

Current Biology, Volume 34

Supplemental Information

**Molecular organization of neuronal cell types
and neuromodulatory systems in the zebrafish telencephalon**

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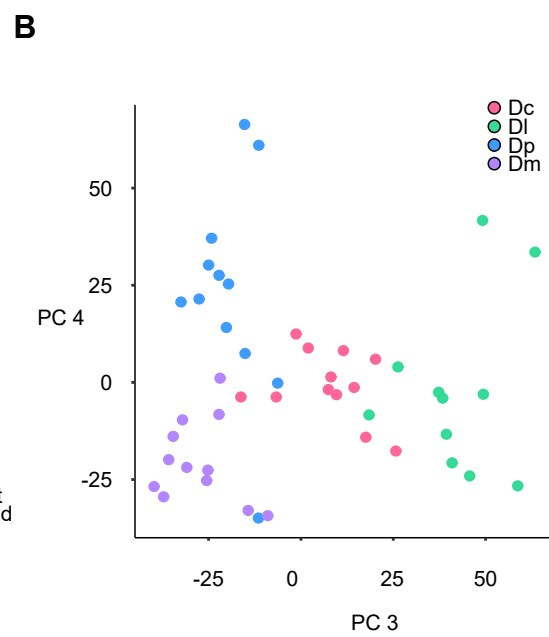
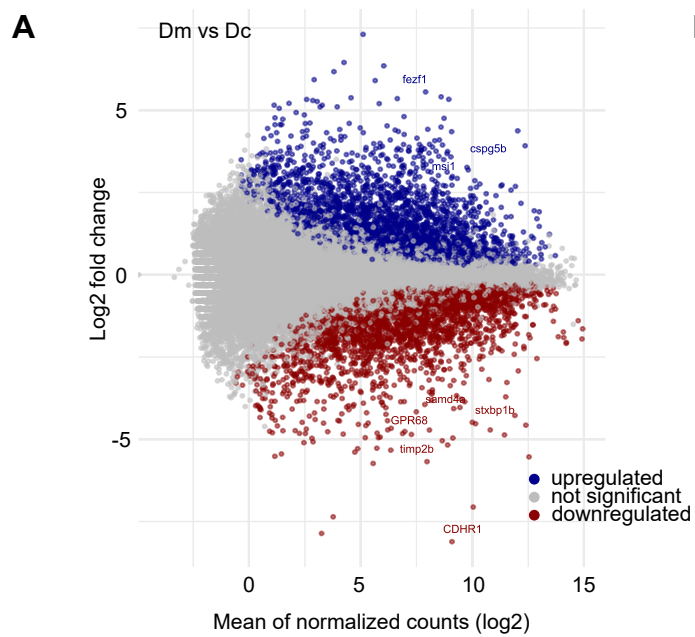


Figure S1. Bulk sequencing of pallial microdissected areas, related to Figure 1.

(A) Example of differential gene expression in two different telencephalic brain areas (Dm and Dc). (B) Projection of gene expression patterns (2.000 most variable genes; same as in Figure 1B) onto PCs 3 and 4. Each data point is one replicate from a pallial region (color-coded). Note separation of Dp and Dm. See Figure 1B for projection of the same data onto PCs 1 and 2.

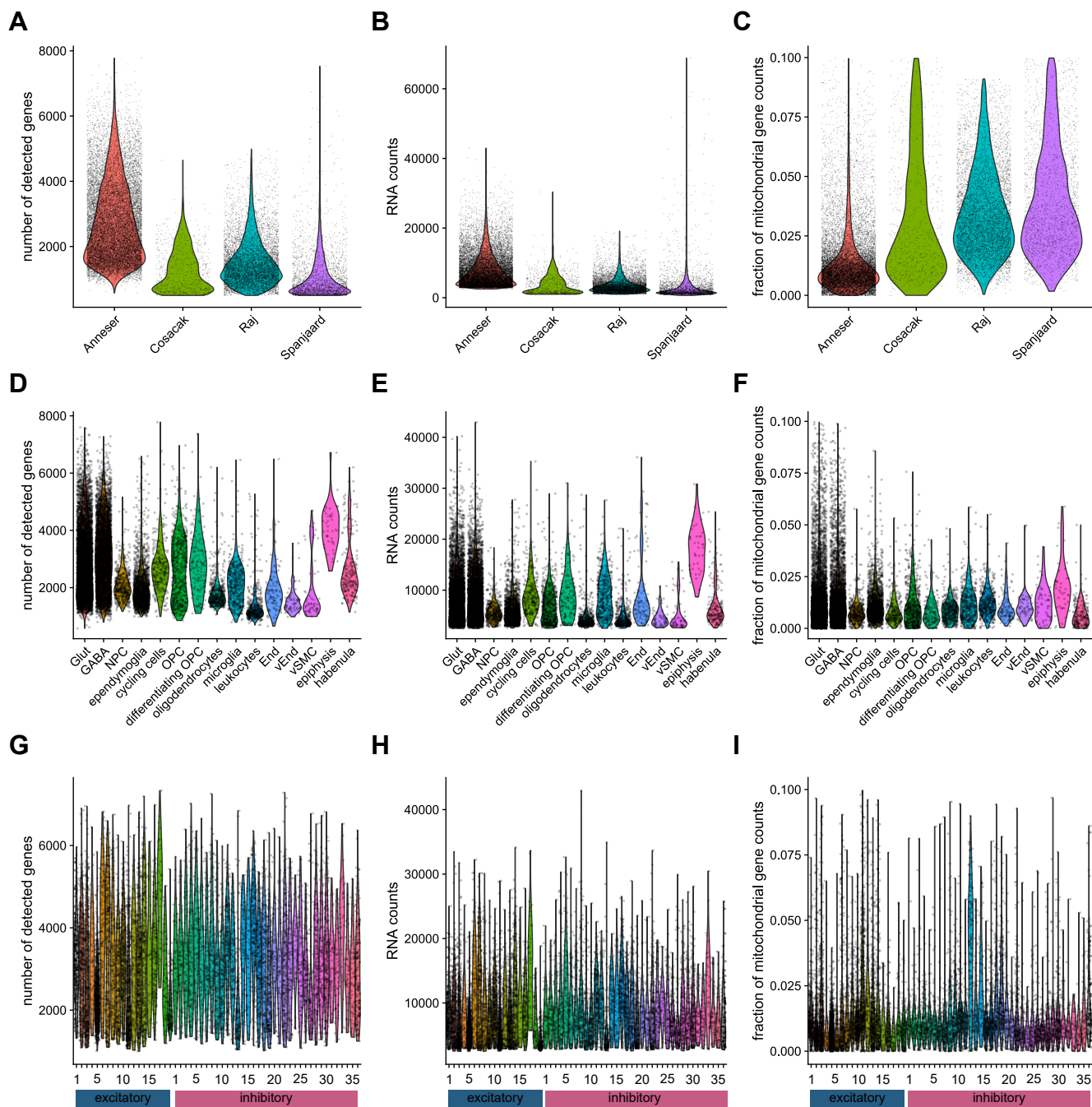


Figure S2. Quality assessment of scRNA-seq dataset, related to Figure 2.

(A-C) Comparison of quality-related parameters of our single-cell RNA sequencing (scRNA-seq) dataset (“Anneser”) to previously published scRNA-seq datasets from the zebrafish telencephalon (“Cosacak”^{S1}, “Raj”^{S2}, “Spanjaard”^{S3}). Each dot represents one sequenced cell; violin plots show distributions. In our dataset, the number of detected genes and RNA counts are comparably high (**A, B**) while the fraction of mitochondrial gene reads, which usually indicates cell death, is relatively low (**C**). **(D-F)** Same quantifications stratified by superclusters in our dataset (Figure 2A). **(E-I)** Same quantification stratified by neuronal subtypes.

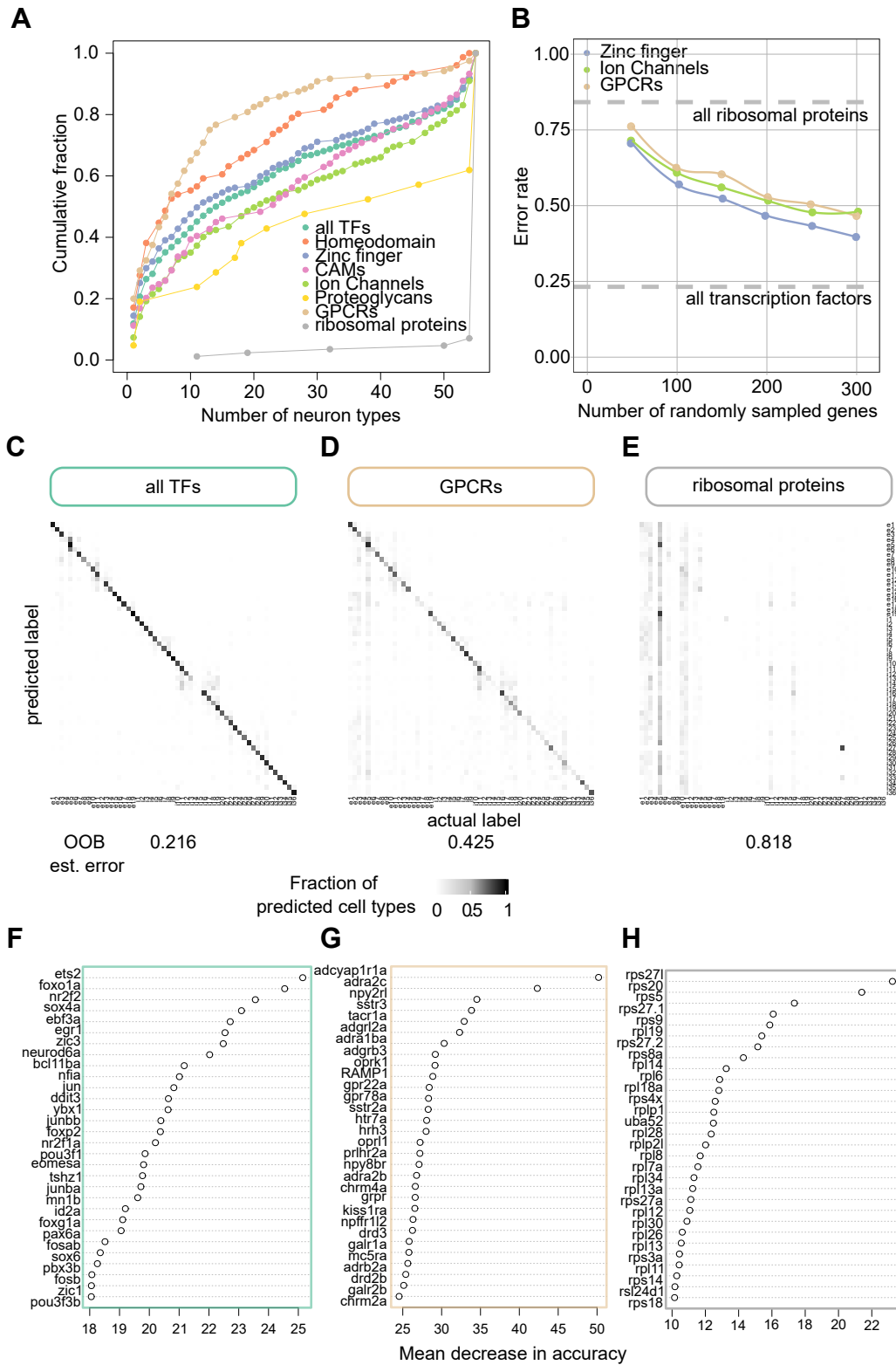


Figure S3. Cell-type specificity of gene expression, related to Figure 2.

(A) Cumulative fraction of information about neuron types contained in expression patterns of different gene families (TFs, transcription factors; Homeodomain TFs; Zinc finger TFs, CAMs, cell adhesion molecules; ion channels; proteoglycans; ribosomal proteins). Information content of individual gene families was assessed by classifying the 55 neuronal cell types which were Louvain-clustered based on the entire transcriptome. This analysis is similar to Figure 1D but follows the computation in Hain et al. 2022^{S4}. (B) Plot of randomly subsampled sets of different gene families and the resulting accuracy of the corresponding random forest classifier, bounded by the accuracy of the ribosomal gene classifier and the classifier trained on the entire set of transcription factors. Different accuracy in random forest classification might thus not only be attributable to degree of heterogeneity in the expression pattern, but simply to the number of genes available per gene family. (C-E) Cell type classification performance of random forest classifiers trained on expression patterns of transcription factors, GPCRs, or ribosomal genes. The number below each plot indicates the out-of-bag error estimate. Note high performance of classifiers trained on TFs and GPCRs but not ribosomal genes. (F-H) Most informative genes for each of the classifier, ranked by mean decrease in accuracy when left out of the classifier.

A

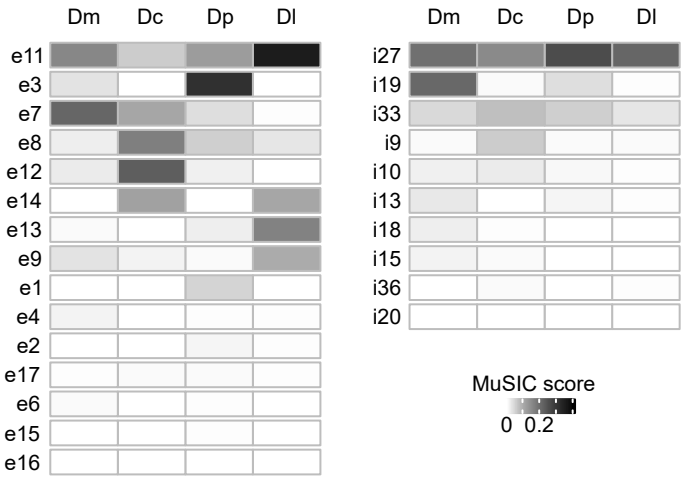
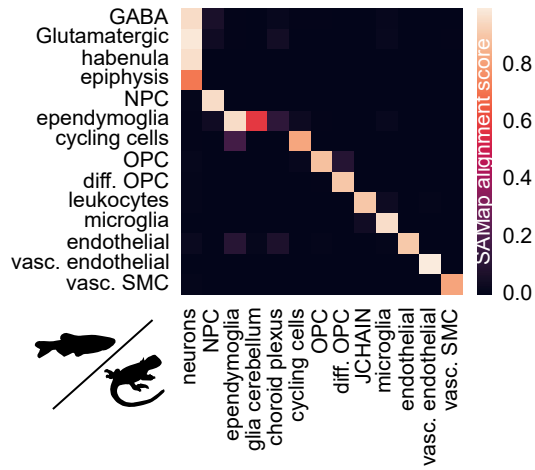
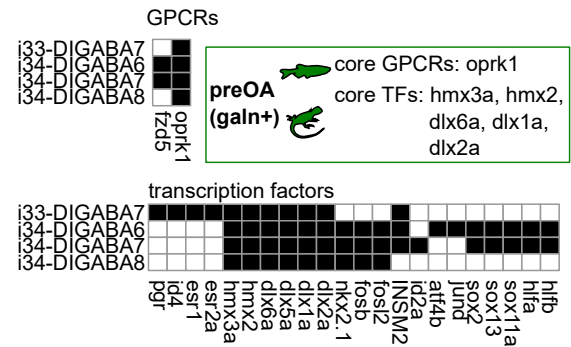


Figure S4. Contribution of individual cell types to pallial areas, related to Figure 4.
(A) Deconvolved contribution of individual cell types to pallial areas, computed by the MuSIC algorithm^{S5}. Heatmaps are separated into glutamatergic and GABAergic cells and sorted by maximum contribution score.

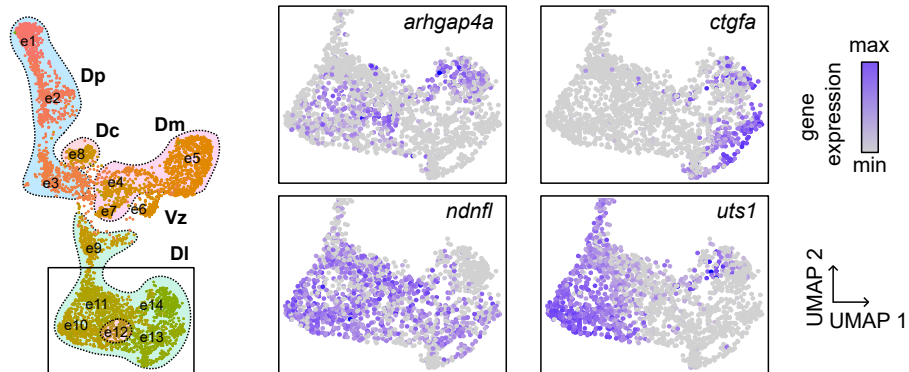
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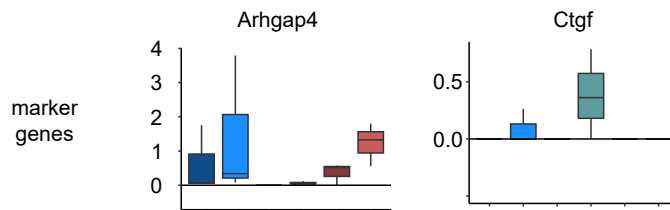
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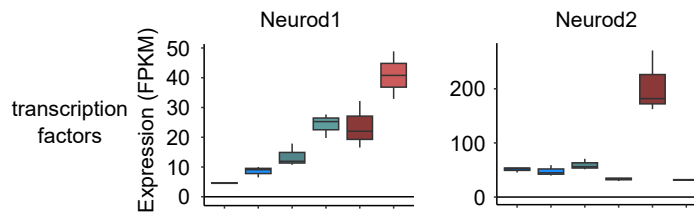
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D



E



F

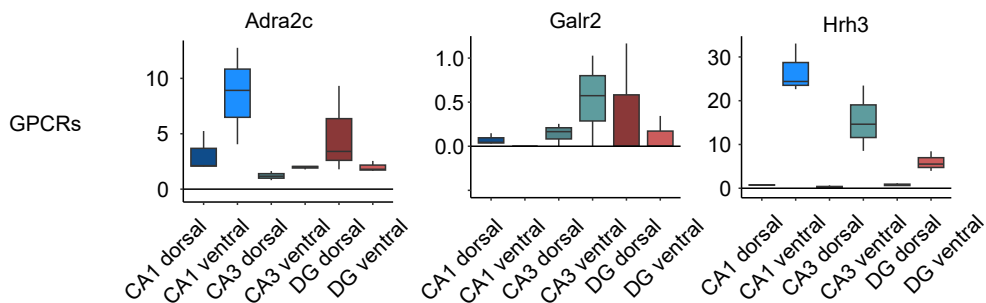


Figure S5. Evolutionary conservation of cell superclusters between zebrafish and bearded dragon, related to Figure 6.

(A) Alignment score between corresponding superclusters in zebrafish and the bearded dragon. The sparsity of the matrix implies specificity of alignment. *vasc.* SMC, vascular smooth muscle cells; OPC, oligodendrocyte precursor cells; *diff.* OPC, differentiating OPC; NPC; neural progenitor cells, JCHAIN, plasma cells. (B) Binarized expression heatmap of GPCRs and TFs significantly contributing to the alignment of galanin-expressing cells in the preoptic area of the zebrafish and hypothalamic galanin+ cells in the lizard. Both appear to be susceptible to modulation via the k-opioid receptor in both species. (C) Enlargement of the area in UMAP space where DI-associated cell types are localized (e10, e11, e13, e14). Expression levels of the four marker genes *arhgap4a*, *ctgfa*, *ndnfl*, and *uts1* within this area are shown. (D – E) Expression levels of the mouse genes orthologous to zebrafish marker genes, TFs and GPCRs enriched in DI taken from the hipposeq dataset^{S6}.

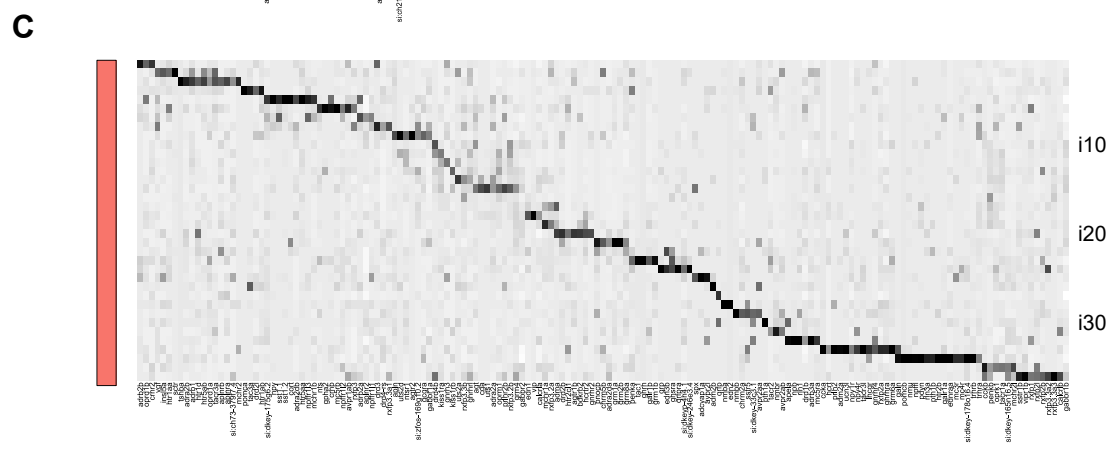
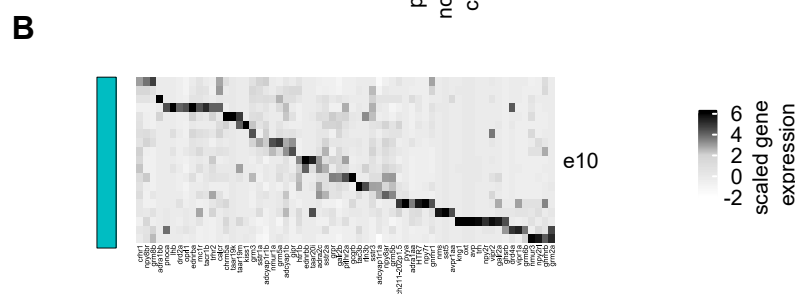
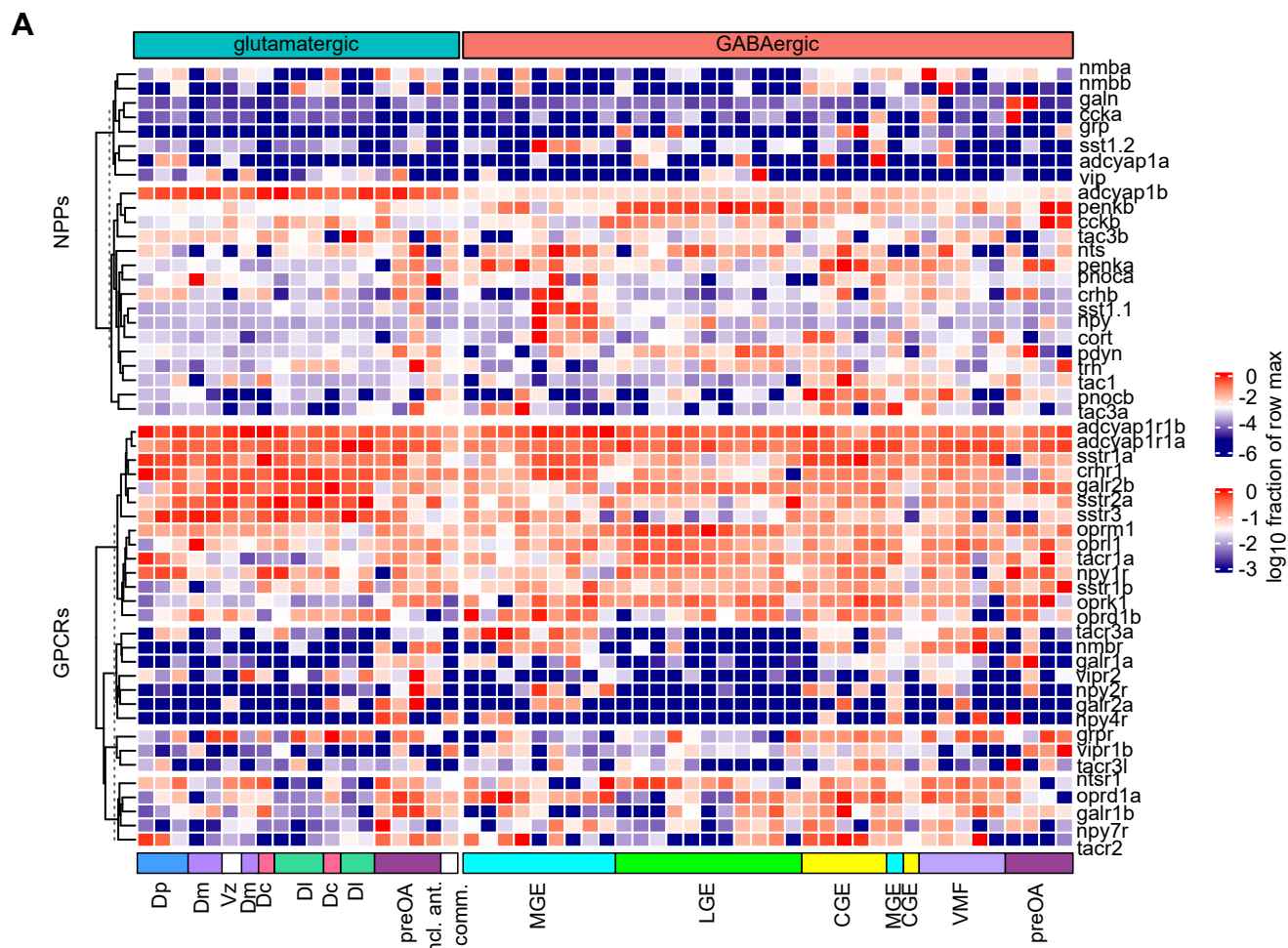


Figure S6. Heterogeneity and specificity of telencephalic neuromodulatory networks, related to Figure 7.

(A) Normalized expression of selected NPPs and GPCRs in GABAergic and glutamatergic cell types identified in our dataset. Expression is shown using a logarithmic color scale after normalization to the maximum in each row. Rows were ordered by hierarchical clustering (left).

(B, C) Enlargements of Figure 7H showing normalized (z-scored) expression levels of all genes associated with neuromodulation separately in glutamatergic and GABAergic cell types. Matrices were sorted by maximum expression per cell type.

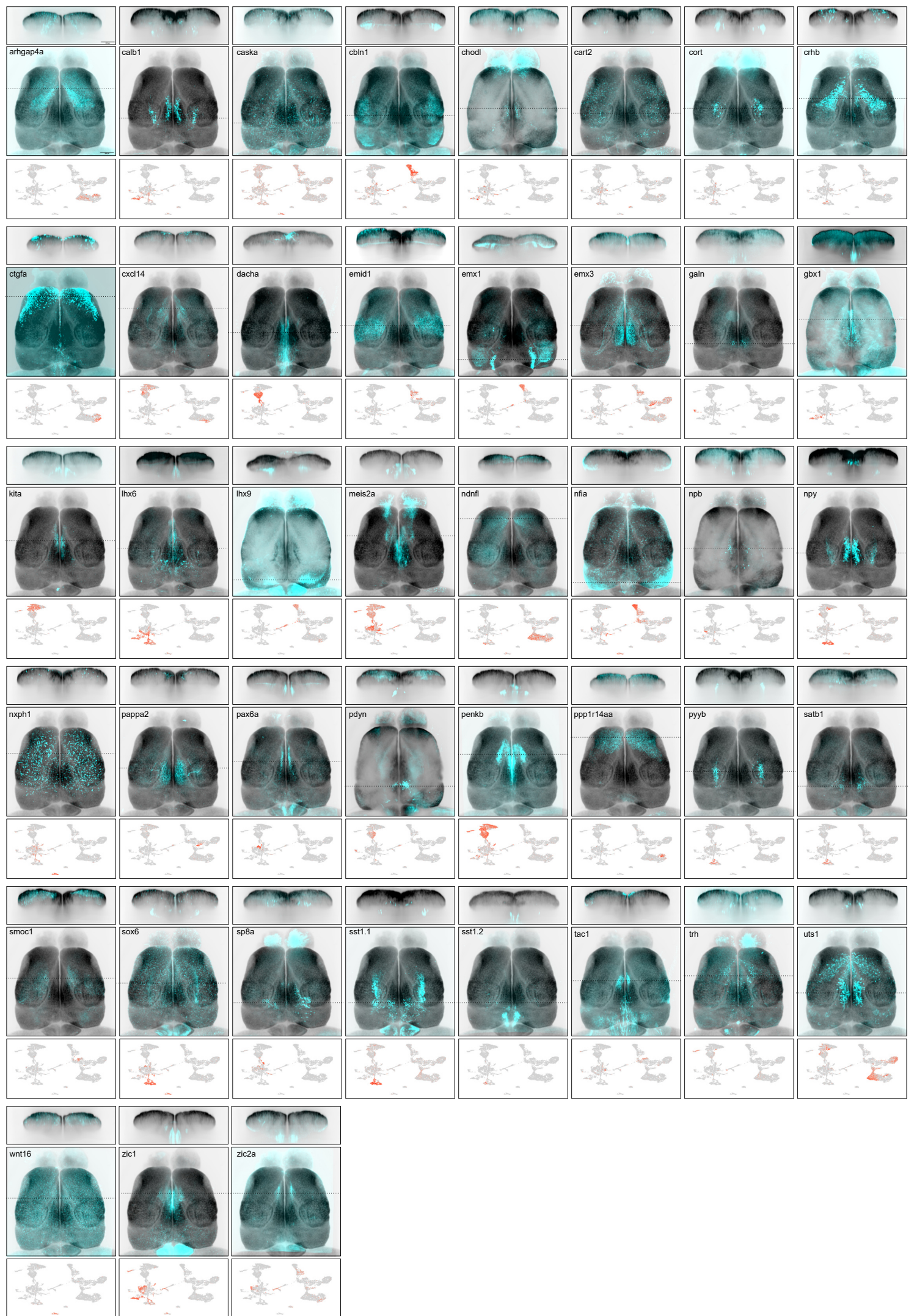


Figure S7. HCR-based visualization of marker genes for individual cell types, related to Figures 4 – 6.

Expression patterns of 43 marker genes in the adult telencephalon. Images show maximum-intensity z-projections (dorsal views) and a coronal section at the anterior-posterior position indicated by the dashed line. Expression patterns were determined by HCR (cyan) and mapped onto a common reference brain using *snap25a* expression for registration (gray background). Scale bar: approximately 200 μm (not precise due to registration of image datasets onto the reference brain). Plots below show gene expression levels in individual cells mapped onto the first two UMAP components (Figure 2C). Genes are ordered alphabetically.

Gene	Dc	DI	Dp	DM
<i>adora1b</i>	-0,95	-0,65	0,80	0,74
<i>adora2aa</i>	0,67	-0,58	-0,24	0,10
<i>adora2b</i>	1,19	-0,75	-0,45	-0,05
<i>adra2b</i>	-0,58	-0,63	0,62	0,54
<i>adra2da</i>	0,17	-0,16	0,82	-0,84
<i>adra2db</i>	0,84	-0,42	0,00	-0,46
<i>adrb1</i>	-0,49	-0,49	-0,36	1,31
<i>agtr2</i>	1,03	-0,17	-0,16	-0,72
<i>aplnra</i>	0,67	-0,08	-0,13	-0,46
<i>aplnrb</i>	1,11	-0,56	-0,23	-0,37
<i>avpr2l</i>	0,44	-0,31	0,29	-0,45
<i>calcrla</i>	1,14	-0,59	-0,71	0,12
<i>chrm2a</i>	1,23	-0,69	-0,04	-0,55
<i>drd1b</i>	0,89	-0,09	-0,34	-0,47
<i>drd3</i>	-0,76	-0,16	0,65	0,25
<i>drd4a</i>	-0,51	-0,64	-0,13	1,23
<i>ednraa</i>	0,72	-0,15	0,17	-0,74
<i>ednrba</i>	-0,47	-0,52	0,13	0,82
<i>fshr</i>	-0,33	-0,65	-0,15	1,07
<i>galr1b</i>	-0,64	-0,87	0,35	1,09
<i>galr2b</i>	-0,23	-0,52	-0,52	1,22
<i>gnrhr1</i>	-0,55	-0,49	0,83	0,17
<i>gnrhr2</i>	1,00	-0,37	-0,33	-0,33
<i>hrh2a</i>	-0,42	-0,43	-0,35	1,17
<i>hrh3</i>	0,84	0,63	-0,81	-0,61
<i>htr1b</i>	-0,70	-0,56	-0,10	1,31
<i>htr1d</i>	-0,11	-0,68	-0,37	1,10
<i>htr2cl1</i>	0,30	-0,72	0,83	-0,47
<i>htr5aa</i>	-0,67	0,28	-0,27	0,68
<i>kiss1ra</i>	1,20	-0,79	-0,66	0,19
<i>lpar1</i>	0,25	0,47	0,24	-0,92
<i>lpar2b</i>	0,78	0,08	-0,16	-0,69
<i>lpar6a</i>	0,80	-0,12	-0,10	-0,59
<i>mc3r</i>	-0,34	-0,33	-0,02	0,66
<i>mtnr1aa</i>	-0,75	-0,73	0,45	0,97
<i>mtnr1al</i>	-0,49	-0,50	-0,38	1,33
<i>mtnr1bb</i>	-0,30	-0,35	-0,14	0,76
<i>npy1r</i>	0,17	-0,22	0,66	-0,63
<i>npy4r</i>	0,65	-0,35	0,07	-0,40
<i>npy7r</i>	-0,45	-0,44	-0,18	1,04
<i>npy8ar</i>	0,98	0,23	-0,74	-0,44
<i>npy8br</i>	1,10	-0,96	0,29	-0,51

<i>opr1b</i>	-0,66	0,93	-0,71	0,51
<i>opr1</i>	0,64	0,00	-0,52	-0,12
<i>p2ry11</i>	0,42	-0,07	0,00	-0,36
<i>prlhr2a</i>	-0,53	-0,25	0,00	0,76
<i>ptger1c</i>	0,87	-0,76	0,57	-0,75
<i>ptger2b</i>	-0,55	0,03	-0,42	0,94
<i>ptger4b</i>	-0,35	-0,11	-0,39	0,84
<i>rxfp2a</i>	-0,36	-0,74	-0,32	1,36
<i>s1pr1</i>	-1,21	0,43	0,03	0,79
<i>tacr1a</i>	-0,58	-0,58	1,01	0,10
<i>tacr2</i>	-0,77	-0,36	0,83	0,27
<i>tacr3a</i>	-0,71	0,17	0,41	0,14
<i>tbxa2r</i>	-0,58	0,07	-0,51	1,04
<i>trhr2</i>	1,23	-0,17	-0,51	-0,56
<i>trhrb</i>	-0,68	0,06	-0,28	0,90
<i>vipr1b</i>	0,61	-0,79	0,56	-0,44
<i>vipr2</i>	1,09	-0,84	-0,06	-0,27

Table S1: list of GPCRs and their normalized bulk expression, related to Figure 1.

SUPPLEMENTAL REFERENCES

- S1. Cosacak, M.I., Bhattarai, P., Reinhardt, S., Petzold, A., Dahl, A., Zhang, Y., and Kizil, C. (2019). Single-Cell Transcriptomics Analyses of Neural Stem Cell Heterogeneity and Contextual Plasticity in a Zebrafish Brain Model of Amyloid Toxicity. *Cell Reports* 27, 1307-1318.e3. 10.1016/j.celrep.2019.03.090.
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- S6. Cembrowski, M.S., Wang, L., Sugino, K., Shields, B.C., and Spruston, N. (2016). Hipposeq: a comprehensive RNA-seq database of gene expression in hippocampal principal neurons. *eLife* 5, e14997. 10.7554/elife.14997.