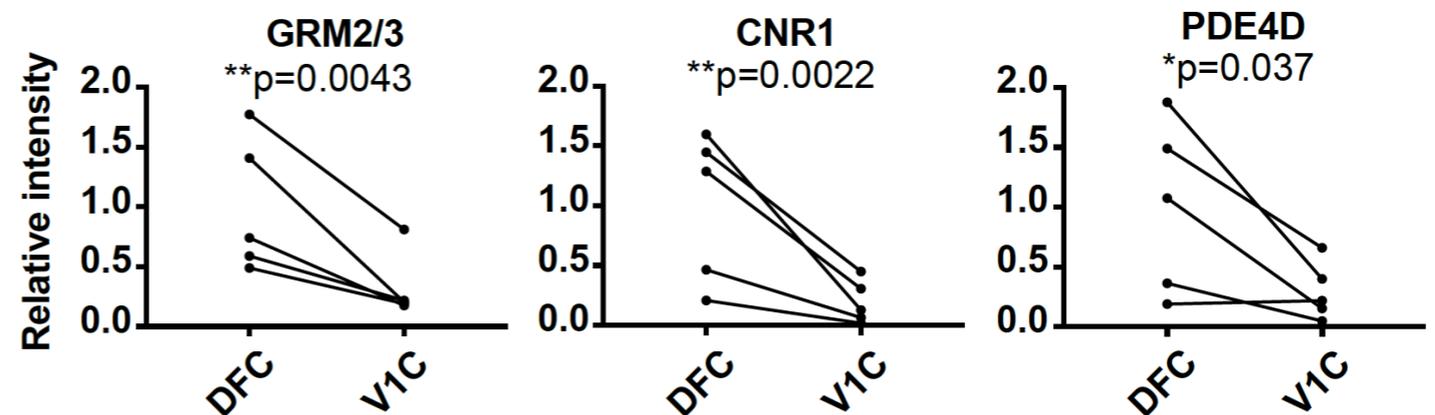
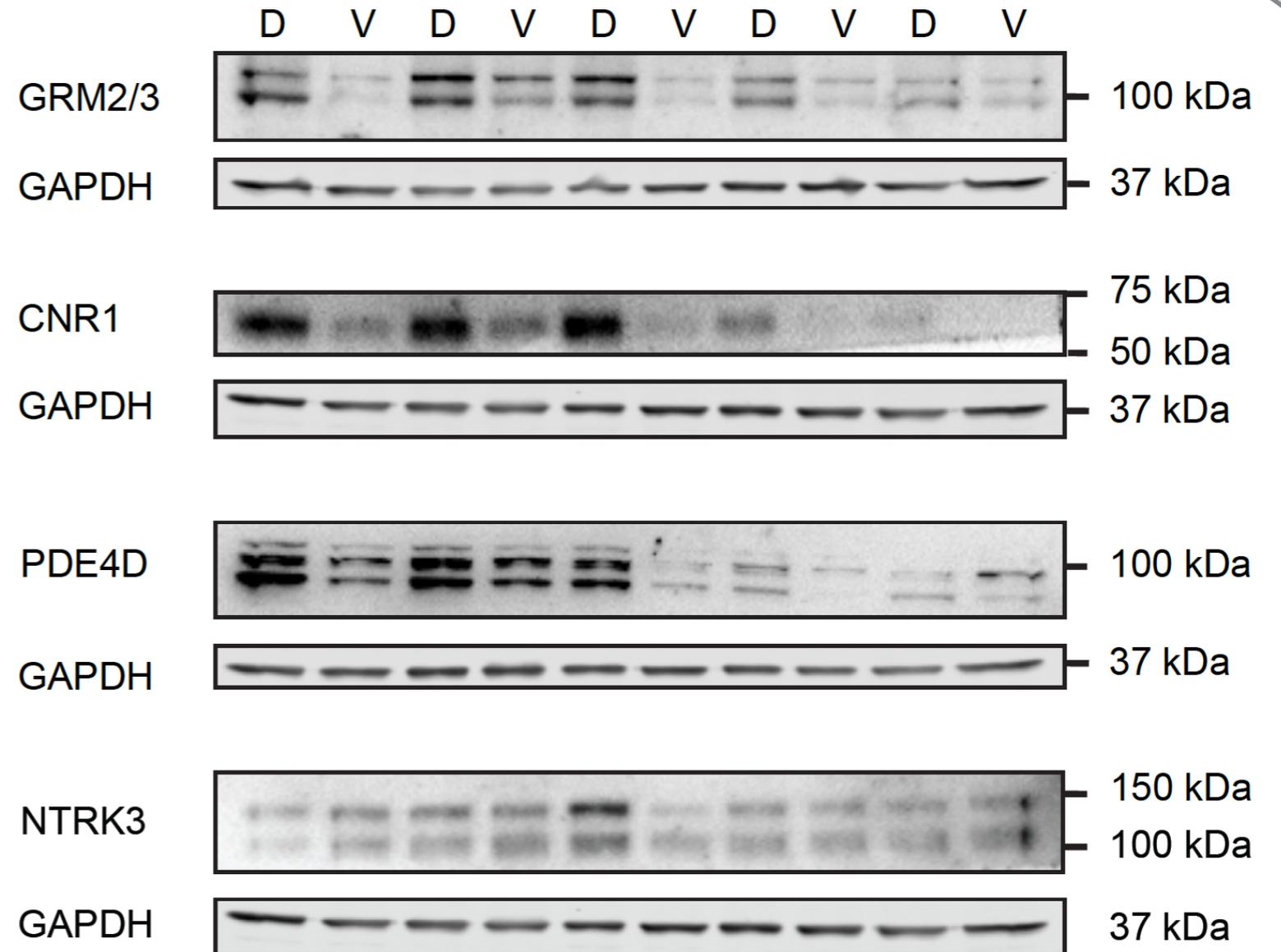
Proteins enriched in DFC vs V1C are functionally related



Comparison of areas with similar cytoarchitecture



Proteins enriched in DFC compared to V1C are functionally related



Talk outline

- Why proteomics?
- BrainSpan/psychENCODE project samples
- Proteomic methods
- Results
- **Conclusions**

Conclusions

- Proteomic level data can reveal differences between regions not seen at the RNA level
- This approach is most powerful in comparing regions with similar cytoarchitecture
- Proteomic data is still biased towards abundant proteins. Future work should consider how best to simplify mixtures to increase the sensitivity of the technique:
 - multiplexing samples for fractionation by liquid chromatography
 - sub cellular fractionation
 - tissue specific references to increase identification rates/identify isoforms
 - cell-type specific proteomics

How can you use this data?

- Increase the number of IDs in your label free experiments through matching between our fractionated libraries
- Use these spectra to select peptides for targeted proteomics
- For assessing protein stoichiometry
- Useful tool for looking at RNA/protein agreement for a new gene of interest
- We also have mouse vs human comparison data - is protein expression similar between humans and your animal model?
- We are hoping to extend this work to isoform level coverage soon

*Data can be found in the supplementary tables here:
<https://www.nature.com/articles/s41593-017-0011-2>*

Thanks!

Nairn Lab - Yale

Angus Nairn

Becky Carlyle

Rob Kitchen

Shannon Leslie

Dan Li

Veronica Musante

Xuehong Shang

Rashaun Wilson



Gerstein Lab - Yale

Rob Kitchen

Jing Zhang

Joel Rozowsky

Mark Gerstein



Sestan Lab

Nenad Sestan

Mihovil Pletikos

Andre Sousa

Yale / NIDA Proteomics

TuKiet Lam

Jean Kanyo

Ted Voss

Navin Rauniyar



Funding: psychENCODE consortium, Yale NIDA Neuroproteomics Center,
NARSAD Young Investigator