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Molecular signatures and cellular diversity during mouse habenula development

Graphical abstract



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In brief

van de Haar et al. characterize cell populations of the developing mouse habenula using single-cell RNA sequencing. They reconstruct the developmental trajectories of molecularly distinct neuronal subsets and establish physiological and connectivity properties of *Cartpt* neurons. Their analysis indicates a possible link between developing habenula neurons and human psychiatric disorders.

Highlights

- scRNA-seq reveals cellular heterogeneity of the developing mouse habenula
- Trajectory inference reconstructs developmental paths of habenula cell types
- *Cartpt* marks physiologically distinct neurons that innervate the dorsal IPN
- Developing habenular cell types align to risk loci of human psychiatric disorders



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Molecular signatures and cellular diversity during mouse habenula development

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SUMMARY

The habenula plays a key role in various motivated and pathological behaviors and is composed of molecularly distinct neuron subtypes. Despite progress in identifying mature habenula neuron subtypes, how these subtypes develop and organize into functional brain circuits remains largely unknown. Here, we performed single-cell transcriptional profiling of mouse habenular neurons at critical developmental stages, instructed by detailed three-dimensional anatomical data. Our data reveal cellular and molecular trajectories during embryonic and postnatal development, leading to different habenular subtypes. Further, based on this analysis, our work establishes the distinctive functional properties and projection target of a subtype of *Cartpt*⁺ habenula neurons. Finally, we show how comparison of single-cell transcriptional profiles and GWAS data links specific developing habenular subtypes to psychiatric disease. Together, our study begins to dissect the mechanisms underlying habenula neuron subtype-specific development and creates a framework for further interrogation of habenular development in normal and disease states.

INTRODUCTION

The habenula (Hb) plays a key role in processing reward information and mediating aversive responses to negative stimuli. It receives input from limbic areas and the basal ganglia through the stria medularis (SM), while its efferent projections form the fasciculus retroflexus (FR) and connect to mid- and hindbrain regions. The Hb has been implicated in several human diseases, such as major depressive disorder (MDD) (Hu et al., 2020; Lecca et al., 2014). How such a small group of neurons underpins a disproportionally large number of (patho)physiological functions is poorly understood but may be explained by their heterogeneous nature. Historically, the Hb is subdivided into two subnuclei: the medial Hb (MHb) and the lateral Hb (LHb). The MHb contains substance P-ergic (SP-ergic) and cholinergic neurons, projects to the interpeduncular nucleus (IPN), and is linked to addiction, anxiety, and depression (Aizawa et al., 2012; Fowler et al., 2011; Herkenham and Nauta, 1977; Molas et al., 2017; Yamaguchi et al., 2013; Zhang et al., 2016). The LHb contains glutamatergic neurons and projects to the rostromedial tegmental nucleus (RMTg), raphe nuclei, and ventral tegmental area (VTA) (Herkenham and Nauta, 1977; Quina et al., 2015). It coordinates responses to aversive stimuli, reward processing, and feeding and is implicated in addiction (e.g., Flanigan et al., 2020; Hong et al., 2011; Lammel et al., 2012; Matsumoto and Hikosaka,

2007; Meye et al., 2016; Proulx et al., 2014; Tian and Uchida, 2015). Several studies reveal further subdivisions of the Hb by identifying clusters of Hb neurons with select molecular signatures. Recent single-cell profiling studies have defined several molecularly distinct subtypes of neurons in the adult Hb (Aizawa et al., 2012; Hashikawa et al., 2020; Pandey et al., 2018; Wagner et al., 2016; Wallace et al., 2020). Interestingly, some of these molecular subtypes are specifically engaged in foot shock responses (Hashikawa et al., 2020) or linked to specific neural activity (Pandey et al., 2018), denoting the functional implications of their diversity. However, the functional properties and connectivity patterns of most molecularly distinct Hb neuron subtypes remain unknown.

Mouse Hb neurons originate from progenitors in prosomere 2 (P2) between E10.5 and E18.5 (Guo and Li, 2019; Vue et al., 2007; Wong et al., 2018). After mitosis, newly born neurons migrate laterally from the ventricular zone (VZ) into the presumptive Hb (Ruiz-reig et al., 2019). There, Hb neurons send out axons instructed by molecular cues in the environment (Funato et al., 2000; Ichijo and Toyama, 2015; Schmidt et al., 2014). While the functional roles and connectivity patterns of the adult mammalian Hb have been studied in great detail, much less is known about the mechanisms underlying mouse Hb morphogenesis (Schmidt and Pasterkamp, 2017). Further, how distinct subtypes of MHb and LHb neurons





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emerge and develop, or contribute to disease, remains unexplored.

Here, we combine single-cell transcriptional profiling of the developing mouse Hb with molecular, anatomical, and functional approaches to unveil developmental trajectories leading to different subtypes of adult Hb neurons. Using these data, we study the developmental gene expression patterns of candidate gene families, identify and characterize a subtype of poorly defined *Cartpt*⁺ Hb neurons, and link clusters of developing Hb neurons to genetic risk variants of psychiatric disease. Together, our data constitute a framework for future studies on Hb development in health and disease.

RESULTS

Three-dimensional analysis of habenula development

To study the development of neuronal heterogeneity in the mouse Hb, we aimed to perform single-cell RNA sequencing (scRNA-seq) at key developmental time points. To supplement published data on Hb development, we performed whole-brain immunohistochemistry, three-dimensional (3D) imaging of solvent-cleared organ (3DISCO) tissue clearing, and fluorescence light sheet microscopy (FLSM) (Adolfs et al., 2021). These data were then used to aid selection of developmental stages for scRNA-seq (Figures 1A and S1).

Neuropilin-2 (NRP2), roundabout-1 (ROBO1), and ROBO3 were used to label Hb neurons and projections (Belle et al., 2014; Giger et al., 2000). Tyrosine hydroxylase (TH) enabled visualization of the midbrain (Mb) dopamine system, an important Hb target and source of afferent projections (Schmidt et al., 2014). From E11 onward, NRP2⁺ neurons were detected in P2. These Hb neurons sent out NRP2⁺/ROBO1⁺/3⁺ axons toward the Mb, forming the FR. At E12, FR axon number had increased and TH⁺ pioneer axons extended to the LHb (Schmidt et al., 2014). At E13, the NRP2⁺ SM projected into the Hb, and the first NRP2⁺/ROBO3⁺ axons crossed the midline of the Mb (Figure S1). At E15, the number of crossing axons had increased, TH⁺ axons were restricted to the FR sheath region, and the NRP2+/ROBO1+ Hb commissure (Hbcom) formed. At E18, Hb size and connectivity had further increased. Based on these data and available literature, different embryonic stages were selected for scRNA-seq.



P4 and P7 were included to cover processes such as synaptogenesis and an adult stage to enable trajectory reconstruction toward adult subtypes.

scRNA-seq identifies different subtypes of habenula neurons during development

Next, we characterized the transcriptomic expression patterns of developing Hb neurons. To specifically isolate Hb neurons, fluorescence-activated cell sorting (FACS) of LacZ⁺ neurons from Brn3a-tauLacZ mice was performed (Figures 1B-1D, S2A, and Table S1). Brn3a, also known as Pou4f1, is a marker of postmitotic Hb neurons (Eng et al., 2001) that is expressed in \pm 80% of MHb and 55% of LHb neurons (Figure S2B). Importantly, endogenous BRN3A expression is not affected in Brn3a-tauLacZ heterozygous mice (Trieu et al., 2003). FACS-based CEL-seq2 scRNA-seq resulted in transcriptional profiles of 5,756 cells across eight ages expressing 38,068 genes (Figures 1E, S2C-S2E, S3A-S3C, and Table S1). Louvain clustering resulted in 14 distinct clusters containing cells derived from different developmental stages (Figures 1E and S3D-S3F). Based on the expression of published marker genes (Chatterjee et al., 2015; Hashikawa et al., 2020; Lee et al., 2017; Lipiec et al., 2019; Wagner et al., 2016; Wallace et al., 2020), we annotated Vim⁺/Sox2⁺ progenitor cells (PCs; cluster 12 and 13), Dcc⁺/Cntn2⁺ immature Hb (iHb) neurons (6 and 8), Gap43⁺/Un5d⁺ LHb neurons (11), Chat⁺ cholinergic ventral MHb neurons (2, 3, 5, 7, and 9), and Tac1⁺ SP-ergic dorsal MHb neurons (0 and 10) (Figure 1E and Table S2). The remaining Robo2⁺ (1) and Pcdh7⁺ (1 and 4) clusters predominantly contained E11 and E12 neurons. Visualization of the top differentially expressed genes (DEGs) per cluster showed that most clusters were characterized by a combination of DEGs, rather than by unique marker genes (Figure 1F). Together, these results unveil a high level of cellular heterogeneity in the developing mouse Hb.

The generated Hb scRNA-seq dataset (DevHb) enables indepth analysis of temporal and subtype-specific expression of candidate genes that instruct important cellular processes. To highlight this potential, we examined the temporal expression of axon guidance and neurotransmission genes (as determined by selected Gene Ontology [GO] terms) (Figures 1G, 1H, and S4A). This revealed, for example, that expression of the axon

Figure 1. Analysis of the developing mouse habenula using FLSM and scRNA-seq

(A) Top: schematic representation of the developing mouse Hb. Bottom: whole-mount immunostaining for NRP2 or TH of mouse embryos followed by 3DISCO and FLSM. Caudal is to the right, dorsal to the top. FR, fasciculus retroflexus; Hb, habenula; SM, stria medularis; P2, prosomere 2.

(G and H) Matrix plot of selected axon guidance- and neurotransmission-related genes per developmental stage.

⁽B) Left: *in situ* hybridization for *Pou4f1/Brn3a* in the E18 mouse Hb (from Allen Brain Atlas). Right: whole-mount β -galactosidase-staining-cleared brain. Top view. (C) Overview of *Brn3a-tauLacZ* locus, in which tauLacZ is incorporated under the *Brn3a* promotor, replacing exon 2 and part of exon 1. Strategy for labeling *Brn3a-tLacZ*⁺ neurons for FACS by β -galactosidase-mediated hydrolysis of fluorescein di- β -D-galactopyranoside (FDG) to fluorescent fluorescein mono- β -D-galactopyranoside (FMG).

⁽D) Developmental and experimental timelines showing sampling stages. n = 2 samples per stage (E12 = 1 sample); N = 2–11 brains per sample (Table S1). E, embryonic; P, postnatal.

⁽E) t-distributed stochastic neighbor embedding (t-SNE) showing clusters as detected by the Louvain algorithm per single cell. Right: t-SNE embedding colorcoded by developmental stage. d, dorsal; i, immature; L/I, lateral; PC, progenitor cell; v, ventral.

⁽F) Dot plot of the scaled and normalized expression of the top 4 DEGs per cluster calculated with Wilcoxon rank-sum test after filtering (minimum normalized expression within cluster = 0.25, maximum normalized expression in other groups = 0.5, minimum fold change = 2).

⁽I) Dot plot of the normalized expression of axon guidance-related genes per cluster.

⁽J) t-SNE embedding showing the expression axon guidance-related genes.

Scale bar: 200 μm (A), 400 μm (B). See Figures S1–S4, Table S1.







Figure 2. Neuronal heterogeneity in the mouse habenula is established at early postnatal stages

(A) Graphical overview of mouse and zebrafish development. Symbols correspond to the respective publications. Arrows indicate sampling time points of each dataset. dpf, days postfertilization; hpf, hours postfertilization.

(B–D) Correlation matrices showing the correlation (r) between clusters of the DevHb (green), Hashikawa (blue), and Wallace datasets (orange). Disc size indicates p value and a dot statistical significance (p < 0.05, Spearman rank order correlation analysis). Labels are referring to the original cluster labels as defined in the respective publications. d, dorsal; i, immature; L/l, lateral; PC, progenitor cell; v, ventral.

guidance receptor *Dcc* decreased toward adulthood, while *Nrp2* expression increased. Further, members of the same axon guidance receptor families often displayed very distinct temporal expression patterns (e.g., EPH receptors). Most neurotransmission-related transcripts were not expressed. If expression was detected, this was predominantly from postnatal stages onward, although a few genes were already detected at E11 (e.g., *Tac1* and *Scl17a*6) (Figures 1H and S4A). In addition, several of the selected genes displayed cluster-specific or -enriched expression (e.g., *Tac1* and *Ephb2*) (Figures 1F, 1I, 1J, and S4B). In all, this analysis revealed unique spatiotemporal and subtype-specific expression patterns that will help to explain how Hb neuron subtypes develop and mature.

Cellular heterogeneity is established during postnatal development

Next, we asked how the clusters sampled from the developing habenula (DevHb) relate to adult Hb clusters reported previously (Hashikawa et al., 2020; Wallace et al., 2020). This was done by correlation of the clusters originally defined in these studies to the DevHb clusters (Figure 2A), using the average expression of DEGs per cluster. First, the two adult datasets were compared and a strong correlation was found (r = 0.8-0.95) (Figure 2B). Correlation analysis of the DevHb clusters to the adult clusters showed that DevHb MHb populations correlated to the reported adult MHb clusters, while DevHb LHb cluster 11 correlated to all adult LHb populations (Figures 2C and 2D). As expected, DevHb clusters containing PCs or iHb (6, 8, 12, and 13) did not correlate well with adult clusters. Using these results and expression of previously determined cluster-specific marker genes (Hashikawa et al., 2020; Wagner et al., 2016; Wallace et al., 2020), we prepared a schematic overview of which cluster(s) correlated and where these clusters were anatomically located in the adult Hb (Figure 2E). This showed that the DevHb dataset captured the previously reported molecular heterogeneity of the adult MHb and confirmed the division of the MHb into dorsal, superior, ventral, and lateral parts. In contrast, only one LHb cluster was found in the DevHb, while multiple adult LHb clusters have been reported (Figures 2C and 2D). This is presumably a result of using the Brn3a-tauLacZ mouse in which the MHb is more extensively labeled then the LHb. The high correlation between the adult LHb populations and DevHb clusters (1, 4, and 6) is most likely due to the fact that genes, such as Pcdh10 and Gap43, are expressed in all Hb neurons during development but become restricted to the LHb at adult stages (Figure S4C).

To assess Hb development cross-species, we compared the developing mouse Hb to the zebrafish Hb (Pandey et al., 2018). The zebrafish Hb is an asymmetric structure subdivided in the dorsal (similar to mouse MHb) and ventral Hb (similar to mouse LHb). Correlation of the DevHb clusters to larval and adult



zebrafish data showed that mouse MHb clusters correlate to dorsal Hb zebrafish clusters and the mouse LHb cluster to ventral Hb zebrafish clusters (Figures S5A and S5B). Further, iHb populations (6 and 8) strongly correlated to the immature larval zebrafish Hb population (La_Hb13), while mouse PCs (12 and 13) correlated to adult zebrafish PCs (Ad_16) (Figures S5A and S5B). Interestingly, the heterogeneity found in the adult zebrafish Hb was already present at 10 days postfertilization (dpf) (Pandey et al., 2018), a stage analogous to mouse postnatal stages (Figure 2A). To determine when Hb heterogeneity is established, cluster occupancy in the DevHb data was plotted per developmental stage (Figure 2F). PC and iHb populations were primarily abundant between E11 and E18 and mostly absent by P4. Clusters that strongly correlated to adult clusters appeared latest at P4 (0-3, 5, 7, and 9-11). Correlation of subclusters per stage showed that adult mouse clusters correlated most strongly to postnatal and not embryonic subclusters (Figure S5C).

Developmental trajectories of mouse habenula neurons are marked by specific gene expression patterns

Next, we wanted to identify the cellular and molecular programs that instruct Hb subtype development. Analysis of the DevHb data using CellRank trajectory inference identified PC cluster 13 as the root of the Hb developmental trajectory (Figures 3A-3C). From PC13, a decrease of the early marker genes Id3 and Vim was found along latent time followed by a trajectory to iHb (yellow) (cluster 6), marked by increased Cntn2 and Dcc expression. Then the trajectory split, leading to LHb neurons and dorsal and ventral subsets of MHb neurons. Dorsal MHb subsets were defined by Calb2. The dorsal MHb-superior (dMHb-S) trajectory led to Tac1⁺ SP-ergic neurons, while dorsal MHb-dorsal (dMHb-D) neurons were Calb1⁺ (Figures 3A and 3B). Both ventral MHb trajectories were marked by the cholinergic marker Chat, but a distinction could be made between Spon1⁺ ventral MHblateral (vMHb-L) and Synpr⁺ ventral MHb-ventral (vMHb-V) neurons (Figures 3A-3C). These trajectory endpoints matched the final cell types defined by correlation of the DevHb to adult clusters (Figures 2B-2D). Interestingly, according to CellRank latent time, cluster occupancy analysis and in situ hybridization (ISH), Tac1⁺ SP-neurons (cluster 0, MHb-S, and E11) arise before Chat⁺ cholinergic neurons (cluster 9, vMHb, and E18) (Figures 2F, 3D, and 3E). To identify molecular signatures of the maturational trajectories and endpoints, expression trends for the top 50 genes whose relative expression level correlates best with the probability of reaching the end stage were plotted along latent time (Figures 4A and 4B). Known PC genes (Dbi, Id3, and Vim) decreased on the trajectory from PC to iHb. Additionally, several trajectory-specific transcripts peaked along latent time, e.g., Dab1 (to LHb), Tenm3 and Pou6f2 (to dMHb-D),

(F) Occupancy plot showing relative abundance of Louvain clusters per developmental stage in the DevHb dataset. See Figure S5.

⁽E) Overview of Hb cell populations and corresponding clusters in the three datasets. This overview is based on the results of the correlation analysis (B–D) and assessment of expression of marker genes, as defined by Wallace et al. (2020) and Hashikawa et at. (2020) (blue: Hashikawa, orange: Wallace, and green: DevHb; [n] cluster number). The MHb is devided in four subsets: superior, dorsal, ventral, and lateral MHb. Because the DevHb dataset contains a limited number of LHb neurons, only one cluster could be detected. For more detailed overview of the adult LHb subset diversity, we refer to Wallace et al. (2020) and Hashikawa et al. (2020). The HbX population is found as a subset of the LHb [11] cluster of the DevHb dataset.



dMHb-D

dMHb-S [0]

[10]





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LHb [11]

NAME:

PC [13]





Figure 3. Trajectory inference reveals multiple developmental trajectories

(A and B) CellRank trajectory inference analysis of the DevHb data identifies one initial (PC) and one intermediate state (iHb) and five trajectory endpoints, plotted on t-SNE embedding. [n] Louvain cluster (A) or represented as dendrogram based (including marker genes) (B). D/d, dorsal; i, immature; L, lateral; PC, progenitor cell; S, superior; V/v, ventral.

(C) Directed partition-based graph abstraction (PAGA) showing the likelihood of cells within the cluster to reach a terminal state (i.e., trajectory endpoint, as depicted in [A]). Pie charts are color-coded according to lineage, as depicted in (B); cells are color-coded according to Louvain cluster. Arrows indicate direction of differentiation and edges the transcriptional similarity.

(D) Marker gene expression (y axis) trend per trajectory position along latent time (x axis) color-coded by trajectory. Adult coronal ISH images from Allen Brain Atlas (ABA).

(E) ISH for Tac1 and Chat in sagittal sections. Green arrows indicate Tac1 from E11 onward. Pink arrows indicate Chat from E18 onward. Tac1 is expressed in the superior medial Hb and Chat in the ventral medial Hb (ABA).

Scale bar: 200 μm (D) and 400 μm (E).

March1 (to dMHb-S), *Nkain3* (to vMHb-V), and *Ebf1* (to vMHb-L). Transcripts peaking at trajectory endpoints were mostly trajectory-specific but in some cases overlapping (e.g., *Syt6* and

Kcng4 marked both dMHb endpoints). Together, our scRNAseq data provide insight into habenula cell-type maturation trajectories and accompanied gene expression dynamics.





Figure 4. Gene expression signatures underlying habenula neuron subtype development

(A) CellRank trajectory inference analysis of the DevHb data (as in Figure 3A). D/d, dorsal; i, immature; L, lateral; PC, progenitor cell; S, superior; V/v, ventral.
 (B) Heatmaps showing temporal expression patterns (calculated using a general additive model) of the top 50 genes, whose relative expression levels correlate best with the probability of reaching the given end stage. Genes were sorted according to their peak at latent time.

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Next, we asked whether gene regulatory networks play a role in driving habenula development and cell type specification. We applied the pySCENIC algorithm to assess a possible role of specific "master" transcription factors (TFs) and effector genes (together defined as a regulon). Regulon activity was primarily associated with developmental stage but less with cellular sub-type, with the exception of PCs (Figures 4C–4E, S5D, and S5E). Interestingly, several regulons displayed specific temporal expression, e.g., early regulons (*Onecut3* and *Cux2*) and late regulons (*Mef2d*), with reported roles in embryonic development and synaptic plasticity, respectively (Figures 4D and 4E). Together, these analyses identify gene regulatory networks that mark, and may instruct, developmental trajectories to adult neuronal subtypes.

Dissection of early habenula (non)neuronal development

The DevHb dataset primarily contains postmitotic Hb neurons because of the use of *Brn3a-tauLacZ* mice. To examine in more in detail (early) Hb development and non-neuronal sub-types, scRNA-seq was performed on the whole E18 Hb. Cluster occupancy analysis indicated that at E18 cells with different maturational stages are present but that these cells are not yet fully mature (Figure 2F).

In many species, the Hb exhibits left-right asymmetry. We therefore collected left and right Hb and determined DEGs (Table S1). In total, 948 cells were analyzed that expressed a total of 28,237 genes; 5 DEGs were found in the right Hb and 11 DEGs in the left Hb (adjusted p < 0.05) (Figure S6A, Tables S3, and S4). However, this differential gene expression was not confirmed by ISH (Allen Brain Atlas), and cells from the left and right Hb showed overlapping Uniform Manifold Approximation and Projection (UMAP) placement (Figure S6B). Therefore, no distinction was made between the left and right Hb in subsequent analyses (referred to as E18 wild type [WT]). Louvain clustering identified nine distinct cell populations in the E18 WT data, which were annotated based on marker gene expression and localization (Figures 5A, S6C, and S6G). A large population of PCs was detected, which marked the trajectory starting point selected by CellRank (Figure 5B), in line with analysis of the DevHb data (Figures 3A–3C). However, in contrast to the DevHb data, analysis of E18 WT data also showed trajectories to oligodendrocyte precursor cells (OPCs), thalamus neurons (Ths) (3), and immature (thalamic) neurons (iNeurs) (6). Thus, the E18 WT dataset provides additional cellular insight into early Hb development and the generation of other (non)neuronal cell types in and around the Hb.

Early development of the mouse Hb remains incompletely understood (Schmidt and Pasterkamp, 2017). Therefore, we analyzed the spatial expression of several genes that marked different parts of the trajectory from PCs to mature Hb neurons. ISH localized the PC markers *Id3* and *Dbi* to the VZ of the Hb lining the third ventricle (Figures 5C and S6H). The Hb-specific trajectory leads from PCs into the iHb cluster (Figure 5B). *Pax6*,

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essential for epithalamus development (Chatterjee et al., 2014), was expressed in PCs and a narrow adjacent region (iHb) (Figure 5D). The iHb cluster was characterized by *Cntn2*, *Dcc*, *Flrt3*, and *Pou4f1/Brn3a* (Figures 5E, 5F, and S7I). Trajectory inference analysis showed decreasing *Cntn2* expression along the trajectory toward more mature Hb clusters (Figure 3D), which was confirmed by ISH (Figure 5F). *In vivo* BrdU pulse labeling confirmed that *Cntn2^{high}* neurons adjacent to the VZ are young neurons and that older *Cntn2^{low}* neurons are located more laterally (Figure 5G). More lateral neurons expressed *Pcp4* (Figure 5H). Thus, the transition from PCs to adult Hb neurons is a topographical process that is accompanied by sequential gene expression changes.

Cartpt expression marks a subset of physiologically distinct habenula neurons

Our data and work of others identify molecularly distinct Hb subtypes. However, whether these transcriptional identities associate with specific connectivity patterns and functional properties remains largely unexplored. To address this point, we aimed to select a molecularly defined subsets of neurons at specific anatomical locations in the Hb and with potentially interesting predicted functions or connectivity patterns. For this selection, we studied the spatial distribution of a large set of genes that label one or a few spatially defined clusters (including neurotransmission-related genes [Figures S4A and S4B]) by ISH or using a newly developed mouse intersectional genetic approach for a few selected genes. This approach relies on a tamoxifen inducible Brn3a-Flp mouse line in combination with the Ai65 reporter line and a Cre line targeting the neuronal subset of interest. Brn3a-Flp targets reporter expression to Brn3a⁺ regions, such as the Hb, while subsequent Cre expression further restricts tdTomato expression to a specific subset of Cre⁺ neurons in the Hb (Figures 6A^{I,II,IV}, S7A, and S7B). This analysis highlighted a specific population of Cartpt⁺ neurons at the boundary of the MHb and LHb, known as HbX and border zone regions (Figures 6A^{III}–6C) (Wagner et al., 2016; Wallace et al., 2020). Cartpt encodes CART peptide, which, for example, affects locomotor activity and has anti-depressant effects (Singh et al., 2021). Interestingly, despite these potent effects, the developmental characteristics, projection targets and functional properties of Cartpt⁺ Hb neurons are unknown. Therefore, we decided to focus on these neurons in subsequent experiments.

To further establish the molecular identity of *Cartpt*⁺ Hb neurons, the DevHb data were analyzed in more detail. This showed that *Cartpt* is expressed in the Hb from P4 onward, as confirmed by ISH (Figure 6D). *Cartpt*⁺ neurons were found in multiple MHb clusters (most prominently in cluster 7 (SP-ergic [*Tac1*⁺]) and 9 (cholinergic [*Chat*⁺ and also *Synpr*⁺]) that contributed to the border zone) and in the LHb cluster (11) (Figures 6E and S4B). *Cartpt*⁺ neurons of LHb cluster 11 (orange) contributed to the HbX and were marked by *Peg10* and *Gpr151* (Figure 6E). Merger

- (D) Regulon expression plotted on new t-SNE embedding (top), including target genes per regulon (bottom).
- (E) Heatmap of z scores of detected regulons per time point computed using pySCENIC. See Figure S5D and S5E.

⁽C) t-SNE embedding based on regulon expression computed with pySCENIC, color-coded by Louvain cluster (top) or developmental stage (bottom). d, dorsal; i, immature; L/I, lateral; PC, progenitor cell; v, ventral.





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of the DevHb data with two adult Hb datasets revealed *Cartpt* expression in adult MHb populations that correlate highly to DevHb clusters 7 and 9 (Figures 2C, 2D, S7C–S7F).

Next, Cartpt-cre:Ai14 mice (referred to here as Cart-cre:Ai14) were used to characterize anatomical and functional properties of Hb Cartpt⁺ neurons. Immunohistochemistry on sections of adult Cart-cre: Ai14 mice showed that, in line with ISH data (Figure 6B), tdTomato⁺ neurons locate to the HbX and border zone from P4 onward (Figure 6F). To establish their precise 3D localization, Cart-cre:Ai14 mice were subjected to 3DISCO staining and FLSM. Quantification revealed that the majority of Cartpt⁺ neurons is not only located at the border of the MHb and LHb but also at the center of the Hb along the rostral-caudal axis (Figures 6G and S7G). As the border zone contained the largest number of Cartpt⁺ neurons, we focused on these cells. Most Cartpt⁺ border zone neurons were Chat⁺ and CHAT⁺ (Figures 6E and S7H), similar to vMHb neurons. Because CHAT+ vMHb neurons project to the IPN (Albanese et al., 1985), we examined whether *Cartpt*⁺ axons also target the IPN. Indeed, in P7 and adult Cart-cre:Ai14 mice, tdTomato⁺ axons were detected in the IPN (Figure 6H), and a subset of CHAT+ axons in the IPN was tdTomato+ (Figure S7I). To unambiguously establish that Cartpt⁺ neurons innervate the IPN, we followed two strategies. First, we assessed the projection targets of Cartpt⁺ Hb neurons using the Brn3a-Flp intersectional approach (Figure 6A). tdTomato⁺ axons were found to project from the Hb to the dorsal IPN (Figure 6I). Second, dextran-488 retrograde tracer was injected in the dorso-lateral IPN of adult Cart-cre:Ai14 mice (Figures 6J and S7J). At 7 days postinjection, a large number of tdTomato⁺/dextran-488⁺ neurons were detected in the border zone (Figure 6K). Quantification showed that at least 35% of the total number of Cartpt⁺ cells projects to the dorsolateral IPN (n = 2 mice; N = 175 neurons) (Figure S7K). Conversely, dextran-488 injection into the VTA, a well-established projection site of LHb neurons, resulted in labeling of LHb, but not of the area of tdTomato⁺ neurons (Figure S7L). Together, these results identify the dorso-lateral IPN as a major target of *Cartpt*⁺ border zone neurons.

Although *Cartpt*⁺ neurons share CHAT expression and projection targets with surrounding *Cartpt*⁻ vMHb Hb neurons, they display a partially distinct molecular profile (e.g., *Cartpt* and *Sema3d* expression). To assess whether these molecular differences reflect distinct physiological properties, we performed whole-cell patch-clamp recordings using *Cart-cre:Ai14*

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mice (n = 13) (Figure 7A). Cartpt⁺ neurons (n = 17) had lower membrane capacitance than MHb (n = 15) and LHb (n = 10) Cartpt⁻ cells (Kruskal-Wallis H test (KW H(2)) = 23.18, p = 0.000; Cartpt⁺ versus MHb Cartpt⁻, p = 0.017; and Cartpt⁺ versus LHb Cartpt⁻, p = 0.000) (Figure 7B). This suggests that Cartpt⁺ neurons are smaller than neighboring MHb and LHb cells, which was confirmed by soma size measurements $(1-way ANOVA F(2,33) = 8.95, p = 0.001; Cartpt^+ versus MHb$ Cartpt⁻, p = 0.002; and Cartpt⁺ versus LHb Cartpt⁻, p = 0.001) (Figure 7C). Cartpt⁺ neurons also exhibited higher membrane resistance (1-way ANOVA F(2,41) = 6.46, p = 0.004;Cartpt⁺ versus MHb Cartpt⁻, p = 0.028; and Cartpt⁺ versus LHb Cartpt⁻, p = 0.001) (Figure 7D). Potential differences in active conductance between Cartpt⁺ and neighboring cells were also examined. Cartpt⁺ neurons exhibited a higher spontaneous firing frequency, as indicated by the number of action potentials fired at rest (KW H(2) = 12.53, p = 0.002; Cartpt⁺ versus MHb Cartp⁻, p = 0.149; and Cartpt⁺ versus LHb Cartpt⁻, p = 0.001). Instead, the action potential threshold was not significantly different (1-way ANOVA F(2,41) = 3.18)p = 0.052) (Figures 7E and 7F). The DevHb dataset showed the presence of hyperpolarization-activated cyclic nucleotidegated (HCN) channels in the Cartpt⁺ cluster (Figure S7M) and, to assess whether the DevHb dataset could be used to predict the presence of active ionic conductance profiles, we measured the occurrence of HCN-mediated voltage sags. Cartpt⁺ cells indeed exhibited voltage sags, and these were more pronounced than in neighboring Cartpt cells (KW H(2) = 8.65, p = 0.013; Cartpt⁺ versus MHb Cartp⁻, p = 0.010; and Cartpt⁺ versus LHb Cartp⁻, p = 0.372) (Figure 7G). Together, these results establish that Cartpt⁺ border zone neurons display a distinct electrophysiological profile, as compared with neighboring neurons. To assess how well one could predict whether or not an Hb neuron in the border zone belongs to the Cartpt⁺ population, unbiased cluster analysis was performed based on cell capacitance and membrane resistance. the two variables that differed most between Cartpt⁺ cells and both MHb and LHb Cartp⁻ cells. This prediction yielded two independent clusters, in which 87.5% of Cartpt⁺ and 75% of Cartpt⁻cells fell into different clusters (Figures 7H–7J).

Overall, these data provide insight into the development and molecular identity of *Ca rtpt*⁺ Hb neurons, unveil a major projection target of these neurons, and show that Cartpt⁺ neurons represent a functionally distinct population in the Hb border zone.

Figure 5. scRNA-seq analysis of the whole E18 habenula

(G) BrdU was injected at E13; embryos were harvested at E13 + 8 h, E15, and E18 and subjected to BrdU immunohistochemistry.

⁽A) Dot plot of the scaled and normalized expression of the top 5 DEG per cluster calculated with Wilcoxon rank-sum test after filtering (minimum normalized expression within cluster = 0.5, maximum expression in other groups = 0.3, and minimum fold change = 1.5). d, dorsal; I, immature; L, lateral; OPC, oligo-dendrocyte precursor cell; PC, progenitor cell; Th, thalamus; v, ventral.

⁽B) CellRank directed PAGA showing the trajectory of E18 mouse Hb cells (E18 WT). Pie charts depict the probability of cells within the cluster to reach a given lineage terminal end stage (right plot). Pie charts are color-coded according to lineage, cells are color-coded according to Louvain cluster. Arrows indicate direction of differentiation and edges the transcriptional similarity. n = 2 samples per hemisphere, plus additional mixed sample; N = 7-8 brains per sample (Table S1).

⁽C–F, H) Left: expression levels of indicated genes plotted on t-SNE embedding of E18 WT data. Right: ISH on sagittal or coronal sections of the E18 Hb (Allen Brain Atlas or this study). Hb, habenula; i, immature; Th, thalamus; VZ, ventricular zone.

Scale bar: 200 μm (C [coronal] and G) and 400 μm (C, D, E, F, and H [sagittal]). All sagittal ISH images are obtained from the Allen Brain Atlas. Hb, habenula; Th, thalamus; VZ, ventricular zone.

See Figure S6.





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Developing habenular subtypes align to GWASidentified risk loci of human disease

Finally, we assessed whether our scRNA-seq data could aid in the identification of cellular subtypes related to Hb-associated diseases, such as psychiatric disorders. By performing MAGMA gene property analysis, in which gene expression levels per cluster are gene properties (de Leeuw et al., 2015; Skene et al., 2018; Watanabe et al., 2019), the convergence of risk variants of a specific trait onto a specific cell type was investigated. For this analysis, psychiatric traits were selected for which involvement of the Hb had been implied (MDD, schizophrenia [SCZ], and bipolar disorder [BIP]) (Hou et al., 2016; Ripke et al., 2014; Wray et al., 2018). Given the involvement of the Hb in feeding behavior, body mass index (BMI) was also included (Locke et al., 2015). Height was included as a control, as no association was suspected (Wood et al., 2014) (Figure 7K). To confirm the validity of our approach, the Tabula Muris dataset was used to show that neurons are implicated in MDD and SCZ, as shown previously (Watanabe et al., 2019) (Figure 7K). Next, developmental Hb cell types (DevHb data) and adult Hb cell types were analyzed. BIP showed no association with the selected clusters, and height only associated with pericytes from the Tabula Muris dataset. BMI showed an association with DevHb clusters 1, 4 (iHb), and 11 (LHb) (Figures 7K and 7L). Interestingly, clusters composed only of adult Hb neuron subtypes did not associate with any of the tested traits and diseases. In contrast, MDD and SCZ were associated with DevHb clusters 1 and 4, and SCZ in addition with clusters 0, 2, 3, 5, 8, and 11 (Figures 7K and 7L). The known MDD risk genes Rbfox1 and Lhx2 (Wray et al., 2018) were highly expressed in clusters 1 and 4, which contain iHb that will mature into Calb2⁺ dMHb neurons and LHb neurons (Figure 7M). Further, expression of the Lhx2 regulon was particularly high at the bridge between cluster 6 (no MDD

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association) and cluster 1 (MDD association) (Figure 7M). Next, we asked whether the detected associations with MDD and SCZ are specific to DevHb neuron subtypes. To test whether the association with MDD and SCZ was Hb development-specific or associated with developing neurons in general, we tested two additional datasets, one containing developing hypothalamus neurons (Kim et al., 2020) and one containing developing Mb dopamine neurons (La Manno et al., 2016). Both datasets were randomly down-sampled to contain a similar number of cells and genes as the developmental Hb dataset. SCZ associated with developing Mb dopamine and hypothalamus neuron subsets, and, for MDD, an association with developing hypothalamic neurons was found (Figures S8A–S8D).

Overall, this analysis shows that both MDD and SCZ are associated with specific developmental Hb subtypes and, to a varying degree, with other developing neuronal subtypes.

DISCUSSION

The mouse Hb is composed of several different subtypes of neurons, but how these subtypes develop, connect, and function remains poorly understood. Progress in this regard is hampered by a lack of knowledge of the developmental gene expression profiles and trajectories of these neuronal subtypes. To address this void, we have combined scRNA-seq with molecular, anatomical, and functional approaches. Our data reveal cellular and molecular trajectories during development leading to different adult Hb subtypes. Further, our work establishes the projection target and distinctive functional properties of a subtype of $Cartpt^+$ Hb border zone neurons. Finally, comparison of DevHb clusters and genome-wide association study (GWAS) data suggests that subtypes of DevHb neurons associate with psychiatric

Figure 6. Molecular signature and projection target of Cartpt⁺ habenula neurons

(A) (I) Schematic of the *Brn3a-mTurquoiseHIS-FIpO*^{ERT2} knock-in strategy. Exon 3 of the endogenous *Brn3a* locus was exchanged with a sequence encoding HIS-tagged mTurquoise for visualization of the transgene and a tamoxifen inducible FIpO enzyme (FIpO^{ERT2}) to allow time-point-specific FIp activity. (II) Schematic of the intersectional genetic approach to specifically label *Cartpt*⁺ habenula neurons through the combinatorial activity of FIp and Cre, from *Brn3a-mTurquoiseHIS-FIpO*^{ERT2} knock-in and *Cart-IRES-cre* mice, respectively, on the Ai65 reporter line. (III) Indication the location of the labeled *Cartpt*⁺ subset within the habenula. (IV) Tamoxifen induction protocol to induce FIpO activity.

(B) Top: ISH for *Cartpt* in an E18 coronal section (Allen Brain Atlas [ABA]). Bottom: location of the border and HbX regions. Arrowheads, different populations of neurons (see E). yellow, HbX; green, substance P border; purple, cholinergic border.

(C) Immunohistochemistry for tdTomato and mTurquiose in coronal sections of P10 *Brn3a-mTurquioseHIS-FlpO*^{ERT2};*Cart-IRES-cre;Ai65* mice (N = 3) to detect *Cartpt*⁺ and *Brn3a-FlpO*⁺ cells in the habenula, respectively. Note that tdTomato *Cartpt*⁺ neurons (red) are detected in the border zone of the mTurquiose⁺ habenula. Boxed area is shown at higher magnification at the right.

(D) Top: violin plot of median Cartpt expression level per developmental stage in the DevHb dataset. Bottom: ISH for Cartpt in sagittal sections (ABA).

(E) Top: gene expression plotted on t-SNE embedding. Bottom: ISH for indicated genes in adult coronal sections.

(F) Immunohistochemistry for tdTomato (and CHAT to mark vMHb) in coronal sections of *Cart-cre:Ai14* mice (N = 3). Dotted line, Hb. White arrowheads, neurites from tdTomato⁺ HbX neurons extending into the MHb.

(G) Whole-mount immunostaining for tdTomato in adult *Cart-cre:Ai14* mice followed by 3DISCO and FLSM showing the 3-dimensional localization of *Cartpt*⁺ neurons within the habenula. *Cartpt*⁺ neurons are indicated with a spot. The third ventricle is masked in yellow and the habenula is masked in gray. Along the rostral-to-caudal axis the habenula is subdivided in 4 equally sized bins. Spots are color-coded per bin: yellow, green, blue, and pink from a rostral-to-caudal perspective. Graphs: distance (x) of each spot from the third ventricle (i.e., along the medial-to-lateral axis). Dotted line, border MHb and LHb. N = 3 mouse/age. (H) Immunohistochemistry for tdTomato, NRP2 (marks Hbcom) and/or PAX7 (marks IPN) in *Cart-cre:Ai14* mice. tdTomato⁺ fibers are observed in the Hbcom, FR, and dorsal IPN.

(I) Immunohistochemistry for tdTomato and mTurquiose in coronal sections of P10 *Brn3a-mTurquoiseHIS-FIpO^{ERT2};Cart-IRES-cre;Ai65* mice (N = 3). tdTomato + axons deriving from the Hb are observed in the dorsal IPN. Boxed area is shown at higher magnification at the right.

(J) Location Dextran-488 tracer injection (yellow) in the dorso-lateral IPN (N = 2 mice). Purple, tdTomato immunohistochemistry.

(K) Representative image of Cartpt⁺/Dextran-488⁺ double-positive neurons (white arrows) in the ventral-lateral MHb following tracer injection as in S8J.

3V, third ventricle; E, embryonic day; FR, fasciculus retroflexus; IPN, interpeduncular nucleus; LHb, lateral habenula; MHb, medial habenula; mTurq, mTurquoise; P, postnatal day; SP, substance P; TAM, tamoxifen; tdTom, TdTomato; wt, wild type. Scale bar: 20 μm (C, lower), 50 μm (I, right), 50 μm (K, lower), 75 μm (C, upper), 100 μm (E–G and H adult Hbcom), and 200 μm (A, D, H, J, and K, upper). See Figure S7.





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disorders (Figure S8G). Our study begins to dissect the molecular and cellular basis of Hb neuron subtype development and constitutes a valuable resource for future studies on Hb development in health and disease.

Cellular diversity in the developing mouse habenula

Previous studies have identified molecularly distinct neuron subtypes in the adult Hb (e.g., Aizawa et al., 2012; Amo et al., 2010; Andres et al., 1999; Hashikawa et al., 2020; Le Foll and French, 2018; Pandey et al., 2018; Wagner et al., 2016; Wallace et al., 2020). However, the ontogeny, projection targets, functional properties, and disease-associated roles of most of these subtypes are unknown. To address this key unresolved question, we performed scRNA-seq on Hb cells sorted from Brn3a-tau-LacZ mice at multiple developmental stages. Analysis of the DevHb and E18 WT data identified a large cluster of Id3⁺/Vim⁺ PCs. BrdU birth dating and neurosphere assays confirmed a population of PCs in the E18 Hb VZ (Figures S8E and S8F). Id3+/Vim+ PCs developed into iHb neurons expressing high levels of Cntn2, Dcc, and Flrt3. Then, the trajectory split into separate axes of the LHb, dMHb, and vMHb. This suggests that mouse LHb and MHb neurons originate from the same PCs. This is interesting, because equivalent zebrafish structures derive from distinct PC populations (Beretta et al., 2013). A high level of cellular heterogeneity was already detected as early as E11 in our data, but mouse Hb subtypes with adult signatures were only present by P4. Therefore, our data establish that Hb subtypes acquire adult transcriptional signatures only at early postnatal stages.

Molecular mechanisms underlying habenula development

Different aspects of Hb development have been studied (e.g., Aizawa et al., 2007; Beretta et al., 2013; Chatterjee et al., 2014; Colombo et al., 2013; Fore et al., 2020; Guglielmi et al., 2020; Hüsken et al., 2014; Kantor et al., 2004; Quina et al., 2009; Ruiz-reig et al., 2019; Schmidt et al., 2014), but how specific neuronal subtypes develop is unknown. Our study begins to address the mechanisms underlying Hb subtype-specific development. For example, we report developmental stage-specific expression or enrichment of axon guidance proteins and receptors in select developmental clusters. This was especially prominent forerythropoietin-producing hepatocellular carcinoma (EPH) receptors and Eph receptor-interacting protein (EPHRINS), a large family of axon guidance proteins involved in topographic mapping of axons (Kania and Klein, 2016). This is interesting because neurons in different Hb subdomains target specific brain (sub)nuclei (Herkenham and Nauta, 1977; Quina et al., 2015; Wallace et al., 2020) and because subdomain-specific EPHRIN/EPH expression is found in these nuclei (García-Guillén et al., 2021).

Previous studies unveiled Hb subtype-specific expression of genes associated with synaptic transmission at adult stages (e.g., Pandey et al., 2018; Wallace et al., 2020). Our data allowed spatiotemporal analysis of the expression of these genes. Expression of synaptic transmission genes was generally first detected at postnatal stages, although striking differences were observed. For example, Tac1 (SP-ergic neurons) was found in the early embryonic dMHb, while Chat (cholinergic neurons) was first detected at postnatal stages in the vMHb. Our work also identified different developmental trajectories leading to mature Hb neuron subtypes. These were marked by trajectoryspecific gene expression trends, which may contribute to the transition of PCs into MHb and LHb neuron subtypes. Further, we identified distinct TF-driven gene regulatory networks at specific developmental stages. Together, these results provide insight into the regulatory networks underlying Hb development. Further studies are needed to examine whether and how these molecular programs establish Hb cellular subtypes.

Linking molecular and functional properties of habenular subtypes

Whether molecularly defined Hb subtypes have distinctive functional properties or connections is incompletely understood. Here, we characterized $Cartpt^+$ neurons that mark the HbX and

Figure 7. Cartpt⁺ neurons in the habenula border zone have distinctive functional properties, and developing habenular subtypes associate with disease

(A) Location of electrophysiological recordings in adult Cart(pt)-cre:Ai14 mice (N = 13; image adapted from Allen Brain Atlas).

(B–G) Cellular capacitance (pF): independent samples Kruskal Wallis: H(2) = 23.18, p = 0.000; pairwise comparisons: $Cartpt^+$ versus MHb $Cartpt^-$, p = 0.001; $Cartpt^+$ versus LHb $Cartpt^-$, p = 0.000; MHb $Cartpt^-$ versus LHb $Cartpt^-$, p = 0.071 (B), cell soma Feret's diameter (representative images are shown): one-way ANOVA: F(2,33) = 8.95, p = 0.001; LSD *post hoc* tests: $Cartpt^+$ versus MHb $Cartpt^-$, p = 0.002; $Cartpt^+$ versus LHb $Cartpt^-$, p = 0.001; MHb $Cartpt^-$, p = 0.002; $Cartpt^+$ versus LHb $Cartpt^-$, p = 0.001; MHb $Cartpt^-$, p = 0.028; $Cartpt^+$ versus LHb $Cartpt^-$, p = 0.001; MHb $Cartpt^-$, p = 0.028; $Cartpt^+$ versus LHb $Cartpt^-$, p = 0.001; MHb $Cartpt^-$, p = 0.028; $Cartpt^+$ versus LHb $Cartpt^-$, p = 0.001; MHb $Cartpt^-$, p = 0.028; $Cartpt^+$ versus LHb $Cartpt^-$, p = 0.001; MHb $Cartpt^-$, p = 0.028; $Cartpt^+$ versus LHb $Cartpt^-$, p = 0.001; MHb $Cartpt^-$, p = 0.028; $Cartpt^+$ versus LHb $Cartpt^-$, p = 0.001; MHb $Cartpt^-$, p = 0.028; $Cartpt^+$ versus LHb $Cartpt^-$, p = 0.001; MHb $Cartpt^-$, p = 0.028; $Cartpt^+$ versus LHb $Cartpt^-$, p = 0.001; MHb $Cartpt^-$, p = 0.029; P = 0.001; MHb $Cartpt^-$, p = 0.029; P = 0.001; MHb $Cartpt^-$, p = 0.029; P = 0.001; MHb $Cartpt^-$, p = 0.029; P = 0.001; MHb $Cartpt^-$, p = 0.029; P = 0.001; MHb $Cartpt^-$, p = 0.002; P = 0.001; P = 0.001; P = 0.002; P = 0.001; P = 0.001; P = 0.002; P = 0.001; P = 0.001; P = 0.002; P = 0.001; P =

(H–J) Recorded cells were classified by a two-step cluster analysis of individual data on cell capacitance and Rmem using the Schwarz's Bayesian criterion. Silhouette measure of cohesion and separation revealed an overall good model fit (average silhouette 0.5). Rmem was the most important predictor of cell clustering (I). 87.5% of all *Cartpt*⁺ and 76% of all *Cartpt*⁻ neurons were classified in accordance to their genotype based on capacitance and Rmem.

(K) MAGMA analysis, cell types are color-coded by dataset. Bar plots indicate signal strength, if dashed line (after Bonferroni correction, p = 0.05/(44*5)) is crossed, signal is considered significant.

(L) t-SNE embedding color-coded by Louvain cluster. Associated clusters per trait are highlighted by dashed lines.

(M) t-SNE embedding showing the relative expression levels of MDD risk genes Rbfox1 and Lhx2, and Lhx2 regulon activity.

AP, action potential; BIP, bipolar disorder; BMI, body mass index; E, embryonic day; FR, fasciculus retroflexus; IPN, interpeduncular nucleus; MDD, major depressive disorder; MOhm, mega Ohm; ms, millisecond; mV, millivolt; pA, pico amper; pF, pico farad; Rmem, membrane resistance; SCZ, schizophrenia. See Figures S7 and S8.

the border zone, an area at the boundary of MHb and LHb (Wagner et al., 2016; Wallace et al., 2020). Cartpt encodes for cocaine- and amphetamine-regulated transcript (CART) peptide, which affects locomotor activity and has anti-depressant effects (Singh et al., 2021). However, Cartpt⁺ Hb neurons, and HbX and border zone neurons in general, remain poorly characterized. Using immunohistochemistry and genetic tools, Cartpt⁺ Hb neurons were detected from early postnatal stages onward. Interestingly, transcripts displaying an expression pattern similar to Cartpt included Sema3d (Wagner et al., 2016; Wallace et al., 2020). SEMA3D is a member of the semaphorin family of guidance proteins, with roles in axon guidance and synaptogenesis (Pasterkamp, 2012), and may play a role in regulating Cartpt⁺ Hb neuron connectivity, for example, in establishing highly distinctive patterns of afferent innervation at the border zone (Herkenham and Nauta, 1977; Shinoda and Tohyama, 1987).

Analysis of Cart-cre:Ai14 and Brn3a-FlpO;Cart-cre;Ai65 mice and retrograde tracing showed that many Cartpt⁺ border zone neurons express CHAT and send their axons to the IPN, both characteristics of MHb neurons. Despite these similarities, distinctive electrophysiological profiles were found for Cartpt+ neurons as compared with surrounding Cartpt⁻ neurons. Differences in cell capacitance and membrane resistance hinted at distinct morphology and *Cartpt*⁺ neurons had a significantly smaller soma. Higher membrane resistance indicates that Cartpt⁺ neurons are more electrotonically compact, which could be mediated by differential potassium channel expression (Tripathy et al., 2017). Finally, Cartpt⁺ neurons displayed prominent voltage sag, in accordance with predictions regarding the presence *Hcn3* and *Hcn4* ion channels in *Cartpt*⁺ neurons. Together, our data show that Cartpt⁺ border zone neurons represent a molecularly and functionally distinct subtype of Hb neurons that send prominent axon projections to dorsal IPN. The functional role of this circuitry is unknown, but the IPN mediates reward processing and depressive-like symptoms (Xu et al., 2018), in line with reported roles of CART peptide (Ahmadian-Moghadam et al., 2018).

A role for developing habenular subtypes in human disease?

Hb dysfunction is linked to depression (Hu and Fitzgerald, 2018) and GWAS of human MDD patients uncovered multiple genetic risk loci (Wray et al., 2018). MAGMA analysis of our data, and other published datasets, revealed a significant association between MDD risk variants and DevHb neurons. In the associated subtypes, high expression of the MDD risk genes Rbfox1 and Lhx2 was observed. Interestingly, Rbfox1 was especially high in young Hb neurons that eventually mature into LHb and Calb2⁺ dMHb neurons. These results unveil a link between the DevHb and MDD and hint at the possibility that changes in the development of LHb and Calb2+ dMHb neurons contribute to the etiology of this psychiatric disorder. iHb populations, and other types of developing neurons, also are associated with SCZ risk genes. This is in line with previous work connecting SCZ and embryonic neuroblasts (Watanabe et al., 2019) but contrasts with data showing lack of association between SCZ and developing neurons (Skene et al., 2018).



In summary, we have harnessed the power of scRNA-seq, in combination with other techniques, to dissect the developmental programs underlying Hb neuron diversity. Our analysis provides examples of cell state markers, developmental trajectories, and transcriptional programs. Further, it also contributes a unique 3D overview of Hb development and the functional characterization of an undefined molecular subtype of Hb neurons. Finally, it reveals that comparison of Hb scRNA-seq and disease GWAS data may provide insight into disease pathogenesis. Together, our data provide an initial framework for dissecting the developmental programs underlying Hb neuron diversity and a starting point for future interrogation of the DevHb in normal and disease states.

Limitations of the study

Although our study provides insight into the development, function, and potential links to disease of Hb neuron subtypes there are a few outstanding issues and considerations:

- (1) Pou4f1/Brn3a is a generally accepted marker of postmitotic Hb neurons. Therefore, we used Brn3a-tLacZ mice to isolate Hb neurons. By choosing this approach, nonneuronal cells, such as microglia and astrocytes, as well as about 50% of the LHb neurons were not included in the analysis. Pou4f1⁻ neurons may have a maturation trajectory distinct from the Pou4f1⁺ neurons of the MHb and LHb. Therefore, future studies are needed to characterize Pou4f1⁻ neurons.
- (2) Our data indicate that Cartpt⁺ neurons are physiologically distinct. Cartpt⁺ neurons project to the IPN, a circuit involved in reward processing and depressive-like symptoms. Given CART's putative role in similar processes, our findings suggest the involvement of Cartpt⁺ neurons in psychiatric disorders. However, further functional studies are needed to test this hypothesis.
- (3) Our MAGMA analysis of the DevHb transcriptomic profile unmasked an association with depression risk genes and links developmental changes in LHb and Calb2⁺ dMHb neurons to MDD. Future experimental work is needed to assess the functional relationship between DevHb neurons and MDD.
- (4) Compared with other previous scRNA-seq studies, the number of cells assessed in our study is relatively low. However, the quality of the cells analyzed is high, which is beneficial when studying a relatively homogeneous (*Brn3a*⁺) neuron population.

STAR***METHODS**

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SUPPLEMENTAL INFORMATION

Supplemental information can be found online at https://doi.org/10.1016/j. ceirep.2022.111029.

ACKNOWLEDGMENTS

We thank Dinos Meletis for input on our manuscript; Eric Turner for *Brn3a-tau-LacZ* mice; and Kiochi Hashikawa, Garret Stuber, Michael Wallace, Bernardo Sabatini, Shristi Pandey, and Alexander Schier for sharing data. This research was supported by the Netherlands Organisation for Scientific Research (ALW-VICI 865.14.004 to R.J.P.), ENW-VENI (016.Veni.192.188 to D.R.), and ERC under the European Union's Horizon 2020 research and innovation programme (804089; ReCoDe to F.J.M.) and partially supported by the NWO Gravitation program BRAINSCAPES: Roadmap from Neurogenetics to Neurobiology (NWO: 024.004.012) and Stichting Parkinson Fonds (to R.J.P.).

AUTHOR CONTRIBUTIONS

L.L.v.d.H. and R.J.P. designed the study and wrote the manuscript with help from all authors. L.L.v.d.H., D.R., and O.G. designed and performed experiments with help from J.E.B., T.E.S., R.E.v.D., Y.A., C.V.D.M., and N.C.H.v.K.; and M.H.B., D.P., and K.W. helped with performing MAGMA analysis. F.J.M. and O.B. designed experiments and aided in data analysis.

DECLARATION OF INTERESTS

The authors declare no competing interests.

INCLUSION AND DIVERSITY

One or more of the authors of this paper self-identifies as an underrepresented ethnic minority in science. One or more of the authors of this paper self-identifies as a member of the LGBTQ+ community.

Received: November 22, 2021 Revised: April 18, 2022 Accepted: June 10, 2022 Published: July 5, 2022

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STAR*METHODS

KEY RESOURCES TABLE

REAGENT or RESOURCE	SOURCE	IDENTIFIER
Antibodies		
goat anti-Neuropilin-2	R&D Systems	Cat#AF567 RRID:AB_2155253
rabbit anti-Tyrosine Hydroxylase	EMD Millipore	Cat#TH152 RRID:AB_390204
goat anti-ROBO1	R&D Systems	Cat#AF1749 RRID:AB_354969
goat anti-ROBO3	R&D Systems	Cat#AF3076 RRID:AB_2181865
rabbit anti-RFP	Rockland	Cat#600-401-379 RRID:AB_2209751
anti-DIG-AP	Sigma	Cat#11207733910 RRID:AB_514500
Rat anti-BrdU	Abcam	Cat#AB6326 RRID:AB_305426
Goat anti-CHAT	EMD Millipore	Cat#AB144P RRID:AB_2079751
Rat anti-RFP	ChromoTek	Cat#5f8-100 RRID:AB_2336064
Chicken anti-GFP	Aves	Cat#GFP-1020 RRID:AB_10000240
Donkey Anti-Goat IgG H&L (Alexa Fluor 647)	Abcam	Cat#ab150135 RRID:AB_2687955
Donkey Anti-Goat IgG H&L (Alexa Fluor 488)	Invitrogen	Cat#A11055 RRID:AB_2534102
Donkey Anti-Rat IgG H&L (Alexa Fluor® 568) preadsorbed	Abcam	Cat#ab175475 RRID:AB_2636887
Donkey Anti-Chicken IgY Antibody (Alexa Fluor® 488)	Jackson ImmunoResearch	Cat#703-545-155 RRID:AB_2340375
Donkey Anti-Rabbit IgG H&L (Alexa Fluor 750)	Abcam	Cat#ab175731
Donkey Anti-Goat IgG H&L (Alexa Fluor 568)	Abcam	Cat#ab175704 RRID:AB_2725786
Donkey Anti-Rat IgG H&L (Alexa Fluor 568)	Abcam	Cat#ab175475 RRID:AB_2636887
Bacterial and virus strains		
DH5-alpha	Invitrogen	Cat#18263012
AAV-hSyn-DIO-mCherry	Addgene / UNC Vector Core	Cat# 50459-AAV5
HSV-hEF1a-LS1L-mCherry (RN413)	from Rachael Neve	N/A
Chemicals, peptides, and recombinant proteins		
HBSS – Ca ²⁺ - Mg ²⁺	Thermo-Fisher	Cat#24020117
MACS Neural Dissociation kit (T)	Miltenyi Biotec	Cat#130-093-231
Hibernate A – Ca ²⁺ - Mg ²⁺	BrainBits	Cat#HACAMG500
B27 supplement	Thermo-Fisher	Cat#17504044
L-glutamine	Thermo-Fisher	Cat#25030081
Bovine Serum Albumine	Jackson	Cat#001-00-162

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REAGENT or RESOURCE	SOURCE	IDENTIFIER
L15 medium	Thermo-Fisher	Cat#11415049
glutaMAX DMEM/F-12	Thermo-Fisher	Cat#10565018
Penicillin-Streptomycin	Thermo-Fisher	Cat# 15140122
EGF recombinant human protein	Thermo-Fisher	Cat# PHG0311
FGF-Basic Recombinant Human Protein	Thermo-Fisher	Cat#PGH0367
HEPES	Thermo-Fisher	Cat#15630106
Methanol	Merck Millipore	Cat#1060092500
Hydrogen peroxide	Merck Millipore	Cat#1072091000
Triton-X-100	Sigma	Cat#x100-500ml
Thimerosal	Gerbu	Cat#1031/USP35
Saponin	Sigma	Cat#S7900
Tetrahydroflurane	Sigma	Cat#186562-1L
Dichloromethane	Sigma	Cat#270997-1L
Dibenzyl ether	Sigma	Cat#108014-1kg
Heparin	Sigma	Cat#H3393
Formaldehyde	Riedel-de Hain	Cat#33200
Glutaraldehyde	Acros Organics	Cat#119989925
Sodium deoxycholate	Sigma	Cat#D6750-100G
K-ferricyanide	Merck	Cat#231-847-6
K-ferrocyanide	Sigma	Cat#p8131-100G
Magnesium Chloride	Merck	Cat#A0344733339
DNase	Roche	Cat#776785
EDTA	Sigma	Cat#E5134-500G
Lithium Chloride	Merck	Cat#7447-41-8
Triethanolamine	Fluka	Cat#90279
Acetic anhydride	Sigma	Cat#A6404
Deionized formamide	ICN	Cat#800686
Ficoll-400	Sigma	Cat#F4375
Polyvinylpyrrolidone	Merck	Cat#7443
BSA-fraction V	ICN	Cat#103703
tRNA baker's yeast	Sigma	Cat#R-6750
SonifiCated Salmon Sperm DNA	Sigma	Cat#D-91565
NBT/BCIP	Boehringer Mannheim	Cat#1681451
Levamisol	Sigma	Cat#L9756
Corn oil	Sigma	Cat#C8267-500ML
Tamoxifen	Sigma	Cat#T5648-1G
Bromodeoxyuridine	Sigma	Cat#B5002
Normal Donkey Serum	Jackson Immunoresearch	Cat#017-000-121
Dextran Alexa Fluor 488	Invitrogen	Cat# D22910
Ketamine Narketan 10	Vétoquinol	GTIN 03605877551210
Medetomidine Sedastart	Ast Farma B.V.	N/A
Lidocaine	B. Braun	N/A
Atipamezole Alzane	Syva	N/A
Carpofen Carpofelican	Dechra	N/A
Sodium pentobarbital Euthanimal		N/A
Fluorsave	VWR international	Cat# 345789-20P
DAPI - 4',6-diamidino-2-phenylindole	Sigma	Cat#D9564

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REAGENT or RESOURCE	SOURCE	IDENTIFIER
Critical commercial assays		
FluoReporter lacZ Flow	Thermo-Fisher	Cat#F1930
Cytometry Kit		
PureLink Quick Gel Extraction kit	Thermo-Fisher	Cat#K210025
pGEM-T-easy vector system	Promega	Cat#A1360
Deposited data		
FastQ devHb	This paper	GEO: GSE188712
Count table devHb	This paper	GEO: GSE188712
FastQ E18WT	This paper	GEO: GSE188712
Count table E18WT	This paper	GEO: GSE188712
Scripts (github)	This paper	https://doi.org/10.5281/zenodo.6536116
Experimental models: Organisms/strains		
Mouse: C57BL/6j	Charles River	Cat#027 IMSR_JAX:000664
Mouse: Brn3a-tauLacZ	gift from prof. Eric Turner Seattle Children's Hospital	N/A
Mouse: Cart-IRES2-cre-D	The Jackson Laboratory	JAX:028533
Mouse: STOP-tdTomato	The Jackson Laboratory	JAX: 007914, RRID:IMSR_JAX:007914
Mouse: Brn3a-mTurquoiseHIS- FlpOERT2	This study	N/A
Mouse: Ai65D	The Jackson Laboratory	JAX: 021875. RRID:IMSR JAX:021875
Oligonucleotides		
See Table S5		N/A
Software and algorithms		
Scanpy		https://github.com/satijalab/seurat
Scanpy Seurat v3		https://github.com/satijalab/seurat https://github.com/theislab/scanpy
Scanpy Seurat v3 scVELO	Bergen et al., 2020	https://github.com/satijalab/seurat https://github.com/theislab/scanpy RRID:SCR_018168
Scanpy Seurat v3 scVELO	Bergen et al., 2020	https://github.com/satijalab/seurat https://github.com/theislab/scanpy RRID:SCR_018168 https://github.com/theislab/scvelo
Scanpy Seurat v3 scVELO CellRank	Bergen et al., 2020 Aibar et al., 2017	https://github.com/satijalab/seurat https://github.com/theislab/scanpy RRID:SCR_018168 https://github.com/theislab/scvelo https://github.com/theislab/cellrank
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Scanpy Seurat v3 scVELO CellRank MapAndGo2 STARmap Python version 3.7.6 R version 4.0.3	Bergen et al., 2020 Aibar et al., 2017 Anna Alemany Python Software Foundation The R Foundation for Statistical Computing	https://github.com/satijalab/seurat https://github.com/theislab/scanpy RRID:SCR_018168 https://github.com/theislab/scvelo https://github.com/theislab/cellrank https://github.com/anna-alemany/ transcriptomics/tree/master/mapandgo RRID:SCR_008394 https://www.python.org/ RRID:SCR_001905 https://www.r-project.org/
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Scanpy Seurat v3 scVELO CellRank MapAndGo2 STARmap Python version 3.7.6 R version 4.0.3 Louvain MAST MAGMA pySCENIC Zen 2 (blue edition) Imspector software version 5.0285.0 Imaris version 8.4-9.4 Fiji version 2.0.0-rc-69/1.520 & version 1.53f51 PatchMaster v2.73 Igor Pro 8.0.4.2	Bergen et al., 2020 Aibar et al., 2017 Anna Alemany Python Software Foundation The R Foundation for Statistical Computing vtraag/louvain-igraph: 0.6.1 Finak et al., 2015 de Leeuw et al., 2015 Aibar et al., 2017 Zeiss LaVision BioTec Bitplane Fiji HEKA Elektronik Wavemetrics	 https://github.com/satijalab/seurat https://github.com/theislab/scanpy RRID:SCR_018168 https://github.com/theislab/sevelo https://github.com/theislab/cellrank https://github.com/anna-alemany/ transcriptomics/tree/master/mapandgo RRID:SCR_008394 https://www.python.org/ RRID:SCR_001905 https://www.r-project.org/ https://doi.org/10.5281/zenodo.595481 RRID:SCR_016340 https://github.com/RGLab/MAST https://github.com/aertslab/pySCENIC N/A
Scanpy Seurat v3 scVELO CellRank MapAndGo2 STARmap Python version 3.7.6 R version 4.0.3 Louvain MAST MAGMA pySCENIC Zen 2 (blue edition) Imspector software version 5.0285.0 Imaris version 8.4-9.4 Fiji version 2.0.0-rc-69/1.520 & version 1.53f51 PatchMaster v2.73 Igor Pro 8.0.4.2 IBM SPSS Statistics version 27	Bergen et al., 2020 Aibar et al., 2017 Anna Alemany Python Software Foundation The R Foundation for Statistical Computing vtraag/louvain-igraph: 0.6.1 Finak et al., 2015 de Leeuw et al., 2015 Aibar et al., 2017 Zeiss LaVision BioTec Bitplane Fiji HEKA Elektronik Wavemetrics IBM Corp	https://github.com/satijalab/seurat https://github.com/theislab/scanpy RRID:SCR_018168 https://github.com/theislab/scvelo https://github.com/theislab/cellrank https://github.com/anna-alemany/ transcriptomics/tree/master/mapandgo RRID:SCR_008394 https://www.python.org/ RRID:SCR_001905 https://www.r-project.org/ https://doi.org/10.5281/zenodo.595481 RRID:SCR_016340 https://github.com/RGLab/MAST https://github.com/aertslab/pySCENIC N/A N/A RRID:SCR_002285 https://imagej.net/Fiji N/A N/A
Scanpy Seurat v3 scVELO CellRank MapAndGo2 STARmap Python version 3.7.6 R version 4.0.3 Louvain MAST MAGMA pySCENIC Zen 2 (blue edition) Imspector software version 5.0285.0 Imaris version 8.4-9.4 Fiji version 2.0.0-rc-69/1.520 & version 1.53f51 PatchMaster v2.73 Igor Pro 8.0.4.2 IBM SPSS Statistics version 27 Graphpad Prism version 9.1.1	Bergen et al., 2020 Aibar et al., 2017 Anna Alemany Python Software Foundation The R Foundation for Statistical Computing vtraag/louvain-igraph: 0.6.1 Finak et al., 2015 de Leeuw et al., 2015 Aibar et al., 2017 Zeiss LaVision BioTec Bitplane Fiji HEKA Elektronik Wavemetrics IBM Corp GraphPad Software	https://github.com/satijalab/seurat https://github.com/theislab/scanpy RRID:SCR_018168 https://github.com/theislab/scvelo https://github.com/theislab/cellrank https://github.com/anna-alemany/ transcriptomics/tree/master/mapandgo RRID:SCR_008394 https://www.python.org/ RRID:SCR_001905 https://www.r-project.org/ https://doi.org/10.5281/zenodo.595481 RRID:SCR_016340 https://github.com/RGLab/MAST https://github.com/RGLab/MAST https://github.com/aertslab/pySCENIC N/A N/A RRID:SCR_007370, https://imaris.oxinst.com/ RRID:SCR_002285 https://imagej.net/Fiji N/A N/A RRID:SCR_002798



RESOURCE AVAILABILITY

Lead contact

Further information and requests for resources and reagents should be directed to and will be fulfilled by the Lead Contact, Jeroen Pasterkamp@umcutrecht.nl).

Materials availability

All unique reagents, plasmids, and transgenic mouse lines generated in this study are available from the lead contact with a completed Materials Transfer Agreement.

Data and code availability

FastQ files and count matrices have been deposited at NCBI GEO and are publicly available as of the date of publication. Accession numbers are listed in the key resources table. All code is deposited at Github: https://github.com/liekevandehaar/Dev_Hb_mouse and is publicly available as of the date of publication. DOI is listed in the key resources table. Any additional information required to reanalyze the data reported in this paper is available from the lead contact upon request.

EXPERIMENTAL MODEL AND SUBJECT DETAILS

Mouse lines

All animal experiments in this study were approved by the (CCD) Centrale Commissie Dierproeven of Utrecht University (CCD license: AVD115002016532) and were in accordance with Dutch law (Wet op de Dierproeven 2014) and European regulations (guideline 2010/ 63/EU). *Brn3a-tauLacZ* (Eng et al., 2001) (gift from prof. Eric Turner), *Cart-IRES2-cre-D* (The Jackson Laboratory, JAX:028533), *Brn3a-mTurquoiseHIS-FIpO*^{ERT2} (this paper), *Ai65D* (The Jackson Laboratory, JAX:021875), *STOP-tdTomato (Ai14)* (The Jackson Laboratory, JAX:007914) and *C57BL/6j* mice (Charles Rivers) were housed at 22 ± 1°C on a wood-chip bedding supplemented with tissue on a 12 h/12 h day/night cycle. Pregnant mothers were housed individually from the moment of observation of the plug (E0.5). Animals were fed *ad libitum* and mice from both sexes were used. Frozen sperm of *Brn3a-tauLacZ* males was provided by prof. Eric Turner. Animals were generated as described (Eng et al., 2001). In short, a tauLacZ reporter cassette was inserted in part of the first and second exon of the *Brn3a* locus. Frozen sperm was recovered and viable mice were generated at Janvier Labs (France). Mice were selected for the presence of the *Brn3a-tauLacZ* locus. Because homozygous animals die shortly after birth, animals were kept heterozygous on a *C57BL/6j* background. Animals were genotyped using the primers listed in Table S5. For the electrophysiological recordings and neuronal tracing experiments mice between P90 and P130 of both sexes were used. For the scRNA-seq experiments E11, E12, E13, E15, E18, P4, P7 and adult (P245 (male) and P259 (female)) were used. *Brn3a-mTurquoiseHIS-FIpO*^{ERT2} characterization was performed at P10. At embryonic and postnatal stages until P10, the sex of the animals was not considered.

Generation of Brn3a-mTurquoiseHIS-FlpOERT2 mice

Brn3a-mTurquoiseHIS-FIpO^{ER72} were generated at the Swiss Federal Institute of Technology Phenomic Center (ETH – EPIC, Zurich, Switzerland). Briefly, a *T2A-mTurquoiseHIS-P2A-FIpO* fragment was commercially synthesized and inserted upstream of Neomycin in a *lox511-Neo-lox511* vector downstream of the *lox511-Neo-lox511* cassette to create the targeting vector. Next, two regions homologous to the *Brn3a* gene (generated by proofreading PCR of BAC clone RPCIB731J17321Q (containing the *Brn3a* WT gene) as a template) were added to the targeting vector. Subsequently, the targeting vector was plasmid was linearized using restriction enzymes and electroporated into ES cells. 24 h later ES cells were screened based for neomycin resistance followed by PCR for the homologous regions. This resulted in 6 positive clones. Blastocysts were then injected with these ES clones and transferred into foster mothers. Three individual rounds of blastocyst injections resulted in a total of 20 chimeric mice. Chimeric offspring was bred one round with *C57BL/6N* animals to establish germline transmission and obtain a total number of 7 founder mice heterozygous for the *Brn3a-mTurquoiseHIS-FIpO^{ER72}* knock-in allele. Animals were genotyped for the presence of the knock-in allele using the primers listed in Table S5 and the following PCR protocol: 98°C 3 min, 35x 95°C 30 s – 56°C 30 s – 72°C 30 s, 72°C 2 min.

Neurosphere generation

Cells were isolated as described previously (Guo et al., 2012). Adult mice were sacrificed by cervical dislocation. Brains were isolated and placed in ice-cold 1x PBS. The Hb and the subventricular zone (SVZ) were dissected in ice-cold L15++ medium (L15 (Thermo-Fisher, Cat#11415049) supplemented with 7.5 mM HEPES (Thermo-Fisher Cat#15630106) and 1 x penicillin/streptomycin (pen/strep; Thermo-Fisher, Cat# 15140122)). Tissue was dissociated using the MACS Neural Dissociation kit (Miltenyi Biotec, Cat#130-093-231) according to the manufacturer's protocol. In short, tissue was incubated in pre-heated dissociation solution (200 µl Trypsin and 1750 µl solution X) at 37°C for 15 min. 20 µl solution Y and 10 µl solution A were added per sample. Tissue was dissociated with a wide-tipped glass pipet. The samples were incubated at 37°C for 10 min. Tissue was further dissociated with a series of 2 glass pipettes with descending tip sizes. 5 ml glutaMAX DMEM/F-12 (Thermo-Fisher, Cat#10565018) supplemented with 1x pen/strep was added to each sample. The suspension was filtered using a 40 µm cell strainer into a 15 ml Falcon tube and centrifuged at 300 x g for



5 min. The cell pellet was washed with L15++ medium and centrifuged at 300 x g for 5 min. Cells were resuspended in glutaMAX DMEM/F-12 (Thermo-Fisher, Cat#10565018) supplemented with 20 ng/ml EGF (Thermo-Fisher, Cat#PHG0311L), 20 ng/ml FGF (Thermo-Fisher, Cat#PGH0367), B-27 (Thermo-Fisher, #17504044) and pen/strep (Thermo-Fisher, Cat#15140148) and kept in a 5% CO₂ incubator at 37°C.

METHOD DETAILS

Sample collection for scRNA-seq

Pregnant Brn3a-tLacZ^{+/-} or C57BL/6j mice were sacrificed by cervical dislocation. Embryos were harvested at embryonic (E) day E11, E12, E13, E15, and E18. The uterine horn was placed in ice-cold 1x PBS. Embryonic brains were dissected in ice-cold medium A (1x HBSS – Ca²⁺ - Mg²⁺ (Thermo-Fisher, Cat#24020117)) supplemented with 1.6 mM HEPES (Thermo-Fisher, Cat#15630106) and 0.63% glucose. Tissue was dissociated using the MACS Neural Dissociation kit (Miltenyi Biotec, Cat#130-093-231) for 30 min according to the manufacturer's protocol. Postnatal (P4 - P7) and adult Brn3a-tLacZ^{+/-} mice were cervically dislocated (only adult: P245-P259) and decapitated. The heads were stored on ice. The brain was removed from the skull and placed in ice-cold HABG medium (Hibernate A - Ca²⁺ - Mg²⁺ (BrainBits, Cat#HACAMG500)) supplemented with 1x B27 (Thermo-Fisher, Cat#17504044) and 0.5 mM L-glutamine (Thermo-Fisher, Cat#25030081)). The Hb was dissected out, manually cut into smaller pieces and placed in 2 ml HABG medium in a 15 ml Falcon tube on ice. Samples were spun down at 200 x g for 1 min. Supernatant was taken and 0.5 ml warm papain diluted in HABG was added. Samples were broken by pipetting 5 times with a P1000 pipet. Samples were incubated horizontally on a shaker for 15 min at 160 RPM at 37°C. 35 μl DNase mix 1 (1 volume enzyme X (DNase) (Neural Dissociation kit (T) (Miltenyi Biotec, Cat#130-093-231)) and 2 volumes buffer Y (Neural Dissociation kit (T) (Miltenyi Biotec, Cat#130-093-231)) was added to the sample. Next, the sample was incubated horizontally on a shaker for 15 min at 160 RPM at 37°C. 135 µl warm DNase mix 2 (30 µl DNase mix 1, 100 µl FBS, 2 µl BSA 1 mg/ml (Jackson, Cat#001-00-162) and 1 µl 0.5 M EDTA) was added to each sample. For complete dissociation, samples were triturated 2-3 times with 3 different glass pipettes decreasing in size. Samples were passed through a 100 μm filter and 5 ml aCSF (92 mM NaCl, 2.5 mM KCl, 1.2 mM NaH₂PO₄, 30 mM NaHCO3, 20 mM HEPES (Thermo-Fisher, Cat#15630106)), 25 mM glucose, 3 mM sodium ascorbate, 2 mM thiourea and 3 mM sodium pyruvate) supplemented with 10% FBS, and 5 mM EDTA was added. Samples were centrifuged at 100 g for 10 min. Supernatant was removed, cells were taken up in 100 µl staining medium (4% FBS and 10 mM HEPES (Thermo-Fisher, Cat#15630106), in 1x PBS), filtered through a 70 µm FACS filter, and placed on ice.

FluorLacZ staining was performed using the FluoReporter lacZ Flow Cytometry Kit (Thermo Fisher Scientific, F1930) according to the manufacturer's protocol. Cell in staining medium and the FDG (fluorescein di- β -D-galactopyranoside) working solution (2 nM FDG in MQ) were pre-warmed for 10 min at 37°C. 100 µl of FDG working solution was added to the cells and rapidly mixed. Cells were placed in a water bath at 37°C for exactly 60 s. To stop the FDG loading 1.8 ml ice-cold staining medium containing 1.5 µM propidium iodide was added to the cells using ice-cold pipettes and cells were stored on ice.

Cells were sorted into 384-wells plates in a BD FACS ARIAII FACS machine with a 100 µm nozzle. Cell were sorted based on DAPInegativity (E18 WT experiment) and DAPI negativity followed by LacZ positivity (*Brn3a-tLacZ* developmental experiment) (Figures S2D and S2E). FACS gating was determined using two negative controls: (1) cortical cells from *Brn3a-tLacZ* mice loaded with FDG, and (2) habenula cells from *Brn3a-tLacZ* mice not loaded with FDG (Figure S2C). Each replicate consists of two 384-well plates each containing an individual sample (Table S1). After sorting, cells were spun down at 1000 rpm for 1 min, placed on dry ice and stored at -80°C.

scRNA-seq sequencing

Cells were sequenced by Single Cell Discoveries (Utrecht, The Netherlands) according to the SORT-seq method based on CEL-Seq2 (Muraro et al., 2016). In short, cells were lysed and the poly-T stretch of the barcoded primers hybridized to the poly-A tail of the mRNA molecules. The DNA-RNA hybrids were converted into cDNA containing the mRNA and primer sequences. All cDNA was pooled and amplified by *in vitro* transcription. Finally, PCR selected for molecules that contain both Illumina adaptors necessary for sequencing. Subsequently, bead cleanup and quality control were performed with the Agilent High Sensitive DNA Kit (5067-4626) and libraries were sequenced by the Utrecht Sequencing Facility (USEQ, Life Sciences Faculty, Utrecht University). Paired end sequencing was performed by Nexseq500.

3DISCO

3DISCO was performed as described previously (Adolfs et al., 2021; Belle et al., 2014). In short, E10, E11, E12 and E13 whole mouse embryos, and E15 and E18 mouse brains were placed in 4% PFA, pH 7.4, for 24 h at 4°C. Samples were washed in 1x PBS for \geq 1.5 h on a shaker at room temperature (RT). Subsequently, they were dehydrated using a series of 50%, 80% and 100% MeOH (Merck Millipore, Cat#1060092500), each step was performed for \geq 1.5 h on a shaker at RT. Samples were bleached overnight at 4°C without shaking in 3% hydrogen peroxide (Merck Millipore, Cat#1072091000) in MeOH. The following day samples were hydrated using a series of 100%, 100%, 80%, 50% MeOH and 1x PBS for \geq 1 h on a shaker at RT. The bleached samples were placed in a 15 ml Falcon tube containing blocking buffer (PBSGT) consisting of 1x PBS with 0.2% gelatin, 0.5% Triton-X-100 (Sigma, Cat#x100-500ml), and 0.01% thimerosal (Gerbu Cat#1031/USP35) at RT on a horizontal shaker at 70 RPM. Incubation time varied



from 3 h (E10-E12) to 24 h (E13-E18). Samples were incubated at 37°C on a horizontal shaker (70 RPM) in a 2 ml Eppendorf tube containing 2 ml PBSGT, 0.1% saponin (Sigma, Cat#S7900), and primary antibody. Antibodies used were goat anti-Neuropilin-2 (NRP2) (R&D Systems, Cat#AF567, 1:1000), rabbit anti-Tyrosine Hydroxylase (EMD Millipore, Cat#TH152, 1:300), goat anti-ROBO1 (R&D Systems, Cat#AF1749, 1:300), and goat anti-ROBO3 (R&D Systems, Cat#AF3076, 1:300). Incubation times varied from 3 days (E10-E13) to 7 days (E15-E18). Samples were washed 6 times for 1 h at RT in a 15ml Falcon tube containing PBSGT on a rotator at 14 RPM. Next, samples were incubated at 37°C for 48 h on a horizontal shaker (70 RPM) in a 2 ml Eppendorf tube containing 2 ml PBSGT, 0.1% saponin (Sigma, Cat#S7900) and secondary antibody protected from the light from this point onwards. Antibodies used were Alexa Fluor 647 donkey anti-goat (Abcam, Cat#ab150135, 1:500) and Alexa Fluor 750 donkey anti-rabbit (Abcam, Cat#ab175731, 1:500). Next, samples were washed 6 times for 1 h at RT in a 15 ml Falcon tube containing PBSGT on a rotator at 14 RPM. After staining, E10 and E11 embryos were embedded in 1% agarose dissolved in 1x PBS. For clearing, tissue samples were placed in a series of 50% (overnight), 80% (1 h), 100% (1 h) and 100% (1 h) tetrahydroflurane with BHT as inhibitor (Sigma, Cat#186562-1L). All steps were performed in a 15 ml Falcon at RT on a horizontal shaker at 70 RPM. To dissolve lipids, samples were placed in 100% dichloromethane (DCM) (Sigma, Cat#270997-1L) for 20 min or until they sank. Finally, samples were incubated overnight in 100% dibenzyl ether (DBE) (Sigma, Cat#108014-1kg) in a 15 ml Falcon tube at RT on a rotator at 14 RPM. Samples were stored in 100% DBE at RT protected from light. Samples were imaged using an Ultramicroscope II (LaVision BioTec) light sheet microscope equipped with an MVX-10 Zoom Body (Olympus), MVPLAPO 2x Objective lens (Olympus), Neo sCMOS camera (Andor) (2560 x 2160 pixels. Pixel size: 6.5 x 6.5 µm2) and Imspector software (version 5.0285.0, LaVision BioTec). Samples were scanned with single-sided illumination, using the horizontal focusing light sheet scanning method and the blend algorithm. The Object lens included a dipping cap correction lens (LV OM DCC20) with a working distance of 5.7 mm. Imaris (version 8.4-9.4, Bitplane) software was used for image processing and analysis.

iDISCO

Cart-IRES2-cre-D+':Ai14+'- mice (E15, P4, P7 and adult) were stained using the iDISCO method described previously (Renier et al., 2016). In brief, samples were dehydrated in a 15 ml Falcon tube containing a series of 20% MeOH, 40% MeOH, 60% MeOH, 80% MeOH, or 100% MeOH, each for 1 h at RT followed by overnight incubation at RT in 66% DCM and 33% MeOH. The following day samples were bleached by two times incubation at RT in 100% MeOH, followed by incubation of the samples at 4°C for 3 h. This was followed by incubation in 2% hydrogen peroxide in MeOH overnight at 4°C. The following day samples were rehydrated using a series of 80% MeOH, 60% MeOH, 40% MeOH, 20% MeOH, each for 1 h at RT. This was followed by washes in 1x PBS, 1 h at 4°C. Next, samples were incubated twice in Ptx.2 (1x PBS and 0.2% Triton-X100) for 1 h at RT and placed in permeabilization solution for 24 h at 37°C. Then, samples were incubated in blocking solution for 4 days at 37°C on a horizontal shaker (70 RPM). Subsequently, samples were incubated in primary antibody solution (PwtH (1 x PBS, 0.2% Triton-X100, 0.1% Heparin (Sigma, Cat#H3393, 10 mg/ml) stock solution, 5% DMSO and 3% normal donkey serum (Jackson Immunoresearch, Cat#017-000-121)) for 7 days at 37°C on a horizontal shaker (70 RPM). Then, samples were washed 5 times for 1 h at RT in a 15 ml Falcon tube containing PtwH. Samples were incubated in secondary antibody solution (PwtH and 3% normal donkey serum (Jackson Immunoresearch, Cat#017-000-121)) for 7 days at 37°C on a horizontal shaker (70 RPM) in a 5 ml Eppendorf. Then, samples were dehydrated using a series of 20% MeOH, 40% MeOH, 60% MeOH, 80% MeOH and 2 times 100% MeOH, each step for 1 h at RT. Then, samples were placed in 66% DCM and 33% MeOH for 3 h at RT, followed by 2 washes in 100% DCM for 15 min at RT each. For clearing, tissue samples were placed in 100% DBE overnight at RT. Samples were stored in 100% DBE at RT until imaging. All washing, dehydration and clearing steps were performed in dark Falcon tubes to protect against light and on a rotator (14 RPM). Primary antibodies used were rabbit anti-RFP (Rockland, Cat#600-401-379, 1:1000) and goat anti-ROBO3 (R&D, AF3076, 1:300). Secondary antibodies used are Alexa Fluor 647 donkey anti-rabbit (Abcam, Cat#ab150135, 1:1000) and Alexa Fluor 568 donkey anti-goat (Abcam, Cat#ab175704, 1:1000). Samples were imaged using an Ultramicroscope II (LaVision BioTec) light sheet microscope equipped with an MVX-10 Zoom Body (Olympus), MVPLAPO 2x Objective lens (Olympus), Neo sCMOS camera (Andor) (2560 x 2160 pixels. Pixel size: 6.5 x 6.5 μm2) and Imspector software (version 5.0285.0, LaVision BioTec). Samples were scanned using single-sided illumination, a sheet NA of 0.148 and a step-size of 2.5 µm using the horizontal focusing light sheet scanning method with 200 steps and using the blend algorithm. The Object lens included a dipping cap correction lens (LV OM DCC20) with a working distance of 5.7 mm. The effective magnification was 8.4208 x (zoom body*objective + dipping lens = 4*2.1052). Imaris (version 8.4-9.4, Bitplane) software was used for image processing and analysis.

Whole-mount X-gal staining

The X-gal staining protocol was based on (Wythe et al., 2013). Embryos were fixed for 24 h in fixative solution containing 2% formaldehyde (Riedel-de Hain, Cat#33200), 0.2% glutaraldehyde (Acros Organics, Cat#119989925), 0.02% sodium deoxycholate (Sigma, Cat#D6750-100G) and 0.01% NP-40 in 1 xPBS. Samples were rinsed 2 times with ice-cold 1x PBS to remove fixative. Samples were permeabilized overnight at 4°C while shaking gently in permeabilization solution containing 0.02% sodium deoxycholate and 0.01% NP-40 in 1x PBS. After permeabilization samples were incubated in staining solution containing 1 mg/ml X-gal, 5 mM K-ferricyanide (Merck, Cat#231-847-6), 5 mM K-ferrocyanide (Sigma, Cat#p8131-100G), 2 mM MgCL₂ (Merck, Cat#A034473339), 0.02% sodium deoxycholate and 0.01% NP-40 in 1x PBS. Incubation was performed at 37°C and in the dark for 3 to 24 h while shaking. Samples were washed 2 times for 15 min with permeabilization solution to remove staining solution. Embryos were





dehydrated in a series of 30%, 50%, 80%, 100% and 100% EtOH. For clearing, embryos were incubated overnight at RT in 100% dibenzyl ether (DBE) (Sigma, Cat#108014-1kg) in a 15 ml Falcon tube on a horizontal shaker at 70 RPM. Embryos were imaged using a Zeiss Primovert inverted microscope using Zen 2 blue edition software (version 2.0.0.0).

In situ hybridization

Probe generation

Primers (Table S5) were designed to amplify the desired probe sequence from total mouse cDNA. PCR products were run in a 1.5% agarose gel and the correct amplicon was purified using the PureLink Quick Gel Extraction kit (Thermo-Fisher Cat#K210025). The purified DNA fragment was ligated into the pGEM-T-easy vector (Promega, Cat#A1360) according to the manufacturer's protocol. Ligation products were transformed into DH5- α cells (Invitrogen, Cat#: 18263012). Upon confirmation by restriction digestion, inserts were sequenced using T7 and SP6 primers. The PCR product was used for SP6/T7 reverse transcription to generate DIG-labeled sense and anti-sense probes as described previously (Brignani et al., 2020). Probes were treated with 2 units of DNase (Roche, Cat#776785) for 15 min at 37°C. Then probes were centrifuged briefly and a solution containing 2 µl of 2.0 M EDTA, pH 8.0 (Sigma, Cat#E5134-500G), 2.5 µl of 4 M LiCl (Merck, Cat#7447-41-8), and 75 µl pre-chilled 100% EtOH was added. Samples were placed for 30 min at -80°C followed by centrifugation at 15,000 x g at 4°C for 15 min. The pellet was washed with 50 µl pre-chilled 70% EtOH and centrifuged at 14,000 RPM at 4°C for 5 min. The pellet was air-dried for 10 min at RT, resuspended in 100 µl Milli-Q water, and stored at -80°C.

In situ hybridization

In situ hybridization was performed as described previously (Brignani et al., 2020). In brief, sections were defrosted and dried for 1.5 h at RT. Subsequently, sections were fixed for 10 min in 4% PFA (Sigma #P-6148) in 1x PBS, pH 7.4. After fixation, sections were washed 3 times for 5 min in 1x PBS. Sections were placed in acetylation solution (1.32% triethanolamine (Fluka Cat#90279), 0.18% HCl and 0.25% acetic anhydride (Sigma Cat#A6404)) for 10 min. After acetylation, sections were washed 3 times for 5 min in 1x PBS. Sections were pre-hybridized for 2 h at RT in hybridization solution (50% deionized formamide (ICN, Cat#800686), 5x saline-sodium citrate (SSC; 0.75 M NaCl, 75 mM NaCitrate, pH 7.0), 5x Denhardts (1 mg/ml Ficoll-400 (Sigma, Cat#F4375)), 1 mg/ml polyvinylpyrrolidone (Merck, Cat#7443), 1 mg/ml BSA-fraction V (ICN Cat#103703), 250 ug/ml tRNA baker's yeast (Sigma, Cat#R-6750), and 50 μg/ml sonificated Salmon Sperm DNA (Sigma, Cat#D-91565) in MQ). DIG-RNA probe was diluted in 150 μl hybridization solution and added to the sections. Sections were covered with NESCO-film and hybridized in a humidified chamber overnight at 68°C. To remove non-bound probe, slides were dipped in 2x SCC (0.3 M NaCl, 30 mM NaCitrate, pH 7.0) at 68°C and then washed in 0.2x SCC (30 mM NaCl, 3 mM NaCitrate, pH 7.0) for 2 h at 68°C. Slides were transferred to RT 0.2x SSC and washed for 5 min. Slides were then washed for 5 min in buffer 1 (100 mM TrisHCl, pH 7.4, 150 mM NaCl) at RT. Sections were incubated for 1 h at RT in buffer 1 supplemented with 10% FCS followed by incubation with anti-DIG-AP primary antibody (Sigma, Cat#11207733910, 1:5000) in buffer 1 containing 1% FCS overnight at 4°C. The following day, sections were washed 3 times for 5 min in buffer 1 and once for 5 min in buffer 2 (100 mM TrisHCI, pH 9.5, 50 mM MgCl2, 100 mM NaCl). Slides were developed in filtered developing solution (NBT/BCIP 1:20 (Boehringer Mannheim, Cat#1681451) and 0.24 mg/ml Levamisol (Sigma, Cat#L9756) in buffer 2). Slides were developed for 2 to 24 h. Sections were mounted with glycerol. Images were obtained using an Axio Imager M2 microscope (Zeiss) and processed in Fiji (Schindelin et al., 2012) (version 2.0.0-rc-69/1.52o).

Stereotactic injections

For retrograde anatomical studies, male and female *Cart-IRES2-cre-D^{+/-}:Ai14^{+/-}* mice (12–14 weeks old) were injected with 0.2–0.5 μ l Alexa Fluor 488-conjugated Dextran (Invitrogen, diluted 1:3 in sterile PBS). Prior to surgery, mice were anaesthetized by intraperitoneal (i.p.) injection of ketamine (Narketan 10, Vétoquinol, 75 mg/kg) and medetomidine (Sedastart, Ast Farma B.V., 1 mg/kg). Lidocaine (B. Braun, 0.1 ml, 10% in saline) was administered subcutaneously under the scalp to provide local anaesthesia. Eye ointment (CAF, Ceva Sante Animale B.V.) was applied before surgery. Throughout surgery, mice were placed under a heating pad to avoid hypothermia. Mice were placed in a Kopf stereotaxic frame apparatus with head position to obtain a flat skull between bregma and lambda (<0.03 mm). A craniotomy was performed using a micro-drill to expose brain tissue. Dextran was delivered intra-IPN (from bregma: AP -3.45 mm; ML 0 mm; DV -5.0 mm, at a 0° angle) via a 31 G stainless steel hypodermic tube (Coopers Needleworks Ltd) connected to a Hamilton syringe controlled by an automated pump (model 220, Uno B.V, Zevenaar, NL) at an injection rate of 0.1 μ /min. Dextran delivery was followed by a 10 min diffusion period. Mice emerged from anesthesia after i.p. injection of atipamezol (Alzane, Syva, 50 mg/kg). All mice received peri- and post-operative care consisting of i.p. injection of carpofen (Carpofelican, Dechra, 5 mg/kg) on the day of surgery, followed by a maximum of 4 days of carpofen (25 mg/l) in the drinking water.

Immunohistochemistry

Embryos were harvested and their brains placed in ice-cold 1x PBS. Pups were decapitated and their brains placed in ice-cold 1x PBS. *Brn3a-mTurquoiseHIS-FlpO*^{ERT2};*Cart-IRES2-cre;Ai65* pups were injected subcutaneously with 100 µL tamoxifen (final dose: 0.05 mg/g; Sigma) dissolved in corn oil (Sigma) at P7 and P8 and sacrificed 48 h post-injection. Pups (P10) and adults were anaesthetized by i.p. injection of sodium pentobarbital (Euthanimal, 0.05 mL (pups) or 0.1 mL (adults)), followed by transcardial perfusion with 4% PFA in 1x PBS, pH 7.4. Brains were fixed in 4% PFA overnight at 4°C. Subsequently, brains were placed for 48 h in 30% sucrose for cryoprotection, frozen in -20°C isopentane and stored at -80°C. Immunohistochemistry was performed as described



previously (Brignani et al., 2020). In brief, 25 μm or 20 μm (*Brn3a-mTurquoiseHIS-FIpO*^{ER72}; *Cart-IRES2-cre;Ai65*) thick cryosections were generated and incubated in blocking solution (4% bovine serum albumin (Sigma Aldrich #05470) and 0.1% Triton x-100 (Merck Millipore #108643)) for 1 h at RT. Sections were incubated overnight at 4°C with primary antibody diluted in blocking solution. Following washes in PBS, sections were incubated with secondary antibody in blocking solution for 1.5 h at RT. Primary antibodies used were rat anti-RFP (1:500, ChromoTek, Cat#5f8-100), chicken anti-GFP (1:1000, Aves, Cat#GFP-1020), goat anti-CHAT (1:100, EMD Millipore, Cat#144P) and goat anti-NRP2 (1:500, R&D Systems, Cat#AF567). Secondary antibodies used were donkey anti-goat Alexa Fluor 488 (1:750, Invitrogen, Cat#A11055), donkey anti-chicken Alexa Fluor 488 (1:500, Jackson ImmunoResearch, Cat#703-545-155) and donkey anti-rat Alexa Fluor 568 (1:500, Abcam, Cat#ab175475). Sections were counterstained with DAPI (Sigma, Cat#D9564) and mounted with Fluorsave (VWR international, Cat# 345789-20). Images were taken using an Axioscope A1 microscope (Zeiss) or AxioImager M2 (Zeiss) and processed in Fiji (Schindelin et al., 2012) (version 2.0.0-rc-69/1.520).

Experimental animals used for anatomical studies were sacrificed 48 h to 4 days post-injection to allow for sufficient retrograde labeling of distant regions. Mice were anaesthetized by i.p. injection of sodium pentobarbital (Euthanimal, 0.1 ml), followed by transcardial perfusion with PBS and 4% PFA in PBS. Brains were dissected and post-fixed in 4% PFA overnight at 4°C. Subsequently, 30 µm sections were generated using a vibratome. Sections were collected in 1 x PBS and stained with DAPI. Sections were mounted in Fluorsave (VWR international, Cat# 345789-20) and Z-stack images were taken using a Zeiss LSM 880 laser-scanning confocal microscope.

BrdU labeling

Bromodeoxyuridine (BrdU) (Sigma, Cat#B5002) was diluted to 15 mg/ml in saline at 37°C for 2 h. Pregnant C57BL/6j mice were injected i.p. with 50 mg BrdU solution per kilogram bodyweight. 8 h to 5 days post-injection embryos were harvested and placed in icecold PBS. Antigen retrieval was performed by incubating the sections in 1 M HCl for 1 h at 37°C followed by incubation in 0.1 M boric acid, pH 8.5, for 10 min at RT. ISH and IHC was performed as described above. Rat anti-BrdU (Abcam, Cat#AB6326, 1:500) and Alexa Fluor 568 donkey anti-rat (Abcam, Cat#ab175475, 1:750) antibodies were used. Sections were obtained using an Axioscope A1 microscope (Zeiss) and images were processed in Fiji (version 2.0.0-rc-69/1.520).

Brain slice preparation for electrophysiological recordings

Male and female *Cart-IRES2-cre-D^{+/-}:Ai14^{+/-}* mice (8-12 weeks old) were anaesthetized by i.p. injection of sodium pentobarbital (Euthanimal, 0.1 ml), followed by transcardial perfusion with ice-cold, carbogen-saturated (95% O₂ and 5% CO₂) slicing solution containing (in mM): 92 choline chloride, 2.5 KCl, 1.2 NaH₂PO₄, 25 NaHCO₃, 20 HEPES, 25 glucose, 20 N-methyl-D-glucamine (NMDG), 10 Na-ascorbate, 2 thiourea, 3 Na-pyruvate, 3.1 N-Acetyl-L-cysteine, 7 MgCl₂.6H₂O and 0.5 CaCl₂.2H₂O (mOsm 300-310; pH 7.3–7.4). After perfusion, mice were decapitated and brains removed rapidly and sectioned in ice-cold carbogen-saturated slicing solution using a vibratome (Leica VT1200S, Leica Microsystems). Coronal slices (thickness 250 µm) containing the Hb were initially recovered in carbogenated slicing solution for 5 min at 37°C and then transferred into a holding chamber with carbogen-saturated artificial cerebrospinal fluid (ACSF), containing (in mM): 92 NaCl, 2.5 KCl, 1.2 NaH₂PO₄, 30 NaHCO₃, 20 HEPES, 25 glucose, 3 Na-ascorbate, 2 thiourea, 3 Na-pyruvate, 2 MgCl₂.6H₂O and 2 CaCl₂.2H₂O (mOsm 300-310; pH 7.3–7.4). Slices were kept at RT for at least 1 h before electrophysiological recordings. Subsequently, slices were transferred to a recording chamber, mounted on a fixed-stage, upright microscope (Olympus BX61W1), mechanically stabilized and continuously superfused (2 ml / min) with carbogenated recording ACSF containing (in mM): 124 NaCl, 2.5 KCl, 1 NaH₂PO₄, 26 NaHCO₃, 5 HEPES, 11 glucose, 1.3 MgCl₂.6H₂O and 2.5 CaCl₂.2H₂O (mOsm 300-310, pH 7.3–7.4). Solution was run through an in-line heater (model SH-27B, connected to a TC-324B controller, Warner Instruments, Hamden, Connecticut, United States)) to perform electrophysiological recordings at 28–30°C.

Electrophysiological recordings

The Hb region was identified in brain slices under a 4x objective. Medial Hb *Cartpt*⁺ cells were identified by tdTomato fluorescence and cells were visualized using infrared video microscopy and differential interference contrast (DIC) optics, under a 40x, 0.8 W water-immersion objective. Patch electrodes were pulled from borosilicate glass capillaries (Warner Instruments, Cat#300053), and had a resistance of 3–5 MΩ when back-filled with intracellular solution. A K-based internal solution was used containing (in mM): 139 K-gluconate, 5 KCl, 2 MgCl₂. 0.2 EGTA, 10 HEPES, 10 Na creatine phosphate, 4 Mg ATP and 0.3 Na GTP (mOsm 300-310, pH 7.3–7.4, liquid junction potential of \sim -12 mV). Signals were low pass filtered with a 2.9 kHz 4-pole Bessel filter and digitized at 20 kHz using an EPC-10 patch-clamp amplifier (HEKA Elektronik GmbH, Lambrecht, Rhineland-Palatinate, Germany) and PatchMaster v2.73 software.

Conventional whole-cell patch-clamp recordings were made on $Cartpt^+$ and $Cartpt^-$ neurons of the MHb ad LHb. After tight-seal (>1G Ω) cell-attached configuration, negative pressure pulses were applied to break into the cellular membrane and go to whole-cell mode. Neurons were first held in voltage clamp at -60 mV for 5 min before the start of experiments. Series resistance was not compensated and was calculated from the capacitive current amplitudes evoked by a 4 mV hyperpolarizing step and neurons that demonstrated stable holding current and series resistance were selected for further analysis. Subsequently, recordings were switched to current clamp mode. To assess membrane properties and excitability of the different neuronal populations, an IV relation was determined for all cells. Neurons were made to step from a baseline of 0 pA current injection of 400 ms to an 800 ms current injection step of -100 pA, which was followed by a return to 0 pA for 1000 ms. Every subsequent sweep had a current injection





step that was 25 pA more depolarized than the previous one. Sweeps were delivered with a 10 s inter-sweep interval. Using data obtained with the IV relation, parameters of interest were calculated. Membrane resistance was calculated according to Ohm's law, using the measured delta of steady-state voltage deflections caused by currents that did not induce active membrane conductance. Voltage sag was calculated as the delta between the minimal membrane potential and the subsequently obtained steady-state membrane potential during the 800 ms -100 pA current step. To determine action potential threshold the first IV sweep in which an action potential was triggered during the current injection step phase of 800 ms was identified and differentiated. In the differentiated waveform, the first +20 mV/ms point was identified and this timepoint was related back to the undifferentiated waveform to obtain the associated membrane potential, as the action potential threshold for the cell. Cellular capacitance was obtained by, in voltage clamp, dividing the membrane time constant tau (obtained by exponential curve fitting of the current recovery after the hyperpolarizing 4 mV test pulse towards steady-state levels) by the series resistance (obtained by dividing hyperpolarizing test pulse voltage by the amplitude of the elicited capacitive current).

QUANTIFICATION AND STATISTICAL ANALYSIS

General statistical analysis

All statistical details of tests can be found in the text or figure legends. When applicable, nonparametric Kruskal Wallis was used for data that were not normally distributed, i.e., to correct for when homogeneity of variance assumption was violated. A test was considered significant when $p \le 0.05$ unless specified otherwise.

External datasets

For comparison we used multiple previously published datasets. Throughout all analyses we used clusters as defined by the original papers. Details are listed below.

Adult mouse Hb; Hashikawa et al.

For the Hashikawa *et al* dataset (Hashikawa *et al.*, 2020) (GEO: GSE137478), a normalized and log-transformed Seurat object that includes the metadata on the cell clustering of the Hb cells, as described in the publication, was imported. The dataset was sequenced on the 10xGenomics platform and consisted of 5,558 cells and 19,726 genes, with a median read depth of 3,666.5 reads/cell.

Adult mouse Hb; Wallace et al.

A Seurat object was created with the Wallace *et al* dataset (Wallace *et al.*, 2020) using the published normalized and log-transformed data matrix (https://doi.org/10.7910/DVN/2VFWF6) and the supplied metadata on the medial and lateral Hb cells. The cells were sequenced on the 10xGenomics platform. The dataset contained 3,379 cells and 25,289 genes, with a median read depth of 4,622.0 reads/cell.

Larval and adult zebrafish Hb; Pandey et al.

Zebrafish Hb datasets were used for inter-species analysis, a larval and adult zebrafish dataset from Pandey *et al* dataset (Pandey et al., 2018). Both datasets were imported by creating a Seurat object with the published normalized and log-transformed data matrix and the supplied metadata of each of the cells. The larval dataset (GEO: GSM2818523) was sequenced on the 10xGenomics platform and SmartSeq2 platform and consisted of 4,233 cells (from 10xGenomics) and 13,160 genes, with a median read depth of 2,565.94 reads/cell. The adult dataset (GEO: GSM2818522) was sequenced on the 10xGenomics platform and consisted of 7,711 cells and 9,582 genes, with a median read depth of 1,596 reads/cell.

Adult mouse brain tissue; Tabula Muris

As an extra control, a scRNA-seq dataset from the Tabula Muris Consortium was tested (Schaum et al., 2018). The consortium performed single cell transcriptomics on cells from twenty organs and tissues, creating a compendium. For this study, FACS was used to collect cells from brain tissue. The dataset was sequenced with NovaSeq and consisted of 7,189 cells and 18,454 genes, with a median read depth of 626,662.0 reads/cell.

Developmental mouse hypothalamus; Kim et al.

In the original publication (GEO: GSE132355), 129,151 developmental hypothalamus cells were sequenced at eight embryonic and four postnatal timepoints using 10xGenomics (Kim et al., 2020). This dataset was sampled to resemble the clustering of the developmental Hb dataset: fourteen clusters with a similar timepoint consistency. Additionally, 3.5% of the cells in each of those clusters were randomly sampled for the MAGMA analysis. The final object consisted of 3,385 cells and 22,187 genes, with a median read depth of 1,454.1 reads/cell.

Developing and adult mouse midbrain

This dataset (GEO: GSE76381) was consisted of six embryonic and one adult timepoint and used Illumina HiSeq (La Manno et al., 2016). It included in total 31 neuronal and non-neuronal cell types. The dataset consisted of 1,761 cells and 24,378 genes, with a median read depth of 8,846.0 reads/cell.

Mapping of scRNA-seq data

Mapping was performed with the MapAndGo2 STARmap pipeline (Alemany, 2019). In short, the lanes of the fastq files were merged, then reads were demultiplexed removing all reads without a CEL-Seq2 barcode. Illumina adaptor sequences and bad quality base



calls from the 3'-end were trimmed off the reads. Then reads were aligned to the reference mouse genome (Ensemble GRC38 release 93) using STAR v-2.5.3a (Dobin et al., 2013; Yates et al., 2020). Then 3 count files were produced: the unspliced file contained reads mapped to intronic regions, the spliced file contained reads mapped to exons and the total file contained both. These three files were the input for the SCANPY pipeline.

scRNA-seq data filtering and normalization

Data were analyzed in Python (version 3.7.6) using the SCANPY package (version 1.4.6.) (Wolf et al., 2018). Reads were UMI-corrected. The Brn3a-tLacZ developmental dataset included 5,756 cells with a total expression of 38,068 genes, from which 85% of the reads were mapped to exonic- and 15% to intronic regions. The cells were sequenced to a median read depth of 8,482.80 reads/cell and a median of 3,156.0 genes/cell. To determine sequencing guality, External RNA Control Consortium (ERCC) spikeins were added to each well (cell) (Jiang et al., 2011). Spike-ins are synthetic RNA with a poly-A tail, which allows their sequencing. By assessing the number of ERCC reads per cell, the sequencing quality of a cell can be determined. The dataset had a median sequencing quality of 14.10% ERCC spike-ins/cell. For further processing all the ERCC genes, together with mitochondrial genes, were removed. We observed two populations of cells, i.e. expressing more or fewer than 1500 genes (Figure S3A). We performed a stringent selection and included cells that expressed >2000 genes. We confirmed that this stringent selection did not cause exclusion of populations that were observed when selecting for cells expressing >1500 genes. Cells that had < 50,000 reads (of which less than: 15% were mitochondrial genes, 20% were ribosomal genes and 30% were ERCC spike-ins, and where at least 85% of genes were mapped to exonic regions) were kept. Besides cell filtering, we also filtered for genes that yielded the highest fraction of counts in each single cell, and genes had to be expressed in more than three cells. After filtering, the dataset contained 3304 cells with a total expression of 30,452 genes with a median read depth of 14,396.69 reads/cell, a median of 4,955.0 genes/cell and a median of 8,45% ERCC spike-ins/cell. The raw matrix with the gene expression was normalized on the total number of transcripts detected multiplied by a scaling factor of 10,000. The data were then log transformed (In) after the addition of a pseudo-count of 1. Additionally, cells that did not express Pou4f1/Brn3a (0.0 > Pou4f1) were removed. These uncorrected data were stored as a count matrix (named uncorrected count matrix) which was used to calculate DEGs, plot gene expression levels and perform downstream analysis.

The E18 WT dataset contained 1920 cells and 36,196 genes. 85% of the reads were spliced and 15% of the reads were unspliced. The median number of genes per cell was 2857. We filtered out ERCC genes, mitochondrial genes, genes that were detected in less than 3 cells and cells that expressed less than 2000 genes. After filtering, the dataset contained 1002 cells and 28237 genes. Next, we removed cells that did not meet the following criteria: number of counts < 50000, mitochondrial reads < 15%, ribosomal reads < 20%, ERCC genes < 30% and coding reads > 85%. After filtering the dataset contained 948 cells and 28,237 genes. The data were normalized to 10,000 reads per cell and log-transformed. Genes with an expression between 0.0125 and 3 and a minimum dispersion of 0.5 were considered highly variable and were kept in the dataset for clustering. Then the effects of total counts per cell and percentage mitochondrial reads were regressed out and data were scaled to a max value of 10.

Batch effect correction for the DevHb dataset using Seurat v3

Seurat v3 integration software (Stuart et al., 2019) requires a Seurat object that is normalized and log-transformed (raw matrix) as input. The CCA, L2-normalization, finding the MNNs and scoring them was executed using the FindTransferAnchors() function, where thirty, fifteen and four MNNs were used for filtering-, scoring- and finding anchors, respectively. The IntegrateData() function merged the data and corrected the data matrix, which was then stored in the integrated assay in the Seurat object. The batch-corrected count matrix was then imported into SCANPY where it replaced the uncorrected count matrix. This matrix was used for downstream clustering.

Principal Component Analysis and clustering

A Principal Component Analysis (PCA) was computed with fifty components. Then the nearest neighbors were calculated using the first forty principal components. Subsequently, t-SNE coordinates were computed using the first forty principal components. Clustering was performed with the Louvain algorithm (Traag, 2017).

Differential gene expression

Marker genes per Louvain cluster were calculated with the Wilcoxon-rank-sum test using the sc.tl.rank_genes_groups() function in SCANPY. This function calculates the DEGs per cluster using the raw and normalized count matrix. In order to find more specific marker genes, as depicted in Figures 1F and 5A, markers genes were filtered using the following parameters: min_in_group_fraction = 0.5, max_out_group_fraction = 0.3, min_fold_change = 1.5. Adjusted p-values of DEGs between the left and right Hb were derived using the sc.get.rank_genes_groups_df() function. All plots show upregulated DEGs ordered by highest z-score.

CellRank trajectory inference

Cellrank trajectory inference was performed on the developmental dataset (Lange et al., 2020). First, it computed terminal states, the number of terminal states was defined by the user (n = 7) using the cr.tl.terminal_states() function. Then, it defined initial states with the cr.tl.initial_states() function. Next, fate maps were computed to define the likelihood that an individual cell reaches a certain fate. Subsequently, the individual fate maps were aggregated to cluster-level fate maps by calculating latent time and running PAGA (Wolf



et al., 2019). Gene expression trends along pseudo- or latent time were calculated based on a general additive model (GAM) with the cr.pl.gene_trends() function. Heatmaps of top 50 genes whose expression levels correlated best with the probability of reaching the terminal stage were plotted using the cr.pl.heatmap() function.

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Correlation analysis

The gene expression patterns of clusters from two datasets (referred to as dataset 1 and 2) were correlated in a correlation analysis, which was based on previously described methods (Tosches et al., 2018). DEG were used for the calculation of pairwise cluster correlations because those suffice as cell type markers, possibly even cross-species. To allow others to replicate our analysis in the most straightforward manner, the complete pipeline described in the original paper was used. As a result, DEG computed with the MAST package using the zlm() function (Finak et al., 2015) were used instead of the DEG calculated in the SCANPY pipeline (Wilcoxon-rank-sum test as described above). The zlm() function uses count data to model the expression based on cluster membership and number of detected genes per cell. First, DEG were computed within each dataset, by calculating among all possible pairs of clusters. Genes were only selected if they were detected in more than 40% of cells, in at least one of the two compared clusters and if there was a minimum average difference of In(2) between the two clusters, together with a false discovery rate (FDR) of 5%. As input, the function requires data formatted as a SingleCellExperiment (Amezquita et al., 2020) such as a Seurat object. For the developmental Hb data, the SCANPY object was made into a Seurat object by exporting the expression matrix, the raw matrix and the metadata. Together, these were transformed into a Seurat object using the function CreateSeuratObject(), where the batchcorrected matrix was added separately. After DEG were determined for each dataset, the intersection of DEG sets for dataset 1 and 2 was taken as input for the correlation analysis. By doing so, genes absent in either one of the datasets were excluded from the analvsis. Gene names in the zebrafish datasets were converted to one-to-one mouse orthologous gene using BioMarts (Kinsella et al., 2011). With the AverageExpression() function in Seurat a data matrix was created that holds the average expression for the DEGs of each cluster. To show the relationship of the genes to each of the clusters, a gene-cell type specificity matrix was calculated. Each average expression was divided by the sum of all, giving a fraction of specificity for all clusters.

Pairwise Spearman rank order correlations were calculated on the gene-cell type specificity matrices, between all clusters for dataset 1 and 2. Significance of correlation coefficients was determined with a permutation test, where the gene expression values were shuffled one thousand times across cell types. Afterwards, the resulting Spearman correlation coefficient rho was calculated with the function:

$$r_{\rm s} = 1 - \frac{6 \Sigma d^2}{n(n^2 - 1)}$$

Where r_s , d and n resemble rho, the difference between ranks and number of genes, respectively. The p value was calculated as the fraction of the absolute value of rho values that were greater than or equal to the value of rho for the non-shuffled data. The correlation coefficients and p values were plotted in a heatmap from the heatmaply⁵⁵ package, where the clusters were ordered by hierarchical clustering, with Euclidean metric for distance calculation and complete-linkage clustering.

pySCENIC

To identify gene regulatory networks that are important during habenula development the pySCENIC pipeline was used. In short, the uncorrected count matrix (that had not been corrected for batch-effects) was used to calculate co-expression models using GRNBoost2. Then, using cisTarget, regulons were calculated and filtered for the presence of a binding region for the respective transcription factor within 10 kb of the transcription start site. Subsequently, using AUCell, regulon activity per cell type was captured in a matrix and integrated with the Anndata object in SCANPY. This was used as input for t-SNE visualizations of the cells (Figure 4C). Regulon specificity scores were calculated per developmental timepoint and Louvain cluster using the regulon_specificity_scores(auc_mtx, adata.obs[])function. To calculate timepoint-specific regulon expression Z-scores were calculated (Figure 4D).

MAGMA gene property analysis

Genome Wide Association Studies (GWAS) summary statistics on MDD (Wray et al., 2018), SCZ (Ripke et al., 2014), BMI (Locke et al., 2015), BIP (Hou et al., 2016) and height (Wood et al., 2014) were analyzed in MAGMA (de Leeuw et al., 2015; Skene et al., 2018) to converge genetic associations to cell types.

The GWAS datasets had to be processed into the same format including the SNP ID, chromosome and base pair number as the first three columns, and the p value (decimal notation) as column "P". A gene-cell type specificity matrix, holding the relative gene expression level per cluster, was calculated. After the matrix was created, uninformative genes were removed from the matrix using an ANOVA test of variance. This analysis was limited to one-to-one human orthologous genes to match GWAS loci. This resulted in the following number of genes for each dataset matrix: 13,330 (DevHb), 11,392 (Hashikawa et al., 2020), 11,217 (Wallace et al., 2020), 14,033 (Schaum et al., 2018), 11,381 (Kim et al., 2020) and 12,214 (La Manno et al., 2016). Gene-property analysis was performed for each GWAS and differently processed scRNA-seq dataset combination, where the input of the function was given as follows: –gene-results [GWAS] –gene-covar [scRNA-seq]. The direction of gene property testing was set to positive (using the flag –model direction = greater). A positive gene-property analysis was performed. The results of the gene-property analysis were visualized in a bar plot, facet wrapped on GWAS type, using ggplot2 (Wickham, 2016). The significance threshold was corrected using a Bonferroni correction.



Quantification of cell body size

To quantify cell-body size (Figure 7C), 2x digital zoom DIC images of $Cartpt^+$ and $Cartpt^-$ neurons were processed in ImageJ. Ferret's diameter (the longest distance between any two points) was measured along each cell-containing ROIs. One-way ANOVA was performed comparing $Cartpt^+$ neurons (n = 15 cells, N = 8 mice), $Cartpt^-$ MHb neurons (n = 14 cells, N = 8 mice) and $Cartpt^-$ LHb neurons (n = 5 cells, N = 2 mice). Results are shown as mean \pm SEM.

Two-step cluster analysis of electrophysiological parameters

Two-step cluster analysis (Figures 7H–7J) was based on the Schwarz's Bayesian criterion, with automatic generation of cluster numbers to avoid biased cluster-member selection. Cell capacitance and membrane resistance were chosen as input variables for clustering, as these parameters showed the largest differences between *Cartpt*⁺ neurons and both non-*Cartpt* populations. Statistical analysis, including cluster analysis, was performed using IBM SPSS Statistics (IBM Corp., Armonk, NY, US, version 27) software.

Quantification of Dextran⁺/Cartpt⁺ neurons

Corresponding to Figures 6K and S7K. For quantification of Dextran⁺/*Cartpt*⁺ cells, representative samples (n = 2 mice, N = 19 slices) containing the border region between the MHb and LHb were imaged and analyzed by an automated in-house Fiji (Schindelin et al., 2012) script that allowed for manual inspection and exclusion of false-positive cell identification. In brief, using 40x maximum projection images, Dextran⁺ cells (n = 175) were identified by sequential application of "Intermodes" and "Max entropy" auto-thresholds, with inclusion parameters of particles size (60–6000) and circularity (0–1.00). ROI masks of detected Dextran⁺ cells were then projected to the TdTomato (*Cartpt*⁺) channel and overlapping signals were established in an unbiased manner. No further statistical tests were applied to these data.

Quantification of mTurquoise⁺ habenula neurons

Corresponding to Figure S7B. For quantification of mTurquoise⁺/DAPI⁺ cells, representative samples (n = 3 mice, N = 10 slices) containing the left and right Hb were imaged at 20x (tiled) and analyzed in Fiji (version 1.53f51; Schindelin et al., 2012). Using maximum intensity projection images, mTurquoise⁺ cells in the LHb and MHb and DAPI+ nuclei in the LHb were identified by manual thresholding followed by "Watershed" processing with inclusion parameters of particle size (10–250) and circularity (0–1.00). DAPI⁺ nuclei in the MHb were manually counted using the "Cell Counter" plugin. No further statistical tests were applied.

3DISCO analysis of Cartpt⁺ neuron localization

Corresponding to Figure 6G. Z-stack light sheet images were merged into an 3D Imaris file. Then, surfaces were created of the 3rd ventricle and the left half of the Hb. This was done in the "surfaces" tool in Imaris, using the "manual creation" option, with "drawing mode = click" in the "surfaces" window. The surfaces were traced in coronal planes every 100 μ m along the rostral-to-caudal axis. Then, the *Cartpt*⁺ RFP⁺ channel was masked using the Hb surface, creating a new masked *Cartpt*⁺ RFP⁺ channel. Subsequently, Cartpt⁺ RFP⁺ neurons were marked with spots. The average diameter of the *Cartpt*⁺ cells was measured in the "slice view" setting. This average diameter was used as input for creating the spots. Spots were created from the *Cartpt*⁺ RFP⁺ channel. As a final check, spots were manually inspected for false positives / false negatives. To localize the *Cartpt*⁺ spots, a reference frame was placed at the most rostral edge of the Hb aligned along the midline of the 3rd ventricle. To measure the length of the Hb along the rostral-to-caudal axis. This was done filtering the spots based on "position Y reference frame", adding an upper and lower boundary per bin. Spots within each bin were color-coded. Then, using the "statistics" option within the *Cartpt*⁺ spots object the "distance from surface = 3rd ventricle" was used to calculate the distance from the ventricle for each spot (medial-to-lateral axis X). These results were exported as .*cvs* and analyzed in Graphpad version 9.1.1. No further statistical tests were applied to these data.

Statistical analysis electrophysiological parameters

Electrophysiological parameters were calculated off-line using Igor Pro 8 (version 8.0.4.2, Wavemetrics, Tigard, OR, US). Statistical tests on these electrophysiological parameters were subsequently performed in IBM SPSS 27. Results were shown as mean \pm SEM. Per experiment statistical "n" and test were indicated in the text or legend.