Comparative single-cell transcriptomic analysis reveals key differentiation drivers and potential origin of vertebrate retina

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Abstract

Despite known single-cell expression profiles in vertebrate retinas, understanding of their developmental and evolutionary expression patterns among homologous cell types remains limited. We examined and compared approximately 230,000 retinal cells from three species and found significant similarities among homologous cell types, indicating inherent regulatory patterns. To understand these shared patterns, we constructed gene regulatory networks for each developmental stage for the three species. We identified 690 regulons governed by 530 regulators across three species, along with 10 common cell type-specific regulators and 16 preserved regulons. RNA velocity analysis pinpointed conserved driver genes and regulators key to retinal cell differentiation in both mouse and zebrafish. Investigation of the origins of retinal cells by examining conserved expression patterns between vertebrate retinal cells and invertebrate *Ciona intestinalis* photoreceptor-related cells revealed functional similarities in light transduction mechanisms. Our findings offer insights into the evolutionarily conserved regulatory frameworks and differentiation drivers of vertebrate retinal cells.

Keywords: Comparative cross-species transcriptomics, Gene regulatory network, RNA velocity, Evolutionary origin, Retina, Single-cell transcriptomics

Introduction

The retina acts as the primary sensory interface between the visual system and the external world and is crucial for vertebrate adaptation to varied ecological niches. Because of its structural complexity and accessibility, the retina is a pivotal model for studying the intricate functions of the central nervous system (CNS) [1,2]. For example, the presence of β-amyloid in both Alzheimer's and age-related macular degeneration not only underscores the interconnected pathways of the retina and CNS, but also highlights the critical role of retina research in understanding and potentially treating CNS disorders [2]. The vertebrate retina comprises one embryonic retinal progenitor cell (RPC) and seven mature cell types that collaborate in visual information processing: rod and cone photoreceptors (PRs), bipolar cells (BCs), horizontal cell (HCs), amacrine cells (ACs), retinal ganglion cells (RGCs) and müller glia (MG) [3]. Notably, certain developmental phases in the zebrafish retina feature ciliary marginal zone (CMZ) progenitors or retinal stem cells (RSCs) akin to RPCs [4]. Unraveling the intricate regulatory mechanisms of the vertebrate retina has been challenging because of the high cellular heterogeneity and the various patterns of cell body distributions and connectivity [5].

The rapid advancement of single-cell RNA-seq (scRNA-seq) has enabled the detailed profiling of thousands of retinal cells, revealing molecular features of diverse cell classes in model organisms such as *Mus musculus* (mouse) [6], *Gallus gallus* (chicken) [7], and *Danio rerio* (zebrafish) [8]. These studies have not only deepened our understanding of cellular identities within the vertebrate retina but also set a foundation for broader comparative analyses. Recent research emphasizes the importance of such comparative studies [7,9], highlighting the shared and unique transcriptomic landscapes across different species. In addition, the field is advancing beyond mere identification of

marker genes, venturing into the exploration of gene regulatory networks. Methods such as SCENIC

[10], which predict gene regulatory networks based on co-expression and transcription factor binding

information, have enabled a deeper insight into the regulatory mechanisms underlying cellular

functions. While these approaches have been applied to the study of mouse retinal cells[11], a

thorough understanding of the conserved regulatory networks in the vertebrate retinal cells is still an

area awaiting further exploration.

Furthermore, identifying genes positively related to retinal cell differentiation, also known as

driver genes, is crucial in understanding retinal development. Previous studies on retinal

differentiation in mouse and zebrafish have primarily utilized methods based on transcriptional

similarity [6,8]. However, these approaches often fall short in capturing the dynamic and complex

cellular changes over time[12], thus limiting our ability to accurately identify cell type-specific driver

genes. Recent advancements have introduced methods grounded in RNA velocity[13,14], which model

the direction and rate of gene expression by using the ratio of unspliced to spliced RNAs, thereby

determining the direction and speed of individual cells. These methods significantly enhance our

ability to predict cellular differentiation trajectories. Combining RNA velocity with transcriptional

similarity enables a more refined strategy for accurately pinpointing driver genes crucial for

developmental processes[15]. Identifying the conserved driver genes between mouse and zebrafish

that underlie retinal differentiation is essential for advancing our understanding of the fundamental

mechanisms governing retinal development across vertebrates.

In addition, considering the complexity of vertebrate retina, exploring corresponding cell types in

simpler invertebrate organisms offers a feasible method to grasp the fundamental regulatory

mechanisms of the vertebrate retina. *Ciona intestinalis* Type A (*Ciona*), one of the closest relatives of

vertebrates, featuring a relatively simplistic nervous system and visual structures, is a fitting model to

investigate the evolutionary origins of the vertebrate retina [16–18]. Although a previous study suggested

that dopamine-synthesizing cells that are located close to *Ciona* photoreceptors have some

vertebrate AC characteristics [19], an in-depth exploration of shared expression patterns between

vertebrate retina and *Ciona* PR-related cells is lacking.

To bridge these gaps, we systematically investigate the transcriptional profiles of vertebrate

retinal cells from three representative species: mouse, chicken, and zebrafish [6–8], focusing on

conserved gene regulatory networks (GRNs) and key driver genes for retinal cell differentiation. We

also extend our analysis by comparing these profiles with the invertebrate *Ciona* [20] to trace the

potential origins of retinal cells.

Results

Overview of the study strategy

We collected and annotated scRNA-seq data from GEO database (Figure 1A). We initially focused

on assessing the conservation of transcriptomes in vertebrate retinal cells across different species

(Figure 1B). Subsequently, we applied SCENIC [10] to discern cell-type-specific regulatory networks

within each species (Figure 1C). We thus identified conserved regulators and regulons across

vertebrates, elucidating evolutionarily preserved regulatory frameworks. In addition, we utilized RNA

velocity analysis to identify conserved driver genes that propel retinal cell differentiation in both

mouse and zebrafish (Figure 1D) and extracted the regulators that drive differentiation by

intersecting regulatory networks (Figure 1E). Finally, we collected scRNA-seq data from *Ciona*

(Figure 1F), identified shared differentially expressed genes (DEGs) between vertebrate retinal cells

and *Ciona* photoreceptor-related cells (Figure 1G), and inferred the potential origin of vertebrate

retinal cells.

Conserved expression patterns in vertebrate retinal cell landscapes

We downloaded scRNA-seq data of retinal cells for the three species from the GEO database (see

Methods). Across the three species, we collected data obtained at 18 different time points (Figures

2A–2F). In total, about 230,000 cells were retained for subsequent analysis, including 100,506

mouse cells (Figure 2A), 86,056 chicken cells (Figure 2B), and 49,311 zebrafish cells (Figure 2C).

The data from all three species comprised seven mature cell types distributed over stages. In the

mouse data, RPCs were divided into three categories: early RPCs, late RPCs, and neurogenic cells,

representing different states of RPCs. RGCs and HCs were predominantly represented in the data

from the early stages of retinal development during Embryonic day 11 to 16 (E11–E16). ACs, cones,

and rods were more represented in the intermediate stages during E18 to postnatal day 2 (E18–P2),

and BCs and MG in the late stages (P5–P14, Figure 2D). Chicken data at Embryonic day 18 (E18)

included all mature cell types except RGCs, which were represented only in the E16 stage. Cells at

E12 represent the early stages of cell development (Figure 2E). Zebrafish data from at 24 hours

post-fertilization (hpf) and 36 hpf comprised of RPCs and RGCs, and the data at the 72 hpf stage

included all retinal cell types. Data at 48 hpf and 14 days post-fertilization (dpf) represented newly

born retinal cells from different regions. Previous reports indicate that the transcriptome profiles and

functions of RPCs and CMZ progenitors are largely congruent [8]. Therefore, we categorized CMZ

progenitor cells as RPCs (Figure 2F).

To evaluate transcriptome conservation across species, we compared cell similarities among the

three species included in this study by first constructing pseudo-cells. This involved combining 10

cells within each cell type to reduce the noise from sequencing, using all the retained cells (see

Methods). The heatmap (Figure 2G) revealed that the overall gene expression patterns in vertebrate

retinal homologous cell types were conserved, indicating that individual cell types displayed less

variation in their transcriptome profiles during evolution than different cell types within the same

species. This observation is consistent with a recent study surveying birds, mammals, reptiles, fish,

and lamprey [9], further substantiating the notion of transcriptome conservation across diverse

vertebrate lineages. All MG shared similar expression patterns with RPCs, indicating their ability to

differentiate into other retinal cell types [21]. Cones and rods shared more similar transcriptional

profiles, which is consistent with their common role in converting light signals to neural signals [22].

Additionally, ACs, HCs, and RGCs demonstrated closer relationships than other cell types, indicating

they share similar expression patterns. Unsupervised clustering of integrated pseudo-cell data

revealed six major clusters (Figure 2H, see Methods). Each cell type was confined to a specific

cluster, and almost all the homologous cell types were grouped into one cluster, indicating the

similarity of expression profiles among homologous cell types (Figure 2I).

Identification and reconstruction of gene regulatory networks (GRNs) in vertebrate retina

cells across species

To construct the GRNs for cells from the three species, we first built the TF-motif ranking database

for chicken and zebrafish (see Methods). Then, we estimated the GRNs for cell types from each

developmental stage from each species using the pySCENIC pipeline [23] (see Methods).

Approximately 2700 regulons were inferred from the 18 datasets. A regulon was defined as a

collection of genes regulated by the same transcription factor (regulator). We calculated the regulon

specificity scores for each regulon and extracted the top 20 cell type-specific regulons. We then

removed low-quality regulons whose regulators were not specifically expressed in one cell type.

Ultimately, 690 cell type-specific regulons were retained for further analysis (Figure 3A). We

identified numerous regulators in various cell types across different stages (Figures S1A–S1G, S2A,

S3A–S3D). Importantly, the regulators identified in mouse align with those discovered in a previous

study [11], despite the study utilized a different method to infer gene regulatory networks. Furthermore,

we reconstructed GRNs for representative cell types in each stage (Figures S1H–S1Q, S2B–S2G,

S3E–S3M), providing detailed maps of the complex interactions among regulators that orchestrate

retinal development and function. To provide a focused discussion, we have selected one

representative stage from each species for an in-depth analysis.

In our study focused on E18 for mouse, we identified both well-established and novel cell

type-specific regulons (Figure 3B). Notably, Sox9 and Hmga2 were active in late RPCs [24,25]. In

neurogenic cells, Atoh7 was verified to be active [26], whereas Tfdp1, Olig2, and Dlx1 appeared to

have novel regulatory functions. Gbx2 was particularly noted in a subtype of gamma-aminobutyric

acid (GABA)ergic ACs [27], and Barhl2 [28] was identified as an active regulator in ACs and RGCs. We

identified several regulators (i.e., Irf5, Lhx5, Pitx1, Pou3f1, Dlx1, and Mafb) that were specifically

active in glycine ACs. Dlx1, essential for RGCs development [29], was also active in glycine ACs.

Established regulators Isl2 and Rxrg were observed in RGCs and PRs [30,31], with Zfp444 and Lcorl as

potential key regulators in PRs.

At the E18 stage in chicken, we identified known and novel retinal cell regulators (Figure 3C).

PAX6 was identified in AC [32]. In BCs, SOX6, DMBX1, and LHX3 were identified as potential active

regulators. HMX3 was found in HCs and MEOX2 in PRs. The known regulator THRB[33] was identified

in cones, and VAX1 in rods. In MG, mouse homologs SOX9 and NR2E1 were verified [24,34], with

SMAD5 and PAX2 as potential regulators.

Our analysis of retinal cell types at 72 hpf in zebrafish (Figure 3D) validated known regulators

and identified novel ones. Key findings include: tcf7l1a [35] and sox9b [36] were found in RPCs along

with the novel regulator, thap12a; in MG, sox10 and jdp2b were identified as potential novel

regulators. In RGCs, our data suggested activity for the novel regulators sox6 and tbr1b, and

corroborated klf7a [37]. We identified roraa as a potential active regulator across all cell types

examined. In ACs and RGCs, pbx3a, nr4a3, sox4b, and pparg were identified. In HCs, the activity of

lhx1a was confirmed [38] and hmx2, hmx3a were identified as novel regulators. In BCs, vsx1 [39],

dmbx1a and rorab were active. Our analysis confirmed the activity of thrb and crx in cones and nrl in

rods, wherein we also identified a novel regulator, mafba.

Based on regulon enrichment in specific cell types, we determined cell type-specific regulons to

reconstruct a GRN. As expected, the regulators within these regulons exhibited pronounced

interactions, implying a collaborative effort to preserve cell identity and foster cell differentiation. In

glycine ACs in the mouse E18 stage, hub regulators such as Tfap2c, Pou3f1, and Lhx5 significantly

influence the GRN (Figure 3E). In cones, Crx, Thrb, Otx2, Zfp444, and Lcorl (Figure 3F) have

multiple regulatory ties. In MG in the chicken E18 stage, SMAD5 and PAX2 are interacting with

SOX9 and JUN (Figure 3G). In the zebrafish 72 hpf HC GRN (Figure 3H), regulators such as

onecut2, lhx1a, and rxraa, hmx3a, form a grouped pattern, suggesting that cooperate in certain

biological functions.

Identification of conserved regulators and regulons in mouse, chicken, and zebrafish retina

Given the conservation and comparability of the transcriptional profiles in retinal cells across the three species, we sought to determine if their GRNs were conserved. First, we identified 530 unique cell type-specific regulators from the three species, yielding 250 regulators from mouse, 56 from chicken, and 224 from zebrafish. Then, we examined the overlap of cell type-specific regulators among these species (Figure 4A). We identified 10 regulators that were consistently active in all three species, including Gbx2-homolog, Lhx9-homolog, and Pbx1-homolog in ACs and Sox9-homolog in MG (Figure S4). Such overlap indicates their evolutionary conservation and pivotal roles in retinal biology. For instance, Gbx2 is expressed in a particular subtype of mouse ACs (Figure 4B) [27], and its homologs are also specifically expressed in ACs. Similarly, the Maf-homolog demonstrated expression in ACs across all three species (Figure 4B), but its role remains to be explored.

We also observed overlaps among two species in distinct retinal cell types. Hmx3 (hmx3a) was highly specific to HCs in chicken and zebrafish (Figure S5A), likely because of the absence of Type 2 HC in mice and humans (Figure S5B) [40]. Similarly, Dmbx1 (dmbx1a) was distinctively expressed in BCs of chicken and zebrafish but not in mouse (Figure S5C). A set of regulators, including Tcf7l1(tcf7l1a) and Pax2 (pax2a), were predominantly linked to RPCs in mouse and zebrafish because of a lack of chicken RPC data (Figures S6A–S6I). Regulators such as Ebf1 (ebf3a) and Lhx4 (lhx4), which were identified as active in mouse and zebrafish, were also expressed in the corresponding chicken cell types (Figures S6J–S6P). regulators such as Tfap2c (TFAP2C) and Eomes (EOMES) were active in both mouse and chicken and were found in zebrafish as well (Figures S7A and S7B). PAX6 (pax6a), SOX6 (sox6), and ONECUT3 (onecut3b), which were

identified as active in chicken and zebrafish, were consistently expressed across all three species

(Figures S7C–E).

After having closely examined the conservation of regulators across the retinas of the three

species, we expanded our analysis to the overarching regulatory networks they orchestrate. We

employed AUCell [10] to measure the activity of a cell type-specific regulon from one species across all

cell types in the other two species and computed the average scaled regulon activity for each cell

type (see Methods). If a cell type-specific regulon presented a high average scaled activity in its

homolog cell type, it was considered a conserved regulon. With an average scaled activity threshold

of 1.6, we found 16 conserved regulons: six from mouse (Figure 4C), four from chicken (Figure 4D),

and six from zebrafish (Figure 4E).

Having identified these 16 conserved regulons, our next aim was to delve deeper into the gene

composition of these regulons to discern which genes show conserved expression patterns across

species. Specifically, we examined the genes within these conserved regulons that were highly

expressed in the homologous cell types across all three species (Figures S8–S11). Among the

conserved regulons, we found that the Lhx4 regulon (Figure S8), which was highly active in mouse

cones at E14 stage (Figure 4C), contained several genes co-expressed with well-known marker

genes such as *Thrb*-homolog and *Crx*-homolog, including *Gabrb3*-homolog, *Pcdh15*-homolog,

Pclo-homolog, and *Prom1*-homolog. Mutation of *GABRB3* has been previously linked to human early

infantile epileptic encephalopathy [41]. Similarly, we observed that the PBX1 regulon (Figure S9),

active in chicken ACs at E12 stage (Figure 4D), showed conserved expression of the *Auts2*-homolog

across all three species. Mutations in *AUTS2* are strongly associated with autism spectrum disorders

[42]. The mafba regulon was active in zebrafish rods at 72 hpf stage (Figure 4E). Genes such as

Tulp1-homolog, *Nr2e3*-homolog, and *Pde6g*-homolog showed cell type-specific expression in rods

across all three species (Figure S10). The VAX1 regulon, which was active in chicken rods at E18

stage (Figure 4D), includes gene ENSGALG00000051404, which was specifically expressed in

chicken rods (Figure S11). Its zebrafish ortholog, si:ch211−113d22.2, shows similar specificity in rods.

A mouse ortholog was not identified.

Identification of key drivers in mouse and zebrafish retinal development

We identified an array of cell type-specific regulators, some of which were conserved across the

three species. This led us to question their pivotal regulatory roles during developmental processes.

To deduce differentiation trajectories for both mouse and zebrafish, we first calculated RNA velocities

using scVelo [14] and visualized them through a two-dimensional uniform manifold approximation and

projection (UMAP) (see Methods). The resulting embedding stream plot demonstrated a pronounced

motion of RPCs in the direction of mature retinal cell types, suggesting that RNA velocities capture

the critical information on key drivers of differentiation (Figures 5A and 5B). Based on combined RNA

velocities and transcriptional similarity, we determined the fate probability of each cell as well as the

driver genes guiding their differentiation into specific cell types (see Methods). We visualized the top

50 correlated driver genes for each cell type in the two species (Figures S12 and S13).

To assess the biological functions of the driver genes across mouse and zebrafish, we

conducted Gene Ontology (GO) analysis focusing on the biological process (BP) category for the top

100 correlated driver genes in each species independently (see Methods, Figure S14). The BPs of

driver genes closely aligned with the specific functions of the corresponding cell types, and

homologous cell types across species shared similar BPs, underscoring the existence of conserved

driver genes.

We further identified two kinds of key differentiation drivers (Figure 5C): 1) differentiation drivers

shared between mouse and zebrafish and 2) regulators that drive differentiation in mouse and

zebrafish.

We examined the intersection of top 100 driver genes in homologous cell types and observed an

overlap, particularly in ACs, RGCs, and cones (Figure 5D). Specifically, 19 shared driver genes were

identified in ACs in both species, followed by 16 in RGCs and 12 in cones. Among the shared driver

genes, *Pax6*-homolog, *Gng3*-homolog, *Stmn2*-homolog, *Rtn1*-homolog, *Syt4*-homolog, and

Rbfox2-homolog drove both AC and RGC development (Figures 5E, 5F, S15A, and S15B),

suggesting that they may be modulated by similar signaling pathways in the early stages of retinal

development. *Celf4*-homolog and *Gad1*-homolog were identified as specific drivers of AC

differentiation (Figures 5E and 5F), whereas *Elavl4*-homolog, *Pou6f2*-homolog, *Ebf1*-homolog, and

Mapt-homolog were specific drivers of RGC differentiation (Figures S15A and S15B). *Guk1*-homolog

and *Cuedc1*-homolog were important drivers in cones (Figures S15C and S15D). Finally,

Car14-homolog and *Glul*-homolog were identified as important drivers of MG differentiation (Figures

S15E and S15F).

To determine cell-type-specific regulators driving retinal cell development, we cross-referenced

the top 100 correlated driver genes for each cell type with cell type-specific regulators identified using

SCENIC. In mice, 37 regulators were identified, including widely known ones such as Nrl for rods and

Rxrg in cones. We also identified regulators that have not been reported to drive differentiation

(Figure 5G). Sox4 expression displayed a consistent upward trend in all cell types as development

progressed. Pou3f1 exhibited a similar increase in expression in all cell types during differentiation,

with the most pronounced rise observed in the RGCs. Prdm1 expression also increased in all cell

types, but the most marked increase was observed in rods. In contrast, Prox1 showed a

differentiation-dependent increase in expression exclusively in BCs. Similarly, Irx2 expression

increased only in the rod during the development. Using a similar approach for zebrafish, 44

regulators were found, including established ones such as klf7a in RGCs and pax6a in ACs. We

discovered regulators not previously reported to regulate differentiation (Figure 5H). The expression

of sox6 increased in both RGCs and cones during development, whereas pbx1a exhibited a

pronounced increase in ACs and RGCs. The expression of hmx2, hmx3a, and onecutl increased

only in HCs, and rorab expression showed an increasing trend in ACs, BCs, HCs, and RGCs as

development progressed.

Notably, some TFs had already been flagged as cell type-specific regulators in our preceding

SCENIC analysis, validating their critical roles in cellular function and identity. For instance, in ACs,

TFs such as Tfap2c-homolog, Tfap2a-homolog, and Barhl2-homolog (Figures 5E and 5F) were not

only identified as shared driver genes but also appeared in the SCENIC-identified list of regulators in

both mouse and zebrafish. Some TFs, such as Tfap2b-homolog, Meis2-homolog, and

Nhlh2-homolog were shared driver genes but not listed as regulators in the SCENIC analysis

(Figures 5E and 5F).

Identification of conserved expression patterns between vertebrate retina and *Ciona*

PR-related cells

To expand our evolutionary understanding beyond the conserved regulatory genes identified in the

mouse, chicken, and zebrafish retinas, we included an invertebrate *Ciona* in our comparative study.

We first identified a set of conserved DEGs in the retinal cells of vertebrates. The top 10 conserved

DEGs are consistently expressed in the corresponding homologous cells (Figure 6A). Then, we

collected scRNA-seq data from *C. intestinalis*, extracted eight PR-related cell types (Figure 6B), and

identified DEGs in these cell types (see Methods). We analyzed the intersection of DEGs between

vertebrate retinal cell types and *Ciona* PR-related cells based on the 5894 orthogroups generated

using Orthofinder (Figure 6C, see Methods) [43]. Specifically, if DEGs in a certain type of *Ciona*

PR-related cells have orthologous DEGs in a certain type of vertebrate retinal cell, these two cell

types may have a homologous relationship. Twenty-one orthologs of the 23 DEGs in vertebrate

retinal cells also displayed differential expression in PR-related cells in *Ciona* (Figure 6D and Table

S1).

Rx has been reported to be specifically expressed in the *Ciona* developing PR [44]. Among the

DEGs in Rx+ aSV cells of *Ciona*, four orthologous genes were specifically expressed in vertebrate

PRs. In *Ciona* Rx+ aSV cells, the genes KH.C2.828 and KH.C1.346 were prominently expressed.

These genes are orthologous to Phosducin (*PDC*) and *Prom1*, respectively, which were specifically

expressed in PR in vertebrates (Figures 6E and S16A). *PDC* modulates G-protein-coupled receptor

signaling by binding to βγ subunits [45], while mutations in *Prom1* have been associated with retinal

degeneration [46]. These findings suggested that *Ciona* Rx+ aSV cells and vertebrate PRs share

similar mechanisms for light transduction. Likewise, the gene KH.S215.4 was selectively expressed

in Rx+ aSV cells and is an ortholog to Lhx4, a key regulator of retinal neural differentiation

specifically expressed in cones and BCs [47,48] (Figure 6F). KH.L152.2 was exclusively expressed in

Rx+ aSV cells and corresponds to Isl2, predominantly expressed in RGCs and cones (Figure S16B).

Together, these results suggest functional similarities between vertebrate PRs and Rx+ aSV cells of

Ciona.

Among the DEGs of Ci-VP+ pSV cells of *Ciona*, six orthologous genes were specifically

expressed in vertebrates RGCs. The gene KH.L71.1 was specifically expressed in Ci-VP+ pSV cells,

and its ortholog *Rbpms2* was a DEG in RGCs (Figure 6G). *Rbpms2*, which is an RNA binding protein

with multiple splicing, has been used as an RGC marker [49]. Similarly, KH.C14.291 is specifically

expressed in Ci-VP+ pSV cells and is an ortholog to *Dach2* (Figure S16C), which was a DEG in

vertebrate RGCs and HCs. Furthermore, KH.L24.10 and KH.C12.301 were both specifically

expressed in Opsin1+ STUM+ aSV and Ci-VP+ pSV, and their orthologs *Ebf1* and *Fstl4* were

specifically expressed in RGCs (Figures S16D and S16E). Ebf1 is a regulator driving the

differentiation of RGCs [50] and Fstl4 is a marker for a subtype of RGCs [51]. Thus, the function of

Ci-VP+ pSV may be related to that of RGCs.

Three DEGs from the inhibitory neuron eminens had orthologs specifically expressed in various

cell types from the vertebrate retina. KH.C2.526 and KH.L50.5 were predominantly expressed in

eminens cells. Its orthologs, *Slc32a1*-homolog and *Slc24a2*-homolog, are specifically expressed in

vertebrate ACs and cones, respectively (Figures 6H and S16F). Slc32a1, also known as Vesicular

GABA transporter (VGAT), is a transporter responsible for   loading inhibitory amino acids, GABA and

glycine, into synaptic vesicles [52]. Slc24a2, also known as NCKX2, is a sodium-calcium-potassium

exchanger protein. It regulates calcium and sodium ion levels, enabling quick recovery and response

to new light stimuli in cones [53]. KH.S761.6 was a DEG for eminens and Ci-VP+ pSV, its ortholog

Gad2-homolog, was specifically expressed in vertebrate ACs and HCs (Figure S16G). Gad2 is a

GABAergic neuron marker [54]. These findings suggest that vertebrate ACs and cones may share

similar functions with *Ciona* eminens cells.

Finally, the KH.C3.298 was found to be specifically expressed in the GLGB+ pSV, its ortholog

Sept3-homolog was the DEG for vertebrate AC and RGC (Figure S16H).

Discussion

In this study, we comprehensively explored the transcriptional landscape of vertebrate retinal cells

across three representative species, unveiling transcriptomic conservation suggestive of shared

regulatory mechanisms. Our detailed investigation into cell type-specific regulatory networks

provided insights into the inherent regulatory dynamics of retinal cells, emphasizing the role of

conserved regulators and regulons in evolutionary terms. Furthermore, comparative analysis with the

invertebrate *Ciona* underscored potential evolutionary relations, hinting at shared ancestral origins of

certain retinal cells. These discoveries not only expand our understanding of retinal cell biology but

also pave the way for broader inquiries into the evolutionary trajectories of complex visual systems.

Our findings corroborate the idea that the transcriptomes of homologous cells are more similar

across species than they are to other cell types within the same species, which is consistent with a

recent study[9]. This not only affirms evolutionary conservation but also suggests the pivotal role of

these cells in visual function across diverse vertebrates. While our study provided insights into

general transcriptomic similarities among homologous cells in vertebrates, we did not delve into the

finer subclasses of these cells. For instance, 40 types of RGCs have been identified in mouse [51], and

more than 30 types in zebrafish [55]. Because we used a broad classification approach, we may have

overlooked nuanced differences or specializations within these subcategories. Future studies based

on more refined cell-type classifications may unveil additional layers of transcriptomic and functional

complexity, providing new avenues for the exploration of retinal biology and evolution.

Our application of SCENIC uncovered 690 high-quality regulons and their corresponding 530

regulators. The smaller number of regulators identified in chicken can be attributed to the fewer time

points analyzed for this species. These regulators showed a high overlap with regulators previously

reported in the literature. For example, previous research has established the importance of Crx,

Otx2, and Thrb in cone specification [56]. Our study extends this by demonstrating their high interaction

with Zfp444 and Lcorl in cones at the mouse E18 stage, suggesting potential roles for these genes in

cone specification (Figure 3F). Additionally, while lhx1a is known as an HC marker gene[40], and

onecut2 is a regulator of HC fate [57], we discovered their interaction with hmx3a and rxraa in zebrafish

HCs at 72 hpf stage (Figure 3H). The functions of hmx3a and rxraa may also be related to the

identity and specification of zebrafish HCs. In summary, this extensive repository of regulons and

regulators, many of which remain to be explored, represents a rich resource for future investigations

into the intricate mechanisms of vertebrate retinal biology.

We pinpointed 10 conserved regulators and 16 conserved regulons in homologous cells across

vertebrates. The relatively modest number of overlapping regulators and regulons can be linked to

the stringent criteria applied, coupled with the intrinsic high noise levels of single-cell RNA-seq data.

The conservation of these regulators suggests their fundamental importance to basic cellular

functions, while also hinting at unexplored, potentially critical roles. For example, Gbx2, although

proven to be expressed in a subtype of GABAergic ACs [27], has yet to have its function established in

the AC of chicken and zebrafish. The role of Maf, a conserved regulator in ACs, emerges as an

intriguing target for future functional verification. The conserved regulons we identified not only

enhance our understanding of the fundamental functions of the retina but offers insights into complex

brain functions. For instance, we observed a conserved expression of Gabrb3 and Auts2 in cones

and ACs across the three species from the conserved regulons (Figures S8 and S9). While these

genes has been linked to human brain diseases like human early infantile epileptic encephalopathy

and autism spectrum disorders [41,42], their specific functions in these diseases remain to be further

investigated. Studying the functions of these conserved genes in a simpler model, such as the

zebrafish retina, provides a more feasible approach than exploring them in the complex environment

of the human brain.

Our RNA velocity analysis in mouse and zebrafish significantly advanced our comprehension of

genes driving retinal differentiation in vertebrates. For instance, we identified conserved driver genes

between mouse and zebrafish, overlapping with previously identified marker genes, and

demonstrated their intricate involvement in establishing cell identity. For example, Gad1, a known

marker for ACs [58], was predominantly expressed in the terminal stages of differentiation (Figure 5E

and 5F); similar patterns were also observed in other driver genes such as Celf4. This not only

corroborates the role of Gad1 as a marker gene but highlights its importance across species in

establishing cell identity. In addition, we uncovered a set of key regulators that drive differentiation,

which allowed us to pinpoint transcription factors and regulatory networks pivotal in guiding cellular

differentiation pathways. Similarly, a recent study reported the regulatory role of Pou3f1 in RGC

differentiation[59]. This finding closely aligns with our results (Figure 5G), supporting the validity of our

identified regulators.

Our comparative analysis with the invertebrate species *Ciona* allowed us to identify DEGs

shared between vertebrate retinal cells and *Ciona* PR-related cells. These shared genes have been

widely used as markers for vertebrate retinal cells though their functions are not yet fully understood.

For instance, *Rbpms2* is an RNA binding protein with multiple splicing. It has been used as a marker

for vertebrate RGCs, but its specific role in these cells is still unknown [49]. Considering the compact

genome of *Ciona* [60], exploring the regulatory programs of their *Ciona* orthologs could yield valuable

insights into their functional conservation and evolutionary dynamics. Furthermore, through these

shared DEGs, we identified corresponding cell types between vertebrate retina and *Ciona*

PR-related cells, such as vertebrate PRs and *Ciona* Rx+ aSV cells, as well as vertebrate RGCs and

Ciona Ci-VP+ cells. This correlation helps us better investigate the simple patterns of information

transmission within retina, offering insights into more complex visual systems. However, it is

important to note that our conclusions about ancestral relationships are drawn from a small number

of conservative genes. Expanding this analysis to include a broader range of genes and other types

of molecular data may provide a more comprehensive view of evolutionary relationships and might

yield additional insights into the ancestral forms of these retinal cells. Future research should

consider a more extensive comparative analysis for a deeper understanding of the evolutionary

history of the vertebrate retina.

Methods

Single-cell RNA-seq data of retinal cells from mouse, chicken, and zebrafish

The FASTQ data from scRNA-seq retinal cells were retrieved from the GEO database for three

species: mouse, chicken, and zebrafish (GSE118614, GSE159107, GSE122680). Specifically, the

mouse dataset spans ten developmental stages: E11, E12, E14, E16, E18, P0, P2, P5, P8, and P14.

Chicken data were collected on E12, E16, and E18. Zebrafish data were collected 24 hours

post-fertilization (hpf), 36 hpf, 48 hpf, 72 hpf, and 14 days post-fertilization (dpf).

CellRanger (Ver. 6.1.2) [61] was used to create original count matrices, using GRCm38, GRCg6a,

and GRCz11 as the reference genomes. These raw count matrices were imported into R and

analyzed using Seurat 4 package [62]. Cells of low quality were identified and removed following

specific criteria. For the data sets including cell type annotation (mouse and chicken cells at E16 and

E18 stage), cells annotated as either non-retinal or low-quality were removed. In the remaining cases

(zebrafish and chicken cells at E12), cells expressing <200 genes and those with mitochondrial

expression proportion >10% were excluded. For each stage without annotated information, filtered

cells were normalized and scaled by *SCTransform* function, followed by principal component

analysis (PCA) using top 3000 HVGs. The top 20 principal components (PCs) were used for

clustering based on a shared nearest-neighbor modularity optimization-based clustering algorithm

with default parameters and range of resolutions from 0.05 to 0.8. Cells were annotated based on

existing marker genes (Table S2). To integrate samples from the same species, the filtered cells were

integrated, normalized, and scaled using *SCTransform* function, followed by PCA using top 3000

HVGs. Next, Canek [63] was used to eliminate the batch effect by the top 20 PCs. Non-linear

dimensionality reduction was conducted on the corrected data using UMAP.

Transcriptional profile similarity analysis

Firstly, to reduce the noise from sequencing, we combined 10 cells within each cell type to construct

pseudo-cells using all the retaining cells. Secondly, we collected 11,000 one-to-one orthologs for the

three species from biomaRt [64] and homologene [65] package using *getLDS* and *homologene* function,

respectively. Then, we employed the orthologs to integrate the pseudo-cell count matrices, which

was used as input to MetaNeighbor. Specifically, we utilized *run_MetaNeighbor_US* function to

identify the similarity between cell-type pairs among retinal cell types using the 391 highly variable

genes (HVGs) generated by *get_variable_genes* function. In brief, MetaNeighbor uses highly

variable genes (HVGs) to calculate Spearman correlations for gene expression across various cell

types in different species, establishing a measure of similarity. The strength of these correlations is

then applied to compute the Area Under the Receiver Operating Characteristic (AUROC), which

determines how well these correlations can distinguish between similar and dissimilar cell types

across species, thus quantifying the reliability of cell type conservation.

For unsupervised clustering, the integrated pseudo-cell count matrix was used as input to Seurat

4 package. The data were log normalized and scaled, followed by PCA using HVGs generated above.

Next, Harmony [66] was used to eliminate the batch effect by the top 20 PCs. The corrected data was

used for clustering based on a shared nearest-neighbor modularity optimization-based clustering

algorithm with default parameters and a resolution of 0.01. Non-linear dimensionality reduction was

conducted using UMAP.

Ranking database development for chicken and zebrafish

We constructed the TF-motif ranking database for the two species according to the SCENIC tutorial

instructions. First, TF lists for chicken and zebrafish were downloaded from the AnimalTFDB

database [67]. Then, the transcription factor binding motifs in both species were obtained from

JASPAR [68], CISBP [69], DBcorrDB [70], and HOCOMOCO [71]. Given that TF binding specificities are

known to be conserved across species [72], the binding information from their human homologs was

also used. Bind motif matrices for 80.6% (914 /1134) TFs in chicken and 44.3% (1128/2547) TFs in

zebrafish were collected. The promoter sequences (upstream 5000bp and downstream 100bp from

the transcription start site) used for ranking were downloaded from Ensembl (GRCg6a, GRCz11).

Finally, promoter sequences were ranked.

GRN reconstruction

We used pySCENIC [23] for the reconstruction of the gene regulatory network. In brief, we generated

co-expression modules based on the scRNA-seq data using GRNBoost2. A ranking database was

used to filter out indirect targets by comparing the preliminary gene set with the transcription factor

binding predictions, retaining only those genes with strong evidence of direct regulation. The

minimum number of genes in a module was set to 8, with all other settings at their default values.

The ranking database for mouse was obtained from cisTarget database. Finally, the activity of each

filtered module in each cell was estimated using AUCell [10]. AUCell estimates the activity of specific

gene sets in individual cells by calculating the area under the curve of the ranked gene expression

distribution, assuming that if the genes in the set are highly active, they should appear earlier in the

ranked list of gene expression. The inferred regulons were filtered by the regulon specificity scores

and the expression specificity its regulator of a regulon. The regulon specificity scores is calculated

by comparing the distribution of a regulon's activity across all cells to its distribution within a specific

cell type, converting the resulting divergence into a score that reflects how specific the regulon is to

that cell type.

Cross-species regulon comparison

We employed cell type-specific regulons for cross-species comparisons (Figures 3B-3D, S1A–S1G,

S2A, and S3A–S3D). First, gene names within these cell-specific regulons were transformed into

their respective homologous gene names in other species, using the previously obtained set of

11,000 one-to-one homologous genes. Subsequently, the AUCell package was used to assess the

activity of each regulon within the cell datasets corresponding to each species. For evaluation, we

incorporated data from mouse cells extracted at four developmental stages: E14, E18, P8, and P14.

In the case of chicken, data was derived from the E16 and E18 stages, while for zebrafish, we

utilized data across all five stages (24, 36, 48, 72 hpf, and 14 dpf). Finally, we computed the mean

z-score of the cell-specific regulons within their homologous cell types.

RNA velocity analysis

First, to generate spliced/unspliced count matrices for samples from mouse and zebrafish, the BAM

files generated by CellRanger (Ver. 6.1.2) were processed using velocyto [13] with the repeat mask

annotation file downloaded from UCSC genome browser (https://genome.ucsc.edu/). Samples from

the same species were then integrated into one loom file. For the zebrafish dataset, cells with

unspliced read proportions lower than 7% were removed. As a result, unspliced reads accounted for

26% of the total in mouse and 14% in zebrafish. Within individual cell types, these proportions

ranged from 19% to 35% in mouse cells and from 9% to 17% in zebrafish cells (Figures S17A and

S17B).

The processed loom files were then used as input to Scanpy [73]. Initially, genes with larger than

10 shared counts were retained. The top 2000 HVGs were then used for PCA, and the top 30 PCs

were used for the calculation of first-/second-order moments for each cell, considering the 30 nearest

neighbors. The moments were used to determine RNA velocities for differentiation using dynamics

model. The transition matrix of cells was estimated by combining gene expression similarity and RNA

velocities using CellRank [15]. The weight for gene expression similarity was set to 0.2. Furthermore,

the initial and terminal states were identified from the transition matrix. In addition, the lineage

probabilities for each cell toward the terminal states of cell types were computed. For each cell type,

we identified the top 100 driver genes through their high correlation with lineage probabilities. For

mouse, these genes had a median correlation of 0.14, p-value $< 2.20 \times 10^{-12}$ (Table S3A); for

zebrafish, the median correlation was 0.27, p-value $< 1.96 \times 10^{-112}$ (Table S3B). To visualize the

trends for driver genes during differentiation, we estimated the Diffusion pseudotime (DPT

pseudotime) for each cell (Figures S17A and S17B). The estimated DPT pseudotime is consistent

with stage information (Figures S17E and S17F). Using this DPT pseudotime to order the cells, we

then fit Generalized Additive Models to the gene expression data, weighting the contribution of each

cell according to fate probabilities.

GO analysis of top 100 cell type-specific driver genes

We employed the *compareCluster* function from the clusterProfiler [74] package for the analysis of

driver genes identified in mouse and zebrafish. The analysis utilized the *enrichGO* function, targeting

Biological processes (BP) with org.Mm.eg.db and org.Dr.eg.db as the gene ontology database for

mouse and zebrafish. We applied the Benjamini-Hochberg method for p-value adjustment, setting

the q-value cutoff at 0.05 for significance. The top 5 significant GO terms for each cell type were

plotted.

Identification of DEGs in vertebrate retina and *Ciona* photoreceptors-related cells

We identified the vertebrate conserved DEGs for the vertebrate retina using the *FindAllMarkers*

function by from Seurat in two datasets, applying criteria of an adjusted p-value <0.05 and setting

'*only.positive = TRUE*' to focus exclusively on positive markers:

1. In the integrated pseudo-cells dataset (Figure 2I), generating pseudo-cell-DEGs.

2. In datasets from each species separately (Figures 2A–C). We took intersections by the 11,000

orthologs to generate ortholog-DEGs.

The final conserved DEGs were defined by retaining genes from pseudo-cell DEGs only if they were

either in the ortholog-DEGs or HVGs from pseudo-cell-DEGs.

We collected scRNA-seq data of *Ciona* at late tailbud II and larval stage from the GEO database

(GSE131155). The single-cell raw count matrices were generated using CellRanger (Ver. 6.1.2) with

KH2013 [60] as the reference genome. Annotations for cellular identities were kindly provided by the

original authors. We focused on cells related to PR, which led to the extraction of eight PR-related

cell types. The selected PR-related cells were normalized and scaled through *SCTransform* function.

Subsequently, PCA was performed to identify the top 20 PCs, which were then used for batch

correction by Canek. The corrected dimensions were further reduced in a non-linear fashion using

UMAP. Marker genes were identified using the *FindAllMarkers* function, applying a threshold of an

adjusted p-value < 0.01 and setting '*only.positive = TRUE*' to focus exclusively on positive markers

Identification of Orthologous gene among mouse, chicken, zebrafish, and *Ciona*

Orthologous gene identification among the four species was conducted using OrthoFinder 2.5.4 [43].

First, we downloaded the protein sequences of the mouse (GRCm38), chicken (GRCg6a), zebrafish

(GRCz11), and Ciona (KH2013) from the Ensembl [75] and Ghost databases [60]

(http://ghost.zool.kyoto-u.ac.jp/download_kh.html). To enhance accuracy and reduce computational

complexity, only the longest transcript variant of a gene was used for multiple sequence alignments

with MAFFT. Subsequently, orthogroups were estimated using the MCL graph clustering algorithm.

2,724 single-copy orthogroups and 6,374 orthogroups containing genes from mouse, chicken,

zebrafish, and *Ciona* were identified. 5,894 orthogroups that contained genes from all species with

less than five copies in each species were retained for further analysis.

Supplementary information

Supplementary Figures

Figures S1-S3. Reconstruction of GRN in mouse (**S1**), chicken (**S2**), zebrafish (**S3**) retinal cells at

various stages. **Figure S4**. Expression of cell type specific regulators that are active in all three

species. **Figure S5**. Expression of Hmx3-homolog and Dmbx1-homolog that are only active in

chicken and zebrafish. **Figure S6**. Expression of regulators that are active in mouse and zebrafish.

Figure S7. Expression of regulators that are only active in mouse and chicken, as well as zebrafish.

and chicken. **Figures S8-S11**. Top expressing genes in the E14_Lhx4 (**S8**), E12_PBX1 (**S9**),

72hpf_mafba (**S10**), E18_VAX1 (**S11**) regulon identified as active in mouse cones, chicken ACs,

zebrafish rods, and chicken rods, respectively. Figures **S12-S13**. Gene expression trends of top 50

driver genes along DPT pesudotime in mouse (**S12**) and zebrafish (**S13**) retina. **Figure S14**. GO

analysis of top 100 cell type specific driver genes in mouse and zebrafish. **Figure S15.** Gene

expression trends of shared driver genes along DPT pesudotime between mouse and zebrafish.

Figure S16 Shared DEGs between vertebrate retinal cells and *Ciona* PR-related cells. **Figure S17**

RNA velocity preprocessing analysis in mouse and zebrafish retinal cells.

Supplementary Tables

Table S1. Share DEGs between vertebrate retinal cells and *Ciona* PR-related cells.

Table S2. Marker genes for annotating vertebrate retinal cells from chicken and zebrafish.

Table S3. Cell type specific driver genes for mouse and zebrafish.

Declarations

Ethics approval and consent to participate

Not applicable

Consent for publication

Not applicable

Availability of data and materials

FASTQ data from scRNA-seq retinal cells were retrieved from the GEO database for the four species:

mouse, chicken, and zebrafish, *Ciona* (GSE118614, GSE159107, GSE122680, GSE131155). Code

generated   for   this   project   is   available   on   GitHub

(https://github.com/xzengComBio/VertebrateRetina_scRNAseq),

Competing interests

The authors declare that they have no competing interests.

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Author's contributions

Conceptualization, X.Z., F.G., T.G.K., and K.N.; Data Curation, X.Z.; Formal Analysis, X.Z.;

Investigation, X.Z., F.G., C.Y., and T.G.K.; Writing – Original Draft, X.Z. and M.L.; Writing – Review &

Editing, X.Z., F.G., Y.C., M.L., T.G.K., and K.N.; Funding Acquisition, T.G.K. and K.N.; Resources,

K.N.; Supervision, T.G.K. and K.N.

References

1.   Lucas-Ruiz, F., Galindo-Romero, C., Albaladejo-García, V., Vidal-Sanz, M., and
Agudo-Barriuso, M. (2021). Mechanisms implicated in the contralateral effect in the central nervous
system after unilateral injury: focus on the visual system. Neural Regen. Res. *16*, 2125.
10.4103/1673-5374.310670.
2.   Wang, L., and Mao, X. (2021). Role of Retinal Amyloid-β in Neurodegenerative Diseases:
Overlapping Mechanisms and Emerging Clinical Applications. Int. J. Mol. Sci. *22*, 2360.
10.3390/ijms22052360.
3.   Stenkamp, D.L. (2015). Development of the Vertebrate Eye and Retina. In Progress in
Molecular Biology and Translational Science (Elsevier), pp. 397–414.
10.1016/bs.pmbts.2015.06.006.
4.   Wan, Y., Almeida, A.D., Rulands, S., Chalour, N., Muresan, L., Wu, Y., Simons, B.D., He, J.,
and Harris, W. (2016). The ciliary marginal zone of the zebrafish retina: clonal and time-lapse
analysis of a continuously growing tissue. Development, dev.133314. 10.1242/dev.133314.
5.   Hoon, M., Okawa, H., Della Santina, L., and Wong, R.O.L. (2014). Functional architecture of
the retina: Development and disease. Prog. Retin. Eye Res. *42*, 44–84.
10.1016/j.preteyeres.2014.06.003.

6.　Clark, B.S., Stein-O'Brien, G.L., Shiau, F., Cannon, G.H., Davis-Marcisak, E., Sherman, T.,
Santiago, C.P., Hoang, T.V., Rajaii, F., James-Esposito, R.E., et al. (2019). Single-Cell RNA-Seq
Analysis of Retinal Development Identifies NFI Factors as Regulating Mitotic Exit and Late-Born
Cell Specification. Neuron *102*, 1111-1126.e5. 10.1016/j.neuron.2019.04.010.

7.　Yamagata, M., Yan, W., and Sanes, J.R. (2021). A cell atlas of the chick retina based on
single-cell transcriptomics. eLife *10*, e63907. 10.7554/eLife.63907.

8.　Xu, B., Tang, X., Jin, M., Zhang, H., Du, L., Yu, S., and He, J. (2020). Unifying Developmental
Programs for Embryonic and Post-Embryonic Neurogenesis in the Zebrafish Retina. Development,
dev.185660. 10.1242/dev.185660.

9.　Hahn, J., Monavarfeshani, A., Qiao, M., Kao, A.H., Kölsch, Y., Kumar, A., Kunze, V.P., Rasys,
A.M., Richardson, R., Wekselblatt, J.B., et al. (2023). Evolution of neuronal cell classes and types in
the vertebrate retina. Nature *624*, 415–424. 10.1038/s41586-023-06638-9.

10. Aibar, S., González-Blas, C.B., Moerman, T., Huynh-Thu, V.A., Imrichova, H., Hulselmans, G.,
Rambow, F., Marine, J.-C., Geurts, P., Aerts, J., et al. (2017). SCENIC: single-cell regulatory
network inference and clustering. Nat. Methods *14*, 1083–1086. 10.1038/nmeth.4463.

11. Lyu, P., Hoang, T., Santiago, C.P., Thomas, E.D., Timms, A.E., Appel, H., Gimmen, M., Le, N.,
Jiang, L., Kim, D.W., et al. (2021). Gene regulatory networks controlling temporal patterning,
neurogenesis, and cell-fate specification in mammalian retina. Cell Rep. *37*, 109994.
10.1016/j.celrep.2021.109994.

12. Tritschler, S., Büttner, M., Fischer, D.S., Lange, M., Bergen, V., Lickert, H., and Theis, F.J.
(2019). Concepts and limitations for learning developmental trajectories from single cell genomics.
Development *146*, dev170506. 10.1242/dev.170506.

13. La Manno, G., Soldatov, R., Zeisel, A., Braun, E., Hochgerner, H., Petukhov, V., Lidschreiber,
K., Kastriti, M.E., Lönnerberg, P., Furlan, A., et al. (2018). RNA velocity of single cells. Nature *560*,
494–498. 10.1038/s41586-018-0414-6.

14. Bergen, V., Lange, M., Peidli, S., Wolf, F.A., and Theis, F.J. (2020). Generalizing RNA velocity
to transient cell states through dynamical modeling. Nat. Biotechnol. *38*, 1408–1414.
10.1038/s41587-020-0591-3.

15. Lange, M., Bergen, V., Klein, M., Setty, M., Reuter, B., Bakhti, M., Lickert, H., Ansari, M.,
Schniering, J., Schiller, H.B., et al. (2022). CellRank for directed single-cell fate mapping. Nat.
Methods, 1–12. 10.1038/s41592-021-01346-6.

16. Dufour, H.D., Chettouh, Z., Deyts, C., De Rosa, R., Goridis, C., Joly, J.-S., and Brunet, J.-F.
(2006). Precraniate origin of cranial motoneurons. Proc. Natl. Acad. Sci. *103*, 8727–8732.
10.1073/pnas.0600805103.

17. Horie, T., Sakurai, D., Ohtsuki, H., Terakita, A., Shichida, Y., Usukura, J., Kusakabe, T., and
Tsuda, M. (2008). Pigmented and nonpigmented ocelli in the brain vesicle of the ascidian larva. J.
Comp. Neurol. *509*, 88–102. 10.1002/cne.21733.

18. Ryan, K., Lu, Z., and Meinertzhagen, I.A. (2016). The CNS connectome of a tadpole larva of
Ciona intestinalis (L.) highlights sidedness in the brain of a chordate sibling. eLife *5*, e16962.
10.7554/eLife.16962.

19. Razy-Krajka, F., Brown, E.R., Horie, T., Callebert, J., Sasakura, Y., Joly, J.-S., Kusakabe, T.G.,
and Vernier, P. (2012). Monoaminergic modulation of photoreception in ascidian: evidence for a
proto-hypothalamo-retinal territory. BMC Biol. *10*, 45. 10.1186/1741-7007-10-45.

20. Cao, C., Lemaire, L.A., Wang, W., Yoon, P.H., Choi, Y.A., Parsons, L.R., Matese, J.C., Wang,

713 W., Levine, M., and Chen, K. (2019). Comprehensive single-cell transcriptome lineages of a

714 proto-vertebrate. Nature *571*, 349–354. 10.1038/s41586-019-1385-y.

715 21. Goldman, D. (2014). Müller glial cell reprogramming and retina regeneration. Nat. Rev.

716 Neurosci. *15*, 431–442. 10.1038/nrn3723.

717 22. Fain, G.L., Hardie, R., and Laughlin, S.B. (2010). Phototransduction and the Evolution of

718 Photoreceptors. Curr. Biol. *20*, R114–R124. 10.1016/j.cub.2009.12.006.

719 23. Van de Sande, B., Flerin, C., Davie, K., De Waegeneer, M., Hulselmans, G., Aibar, S., Seurinck,

720 R., Saelens, W., Cannoodt, R., Rouchon, Q., et al. (2020). A scalable SCENIC workflow for

721 single-cell gene regulatory network analysis. Nat. Protoc. *15*, 2247–2276.

722 10.1038/s41596-020-0336-2.

723 24. Poché, R.A., Furuta, Y., Chaboissier, M.-C., Schedl, A., and Behringer, R.R. (2008). Sox9 is

724 expressed in mouse multipotent retinal progenitor cells and functions in Müller Glial cell

725 development. J. Comp. Neurol. *510*, 237–250. 10.1002/cne.21746.

726 25. Parameswaran, S., Xia, X., Hegde, G., and Ahmad, I. (2014). *Hmga2* regulates self-renewal of

727 retinal progenitors. Development *141*, 4087–4097. 10.1242/dev.107326.

728 26. Ghiasvand, N.M., Rudolph, D.D., Mashayekhi, M., Brzezinski, J.A., Goldman, D., and Glaser, T.

729 (2011). Deletion of a remote enhancer near ATOH7 disrupts retinal neurogenesis, causing NCRNA

730 disease. Nat. Neurosci. *14*, 578–586. 10.1038/nn.2798.

731 27. Kerstein, P.C., Leffler, J., Sivyer, B., Taylor, W.R., and Wright, K.M. (2020). Gbx2 Identifies

732 Two Amacrine Cell Subtypes with Distinct Molecular, Morphological, and Physiological Properties.

733 Cell Rep. *33*, 108382. 10.1016/j.celrep.2020.108382.

734 28. Ding, Q., Chen, H., Xie, X., Libby, R.T., Tian, N., and Gan, L. (2009). BARHL2 Differentially

735 Regulates the Development of Retinal Amacrine and Ganglion Neurons. J. Neurosci. *29*, 3992–4003.

736 10.1523/JNEUROSCI.5237-08.2009.

737 29. Zhang, Q., Zagozewski, J., Cheng, S., Dixit, R., Zhang, S., De Melo, J., Mu, X., Klein, W.H.,

738 Brown, N.L., Wigle, J.T., et al. (2017). Regulation of *Brn* 3b by *Dlx* 1 and *Dlx* 2 is required for

739 retinal ganglion cell differentiation in the vertebrate retina. Development, dev.142042.

740 10.1242/dev.142042.

741 30. Triplett, J.W., Wei, W., Gonzalez, C., Sweeney, N.T., Huberman, A.D., Feller, M.B., and

742 Feldheim, D.A. (2014). Dendritic and axonal targeting patterns of a genetically-specified class of

743 retinal ganglion cells that participate in image-forming circuits. Neural Develop. *9*, 2.

744 10.1186/1749-8104-9-2.

745 31. Roberts, M.R., Hendrickson, A., McGuire, C.R., and Reh, T.A. (2005). Retinoid X Receptor γ Is

746 Necessary to Establish the S-opsin Gradient in Cone Photoreceptors of the Developing Mouse Retina.

747 Investig. Opthalmology Vis. Sci. *46*, 2897. 10.1167/iovs.05-0093.

748 32. Hsieh, Y.-W., and Yang, X.-J. (2009). Dynamic Pax6 expression during the neurogenic cell

749 cycle influences proliferation and cell fate choices of retinal progenitors. Neural Develop. *4*, 32.

750 10.1186/1749-8104-4-32.

751 33. Enright, J.M., Lawrence, K.A., Hadzic, T., and Corbo, J.C. (2015). Transcriptome profiling of

752 developing photoreceptor subtypes reveals candidate genes involved in avian photoreceptor

753 diversification. J. Comp. Neurol. *523*, 649–668. 10.1002/cne.23702.

754 34. Corso-Díaz, X., and Simpson, E.M. (2015). Nr2e1 regulates retinal lamination and the

755 development of Müller glia, S-cones, and glycineric amacrine cells during retinogenesis. Mol. Brain

756 *8*, 37. 10.1186/s13041-015-0126-x.

35. Young, R.M., Hawkins, T.A., Cavodeassi, F., Stickney, H.L., Schwarz, Q., Lawrence, L.M.,
Wierzbicki, C., Cheng, B.Y., Luo, J., Ambrosio, E.M., et al. (2019). Compensatory growth renders
Tcf7l1a dispensable for eye formation despite its requirement in eye field specification. eLife *8*,
e40093. 10.7554/eLife.40093.
36. Yokoi, H., Yan, Y.-L., Miller, M.R., BreMiller, R.A., Catchen, J.M., Johnson, E.A., and
Postlethwait, J.H. (2009). Expression profiling of zebrafish sox9 mutants reveals that Sox9 is
required for retinal differentiation. Dev. Biol. *329*, 1–15. 10.1016/j.ydbio.2009.01.002.
37. Veldman, M.B., Bemben, M.A., Thompson, R.C., and Goldman, D. (2007). Gene expression
analysis of zebrafish retinal ganglion cells during optic nerve regeneration identifies KLF6a and
KLF7a as important regulators of axon regeneration. Dev. Biol. *312*, 596–612.
10.1016/j.ydbio.2007.09.019.
38. Rao, M.B., Didiano, D., and Patton, J.G. (2017). Neurotransmitter-Regulated Regeneration in
the Zebrafish Retina. Stem Cell Rep. *8*, 831–842. 10.1016/j.stemcr.2017.02.007.
39. Letelier, J., Buono, L., Almuedo-Castillo, M., Zang, J., Mounieres, C., González-Díaz, S.,
Polvillo, R., Sanabria-Reinoso, E., Corbacho, J., Sousa-Ortega, A., et al. (2023). Mutation of vsx
genes in zebrafish highlights the robustness of the retinal specification network. eLife *12*, e85594.
10.7554/eLife.85594.
40. Boije, H., Shirazi Fard, S., Edqvist, P.-H., and Hallböök, F. (2016). Horizontal Cells, the Odd
Ones Out in the Retina, Give Insights into Development and Disease. Front. Neuroanat. *10*.
10.3389/fnana.2016.00077.
41. Papandreou, A., McTague, A., Trump, N., Ambegaonkar, G., Ngoh, A., Meyer, E., Scott, R.H.,
and Kurian, M.A. (2016). *GABRB3* mutations: a new and emerging cause of early infantile epileptic
encephalopathy. Dev. Med. Child Neurol. *58*, 416–420. 10.1111/dmcn.12976.
42. Oksenberg, N., Stevison, L., Wall, J.D., and Ahituv, N. (2013). Function and Regulation of
AUTS2, a Gene Implicated in Autism and Human Evolution. PLoS Genet. *9*, e1003221.
10.1371/journal.pgen.1003221.
43. Emms, D.M., and Kelly, S. (2019). OrthoFinder: phylogenetic orthology inference for
comparative genomics. Genome Biol. *20*, 238. 10.1186/s13059-019-1832-y.
44. Oonuma, K., and Kusakabe, T.G. (2019). Spatio-temporal regulation of Rx and mitotic patterns
shape the eye-cup of the photoreceptor cells in Ciona. Dev. Biol. *445*, 245–255.
10.1016/j.ydbio.2018.11.011.
45. Lee, B.Y., Thulin, C.D., and Willardson, B.M. (2004). Site-specific Phosphorylation of
Phosducin in Intact Retina. J. Biol. Chem. *279*, 54008–54017. 10.1074/jbc.M405669200.
46. Cehajic-Kapetanovic, J., Birtel, J., McClements, M.E., Shanks, M.E., Clouston, P., Downes,
S.M., Charbel Issa, P., and MacLaren, R.E. (2019). Clinical and Molecular Characterization of
PROM1 -Related Retinal Degeneration. JAMA Netw. Open *2*, e195752.
10.1001/jamanetworkopen.2019.5752.
47. Buenaventura, D.F., Corseri, A., and Emerson, M.M. (2019). Identification of Genes With
Enriched Expression in Early Developing Mouse Cone Photoreceptors. Investig. Opthalmology Vis.
Sci. *60*, 2787. 10.1167/iovs.19-26951.
48. Dong, X., Yang, H., Zhou, X., Xie, X., Yu, D., Guo, L., Xu, M., Zhang, W., Liang, G., and Gan,
L. (2020). LIM-Homeodomain Transcription Factor LHX4 Is Required for the Differentiation of
Retinal Rod Bipolar Cells and OFF-Cone Bipolar Subtypes. Cell Rep. *32*, 108144.
10.1016/j.celrep.2020.108144.

49. Rodriguez, A.R., De Sevilla Müller, L.P., and Brecha, N.C. (2014). The RNA binding protein
RBPMS is a selective marker of ganglion cells in the mammalian retina: RBPMS identifies retinal
ganglion cells. J. Comp. Neurol. *522*, 1411–1443. 10.1002/cne.23521.
50. Chuang, J.-H., Yarmishyn, A.A., Hwang, D.-K., Hsu, C.-C., Wang, M.-L., Yang, Y.-P., Chien,
K.-H., Chiou, S.-H., Peng, C.-H., and Chen, S.-J. (2018). Expression profiling of cell-intrinsic
regulators in the process of differentiation of human iPSCs into retinal lineages. Stem Cell Res. Ther.
9, 140. 10.1186/s13287-018-0848-7.
51. Rheaume, B.A., Jereen, A., Bolisetty, M., Sajid, M.S., Yang, Y., Renna, K., Sun, L., Robson, P.,
and Trakhtenberg, E.F. (2018). Single cell transcriptome profiling of retinal ganglion cells identifies
cellular subtypes. Nat. Commun. *9*, 2759. 10.1038/s41467-018-05134-3.
52. Wojcik, S.M., Katsurabayashi, S., Guillemin, I., Friauf, E., Rosenmund, C., Brose, N., and Rhee,
J.-S. (2006). A Shared Vesicular Carrier Allows Synaptic Corelease of GABA and Glycine. Neuron
50, 575–587. 10.1016/j.neuron.2006.04.016.
53. Sharon, D., Yamamoto, H., McGee, T.L., Rabe, V., Szerencsei, R.T., Winkfein, R.J., Prinsen,
C.F.M., Barnes, C.S., Andreasson, S., Fishman, G.A., et al. (2002). Mutated Alleles of the Rod and
Cone Na-Ca K- Exchanger Genes in Patients with Retinal Diseases. *43*.
54. Huang, W., Xu, Q., Liu, F., Su, J., Xiao, D., Tang, L., Hao, Z.-Z., Liu, R., Xiang, K., Bi, Y., et al.
(2022). Identification of TPBG-Expressing Amacrine Cells in DAT-tdTomato Mouse. Investig.
Opthalmology Vis. Sci. *63*, 13. 10.1167/iovs.63.5.13.
55. Kölsch, Y., Hahn, J., Sappington, A., Stemmer, M., Fernandes, A.M., Helmbrecht, T.O., Lele, S.,
Butrus, S., Laurell, E., Arnold-Ammer, I., et al. (2021). Molecular classification of zebrafish retinal
ganglion cells links genes to cell types to behavior. Neuron *109*, 645-662.e9.
10.1016/j.neuron.2020.12.003.
56. Musser, J.M., and Arendt, D. (2017). Loss and gain of cone types in vertebrate ciliary
photoreceptor evolution. Dev. Biol. *431*, 26–35. 10.1016/j.ydbio.2017.08.038.
57. Sapkota, D., Chintala, H., Wu, F., Fliesler, S.J., Hu, Z., and Mu, X. (2014). Onecut1 and
Onecut2 redundantly regulate early retinal cell fates during development. Proc. Natl. Acad. Sci. *111*.
10.1073/pnas.1405354111.
58. Yan, W., Laboulaye, M.A., Tran, N.M., Whitney, I.E., Benhar, I., and Sanes, J.R. (2020). Mouse
Retinal Cell Atlas: Molecular Identification of over Sixty Amacrine Cell Types. J. Neurosci. *40*,
5177–5195. 10.1523/JNEUROSCI.0471-20.2020.
59. Fries, M., Brown, T.W., Jolicoeur, C., Boulan, B., Boudreau-Pinsonneault, C., Javed, A., Abram,
P., and Cayouette, M. (2023). Pou3f1 orchestrates a gene regulatory network controlling
contralateral retinogeniculate projections. Cell Rep. *42*, 112985. 10.1016/j.celrep.2023.112985.
60. Satou, Y., Mineta, K., Ogasawara, M., Sasakura, Y., Shoguchi, E., Ueno, K., Yamada, L.,
Matsumoto, J., Wasserscheid, J., Dewar, K., et al. (2008). Improved genome assembly and
evidence-based global gene model set for the chordate Ciona intestinalis: new insight into intron and
operon populations. Genome Biol. *9*, R152. 10.1186/gb-2008-9-10-r152.
61. Zheng, G.X.Y., Terry, J.M., Belgrader, P., Ryvkin, P., Bent, Z.W., Wilson, R., Ziraldo, S.B.,
Wheeler, T.D., McDermott, G.P., Zhu, J., et al. (2017). Massively parallel digital transcriptional
profiling of single cells. Nat. Commun. *8*, 14049. 10.1038/ncomms14049.
62. Hao, Y., Hao, S., Andersen-Nissen, E., Mauck, W.M., Zheng, S., Butler, A., Lee, M.J., Wilk,
A.J., Darby, C., Zager, M., et al. (2021). Integrated analysis of multimodal single-cell data. Cell *184*,
3573-3587.e29. 10.1016/j.cell.2021.04.048.

63. Loza, M., Teraguchi, S., Standley, D.M., and Diez, D. (2022). Unbiased integration of single cell
transcriptome replicates. NAR Genomics Bioinforma. *4*, lqac022. 10.1093/nargab/lqac022.
64. Durinck, S., Moreau, Y., Kasprzyk, A., Davis, S., De Moor, B., Brazma, A., and Huber, W.
(2005). BioMart and Bioconductor: a powerful link between biological databases and microarray
data analysis. Bioinformatics *21*, 3439–3440. 10.1093/bioinformatics/bti525.
65. Sayers, E.W., Bolton, E.E., Brister, J.R., Canese, K., Chan, J., Comeau, D.C., Connor, R., Funk,
K., Kelly, C., Kim, S., et al. (2022). Database resources of the national center for biotechnology
information. Nucleic Acids Res. *50*, D20–D26. 10.1093/nar/gkab1112.
66. Korsunsky, I., Millard, N., Fan, J., Slowikowski, K., Zhang, F., Wei, K., Baglaenko, Y., Brenner,
M., Loh, P., and Raychaudhuri, S. (2019). Fast, sensitive and accurate integration of single-cell data
with Harmony. Nat. Methods *16*, 1289–1296. 10.1038/s41592-019-0619-0.
67. Shen, W.-K., Chen, S.-Y., Gan, Z.-Q., Zhang, Y.-Z., Yue, T., Chen, M.-M., Xue, Y., Hu, H., and
Guo, A.-Y. (2023). AnimalTFDB 4.0: a comprehensive animal transcription factor database updated
with variation and expression annotations. Nucleic Acids Res. *51*, D39–D45. 10.1093/nar/gkac907.
68. Castro-Mondragon, J.A., Riudavets-Puig, R., Rauluseviciute, I., Berhanu Lemma, R., Turchi, L.,
Blanc-Mathieu, R., Lucas, J., Boddie, P., Khan, A., Manosalva Pérez, N., et al. (2022). JASPAR
2022: the 9th release of the open-access database of transcription factor binding profiles. Nucleic
Acids Res. *50*, D165–D173. 10.1093/nar/gkab1113.
69. Weirauch, M.T., Yang, A., Albu, M., Cote, A.G., Montenegro-Montero, A., Drewe, P.,
Najafabadi, H.S., Lambert, S.A., Mann, I., Cook, K., et al. (2014). Determination and Inference of
Eukaryotic Transcription Factor Sequence Specificity. Cell *158*, 1431–1443.
10.1016/j.cell.2014.08.009.
70. Grau, J., Grosse, I., Posch, S., and Keilwagen, J. (2015). Motif clustering with implications for
transcription factor interactions (PeerJ PrePrints).
71. Kulakovskiy, I.V., Vorontsov, I.E., Yevshin, I.S., Sharipov, R.N., Fedorova, A.D., Rumynskiy,
E.I., Medvedeva, Y.A., Magana-Mora, A., Bajic, V.B., Papatsenko, D.A., et al. (2018).
HOCOMOCO: towards a complete collection of transcription factor binding models for human and
mouse via large-scale ChIP-Seq analysis. Nucleic Acids Res. *46*, D252–D259. 10.1093/nar/gkx1106.
72. Nitta, K.R., Jolma, A., Yin, Y., Morgunova, E., Kivioja, T., Akhtar, J., Hens, K., Toivonen, J.,
Deplancke, B., Furlong, E.E.M., et al. (2015). Conservation of transcription factor binding
specificities across 600 million years of bilateria evolution. eLife *4*, e04837. 10.7554/eLife.04837.
73. Wolf, F.A., Angerer, P., and Theis, F.J. (2018). SCANPY: large-scale single-cell gene
expression data analysis. Genome Biol. *19*, 15. 10.1186/s13059-017-1382-0.
74. Yu, G., Wang, L.-G., Han, Y., and He, Q.-Y. (2012). clusterProfiler: an R Package for
Comparing Biological Themes Among Gene Clusters. OMICS J. Integr. Biol. *16*, 284–287.
10.1089/omi.2011.0118.
75. Cunningham, F., Allen, J.E., Allen, J., Alvarez-Jarreta, J., Amode, M.R., Armean, I.M.,
Austine-Orimoloye, O., Azov, A.G., Barnes, I., Bennett, R., et al. (2022). Ensembl 2022. Nucleic
Acids Res. *50*, D988–D995. 10.1093/nar/gkab1049.

Figure Legends

Figure 1 Workflow illustrating the transcriptional analysis of vertebrate retinal cells and their

comparison with the invertebrate species *Ciona*.

A scRNA-seq data from various stages of three representative species are collected and annotated.

B Transcriptome similarity analysis showing transcriptome conservation among homologous retinal

cells in vertebrates. **C** GRN analysis indicating cell type-specific regulatory networks and evolutionarily

preserved regulatory frameworks within each vertebrate species. **D** RNA velocity-based analysis

revealing conserved driver genes responsible for retinal cell differentiation in mouse and zebrafish.

E Extraction of key differentiation-driving regulators through intersecting regulatory networks and

RNA velocities. **F** scRNA-seq data from *Ciona* are collected. **G** Identification of genes with shared

expression between vertebrate retinal cells and PR-related cells in *Ciona*, implying a potential

ancestral link in their origin.

Figure 2 Comparative single-cell transcriptomic landscapes of retinal cells in mouse, chicken,

and zebrafish

A-C UMAP plots of retinal scRNA-seq data from mouse (**A**, N = 100,506), chicken (**B**, N = 86,056),

and zebrafish (**C**, N = 49,311). **D-F** Proportions of cell types within each stage for mouse (**D**), chicken

(**E**), and zebrafish (**F**). Red indicates a high proportion, blue indicates a low proportion. **G** Heatmap of

AUROC scores computed using MetaNeighbor, indicating similarity among cell types based on

orthologous gene expression from mouse, chicken, and zebrafish. In the heatmap, cell type prefixes

denote their origin: 'M-' for mouse, 'C-' for chicken, and 'Z-' for zebrafish. Red indicates high similarity,

blue indicates low similarity. **H** Unsupervised clustering of pseudo-cells into six clusters displayed in

UMAP. **I** Heatmap displaying the proportion of each cell type within the 6 clusters identified in **H**. Red

indicates a high proportion; blue indicates a low proportion.

Figure 3 Gene regulatory network analysis in vertebrate retinal cells

A Number of identified and filtered regulons from scRNA-seq data across different stages in mouse,

chicken, and zebrafish. Bar represents the number of regulons that have been filtered and identified

from scRNA-seq data at distinct stages in the respective species. **B−D** Heatmaps of scaled regulon

activity across different species and stages, including (**B**) mouse at E18, (**C**) chicken at E18, and (**D**)

zebrafish at 72hpf. Each row represents a distinct regulon and each column denotes a specific cell

type. Representative regulons for each cell type are highlighted in the right panel. Each regulon is

named following the format "TF (N)", where "TF" represents the regulator and "N" denotes the

number of target genes in that regulon. Red indicates high activity, blue indicates low activity. **E-H**

GRN derived from the regulons highlighted in **B-D**. (**E**) mouse E18 glycine ACs, (**F**) mouse E18

cones, (**G**) chicken E18 MG, and (**H**) Zebrafish 72hpf HC. The networks show the interplay among

genes and regulators within the respective cell types. Purple nodes represent regulators, gray nodes

represent target genes. Red edges indicate regulatory interactions between regulators, gray edges

highlight regulatory relationships of regulators with other genes. The layout of the networks is based

on centrality, emphasizing the significance of certain nodes within the GRN.

Figure 4 Cross-species analysis of active regulators and regulons in homologous retinal cells

A Venn diagram illustrating the overlap of regulators in mouse, chicken, and zebrafish. **B** UMAP

visualizations of Gbx2 (top) and Maf (bottom) among the shared regulators that are specifically

actived in ACs in (**A**). Order from left to right: mouse, chicken, zebrafish. Blue indicates high

expression, grey indicates low expression. **C-E** Average scaled activity of cell type-specific regulons

generated from mouse (**C**), chicken (**D**) and zebrafish (**E**) in the homologous cell type from other two

species. Dashed lines at 1.6 (both horizontal and vertical) indicate the thresholds for conserved

regulons.

Figure 5 Analysis of RNA velocities, shared driver genes, and regulators that drive

differentiation in retinal cell development of mouse and zebrafish

A, B UMAP-based projections of RNA velocities for retinal cell differentiation in mouse (**A**) and

zebrafish (**B**). **C** Venn diagram illustrating the relationships among mouse driver genes, mouse

regulators, zebrafish regulators, and zebrafish driver genes. The intersections highlight (1) shared

driver genes between mouse and zebrafish and (2) cell type-specific regulators that drive retinal cell

development in mouse and zebrafish. **D** Bar chart representing the number of shared driver genes

among the top 100 cell type-specific driver genes in both mouse and zebrafish. **E, F** Expression

heatmaps of shared driver genes in ACs across diffusion pseudotime for mouse (**E**) and zebrafish (**F**).

G, H Cell type-specific regulators that drive differentiation in mouse (**G**) and zebrafish (**H**).

Figure 6 Exploration of the origin of the vertebrate retina

A Heatmap of the top 10 conserved DEGs in homologous cell types of the vertebrate retina. Red

indicates high expression, blue indicates low expression. **B** UMAP visualization of scRNA-seq data

from *Ciona* PR-related cells. **C** Venn diagram showing overlapping DEGs between vertebrate retina

cells and Ciona PR-related cells, used to infer the origin of vertebrate retina. **D** Sankey diagram

showing shared DEGs between *Ciona* PR-related cells and vertebrate retinal cells. **E-H** UMAP

visualizations of gene expression levels across the four species. Each panel represents the

expression of a specific gene: (**E**) *Pdc*-orth, (**F**) *Lhx4*-orth, (**G**) *Rbpms2*-orth, and (**H**) *Slc32a1*-orth.

Order from left to right: mouse, chicken, zebrafish, and *Ciona*. Blue indicates high expression, gray

indicates low expression.