Supplementary Information for

A high-risk retinoblastoma subtype with stemness features, dedifferentiated cone states and neuronal/ganglion cell gene expression

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Supplementary Fig. 1 | Multi-omics classification of a series of 72 retinoblastomas.

a, GISTIC plot for copy number alteration data. GISTIC plot of the 72 retinoblastoma
 specimens, integrating frequency and amplitude to identify significant amplifications (red) and
 deletions (blue) across the genome (ordered by chromosome). These significant amplifications
 and deletions were used for copy number alteration data clustering (Fig. 1a).

5 b, Cluster-of-clusters classification. Cluster-of-clusters analysis for 72 retinoblastoma cases, 6 based on three unsupervised partitions, each partition being obtained using a different 7 genomic platform (transcriptomic, methylomic, and copy number alteration data). A [0;1] 8 normalized co-classification matrix was derived from the 3 initial partitions. Hierarchical 9 clustering was then performed with inter-individual distance defined as (1 - co-classification 10 score) and complete linkage. This analysis identified two major groups of 27 and 37 samples, 11 and one ambiguous/unclassified group of 8 samples. The annotations below the dendrogram represent the clusters of samples defined independently by consensus clustering of all three 12 13 genomic datasets.

14 c, Centroid-based classification. This is a schematic representation of the centroid 15 classification methodology used. Each point represents a sample. Samples belonging to cluster-of-clusters 1 are in yellow (n=27), cluster-of-clusters 2 in blue (n=37), and yet 16 17 unclassified samples in black (n=8). On the X-axis is the difference between each sample's 18 correlation to the cluster-of-clusters 1 transcriptomic centroid and the cluster-of-clusters 2 19 transcriptomic centroid. The Y-axis represents the same information for the methylomic 20 centroids. Samples with one dataset missing have this difference set to 0 (X coordinate=0 for 21 missing transcriptome, Y coordinate=0 for missing methylome), and the corresponding data 22 points are circled in red. Centroid correlation differences below 0.2 were considered to be 23 outliers, and are in the gray areas. Gray areas therefore contain outliers or samples with one 24 missing dataset. All cluster-of-clusters 1 samples re-classified correctly in methylomic centroid 25 1 and transcriptomic centroid 1 (or one of the two if a dataset is missing). All but one cluster-26 of-clusters 2 samples re-classified correctly also. The last one (RB208) was an outlier in the 27 methylomic dataset with no transcriptomic data available, and was thus set to unclassified. For 28 the samples with no cluster-of-clusters attribution (black points): RB52, RB23, RB50 and RB33 29 were assigned to the first group; RB22, and RB209 were assigned to the second; RB61 was 30 an outlier in both datasets, and RB60 had a discrepancy in the transcriptomic centroid-based 31 classification and the methylomic centroid-based classification, RB61 and RB60 therefore 32 remained unclassified. Ultimately 31 samples were assigned to the first group, 38 to the 33 second, and 3 (RB208, RB60, RB61) remained unclassifed.

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Supplementary Fig. 2 | Distribution of the differentially methylated CpG between the two subtypes in relation to CpG islands.

- a, Distribution of the hypomethylated CpGs (upper panel) and the hypermethylated CpGs
- 2 (lower panel) in subtype 2 as compared to subtype 1, by CpG content and neighborhood
- 3 context. Hypermethylated CpGs were more frequently located within CpG islands than
- 4 hypomethylated CpGs (751 of 2087 versus 527 of 4520, $p=193 \times 10^{-119}$, Chi² test) (related to
- 5 Figure 2F).

6 b, Boxplot comparing the number of differentially methylated CpGs with high levels of

- 7 methylation (beta-value within 0.8 to 1) located in CpG Islands in subtype 1 (n=27), subtype 2
- non-*MYCN*-amplified (n=32), and subtype 2 *MYCN*-amplified (n=4) tumors (related to Figure
 2g upper panel).
- 10 c, Boxplot comparing the number of differentially methylated CpGs with low levels of
- 11 methylation (beta-value within 0 to 0.2) located outsideCpG Islands in subtype 1 (n=27),
- 12 subtype 2 non-MYCN-amplified (n=32), and subtype 2 MYCN-amplified (n=4) tumors (related
- to Figure 2g bottom panel).
- **b**, **c**, In the boxplots, the central mark indicates the median and the bottom and top edges of the box the 25th and 75th percentiles. Whiskers are the smaller of 1.5 times the interquartile range or the length of the 25th percentiles to the smallest data point or the 75th percentiles to the largest data point. Data points outside the whiskers are outliers. Significance was tested by two-sided Wilcoxon test, $p=1.9\times10^{-7}$.



Supplementary Fig. 3 | Stemness Indices estimated by different signatures.

- **a**, Boxplot of stemness indices in the two subtypes of retinoblastoma (subtype 1 tumors: n=26,
- 2 subtype 2 tumors: n=31) estimated by different signatures (from left to right: Miranda et al.
- 3 2018³³, Shats *et al.* 2011³⁵, Smith *et al.* 2018³⁶). In the boxplots, the central mark indicates the
- median and the bottom and top edges of the box the 25th and 75th percentiles. Whiskers are
 the smaller of 1.5 times the interquartile range or the length of the 25th percentiles to the
- smallest data point or the 75th percentiles to the largest data point. Data points outside the
 whiskers are outliers. Data points outside the whiskers are outliers. Statistical tests are twosided.
- 9 b, Scatter plot showing the correlation between stemness indice estimated by Malta et al.'s
 10 method (y-axis) or by other signatures (x axis, from left to right: Miranda *et al.* 2018, Shats *et*
- 11 *al.* 2011, Smith *et al.* 2018). Two-sided Spearman's correlation test was applied, rho and p
- 12 value are shown.













Horizontal marker expression Signif log2FC Cell exp. TFAP2A H/A Π TFAP2B H/A ns NKAIN2 0 Н SOSTDC1 ns н GRIA3 н 1 SLC32A1 ns H/A 0 ZNF385D ESRRB * H/A * H/R GAD2 H/A ns ** 0 MEGF11 н CACNG3 ns н H H LHX1 TMOD1 ns ns 1 PLXNA2 H/A ns * ELAVL4 H/G/A 0 ONECUT1 **** CABP1 ns H H/A/B CABP1 ns CACNG7 ns н H/C ONECUT3 ns H H DRD2 TNR ns ns CNTNAP2 ns Ĥ LNP1 Н C1QL1 ns н 2 å 6 expression RET21 EB. 1 2 3 -1 0



Bipolar marker expression

- C: Cone
- R: Rod PanPh: Pan-Photoreceptor
- G: Ganglion
- M: Müller glia
- A: Amacrine
- H: Horizontal
- B: Bipolar





Expression in normal developing retina of ganglion markers upregulated in subtype 2 tumors

Supplementary Fig. 4 | Retinal cell gene expression analysis in the two subtypes of retinoblastoma.

- a, The expression of retinal cell-associated genes is presented as a heatmap, for three fetal 1 2 retinas, 26 subtype 1 tumors and 31 subtype 2 tumors. A heatmap was constructed for each 3 retinal cell type, indicating the level of expression, the statistical significance (Signif.) and the 4 log2 fold change (log2FC) of expression between subtype 2 and subtype 1 tumors. It is also 5 indicated whether the gene is expressed by more than one retinal cell type (Cell exp.). The 6 complete list of markers is given in Supplementary Table 3. Limma moderated t-test was used 7 for the analysis of gene expression, BH correction was applied, exact p values are provided in 8 Supplementary Table 3. 9 b, Expression of ganglion markers, displayed in Figure 3e, in the normal developing retina
- 10 according to cell types (single-cell RNA-seq data from Lu *et al.*, 2020³⁸). For each gene and
- each cell type, a pseudo-dot plot is provided. At each age (x-axis), the dot size is proportional
- 12 to the percentage of expressing cells (i.e. non-zero counts) and its y-coordinate indicates the
- 13 mean expression.
- 14 For the markers of all cell types, the online tool can be used to visualize their expression:
- 15 cone: https://retinoblastoma-retina-markers.curie.fr/cone,
- 16 rod: https://retinoblastoma-retina-markers.curie.fr/rod,
- 17 pan-photoreceptor: https://retinoblastoma-retina-markers.curie.fr/pan-photoreceptor,
- 18 ganglion: <u>https://retinoblastoma-retina-markers.curie.fr/ganglion</u>,
- 19 muller glia: <u>https://retinoblastoma-retina-markers.curie.fr/muller</u>,
- 20 amacrine: https://retinoblastoma-retina-markers.curie.fr/amacrine,

- 21 horizontal: <u>https://retinoblastoma-retina-markers.curie.fr/horizontal</u>,
- 22 bipolar: https://retinoblastoma-retina-markers.curie.fr/bipolar



CRX

ARR3

d

OTX2



EBF3

CR)









Supplementary Fig. 5 | Data related to single-cell RNA-seq of a subtype 2 retinoblastoma (RBSC11).

- 1 a, Immunohistochemical staining of CRX (photoreceptor marker), ARR3 (late cone marker)
- 2 and EBF3 (ganglion cell marker) for a new case of retinoblastoma (RBSC11).

- **b**, Quality control of single-cell analysis for RBSC11: UMI counts, number of genes and
 percentage of mitochondrial genes.
- c, Heatmap of SingleR annotation scores derived by reference to the HPCA dataset with
 clusters superimposed for 1198 single cells in RBSC11.
- 7 d, Expression of selected retinal/neuronal genes in RBSC11 shown in 2D t-SNE plots: early
- 8 photoreceptor markers (OTX2, CRX, THRB, RXRG); late cone markers (ARR3, GUCA1C);
- 9 ganglion/neuronal cell markers (*GAP43*, *SOX11*, *UCHL1*, *DCX*, *EBF3*).
- 10 e, Expression of selected genes in the normal developing retina according to cell types (single-
- 11 cell RNA-seq data from Lu *et al.*, 2020³⁸) (left panels). For each gene and each cell type, a
- 12 pseudo-dot plot is provided. At each age (x-axis), the dot size is proportional to the percentage
- 13 of expressing cells (i.e. non-zero counts) and its y-coordinate indicates the mean expression.
- 14 Expression of these genes in the seven cell clusters of retinoblastoma RBSC11 (right panels).
- 15 For each cluster, the dot size is proportional to the percentage of expressing cells (i.e. non-
- 16 zero count) and its y-coordinate indicates the mean expression.
- 17 The selected genes are representative of the different cell clusters found in tumor RBSC11.
- 18 Clusters 0 and 2, expressed early photoreceptor/cone markers (e.g. *CRX*) and 19 neuronal/ganglion cell markers (e.g. *EBF3*, *GAP43*).
- 20 Clusters 1 and 4, expressed early photoreceptor/cone markers (e.g. *CRX*) and late cone 21 markers (e.g. *ARR3*).
- 22 Clusters 2 and 4 correspond to G2/M cells (expressing *MKI67*)
- 23 Cluster 3 corresponds to hypoxic cells (of both tumor cell populations) and expressed BNIP3.
- 24 Clusters 5 and 6 correspond to normal cells, macrophage/microglia (cluster 5 expressing
- 25 *CD14*), T-lymphocytes (cluster 6 expressing *CD3D*).
- 26 f, Co-expression of CRX/EBF3/GAP43 in the retinoblastoma sample (RBSC11) (upper panels)
- and the normal developing retina (lower panels). The bar plots represent the abundance of the
- co-expression pattern. When the number of cells displaying co-expression is not zero, the
- 29 proportion and absolute number of co-expressing cells are displayed. For the retinoblastoma
- 30 sample, cells co-expressing the three genes (*CRX/EBF3/GAP43*) are shown in 2D t-SNE plots.
- 31 The plots can be retrieved from: <u>https://retinoblastoma-retina-markers.curie.fr/coexp-ExtDat</u>.



Supplementary Fig. 6 | Expression of *TFF1* and *EBF3* in the normal developing retina and in the two subtypes of retinoblastoma.

1 a, Expression of TFF1 and EBF3 and of two cone photoreceptor genes (CRX, ARR3) in the 2 normal developing retina according to cell types (single cell RNA-seq data from Lu et al., 3 2020³⁸) (left panels). For each gene and each cell type, a pseudo-dot plot is provided. At each 4 age (x-axis), the dot size is proportional to the percentage of expressing cells (i.e. non-zero 5 counts) and its y-coordinate indicates the mean expression. TFF1 is not expressed in the normal developing retina. Expression of these genes in the seven cell clusters of 6 7 retinoblastoma RBSC11 (right panels). For each cluster, the dot size is proportional to the 8 percentage of expressing cells (i.e. non-zero count) and its y-coordinate indicates the mean 9 expression.

b, Volcano plots showing that TFF1 and EBF3 are among the most significantly upregulated
genes in subtype 2 retinoblastoma (blue) compared to subtype 1 (gold) in our series and in
two independent transcriptomic retinoblastoma datasets (Kooi et al.¹⁸, McEvoy et al.¹⁶). For
the subtype assignment of the tumors of these two series, see Methods and Supplementary
Fig. 7.
c, Boxplots representing the expression of TFF1 and EBF3 in the two subtypes using three

independent datasets (this report, Kooi et al.¹⁸, McEvoy et al.¹⁶). In the boxplots, the central mark indicates the median and the bottom and top edges of the box the 25th and 75th percentiles. Whiskers are the smaller of 1.5 times the interquartile range or the length of the 25th percentiles to the smallest data point or the 75th percentiles to the largest data point. Data

20 points outside the whiskers are outliers.



Note: a: Fisher's exact test, b Wilcoxon test, c: Chi-square test

Supplementary Fig. 7 | The two retinoblastoma subtypes were different in clinical and molecular features in two additional independent datasets.

- 1 The two retinoblastoma subtypes were identified in two additional independent retinoblastoma
- 2 datasets (Kooi et al. 2015¹⁸, McEvoy et al. 2011¹⁶) using our centroid-based transcriptomic
- 3 predictor (see Methods). For each dataset, clinical (*RB1* germline mutation status, age at
- 4 diagnosis) and molecular (stemness, E2F targets, MYC pathways, interferon responses,
- 5 estimation of abundance of various immune cells) features, and the expression of cone and

6 ganglion markers, and of TFF1 were compared between the two subtypes. The features 7 characteristic of each subtype identified in our initial series were found in these two 8 independent series. For germline mutation status and age, the statistical tests used to evaluate 9 the difference between the two subtypes are indicated in the Wilcoxon test was used to 10 evaluate the differences in HALLMARK pathway meta-scores and MCP counter-estimated 11 immune cell abundance between the two subtypes of retinoblastoma; Limma moderated t-test 12 was used for the analysis of gene expression, significance based on adjusted p values is 13 shown. log₂ fold-changes in expression between subtype 2 and subtype 1 are also shown. *p*≥0.1 (ns), *p*<0.1 (.), *p*<0.05 (*), *p*<0.01 (**), *p*<0.001 (***), *p*<0.0001 (****). 14