Spatiotemporal analysis of glioma heterogeneity reveals Col1A1 as an actionable 1 target to disrupt tumor mesenchymal differentiation, invasion and malignancy 2

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- Andrea Comba^{1,2,3}, Syed M. Faisal^{1,2,3}, Patrick J. Dunn^{1,2,3}, Anna E. Argento^{1,2}, Todd C. Hollon¹, Wajd 4
- N. Al-Holou¹, Maria Luisa Varela^{1,2,3}, Daniel B. Zamler^{1,2,3}, Gunnar L Quass⁷, Pierre F. Apostolides^{7,8}, 5
- Christine E. Brown⁹, Phillip E. Kish^{1,5}, Alon Kahana⁵, Celina G. Kleer^{3,6}, Sebastien Motsch⁴, Maria G 6

Castro^{1,2,3}, Pedro R. Lowenstein^{1,2,3,10}, *

- 7
- ¹Dept. of Neurosurgery, University of Michigan Medical School, Ann Arbor, 48109, MI, USA
- 8 9 ²Dept. of Cell and Developmental Biology, University of Michigan Medical School, Ann Arbor, 48109 MI, USA
- 10 ³Rogel Cancer Center, University of Michigan Medical School, Ann Arbor, 48109, MI, USA
- ⁴School of Mathematical and Statistical Sciences, Arizona State University, Tempe, AZ, USA 11
- 12 ⁵Ophthalmology & Visual Science, University of Michigan Medical School, Ann Arbor, 48109 MI, USA
- 13 ⁶Dept. of Pathology, University of Michigan Medical School, Ann Arbor, MI 48109, USA
- 14 ⁷Kresge Hearing Research Institute, Department of Otolaryngology-Head & Neck Surgery, University of
- 15 Michigan Medical School, Ann Arbor, MI 48109, USA
- 16 ⁸Dept. of Molecular & Integrative Physiology, University of Michigan Medical School, Ann Arbor, MI 48109, 17 USA.
- 18 ⁹Departments of Hematology & Hematopoietic Cell Transplantation and Immuno-Oncology, City of Hope, 19 Duarte, CA, USA.
- 20 ¹⁰ Dept. of Biomedical Engineering, University of Michigan, Ann Arbor, MI 48109,
- 21 *Correspondence to: pedrol@umich.edu
- 22

23 ABSTRACT

24 Intra-tumoral heterogeneity and diffuse infiltration are hallmarks of glioblastoma that challenge 25 treatment efficacy. However, the mechanisms that set up both tumor heterogeneity and invasion 26 remain poorly understood. Herein, we present a comprehensive spatiotemporal study that aligns 27 distinctive intra-tumoral histopathological structures, oncostreams, with dynamic properties and a 28 unique, actionable, spatial transcriptomic signature. Oncostreams are dynamic multicellular 29 fascicles of spindle-like and aligned cells with mesenchymal properties, detected using ex vivo 30 explants and *in vivo* intravital imaging. Their density correlates with tumor aggressiveness in 31 genetically engineered mouse glioma models, and high grade human gliomas. Oncostreams 32 facilitate the intra-tumoral distribution of tumoral and non-tumoral cells, and the invasion of the 33 normal brain. These fascicles are defined by a specific molecular signature that regulates their 34 organization and function. Oncostreams structure and function depend on overexpression of 35 COL1A1. COL1A1 is a central gene in the dynamic organization of glioma mesenchymal transformation, and a powerful regulator of glioma malignant behavior. Inhibition of COL1A1 36 37 eliminated oncostreams, reprogramed the malignant histopathological phenotype, reduced 38 expression of the mesenchymal associated genes, induced changes in the tumor microenvironment 39 and prolonged animal survival. Oncostreams represent a novel pathological marker of potential 40 value for diagnosis, prognosis, and treatment.

42 INTRODUCTION

High grade gliomas (HGG) are the most prevalent and malignant brain tumors. They grow rapidly, invade surrounding normal brain, and recur within 12 months. Median survival is 18-20 months, in spite of current standard of care^{1,2}. Despite some notable successful outcomes from the large cancer sequencing programs, which identified driver genes in a number of cancers, effective therapeutically actionable breakthroughs have not yet been identified in HGG³⁻⁷.

48 HGG are highly heterogeneous at the histological, cellular, and molecular level. Heterogeneity of 49 HGG is illustrated in addition by characteristic pathological structures such as pseudopalisades, microvascular proliferation, and areas of hypoxia and necrosis^{2,8}. The molecular characterization of 50 glioma heterogeneity identified three main molecular signatures: proneural, mesenchymal, and 51 classical^{4,9}. However, later studies demonstrated that all three transcriptomic signatures are 52 expressed within individual tumors^{5,10,11}. Rather than outright glioma subtypes, the consensus 53 54 proposes that individual tumors are enriched in particular molecular subtypes. Thus, studies have correlated histological features with genetic alterations and transcriptional expression patterns. For 55 56 example, highly aggressive histological features such as hypoxic, perinecrotic and microvascular 57 proliferative zones have been associated with the mesenchymal molecular signature and worse prognosis⁸. However, the molecular classification has only minor clinical impact. Thus, alternative 58 59 classification schemes using a pathway-based classification are currently being considered¹². How 60 these new classifications will deal with tumor heterogeneity remains to be explored. Moreover, different microenvironmental, metabolic, and therapeutic factors drive transitions of the GBM 61 62 transcriptomic signature, particularly transitions to mesenchymal states. It is important to note that 63 glioblastoma plasticity explains the high degree of tumor heterogeneity and prompts the selection of new clones at recurrence or therapy resistance¹³⁻¹⁶. It has been established that intra-tumoral 64 heterogeneity is represented by four main cellular states, the progenitor, astrocyte, 65 oligodendrocyte, and mesenchymal like-state, which represent tumor plasticity and are affected by 66 the tumor microenvironment¹⁵. 67

Tumoral mesenchymal transformation is a hallmark of gliomas^{13,17,18}. A mesenchymal phenotype is defined by cells with spindle-like, fibroblast-like morphology associated with alterations in their dynamic cellular organization leading to an increase in cell migration and invasion^{19,20}. The mesenchymal phenotype is controlled by particular transcription factors and downstream genes related to the extracellular matrix (ECM), cell adhesion, migration, and tumor angiogenesis^{18,21,22}.

However, the cellular and molecular mechanisms that regulate mesenchymal transformation in
gliomas, especially concerning the mesenchymal features of invasive cells, has remained elusive.
Integrating morphological features, spatially resolved transcriptomics, and cellular dynamics
resulting from mesenchymal transformation, growth, and invasion are thus of great relevance to
our understanding of glioma progression^{13,20}.
Cell migration is essential to continued cancer growth and invasion. Morphological and biochemical

changes that occur during mesenchymal transformation allow glioma cells to move throughout the tumor microenvironment and invade the adjacent normal brain. Tumor cells also migrate along blood vessels, white matter tracks, and the subpial surface. Within the tumor microenvironment, aligned extracellular matrix fibers help guide the movement of highly motile mesenchymal-like cancer cells ²³⁻²⁵.

Our study reveals that malignant gliomas, both high grade human gliomas and mouse glioma models, display regular distinctive anatomical multicellular fascicles of aligned and elongated, spindle-like cell. We suggest they are areas of mesenchymal transformation. For the sake of simplicity in their description throughout the manuscript, we have named these areas 'oncostreams'.

89 Using time lapse laser scanning confocal imaging ex vivo, and multiphoton microscopy in vivo of high grade glioma explants we demonstrated that oncostreams are organized collective dynamic 90 91 structures; they are present at the tumor core and at areas of tumor invasion of the normal brain. 92 Collective motion is a form of collective behavior where individual units' (cells) movement is 93 regulated by local intercellular interactions (i.e., attraction/repulsion) resulting in large scale 94 coordinated cellular migration ²⁶⁻³⁰. Collective motion plays an essential role in embryogenesis and wound healing^{27,31-33}. Emergent organized collective motion patterns could help explain so far 95 96 poorly understood tumoral behaviors such as invasion, metastasis, and especially recurrence^{27,31}.

97 Studies of tumor motility have concentrated on the behavior of glioma cells at the tumor invasive 98 border^{27,34,35}. Potential motility at the glioma core has not been studied in much detail so far. This 99 study challenges the conventional belief that cells in the central core are non-motile and indicate 100 that the glioma core displays collective migratory patterns. This would suggest that the capacity of 101 gliomas to invade and grow, results from phenomena occurring at the tumor invasive border, and 102 from the overall capacity of gliomas to organize collective motion throughout the tumor mass, from 103 the tumor core to the tumor invasive border.

104 To study the molecular mechanisms underlying oncostream organization and function we used laser 105 capture microdissection (LCM) followed by RNA-sequencing and bioinformatics analysis. We discovered that oncostreams are defined by a mesenchymal transformation signature enriched in 106 107 extracellular matrix related proteins, and which suggest that Collagen1A1 (COL1A1) is a key determinant of oncostream organization. Inhibition of COL1A1 within glioma cells lead to 108 109 oncostream loss and reshaping of the highly aggressive phenotype of HGG. These data indicate that 110 COL1A1 is likely to constitute the tumor microenvironment scaffold, and to serve to organize areas 111 of collective motion in gliomas.

112 COL1A1 has been shown previously to be a major component of the extracellular matrix in different 113 cancers, including glioma, and has been reported to promote tumor growth and invasion^{36,37}. Alternatively, some data suggest that collagen fibers could be passive barriers to resist tumor cell 114 115 infiltration or provide biophysical and biochemical support for cell migration. Some studies reported 116 that density of COL1A1 inversely correlates with glioma patient's prognosis. However, other studies 117 showed that either increased or decreased deposition of collagen could be associated with increased tumor malignancy³⁸⁻⁴⁰. Therefore, it is important to further determine the role of COL1A1 118 119 in glioma invasion and continued growth.

120 This study provides a comprehensive study of the histological, morphological, and dynamic 121 properties of glioma tumors. In addition, we uncover a novel characterization of the molecular 122 mechanisms that define intra-tumoral mesenchymal transformation in gliomas and discuss their 123 therapeutic implications. Oncostreams are anatomically and molecularly distinctive, regulate glioma 124 growth and invasion, display collective motion, and are regulated by the extracellular matrix, 125 specially by COL1A1. Inhibiting COL1A1 within glioma cells is a potential therapeutic strategy to 126 mitigate glioma mesenchymal transformation, intra-tumoral heterogeneity, and thus, reduce 127 deadly glioma invasion and continued growth.

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135 **RESULTS**

136 Intra-tumoral multicellular fascicles of elongated and aligned cells in gliomas: oncostreams

High grade gliomas (HGG) are characterized by anatomical, cellular and molecular heterogeneity which determines, in part, tumor aggressiveness and reduces treatment efficacy^{5,7,11}. Histopathological analysis of mouse and human gliomas revealed the presence of frequent distinct multicellular fascicles of elongated (spindle-like) and aligned cells (≈5-30 cells wide) distributed throughout the tumors. These structures resemble areas of mesenchymal transformation which we describe as "oncostreams" (Fig. 1A-B).

143 To study the presence and morphological characterization of oncostreams, we examined 144 histological sections from various mouse glioma models as well as human glioma specimens (Fig. 1 145 A-B). We determined the existence of oncostreams in genetically engineered mouse models 146 (GEMM) of glioma including NPA (Nras, shP53, shATRx) and NPD (Nras, shP53, PDGFβ) and other 147 implantable models (GL26) (Fig. 1A and Supplementary Fig. 1A-B). Moreover, human glioma 148 samples from primary resections and a xenograft glioma model, SJGBM2, established the presence 149 of these multicellular structures in human tissue (Fig. 1B and Supplementary Fig. 1C). 150 Morphological analysis determined that cells within histological areas corresponding to 151 oncostreams have an aspect ratio of 2.63±0.19 (elongated or spindle-like cells) compared to the surrounding tissue where cells have an aspect ratio of 1.37±0.12 (round cells), both in mouse and 152 153 human gliomas as shown in Fig. 1C and Supplementary Fig. 1D. We also determined that elongated 154 cells within oncostreams are nematically aligned with each other, whereas outside of oncostreams, 155 cell orientations are not aligned (Fig. 1D and Supplementary Fig. 1E).

156 To gain insight into the cellular features of oncostreams we asked if they are homogeneous or 157 heterogeneous multicellular structures. We observed that in GEMM of gliomas, oncostreams are 158 formed by GFP+ tumor cells, and are enriched in other tumor microenvironment cells such as 159 ACTA2+ mesenchymal cells, Iba1+ and CD68+ tumor associated microglia/macrophages cells, Nestin+ cells and GFAP+ glial derived cells (Fig. 1, E-G, and Supplementary Fig. 2 A-D). The 160 161 quantification of mesenchymal cells (ACTA2+), and tumor associated microglia/macrophages (TAM) 162 cells (CD68+ and Iba1+) showed a significant enrichment of these populations within oncostreams 163 compared to the surrounding areas (Fig. 1E-G). Moreover, non-tumoral cells within oncostreams 164 were positively aligned along the main axes of oncostreams, and with tumor cells in mouse gliomas

(Fig. 1H). This suggests that oncostreams are mesenchymal-like structures which interact with TAM and mesenchymal cells.

To test if oncostreams form along existing brain structures, we evaluated their co-localization with white matter tracts. Although, occasional positive immune-reactivity (Neurofilament-L) was present within some areas of the tumors, oncostream fascicles were not preferentially organized along brain axonal pathways (Supplementary Fig. 1F). These data indicate that oncostreams are fascicles of spindle-like aligned cells within glioma tumors, which contain tumor and non-tumor cells.

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Oncostream density positively correlates with tumor aggressiveness and poor prognosis in mouse and human gliomas

175 Oncostreams are unique histological features that contribute to intra-tumoral heterogeneity 176 suggesting a potential role in glioma progression and malignancy. To understand whether the 177 presence of oncostreams correlates with tumor aggressiveness and clinical outcomes, we generated 178 genetically engineered tumors of different malignant behaviors using the Sleeping Beauty 179 Transposon system. These models reproduce the malignant histopathological features of gliomas as demonstrated in previous studies⁴¹⁻⁴⁴. We induced tumors harboring two different genotypes: 180 181 (1) Activation of RTK/RAS/PI3K pathway, in combination with p53 and ATRX downregulation (NPA), and, (2) RTK/RAS/PI3K activation, p53 downregulation, ATRX downregulation, and mutant IDH1-182 183 R132 expression (NPAI) (Fig. 2A). IDH1-wild-type tumors (NPA) display a highly malignant 184 phenotype and worse survival prognosis (Mediam survival (MS): 70 days), compared with tumors 185 harboring the IDH1-R132R mutation, NPAI, (MS: 213 days) (Fig. 2B). This outcome reproduces 186 human disease, as patients with IDH1-mutant tumors also have prolonged median survival^{1,44,45}. 187 Tumor histopathological analysis showed a positive correlation between the density of oncostreams 188 and tumor malignancy (Fig. 2C- D). NPA (IDH1-WT) tumors exhibited larger areas of oncostreams 189 within a highly infiltrative and heterogeneous glioma characterized by abundant necrosis, microvascular proliferation, pseudopalisades and cellular heterogeneity as described before ^{43,44}. 190 191 Conversely, NPAI (IDH1-Mut) tumors display a very low density of oncostreams and a homogenous 192 histology mainly comprised of round cells, low amounts of necrosis, no microvascular proliferation, 193 absence of pseudopalisades and less invasive borders (Fig. 2C and Supplementary Fig. 5).

Further, to objectively identify and quantify tumor areas covered by oncostreams, we trained a fully convolutional neural network (fCNN) (**Supplementary Fig. 3 and 4A**). Our deep learning analysis

196 found that oncostreams occupied 15.28 ± 6.10% of the area in NPA tumors compared with only 1.18

± 0.81 % in NPAI tumors (Fig. 2C and D, and Fig. Supplementary 5A and B). Cellular alignment
 analysis validated the presence or absence of oncostreams (Fig. 2E).

199 To determine whether oncostreams are linked to glioma aggressiveness in human patients, we 200 evaluated a large cohort of TCGA glioma diagnostic tissue slides from the Genomic Data Commons 201 Portal of the National Cancer Institute. We visually examined 100 TCGA-glioblastoma multiforme 202 tissue sections (WHO Grade IV) and 120 TCGA-low grade glioma tissues (WHO Grade II and III) using 203 the portal's slide image viewer (Supplementary Table 1). Oncostreams were present in 47% of 204 TCGA-GBM grade IV tumors tissue, in 8.6 % of TCGA-LGG grade III, and were absent from TCGA-LGG 205 grade II (Fig. 3A-C and Supplementary Table 2), consistent with tumor aggressiveness (http://gliovis.bioinfo.cnio.es)⁴⁶. We then determined the presence of oncostreams across known 206 molecular subtypes of HHG (Grade IV)⁴. We found oncostream fascicles in 59.4% of Mesenchymal 207 208 (MES), 53.6% of Classical (CL) subtypes and only 26.7% of Proneural (PN) (Fig. Supplementary 6A). 209 Finally, we evaluated oncostreams presence related to IDH status and 1p 19g co-deletion in LGG 210 (Grade III). Oncostreams were present in 16.6% of IDH-WT subtype, 5% of IDHmut-non-codel and 211 absent from IDHmut-codel subtype (Fig. Supplementary 6B). These analyses suggest that 212 oncostream presence is higher in Mesenchymal and Classical subtypes and correlates with IDH-WT 213 status, and thus with a poor prognosis.

214 To validate the histological identification, we examined H&E images using our deep learning 215 algorithm (Fig. Supplementary 4B). We observed a strong concordance (>84%) between machine 216 learning and the manual histological identification of oncostreams (Table Supplementary 3). 217 Oncostream presence and their segmentation by deep learning is illustrated in Fig. 3C and 218 Supplementary Fig. 7 and 8. Additionally, alignment analysis of glioma cells confirmed the existence 219 of fascicles of elongated, mesenchymal-like cells in human gliomas (Fig. 3D). Thus, our deep learning 220 algorithm validates our histological identification of oncostreams and confirms that the density of 221 oncostream fascicles positively correlates with glioma aggressiveness.

The analysis of cellular heterogeneity showed that non-tumoral cells such as Iba1+ macrophages/microglia and GFAP+ glial derived cells were positively aligned within oncostreams tumoral cells (SOX+) in human HGG (**Fig. 3E**). Conversely, we detected that low grade gliomas (LGG) exhibited homogenous round cells, GFAP+ and Iba1+ cells throughout the tumor with no defined orientation or alignment (**Fig. 3F**).

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228 Oncostreams are defined by a distinctive spatial transcriptome signature

To determine whether oncostreams fascicles are characterized by a specific gene expression profile, we performed a spatially-resolved transcriptomic analysis using laser capture microdissection (LCM) coupled to RNA sequencing (RNA-Seq). Oncostreams were dissected according to their morphological characteristics defined above. Surrounding areas of homogenous rounded cells were selected as non-oncostreams areas (control) (**Fig. 4A**). RNA-Seq analysis detected a set of 43 differentially expressed (DE) genes; 16 genes were upregulated and 27 downregulated within oncostreams (**Fig. 4 B-C and Table Supplementary 4**).

236 Functional enrichment analysis of DE genes, performed using the I-PathwayGuide platform (Advaita 237 Corporation, MI, USA), showed that False Discovery Rate (FDR) corrected gene ontology (GOs) 238 terms were associated with migration and extracellular matrix biological process. GOs such as 239 "positive regulation of motility" "positive regulation of cell migration", "collagen catabolic 240 processes" and "extracellular matrix organization" were the most over-represented biological 241 processes (Fig. 4D and Table Supplementary 5). The upregulated DE genes within the relevant GOs 242 include: COL1A1, MMP9, MMP10, ACTA2, ADAMTS2, CDH5, CYR61, PLP1 and those downregulated 243 were ENPP2, AKAP12, BDKRB1 (Fig. 4E and Fig. Supplementary 9). Significant DE genes shared by 244 related GOs are shown in Supplementary Fig. 9. These data indicate that oncostreams can be 245 identified by a specific gene expression set and suggest a distinct role for oncostreams as intra-246 tumoral mesenchymal-like migratory assemblies within glioma tumors.

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248 **COL1A1** contributes to oncostream organization in high-grade gliomas

249 Histopathologically, oncostreams are spindle-like multicellular fascicles with a defined DE gene 250 expression signature enriched in mesenchymal genes. The GO ontology analyses suggests a central 251 role of collagen catabolic process and extracellular matrix organization in oncostreams function. To 252 understand the molecular mechanisms that regulate oncostream organization and function, we 253 identified critical genes using network analysis. Network interactions revealed that COL1A1 is a hub 254 gene, one of the most highly connected nodes, representing a potential regulator of the network's 255 signaling pathways and biological functions (Fig 5A and Fig. Supplementary 10A). We found that 256 the most relevant COL1A1 related pathways include: Focal Adhesion, Extracellular Matrix

Organization and Integrin Signaling pathways (Fig. Supplementary 10B-C and Table Supplementary
7 and 8).

259 To analyze the role of COL1A1 in oncostream organization, we analyzed COL1A1 expression by 260 immunofluorescence analysis. The COL1A1 gene encodes for the alpha-1 chain of type I collagen fibers. We observed that collagen fibers were aligned within oncostreams and overexpressed in 261 262 more aggressive NPA (IDH1-WT) gliomas compared with NPAI (IDH1-Mut) tumors. COL1A1 263 expression was significantly lower and only found surrounding blood vessels in NPAI (IDH-Mut) 264 tumors (Fig. 5B-C). Correspondingly, human GBM glioma tumors (IDH1-WT) with high oncostream densities showed prominent alignment of collagen fibers along these fascicles and higher COL1A1 265 266 expression compared to LGG (IDH1-Mut) (Fig. 5D and E).

267 Moreover, TCGA-glioma data indicate that COL1A1 has differentially higher expression in GBM 268 histological Grade IV. LGG IDH-WT tumors display higher expression of COL1A1 than IDH1-Mutant. 269 Within the GBM molecular subtype classification^{4,9}, the Mesenchymal group shows higher 270 expression of COL1A1 than the Proneural and Classical groups (Fig. Supplementary 11A); the 271 COL1A1 gene is clearly associated with the mesenchymal subtype. Analysis of patient survival 272 related to COL1A1 expression showed that mesenchymal GBM subtype displayed a significantly 273 shorter survival (MS: 10.4 months) for COL1A1 high tumors, compared to COL1A1 low (MS: 17.9 274 months) tumors. Classical and Proneural subtypes did not show survival differences associated to 275 COL1A1 expression (Fig. Supplementary 11B). Thus, oncostreams represent intra-tumoral 276 mesenchymal-like structures organized along collagen fibers.

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278 COL1A1 depletion leads to oncostream loss, tumor microenvironment (TME) remodeling and 279 increases in median survival

280 To evaluate the functional role of COL1A1 in oncostream formation we generated a COL1A1-281 deficient genetically engineered mouse glioma model. We generated COL1A1 wildtype, and COL1A1 282 knock-down tumors with different genetic backgrounds (Fig Supplementary 12A-C). COL1A1 283 downregulation increased median survival (MS) (Fig. 5F and G). The knockdown of COL1A1 in NPA 284 tumors (NPAshCOL1A1) increased survival to MS: 123 days, compared to NPA control tumors (MS: 285 68 days) (Fig. 5F). Similarly, COL1A1 knockdown in NPD tumors harboring PDGFB ligand upregulation 286 (NPDshCOL1A1), also exhibited an increased median survival (MS: 98 days) compared to the NPD 287 controls (MS: 74 days) (Fig. 5G).

288 To further analyze the effects of COL1A1 downregulation, we evaluated the histopathological 289 features of glioma tumors, quantified oncostream density using deep learning analysis and 290 evaluated COL1A1 expression within glioma tissues (Fig. 5 H-I). We observed that NPA tumors with 291 COL1A1 downregulation showed a significant reduction of COL1A1 immunoreactivity within tumors; 292 it was only maintained in small areas surrounding blood vessels (Fig. 5J-K). COL1A1 inhibition led to 293 oncostream loss and reprogramming of the histopathological tumoral characteristics as evidenced 294 by homogenous round cell morphology, resembling low grade tumors (Fig. 5J-K). Downregulation 295 of COL1A1 in NPD tumors appeared less effective, with large areas of remaining COL1A1 (Fig. %H-296 I). Nonetheless, COL1A1 was downregulated within tumor cells and oncostream dismantling was 297 significant compared to NPD control. Some oncostream areas remained associated with blood 298 vessels which displayed significant amounts of COL1A1 (Fig. 5J-K).

299 We analyzed the effect of COL1A1 depletion on the intrinsic properties of tumoral cells. In vitro

300 studies showed that COL1A1-knockdown cells exhibited a significantly decreased cell proliferation

301 and cell migration compared to controls (Supplementary Fig. 13 A-D). Also, we observed that

303 progression when compared to controls (Fig. Supplementary 13E). In vivo, genetically engineered

intracranial implantation of COL1A1-knockdown cells resulted in decreased tumor growth and

304 COL1A1 knockdown tumors displeyed decreased cell proliferation (PCNA+ cells) (Fig. 6A-B),

305 increased apoptosis via activation of Cleaved-Caspase 3, and downregulation of the anti-apoptotic

306 protein Survivin **(Supplementary Fig. 14A-C)**.

Furthermore, to determine whether COL1A1 downregulation within glioma cells modifies the glioma TME we analyzed changes in tumor associated macrophages (TAM), endothelial cells and mesenchymal cells. We found that COL1A1 knockdown tumors exhibited a decreased recruitment of CD68+ TAM (**Fig. 6C-D**), impaired CD31+ endothelial vascular proliferation (**Fig. 6E-F**) and diminished ACTA2+ perivascular mesenchymal cells (**Fig. 6G-H**). Moreover, inhibition of COL1A1 within glioma cells led to downregulation of fibronectin expression, a mesenchymal associated extracellular matrix protein (**Fig. Supplementary 14E-F**) that is associated with a more aggressive

314 phenotype.

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These preclinical animal models knocked down the expression of COL1A1 from the earliest stages of tumor development. Further, to evaluate the effects of the pharmacological degradation of deposited collagen fibers in highly malignant tumors we analyzed explants of brain tumor sections treated with collagenase. We observed that collagenase treatment decreased reticular fibers

(general collagen staining), reduced COL1A1 expression and disassemble fibers' alignment along tumoral cells and caused oncostreams depletion in a dose dependent manner (Fig. Supplementary 15A-D). These data indicate that oncostream organization and functions are regulated by COL1A1. COL1A1 knockdown within glioma cells decreased oncostream formation, reprogramed glioma mesenchymal transformation and remodeled the glioma TME, thus increasing animal survival. COL1A1 inhibition represents a novel approach for future translational development.

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326 Oncostreams' mesenchymal patterns reveal intra-tumoral collective motion in gliomas

327 GO analysis indicates that biological processes such as positive regulation of motility/migration are 328 enriched within oncostream fascicles. Overexpression of extracellular matrix (ECM)-associated 329 proteins suggest a potential role of COL1A1 fibers in regulating oncostreams' motility. To study if 330 oncostreams represent migratory structures within glioma tumors, we established a physiologically 331 viable explant brain tumor slice model containing a high density of oncostreams (Fig. 7A). The 332 movement of glioma cells expressing green fluorescent protein (GFP), within the thickness of each 333 explant, was visualized using time-lapse confocal imagining and tracked using Fiji's plug-in Track-334 Mate (Fig. 7A-C).

335 Migration analyses show complex glioma cell dynamics throughout the tumor core. The glioma 336 tumor core displays groups of cells (within particular zones) with similar nematic orientation and 337 displaying complex movement patterns (Fig. 7D and Fig. Supplementary 16A) and, which represent collective motion^{27,29-31}. Angle velocity distribution indicated the existence of three patterns of 338 339 collective motility shown schematically in Fig. 7D and F: in 'Zone A' cells don't have a preferred 340 direction, in 'Zone B' cells move in opposite directions (~ 135° and 315°), and in 'Zone C' all cells 341 move with a predominant preferred direction ($\sim 45^{\circ}$) (Fig. 7D and F). We named these patterns 342 'swarm' (Zone A), 'stream' (Zone B), or 'flock' (Zone C) (Fig. 7G). They were classified by likelihood 343 analysis: the distribution of the angle velocity is constant in a *swarm* (all angle velocity are equally 344 probable), bi-modal in a stream (cells are moving in equal but opposite directions), and uni-modal 345 in a *flock* (cells move in one direction) (Fig. 7H). These patterns were observed in all tumor slices 346 examined (Fig. Supplementary 18, 19 and 20). Average cell speeds differed among the three 347 patterns (Fig. 7E, and Supplementary 18, 19 and 20). In the tumor core, swarms moved faster and 348 without orientation, followed by directionally moving flocks and streams (Fig. Supplementary 27). 349 To determine which of these collective motion patterns match oncostream histological features, we

350 analyzed H&E sections corresponding to imaged organotypic slices (Fig. Supplementary 16B). Cells 351 within histological areas corresponding to streams and flocks have an aspect ratio of 2.2 and 2.7, 352 respectively, (spindle-like cells), while those within areas corresponding to *swarms* have an aspect 353 ratio of 1.2 (round cells) (Fig. Supplementary 16C-D). Moreover, elongated cells within streams and 354 flocks are nematically aligned with each other, whereas round cells within swarms are not (Fig. Supplementary 16E). As predicted by our *in silico* model, ⁴⁷ these results suggest that cell shape, or 355 356 eccentricity, is driving feature in the organization of collective motion patterns (Fig. Supplementary 357 **16F**). Therefore, taking into account cell shape and alignment, we define oncostreams as the 358 histological expression of collective motion patterns (streams and flocks). Notice that only the 359 dynamic analysis of collective motion can differentiate between streams and flocks. At the 360 histological level both appear as oncostreams.

361 In collective motion of flocks, interactions among individual cells are sufficient to propagate order throughout a large population of starlings⁴⁸. To define if oncostream migration patterns recall 362 363 organized collective motion behavior, we analyzed the organization of the cells by performing local 364 pair-wise correlation analysis (relative position and pair directional correlation) by tumor zones (Fig. 365 **Supplementary 17A-C**). These analyses indicate the spatial correlation of location and alignment 366 between individual cells. We observed that within *swarms* cells are more separated, as neighbors are located at 20-40 µm. Streams and flocks have higher cell density, and the nearbest neighbors 367 are closer, at 20-30 μm (Fig. Supplementary 17E and S18, S19, S20). Pair-wise directional correlation 368 369 with nearby neighbors showed that cell movement is positively correlated in all patterns at 370 distances between 10-50 μm , with higher correlation left-to-right for streams (≈ 0.2), left-to-371 right/front-to-back for *flocks* (≈0.2-0.4), and a lower correlation for *swarms* (≈0.1) (Fig. 372 Supplementary 17F and S18, S19, S20). We ascertained that tumor cells within oncostreams migrate in a directional manner ("streams $(\uparrow\downarrow)$ " and "flocks $(\uparrow\uparrow)$ "), while non-oncostream cells 373 374 move randomly without directional alignment as "swarms". Thus, our analyses strongly indicate 375 that within the tumor core of high grade glioma cells are dynamically heterogeneous and display 376 organized collective migratory behavior associated with tumor histological and genetic features.

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378 Oncostreams increase the intratumoral spread of tumoral and non-tumoral cells

Pair-wise correlation analysis showed that oncostream glioma cells are collectively organized. To

380 test the underlying nature of collective oncostream motility, we analyzed adherent junction

381 markers. Tumors with oncostreams were negative for E-cadherin, whereas N-cadherin was strongly 382 expressed (Fig. Supplementary 21A), suggesting that these fascicles move in a manner akin to 383 collective migration of mesenchymal cells of the neural crest^{33,49}. Although, no difference in N-384 cadherin were found within oncostreams and the surrounding areas, N-cadherin was elevated in 385 TCGA-GBM (Grade IV) tumors compared to TCGA-LGG (Grade III and II). High levels of N-cadherin 386 correlate with lower survival in HGG patients and mesenchymal transformation (Fig. 387 Supplementary 21B-C). 388 On the other hand, oncostream growth and motility is unlikely to be due to glioma proliferation. 389 BrdU staining showed no differences between oncostream and non-oncostream regions, and in the 390 oncostreams, the mitotic plane was always perpendicular to the main axis as expected (Fig.

391 **Supplementary 21D-E).** These results are also supported by the RNA-Seq data of dissected

392 oncostreams, where proliferation genes were not differentially expressed (**Fig. 4 A-C**).

393 Collective motion could affect the distribution of other cells within the tumor. Since oncostreams 394 are heterogeneous, we inquired about their pro-tumoral role by potentially spreading cells 395 throughout the tumor. We designed co-implantation experiments using human glioma stem cells 396 (MSP-12), and highly aggressive and oncostream-forming glioma cells (GL26) co-implanted into 397 immunosuppressed mice. Implantation of MSP-12 cells alone generated slow-growing tumors 398 (median survival of 6-8 months). At 21 days post-implantation, MSP-12 cells remained restricted to 399 the injection area with an average distance of $28.9\pm7.73 \ \mu m$ from the actual injection site. 400 Surprisingly, when MSP-12 cells were co-implanted with GL26-citrine cells, MSP-12 cells spread 401 throughout the tumor, moving along oncostreams to much longer distances (83.7±23.74 μm) from 402 the injection site (Fig. 7I-K). Cellular cytoplasmic processes from MSP-12 cells implanted alone 403 displayed a random distribution. However, in co-implanted tumors, such processes from MSP-12 404 cells are completely aligned with glioma GL26 cells within oncostreams (Fig. 7K-L and Fig. 405 Supplementary 21F). These results strongly suggest that oncostreams function as intra-tumoral 406 highways facilitating the rapid distribution of slow-moving glioma cells and/or non-tumor cells 407 throughout the tumor mass. These findings could help explain the dispersal and intratumoral 408 mixing of diverse clonal populations as demonstrated in previous studies, supporting an important 409 potential role of oncostreams in determining spatial cellular heterogeneity.

411 Dynamic interactions at the tumor border: oncostreams foster glioma aggressiveness through

412 collective invasion of the normal brain parenchyma

Furthermore, we asked whether oncostreams participate in glioma invasion. The analysis of 413 414 histological sections showed that multicellular fascicles of elongated and aligned cells are found 415 invading from the tumor border into the normal brain parenchyma (Supplementary Fig. 22A). 416 Formation of streams around blood vessels was also observed (Supplementary Fig. 22A). These 417 patterns of invasion are also detected using our deep learning methods (Fig. Supplementary 22B). 418 We then used our glioma explant model to analyze the invasion dynamics by time-lapse confocal 419 imaging at the tumor border (Fig. 8A). We implanted glioma NPA GFP+ cells into tdTomato (mT/mG) 420 mice so tumor borders could be delineated. We observed that glioma cells that extended from the 421 tumor border to the normal brain parenchyma used different dynamic patterns, moving as isolated 422 random cells and/or as collective migratory structures moving directionally, and resembled 423 oncostream structures similar to those in the tumor core (Fig. 8B-G and Fig. Supplementary 23-25). 424 To objectively distinguish between different dynamic patterns, we determined the angle velocity 425 distribution, and the likelihood that distributions corresponded to either a stream, a flock, or a 426 swarm. We found streams along the perivascular niche or invading brain parenchyma without 427 following any pre-existing brain structures, as well as cells invading as *flocks*, and *swarms* (Fig. 8E-F, 428 and Fig. Supplementary 23-25 A, C and D). Glioma cells moving along blood vessels or directly into 429 the brain as single cells is consistent with previous studies¹⁰. The correlation of position and velocity 430 supports the existence of invading collective motion structures in NPA tumors with high expression 431 of COL1A1 (Fig. S22D-E). We also determined the participation of collagen fibers in oncostreams 432 invasion. Immunofluorescence analysis on explant slices showed that collagen fibers are aligned 433 along multicellular fascicles of glioma cells invading the normal brain. These data show how collagen 434 fibers serve as scaffolds for collective tumoral cell invasion (Fig. Supplementary S26).

Our data indicate the existence of a complex framework of collective motion patterns at the glioma
border, that is consistent with previous descriptions³⁴. Although the patterns observed at the NPA
tumor border are similar to those of the tumor core, cell speed differed between the areas. Cells in
the tumor core displayed significantly lower average speeds (*stream*: 4.26; *flock*: 5.95, *swarm*: 6.27
µm/hr) compared to cells at the tumor border or those invading the normal parenchyma (*stream*:
7.95; *flock*: 7.55, *swarm*: 8.01 µm/hr) (Fig. Supplementary 27A-B).

441	Then, we asked whether the knockdown of COL1A1 in NPA gliomas affects changes in the patterns
442	of migration and invasion. Analysis of tumor cells (GFP+) at the tumor borders of GEMM of gliomas
443	comparing NPA and NPA-shCOL1A1 showed a difference in the apparent invasion patterns. The
444	analysis of tumor borders revealed an increase in the sinuosity of NPA tumors, a finding compatible
445	with NPA tumors exhibiting a higher proportion of collective invasion into the normal brain when
446	compared to NPA-shCOL1A1 tumors (Fig. 8G-I).
447	Moreover, the time lapse-confocal imaging and migration analysis of NPA-shCOL1A1 explants
448	showed that tumor cells invade the normal brain parenchyma as isolated cells (Supplementary Fig.
449	29-32). Velocity angle, velocity vector and the likelihood analysis indicated that the overall
450	distribution corresponded predominantly to <i>swarm</i> random patterns (Supplementary Fig. 29-32 D,
451	E, F). Further analysis of Relative Position Correlation and Pairwise correlation supports the
452	presence of low density of cells compatible with single cell invasion patterns in NPAshCOL1A1
453	tumors with low expression of collagen (Supplementary Fig. S29-32 G,H).
454	We conclude that oncostreams (streams and flocks) are organized collective migratory structures
455	enriched in COL1A1 that participate in the dynamic organization of the tumor microenvironment
456	within the tumor core and at the tumor invasive border of high-grade gliomas, and facilitate invasion
457	into the normal brain, impacting the malignant behavior of gliomas. Depletion of collagen1A1
458	eliminates oncostreams and their associated functions.
459	
460	Intravital imaging of glioma reveals the existence of oncostreams collective motion patterns in

461 vivo and their contribution to invasion

462 To analyze further, the previously described collective migration patterns of glioma cells we
 463 performed high resolution time lapse intravital imaging using *in vivo* two photon microscopy. NPA

464 glioma cells were intracranially implanted in the brain of tdTomato (mT/mG) mice. To visualize the

465 migration of the cells we established a cranial window above the injection site. After 7-15 days of

466 tumor development, we acquired time lapse images of the tumor core and border with a time frame

467 of 5 minutes for 8-12 hours periods (**Fig. 9A**). The movement of GFP+ glioma cells was tracked using

468 Fiji's plug-in Track-Mate (Fig. 9B-C). The analysis of cell migration *in vivo* showed that the glioma

469 tumor core exhibits organized, nematically aligned cells moving collectively. Angle velocity

470 distribution analysis determined the existence of 'stream' collective motion patterns as shown for

471 example in Fig. 9E, illustrating that nematically aligned cells are moving in opposite directions

472	(~ 65° and 250°). To corroborate the existence of 'streams' patterns we classified the movement by
473	likelihood analysis. For 'streams' the distribution of the angle velocity and velocity vectors showed
474	a bi-modal distribution (cells were moving in equal but opposite directions) same as we observed in
475	the explant models (Fig. 9F and Supplementary Fig. S33A). The analysis of cell speeds showed a
476	mean speed of 3.44 μ m/hour (Fig. 9G), in concordance with the speed found for streams in the
477	tumor core of the explant model (Fig. 7E).
478	To analyze the invasion of glioma cells, we focused the movement analysis at the tumor border (Fig
479	9G-H). We observed that cells at the tumor border displayed 'stream' collective dynamic patterns
480	(Fig. 9I, Zone A). Cells that spread from the tumor border to the normal brain (i.e., invasion) used
481	two different collective dynamic patterns: 'streams' (e.g., Zone B) and 'flocks' (e.g., Zone C) (Fig. 9I).
482	These motion patterns were determined by angle velocity distribution analysis, and likelihood
483	distribution analysis (Fig. 9K-L and Supplementary Fig. S33 B). Similar patterns were also observed
484	in further movies (Fig. S33 C-H). The collective motion patterns found <i>in vivo</i> resembled the
485	collective motions patterns described in the <i>ex vivo</i> explant model. Our results show that glioma
486	cells are organized in collective dynamic patterns at the tumor core and the tumor invasive border,
487	in tumor explants and in <i>in vivo</i> intravital models of gliomas, analyzed by two photon microscopy.
100	

488

489 **DISCUSSION**

490 Mesenchymal transformation is a hallmark of tumor heterogeneity that is associated with a more 491 aggressive phenotype and therapeutic resistance^{13,18,21}. Mesenchymal transformation involves 492 fibroblast-like morphological changes associated with active migration and gain of expression of 493 mesenchymal genes as previously described^{21,22}.

Herein we present a comprehensive study that defines the morphological, cellular, dynamic, and molecular properties of multicellular mesenchymal-like structures within gliomas. These structures are fascicles of aligned spindle-like cells found throughout the tumors and represent areas of mesenchymal transformation. We interpret these structures to be the histological expression of areas of collective motion of glioma cells. For the sake of simplicity, we have referred to these areas of mesenchymal transformation as oncostreams.

500 Oncostreams are areas of mesenchymal transformation and are identified histologically as fascicles 501 of aligned and elongated cells. When examined dynamically, we found that tumor cells move by 502 collective motion within the tumor core and at the invading border. The capacity to identify areas

503 of collective motion in histological sections has allowed us to characterize the molecular 504 organization of such dynamic structures. We thus describe the overall molecular mechanisms that 505 govern the organization and function of these structures and demonstrate the causal role of 506 individual mediators. Surprisingly, we discovered that COL1A1 is central to the structural and 507 dynamic characteristics of oncostreams. Indeed, the loss of COL1A1 expression from tumor cells 508 disrupts the structural and functional characteristics of oncostreams, resulting in a complete loss of 509 mesenchymal areas within gliomas and a reduction in glioma malignant behavior (**Fig. 10**).

510 The analysis of the gene ontologies over-represented within oncostreams indicates that 511 oncostreams denote areas enriched for "positive regulation of cell migration", and in mesenchymal 512 related genes. Interestingly, COL1A1 appeared as a central hub of oncostream organization and 513 mesenchymal transformation. We postulate that oncostreams are the histopathological expression 514 of patterns of collective motion (i.e., streams and flocks) in high grade glioma tumors. Different 515 strategies of cell migration encountered in our gliomas are reminiscent of migratory characteristics observed during embryonic development^{31,32,49}. In developmental biology, collective motion is 516 517 represented by cells moving together in clusters, sheets, streams, or other multicellular arrangements^{28,31,32}. 518

519 Our studies of oncostream dynamics at the tumor core are compatible with the results of Ralitza et 520 al^{50} . This group studied *ex-vivo* explant slices of spontaneous intestinal carcinoma, and showed that 521 cells within the tumor core were highly dynamic and display directionally correlated cell motion⁵⁰, 522 similar to our results described herein. Recent *in silico* based mathematical modelling of glioma cell 523 dynamics by our group, showed that only elongated cells, but not spherical cells, are able to form 524 organized aligned cellular structures in a cell-density dependent manner⁴⁷. Our modeling studies 525 strongly support our *in-vivo* and *ex-vivo* data described in this manuscript.

Moreover, it has been described that increased matrix cross-linking, enzymatic remodeling and parallel orientation of matrix collagen fibers stiffens tissue, modifies cell morphology and promotes cell migration and invasion^{36,38,39,51,52}. Our results support the proposal that oncostreams serve as highways to spread tumor, and non-tumor cells, throughout the tumor. Indeed, oncostream fascicles contain higher amounts of macrophages/microglia and mesenchymal cells. Dispersal of tumor and non-tumoral cells throughout the tumors could help explain the mixing of different clonal populations seen in molecular studies of high-grade gliomas¹⁰.

533 This study contributes to explaining how a particular feature of intratumoral heterogeneity, namely 534 mesenchymal transformation, affects HGG progression. Our data indicate that the density of 535 oncostreams plays a potential role in overall glioma malignant behavior in mouse and human 536 gliomas.

537 Spatially resolved transcriptional analysis using laser capture microdissection provided novel 538 insights into the molecular mechanisms that regulate oncostream functions. Oncostreams were 539 defined by a unique transcriptomic signature that matched our immunohistochemical studies. 540 COL1A1 overexpression within oncostreams was complemented with the overexpression of 541 extracellular matrix proteins such as MMP9, MMP10, ADAMTS2, which are known to remodel and 542 participate in the reorganization of collagen fibers. Oncostream fascicles were correspondingly 543 enriched in COL1A1 when assessed by immunohistochemistry.

544 Within the extracellular matrix, collagen fibers constitute a scaffold for the organization of the 545 tumor microenvironment and thus promote tumor infiltration and invasion. While collagen was 546 previously thought to be a passive barrier that could reduce tumor invasion, it has now been shown 547 that collagen fibers can serve as mechanical and biochemical tracks that facilitate cellular migration and tumor progression^{36-38,53}. Previously, multi-cancer computational analysis found that within a 548 549 mesenchymal transformation signature in different cancers including gliomas, COL1A1 was one of the top differentially expressed genes^{18,22,54}. COL1A1 is overexpressed in high grade malignant 550 gliomas and its expression levels are inversely correlated with patient survival⁵⁵ as indicated in 551 https://www.cancer.gov/tcga. In our mouse glioma models and in human gliomas, tumors with 552 553 higher density of oncostreams also express higher levels of COL1A1. COL1A1 is a consistently 554 differentially expressed gene in the glioma mesenchymal signature identified in malignant gliomas and in glioma stem cells as described in previous studies^{4,9,56}. Overall, our data are in agreement 555 556 with a recent study by Puchalski et al., which assigned genetic and transcriptional information to 557 the most common morphological hallmarks of a glioma, emphasizing the importance of integrative histo-molecular studies⁸. 558

559 Surprisingly, our data indicate a remarkable plasticity of the mesenchymal phenotype in gliomas, 560 similar to other studies^{13,15}. Genetic inhibition of COL1A1 within glioma cells depleted COL1A1 from 561 tumors, eliminated oncostream structures, reduced the glioma malignant phenotype, and 562 prolonged animal survival. Our findings are comparable with results from various studies that 563 investigated the *in-vitro* and *in-vivo* consequences of collagen depletion, inhibition of collagen cross-

564 linking or collagen synthesis inhibition on normalizing tumor ECM. In these studies, inhibition of 565 collagen led to changes in the ECM which improved drug penetration, efficacy, as well as tumor access of therapeutic nano-particles or gene based therapies⁵⁷⁻⁶¹. In addition, COL1A1 inhibition 566 567 within glioma cells induced cell intrinsic and extrinsic changes in the TME. COL1A1 inhibition not 568 only inhibits tumor cell proliferation and migration but also decreased the infiltration of 569 microglia/macrophages, endothelial cells proliferation, and perivascular mesenchymal-like cells. As 570 previously shown by other studies, glioblastomas exhibit a complex interaction between tumoral 571 and non-tumoral cells including macrophages, immune cells, endothelial cells that influence tumor 572 growth, transformation, and invasion affecting the response to treatment^{9,13,62}. However, a major 573 remodeling of the tumor mesenchymal phenotype in response to inhibition of COL1A1 has not been 574 described earlier. 575 Moreover, we found that multicellular oncostream fascicles are detected in both ex vivo and in 576 vivo glioma models, and that oncostreams facilitate tumor cell invasion thereby increasing glioma 577 aggressiveness. Our findings strongly support the importance of collective motility of cancer cells in 578 the progression of tumor growth and invasion of normal brain parenchyma, as evidenced by earlier studies of normal and pathological conditions^{27,34,35,63-67}. 579 580 In summary, our observations suggest that oncostreams are morphologically and molecularly 581 distinct structures that represent areas of collective motion that contribute to tumor growth and

invasion. These malignant dynamic structures overexpress COL1A1. COL1A1 knockdown eliminates oncostreams, reduces the mesenchymal phenotype, modifies the TME and delays tumor progression. Our findings open new paths to understanding tumor mesenchymal transformation and its therapeutic treatment. We propose that depletion of COL1A1 within oncostreams is a promising approach to reprogram mesenchymal transformation in glioma tumor as a novel therapeutic approach, and thus reduce the glioma malignant phenotype.

588

589 METHODS

590 Glioma cell lines and culture conditions:

591 Mouse glioma cells (NPA, NPD, NPAshCol1A1, NPDshCol1A1 and GL26) and human glioma cells 592 (MSP-12, SJGBM2) were maintained at 37 °C with 5% CO2 and their respective media as described 593 before⁴¹⁻⁴⁴. Mouse NPA, NPD, NPAshCol1A1, NPDshCol1A1 neurospheres were derived from 594 genetically engineered tumor using the Sleeping Beauty (SB) transposase system as previously

595 described⁴¹⁻⁴⁴. Mouse GL26 glioma cells were generated by Sugiura K and obtained from the frozen

596 stock maintained by the National Cancer Institute (Bethesda, MD)²⁵. MSP-12 human glioma cell lines

597 were provided by Christine Brown, City of Hope, and SJGBM2 human glioma cells were provided by

598 Children's Oncology Group (COG) Repository, Health Science Center, Texas Tech University.

599 Intracranial implantable syngeneic mouse gliomas:

600 Glioma tumors were generated by stereotactic intracranial implantation into the mouse striatum of 601 3.0 x 10⁴ mouse glioma cells (either, NPA, NPD or, GL26) in C57BL/6 mice, or human glioma cells in immune-deficient NSG mice (SJGBM2) as described before^{42-44,68}. To test whether oncostream 602 603 tumor cells help move other cells throughout the tumor we generated a co-implantation glioma 604 model by intracranial implantation of highly malignant GL26-citrine cells with low aggressive human 605 MSP12 glioma cells at a ratio of 1:30 (1,000 GL26-citrine cells and 30,000 MSP12 cells) in immune-606 deficient NSG mice. As controls, NSG mice were implanted with 30,000 MSP12 cells alone or 1,000 607 GL26-citrine cells alone as controls. Experiments were conducted according to the guidelines 608 approved by the Institutional Animal Care (IACUC) and Use Committee at the University of Michigan. Stereotactic implantation was performed as previously described⁴². 609

610 Genetically engineered mouse glioma models (GEMM)

We used genetically engineered mouse glioma models for survival analysis and histopathological analysis. Murine glioma tumors harboring different genetic drivers were generated using the Sleeping Beauty (SB) transposon system as described before⁴¹⁻⁴⁴. Genetic modifications were induced in postnatal day 1 (P01) male and female wild-type C57BL/6 mice (Jackson Laboratory), according to IACUC regulations. shRNA targeting the COL1A1 gene was cloned as describe in detail in Supplementary Methods.

617 Analysis of oncostreams in human glioma tissue

Oncostream presence was analyzed in unidentified H&E sections of paraformaldehyde-fixed paraffin-embedded (PFPE) human glioma samples obtained from primary surgery from the University of Michigan Medical School Hospital. To determine the presence of oncostreams in a large cohort of human glioma tissues we used the biospecimens from "The Cancer Genome Atlas Research Network" (TCGA) from the Genomic Data Commons Data Portal, National Cancer Institute, NIH (https://portal.gdc.cancer.gov). We analyzed primary Glioblastoma multiforme (TCGA-GBM) and Low-Grade Glioma (TCGA-LGG) databases. We selected cases that have available the Slide

Image and diagnostic Slides. The diagnostic slides are available for TCGA-GBM: 389 patients and
 TCGA-LGG: 491 patients. The presence of oncostreams was scored on 100 TCGA-GBM Grade IV
 tissue samples and 120 TCGA-LGG samples.

628 Cell aspect ratio and alignment analysis in H&E tumor sections

629 Images were obtained using bright-field microscopy of H&E stained paraffin sections (Olympus BX53

- 630 Upright Microscope). Tumors were imaged using 40X and 20X objectives. Images were processed
- 631 using the program ImageJ as indicated in detail in supplementary methods.

632 Deep learning analysis for oncostreams detection on H&E staining of glioma tissue

633 A fully convolutional neural network (fCNN) was trained in order to identify and segment 634 oncostreams in histologic images⁶⁹. We implemented a U-Net architecture to provide semantic segmentation of glioma specimens using deep learning⁷⁰⁻⁷². Our oncostream dataset consisted of 635 images from mouse tissues and open-source images from The Cancer Genome Atlas (TCGA). A total 636 637 of 109 hematoxylin and eosin (H&E) stained histologic mouse images and 64 from TCGA were 638 reviewed and oncostreams were manually segmented by the study authors (AC, A.E.A and P.R.L.). Images from both datasets were then augmented by randomly sampling regions within each image 639 640 to generate unique patches (~ 300 patches/image). The location and scale of each patch was 641 randomly chosen to allow for oncostream segmentation to be scale invariant. The analysis is 642 explained in further detail in Supplementary Methods.

643 Immunohistochemistry on paraffin embedded brain tumors

644 This protocol was performed as described before⁴² and as is detailed in Supplementary Methods. 645 Primary antibodies were diluted at the concentration indicated in Supplementary Table 8. Images 646 were obtained using bright-field microscopy from five independent biological replicates (Olympus BX53 Upright Microscope). Ten different fields of each section were selected at random for study 647 648 to include heterogeneous tumor areas. For immunofluorescence on paraffin embedded sections 649 from brain tumors images were acquired with a laser scanning confocal microscope (LSM 880, Axio 650 Observer, Zeiss, Germany). Integrated density was determined for the analysis of Col1a1 expression 651 using Image J. For immunohistochemistry on vibratome brain tumor sections were left in 4% 652 paraformaldehyde fixation for 48 hours and then transferred to PBS 0.1% sodium azide for an 653 additional 24 hours at 4°C. A Leica VT100S vibratome was used to obtain 50 µm coronal brain sections. The immunohistochemistry protocol was performed as previously described^{73,74}. 654

655 Laser capture microdissection (LCM) of brain tumors

656 Malignant glioma tumors were induced by intracranial implantation of dissociated NPA 657 neurospheres in C57BL/6 mice as described above. LCM approach to analyze differential mRNA 658 expression of intra-tumoral glioma heterogeneity was performed as described elsewhere⁷⁵.

659 **RNA-Sequencing and bioinformatics analysis**

660 RNA was isolated for laser microdissected tissues using the RNeasy Plus Micro Kit following the 661 manufacturer recommendations (Qiagen). Before library preparation, RNA was assessed for quality 662 using the TapeStation System (Agilent, Santa Clara, CA) using manufacturer's recommended 663 protocols. We obtained a RIN between 6 to 7 after laser microdissection of glioma tissue. A RIN of 664 6 was determined to be suitable for cDNA library preparation. 0.25 ng to 10 ng of total RNA was 665 used for cDNA library preparation using a kit suitable for RNA isolation at pico-molar concentrations 666 (MARTer Stranded Total RNA-Seq Kit v2 - Pico Input Mammalian) following manufacturer recommended protocol (Clontech/Takara Bio #635005). Sequencing was performed by the UM DNA 667 668 Sequencing Core, using the Illumina Hi-Seq platform and Bioinformatic analysis were executed by 669 the UM Bioinformatics Core. Differentially expressed genes of all tumors were used for gene 670 ontology (GO), Pathways analysis and genes analysis using iPathwayGuide (Advaita Corporation 671 2021). Network analysis of the DE genes were achieved using Cytoscape and Reactome App. 672 Network was clustered by Reactome Functional Interaction (FI). Analysis of the expression of 673 COL1A1 in normal tissue and in human gliomas were performed using the dataset of TCGA-GBM and TCGA-LGG from Gliovis (http://gliovis.bioinfo.cnio.es)⁴⁶. 674

Tumor explant brain slice culture glioma model and time-lapse confocal imaging

676 For the analysis of glioma dynamics, we generated tumors by intracranial implantation of 3×10^4 677 NPA neurospheres which were used to carry out a 3D explant slice culture glioma model. C57BL6 678 mice were used for the dynamic analyses of the tumor core and B6.129(Cg)-679 Gt(ROSA)26Sortm4(ACTB-tdTomato,-EGFP)Luo/J- transgenic mice (Jackson laboratory, STOCK 680 007676) were used for invasion analysis. Mice were euthanized at 19 days' post-implantation for 681 NPA tumors and 31 days' post-implantation for NPAshCOL1A1 tumors. Brains were removed, 682 dissected, and embedded in a 4% agarose solution and kept on ice for 5 minutes. Then, brains were 683 submerged in ice-cold and oxygenated media (DMEM High-Glucose without phenol red, GibcoTM, 684 USA) and sectioned in a Leica VT100S vibratome (Leica, Buffalo Grove, IL) set to 300 µm in the z-

685 direction. All steps were performed under sterile conditions in a BSL2 laminar flow hood. Brain tumor sections were transferred to laminin-coated Millicel Cell Culture Insert (PICM0RG50, 686 Millipore Sigma, USA). Tumor slices were maintained in D-MEM F-12 media supplemented with 25% 687 688 FBS, Penicillin-Streptomycin 10.000 U/ML at 37 °C with a 5% CO2 atmosphere for 24 hours. After 24 689 hours' media was replaced with DMEM-F12 media supplemented with B27 2%, N2 1%, Normocin 0.2 %, Penicillin-Streptomycin 10.000 U/ML and growth factors EGF and FGF 40 ng/ml. For time-690 691 lapse imaging slices were placed in an incubator chamber of a single photon laser scanning confocal 692 microscope model LSM 880 (Carl Zeiss, Jena, Germany) at 37 °C with a 5% CO2. Images were 693 obtained every ten minutes for 100-300 cycles. Following movie acquisition, sections were fixed in 694 4% paraformaldehyde (PFA) for 2 days. Fixed sections were embedded in 2% agarose for H&E and 695 immunohistochemistry analysis. Sections were processed and embedded in paraffin at the 696 University of Michigan Microscopy & Image Analysis Core Facility using a Leica ASP 300 paraffin 697 tissue processor/Tissue-Tek paraffin tissue embedding station (Leica, Buffalo Grove IL). Tumor 698 explants were used for collagenase treatment. Sections were then treated for 48 hours with 699 collagenase (C2399, MilliporeSigma, USA) at a concentration of 5, 10, or 15 units/ml or vehicle 700 control. Following treatment, sections were fixed in 4% paraformaldehyde (PFA) for 2 days.

701

702 **Cranial window implantation and two photon intravital live imaging** *in vivo*:

703 A craniotomy and cranial window implantation were performed following previously described 704 protocols by us and others^{25,76,77}. The protocol was conducted according to the guidelines approved 705 by the Institutional Animal Care (IACUC) and Use Committee at the University of Michigan. Briefly, 706 mice were anesthetized and placed in a stereotactic frame. A craniotomy of 3x3 mm size was made 707 over the right hemisphere in between bregma and lambda. 5x10⁴ GFP⁺ NPA glioma cells were 708 intracranially implanted at 0.8mm deep in the center of the craniotomy into the brain cortex of 709 B6.129(Cg)-Gt(ROSA)26Sortm4(ACTB-tdTomato-EGFP)Luo/J mice. The cranial window was created 710 with round microscope cover glass, and a metal head bar was positioned on the skull posterior to 711 the cranial window. One week post tumor cells injection and cranial window implantation, intravital 712 live imaging was performed using a two-photon microscope (Bruker Technology) with a 20X water 713 immersion objective (Olympus, NA 1.0) for 8-12 hours. The detailed methodology is available in 714 Supplementary Methods.

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- 716

717 Mathematical analysis of tumor cell movement

To determine the movement of cells in different areas of the tumor we performed localized statistical analysis in different zones of the tumor. We selected localized areas based on the organization of cells in clusters, group of cells moving together with similar distribution. Raw data of 4 movies from the tumor core and 4 movies from the tumor border were analyzed for 293 cycles (core) and 186 cycles (border) for a frame rate of Δ t = 10 min between image acquisition. To track cell motion, we used the software Fiji with the plugin TrackMate⁷⁸. Analysis was performed as indicated in detail in Supplementary Methods.

725 **Classification of glioma migration patterns**

726 To classify the collective cellular motion behavior of the three types of patterns called flock, stream, 727 and swarm illustrated in Supplementary Figure S10F we used as criteria the orientation of each cell 728 described by its unique angle velocity denoted θ i. More precisely, we transformed the Angle 729 Velocity Distribution graph into a histogram where we examined the distribution of all the values 730 θi. A schematic representation of these distributions is depicted in Figure 4G. Considering a data-731 set $\theta n = 1...N$ of orientations where N is the total number of cells, $\theta n \in [0, 2\pi]$ is the direction of the 732 cell n. We tested three types of distributions p to describe the dataset and gave a likelihood in each case as described in Supplementary Methods. The Akaike Weight (AW) indicates which pattern has 733 the highest likelihood in each experimental situation⁷⁹. 734

735 **Statistical Analysis**

All in vivo experiments were performed using independent biological replicates, as indicated in the 736 737 text and figures for each experiment. Data are shown as the mean ± SEM. Any difference was 738 considered statistically significant when p < 0.05. In experiments that included one variable, the one-way ANOVA test was used. In experiments with two independent variables, the two-way 739 740 ANOVA test was employed. A posterior Tukey's multiple comparisons test was used for mean 741 comparisons. Student's t-test was used to compare unpaired data from two samples. Survival data 742 were entered into Kaplan-Meier survival curves plots, and statistical analysis was performed using 743 the Mantel log-rank test. Median survival is expressed as MS. Significance was determined if p<0.05. 744 All analyses were conducted using GraphPad Prism (version 8.0.0) or SAS (2021 SAS Institute, Cary, 745 NC). Each statistical test used is indicated within the figure legends.

747 **Data availability:** All data associated with this study are in the paper and/or the Supplementary 748 Information. RNA-Seq date was deposited NCBI's Gene Expression Omnibus (GEO) with identifiers 749 GSE188970. Further information and requests for resources and reagents should be directed to and

750 will be fulfilled by the corresponding authors P.R. Lowenstein.

751

752 Code Availability: The analysis of oncostreams in mouse and human glioma tissue was performed using U-Net architecture to provide semantic segmentation of specimens using deep learning. 753 754 Public GitHub repository for the project code can be found at 755 https://github.com/MLNeurosurg/DeepStreams.

Analysis of glioma cells dynamics was performed using the Julia Programing Language. Link for this
 project Script and their dependencies can be found at public GitHub repository
 <u>https://github.com/smotsch/analysis_glioma</u>.

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968

969 **AUTHORS INFORMATION**

- 970 Affiliations:
- 971

972 Dept. of Neurosurgery, University of Michigan Medical School, Ann Arbor, 48109, MI, USA

- Andrea Comba, Syed M. Faisal, Patrick J. Dunn, Todd C. Hollon, Wajd N. Al-Holou, Maria Luisa Varela,
 Daniel B. Zamler, Anna E. Argento, Phillip E. Kish, Maria G Castro, Pedro R. Lowenstein
- 975
- 976 Dept. of Cell and Developmental Biology, University of Michigan Medical School, Ann Arbor,
 977 48109 MI, USA
- Andrea Comba, Syed M. Faisal, Patrick J. Dunn, Maria Luisa Varela, Daniel B. Zamler, Anna E.
 Argento, Maria G Castro, Pedro R. Lowenstein
- 980
- 981 Rogel Cancer Center, University of Michigan medical School, Ann Arbor, 48109, MI, USA
- Andrea Comba, Syed M. Faisal, Patrick J. Dunn, Maria Luisa Varela, Daniel B. Zamler, Maria G Castro,
 Pedro R. Lowenstein
- 984
- 985 School of Mathematical and Statistical Sciences, Arizona State University, Tempe, AZ, USA
- 986 Sebastien Motsch
- 987
- 988 Ophthalmology & Visual Science, University of Michigan Medical School, Ann Arbor, 48109 MI,
 989 USA
- 990 Phillip E. Kish, Alon Kahana

991

992 Dept. of Pathology, University of Michigan Medical School, Ann Arbor, MI 48109, USA

- 993 Celina G. Kleer
- 994

998

995 Kresge Hearing Research Institute, Department of Otolaryngology-Head & Neck Surgery, 996 University of Michigan medical School, Ann Arbor, MI 48109, USA

- 997 Gunnar L Quass, Pierre F. Apostolides
- 999 Dept. of Molecular & Integrative Physiology, University of Michigan medical School, Ann Arbor,
- 1000 **MI 48109, USA**
- 1001 Pierre F. Apostolides
- 1002

1003Departments of Hematology & Hematopoietic Cell Transplantation and Immuno-Oncology, City1004of Hope, Duarte, CA, USA

- 1005 Christine E. Brown
- 1006

1007 Contributions: Conception and design: A. Comba, M.G. Castro, P. R. Lowenstein. Development of 1008 methodology: A. Comba, M.S. Faisal, P. J. Dunn, A. E. Argento, T. Hollon, W.N. Al-Holou, M.L. Varela, 1009 D.B. Zamler, S. Motsch, P. R. Lowenstein. Acquisition of data, analysis, and interpretation: A. Comba, 1010 M.S. Faisal, P. J. Dunn, A. E. Argento, T. Hollon, M.L. Varela, D.B. Zamler, M.G. Castro, S. Motsch, P. 1011 R. Lowenstein. Human histopathology analysis and identification of oncostreams: A. Comba, C. 1012 Kleer, A. E. Argento, P. R Lowenstein. Laser microdissection protocol: A. Comba, P. R. Lowenstein, 1013 P. E. Kish, Alon Kahana. Development and establishment of intravital imaging using multiphoton 1014 microscopy: Comba, M.S. Faisal, P. R. Lowenstein, G. L. Quass, P. F. Apostolides. Development and 1015 experimental assistance with human glioma cell lines. development and assistance: C.E. Brown. 1016 Manuscript writing: A. Comba, S. Motsch, P. R. Lowenstein. Administrative, technical, or material 1017 support (i.e., reporting or organizing data, constructing databases): A. Comba, P. R. Lowenstein. 1018 Study supervision: M. G. Castro and P. R. Lowenstein. All authors reviewed the final version of the 1019 manuscript.

- 1020
- 1021 **Corresponding author:**
- 1022 Correspondence to Pedro R. Lowenstein
- 1023

1024 ETHICS DECLARATIONS

- 1025 **Competing interests:** All authors of this paper declare no potential conflicts of interest.
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1032 Figure Legends

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1034 Fig. 1. Oncostreams are multicellular fascicles present in mouse and human gliomas

1035 A) Representative 5 µm H&E microtome sections from gliomas showing that fascicles of spindle-like 1036 glioma cells (oncostreams, outlined by the dotted line) are present in a GEMM of gliomas NPA 1037 (NRAS/shATRx/shp53) and the GL26 intracranial implantable model of glioma. Scale bars: 50 µm. B) 1038 Representative H&E microtome sections of human glioma and human xenografts showing the 1039 presence of oncostreams. Scale bar: 20 µm. C-D) Histograms showing the cellular shape analysis 1040 (aspect ratio) (C) and angle orientation (alignment) for the corresponding images (D) shows areas of 1041 oncostreams (OS) formed by elongated and aligned cells and areas with no oncostreams (No-OS) as 1042 rounded and not-aligned cells. E-G) Immunostaining show that tumor cells, mesenchymal cells 1043 (ACTA2+), microglia/macrophages (IBA1+ and CD68+), are aligned within, the main orientation axis of oncostreams. Scale bar: 20 μm. Bar graphs show the quantification of ACTA+, n=3 (E), IBA1+, n=5 1044 1045 (F) and CD68+, n=5 (G) cells within oncostreams areas in NPA tumors. 6-13 areas of oncostreams per 1046 tumor section per animal were imaged. Error bars represent ± SEM; unpaired t-test analysis, 1047 *p<0.05, **p<0.001 . H) Angle orientation shows the alignment of ACTA+, IBA1+ and CD68+ cells 1048 within oncostreams for the corresponding images.

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1050 Fig. 2. Oncostreams density positively correlates with tumor aggressiveness in GEMM of gliomas.

1051 A) Genetic makeup of NPA and NPAI tumors. B) Kaplan–Meier survival curves of NPA and NPAI mouse 1052 gliomas show that animals bearing IDH1-R132H mutant tumors (NPAI) have prolonged median 1053 survival (MS): NPA (MS: 86 days; n: 18) versus NPAI (MS: 213 days; n:12). Log-rank (Mantel-Cox) test; 1054 ****p<0.0001. C-D) Deep learning method for oncostream detection in H&E stained mouse glioma 1055 sections: C) Representative images of oncostreams manually segmented on H&E stained sections of 1056 NPA gliomas and NPAI tumors. The output of our trained model for each image is shown below (probability heat maps), for tissues containing oncostreams (NPA), and without oncostreams (NPAI), 1057 1058 scale bar = 100 μm. D) 10-14 random fields per tumor section per animal were imaged, n=9 NPA and 1059 n=12 NPAI, and quantified using deep learning analysis. Error bars represent ± SEM; unpaired t-test 1060 analysis, *p<0.05. E) Angle histogram plots show aligned cells in NPA tumors vs non-aligned cells in 1061 NPAI tumors for the representative images showed in figure.

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1063 Fig. 3. The density of oncostreams positively correlates with tumor aggressiveness in human 1064 gliomas. A) TCGA tumors were analyzed from different grade: GBM-Grade IV (100 tumors). LGG-1065 Grade III (70 tumors) and LGG-Grade II (50 tumors). Pie charts show percentage of tumors displaying 1066 oncostreams in relation to tumor grade. Oncostreams are present in 47% of GBM grade IV tumors, 8.6 % of LGG grade III, and are absent from LGG grade II. B) Manual identification of oncostreams in 1067 1068 H&E images are shown for human gliomas with WHO grades IV, III, II from TCGA. C) Deep learning 1069 analysis for human gliomas. Our algorithm was able to detect oncostreams in grade IV and III gliomas 1070 but not in grade II gliomas. D) Angle histogram plots show the alignment of cells in H&E histology 1071 sections of Grade IV and Grade III gliomas' oncostreams and random alignment in grade II glioma 1072 sections lacking oncostreams. Angle histogram correspond to the representative images. E-F) 1073 Immuno-fluorescence staining of SOX2+ tumor cells (green), glial fibrillary acidic protein (GFAP+) 1074 cells (red), and microglia/macrophage (IBA1+) cells (red) in high-grade human glioblastoma (GBM) 1075 (WHO Grade IV), IDH-WT (E) and in low-grade glioma (LGG) (WHO Grade III), IDH-mutant (F), 1076 showing oncostreams heterogeneity and cellular alignment in human high-grade gliomas but not in 1077 low grade gliomas. Scale bars: 50 µm.

1078 Fig. 4. Oncostreams are defined by a unique gene expression signature related to mesenchymal 1079 transformation and migration. A) Schematic representation of spatial transcriptomic analysis of 1080 glioma oncostreams using Laser Capture Microdissection (LCM). Glioma tumors were generated by 1081 intracranial implantation of NPA tumor cells in C57BL6 mice (a). Oncostream areas (red outline) were 1082 identified and dissected from surrounding glioma tissue (black outline) in mouse glioma samples 1083 using a LCM microscope (b-c). B) A volcano plot displays differentially expressed (DE) genes from oncostream vs no-oncostream areas. DE genes were selected based on a fold change of \geq 1.5 and a 1084 1085 q-value (false discovery rate (FDR) corrected p-value) of ≤ 0.05 . Upregulated genes (red dots) and 1086 downregulated genes (green dots) are shown. Relevant genes related to mesenchymal migration are 1087 labeled on the graph. C) Heat map illustrates gene expression patterns for oncostream vs no-1088 oncostream areas in NPA glioma tumors (n=3 biological replicates/group). Differentially upregulated 1089 genes (16) are represented in red and downregulated genes (n=27) are represented in green (q-1090 value \leq 0.05 and fold change \geq ± 1.5). **D)** Functional enrichment analysis of overrepresented GO 1091 terms (biological processes) obtained when comparing oncostream vs no-oncostream DE genes. p-1092 value corrected for multiple comparisons using the FDR method. Cutoff FDR<0.05. Blue: 1093 Downregulated GOs. Red: upregulated GOs. E) Bar graphs show DE genes annotated to the most 1094 relevant enriched GOs biological process: "Positive regulation of cell motility", "Regulation of cell 1095 migration" and "Collagen metabolic process."

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1097 Fig. 5. COL1A1 is a central hub in oncostream organization and glioma malignancy. A) Network 1098 analysis of the DE genes comparing oncostreams versus no-oncostreams DE genes. Genes with a 1099 higher degree of connectivity are highlighted with larger nodes. Clusters of nodes with the same 1100 color illustrate modules of highly interacting of genes in the network. B) Immunofluorescence 1101 analysis of COL1A1 expression in GEMM of glioma tissues comparing NPA (IDH1-WT) vs NPAI 1102 (IDH1mut). Representative confocal images display COL1A1 expression in green (Alexa 488) and 1103 nuclei in blue (DAPI). Scale bar: 50 µm. C) Bar graphs represent COL1A1 quantification in terms of 1104 fluorescence integrated density. NPA n=5 and NPAI n=6 animals for each experimental condition 1105 were used for the analysis. Ten fields of each tumor section were selected at random. Error bars 1106 represent ±SEM. t-test, **p<0.01. D) Immunofluorescence analysis of COL1A1 expression in human 1107 GBM and LGG tumors. COL1A1 expression in green (Alexa 488) and nuclei in blue (DAPI). Scale bar: 1108 50 µm. E) Bar graphs represent COL1A1 quantification as fluorescence integrated density. 5 (LGG) 1109 and 8 (GBM) tumor samples were used for the analysis. Ten fields of each tumor section were 1110 selected at random. Error bars represent ±SEM. t-test, **p<0.01. F-G) GEMM of glioma with COL1A1 1111 inhibition. F) Kaplan–Meier survival curve comparing NPA (MS: 68 days; n: 14) vs NPAshCOL1A1 (MS: 1112 123 days; n: 28) G) Kaplan–Meier survival curve comparing NPD (MS: 74 days; n=15) versus 1113 NPDshCOL1A1 (MS: 98 days; n=17). Log-rank (Mantel-Cox) test. **** p<0.0001, **p<0.0126. H) 1114 Immunofluorescence analysis of COL1A1 expression in GEMM of glioma controls (NPA and NPD) and 1115 Col1A1 downregulation (NPAshCOL1A1 and NPDshCOL1A1). Representative confocal images of 1116 COL1A1 expression in green (Alexa 488) and nuclei in blue (DAPI). Scale bar: 50 µm. I) Bar graphs 1117 represent COL1A1 quantification in terms of fluorescence integrated density. 5-7 tumor samples for 1118 each experimental condition were used for the analysis. Ten fields of each tumor section were 1119 selected at random. Error bars represent ±SEM. t-test, **p<0.01. J) Representative images of the 1120 histopathological identification of oncostreams in H&E tissue sections comparing the COL1A1 1121 knockdown tumors with their respective controls. Scale bars: 50 µm. K) Quantitative analysis of 1122 oncostream areas using deep learning analysis. 4-12 tumor samples for each experimental condition 1123 were used for the analysis. Error bars represent \pm SEM; unpaired t-test analysis, *p<0.05.

1124 Fig. 6. Knockdown of COL1A1 within glioma cells modifies the tumor microenvironment

1125 Immunohistochemical analysis (A, C and E) of GEMM of glioma controls (NPA and NPD) and COL1A1 1126 downregulation (NPAshCOL1A1 and NPDshCOL1A1). A) Representative images of PCNA expression. Scale bar: 20 μ m. **B)** Bar graphs represent the quantification of PCNA+ cells numbers (cells/mm²) 1127 1128 using QuPath positive cell detection. Error bars represent ±SEM, (NPA: n=8, NPAshCOL1A1: n=5, NPD: n=8, NPDshCOL1A1: n=4), t-test, *p<0.05. C) Representative images of CD68 expression. Scale 1129 bar: 20 μm. **D)** Bar graphs represent CD68+ cell quantification (cells/mm²) using QuPath positive cell 1130 detection. Error bars represent ±SEM, (NPA: n=8, NPAshCOL1A1: n=5, NPD: n=6, NPDshCOL1A1: 1131 1132 n=4), t-test, *p<0.05, ns: no significant. E) Representative images of CD31 expression. Scale bar: 20 μ m. F) Bar graphs represent CD31+ cells quantification (cells/mm²) using QuPath positive cells 1133 detection. Error bars represent ±SEM, (NPA: n=8, NPAshCOL1A1: n=5, NPD: n=6, NPDshCOL1A1: 1134 1135 n=4), t-test, **p<0.01, *p<0.05. G) Immunofluorescence analysis of GEMM of glioma controls (NPA 1136 and NPD) and COL1A1 downregulation (NPAshCOL1A1 and NPDshCOL1A1). Representative images 1137 of ACTA2 expression in red (Alexa 555) and nuclei in blue (DAPI). Scale bar: 50 µm. H) Bar graphs 1138 represent ACTA2 quantification in terms of fluorescence integrated density. Error bars represent 1139 ±SEM, (NPA: n=6, NPAshCOL1A1: n=5, NPD: n=4, NPDshCOL1A1: n=3), t-test, **p<0.001, ns: no 1140 significant.

1141

1142 Fig. 7. Collective dynamics of oncostreams increase cell spreading within the tumor core. A) 1143 Experimental setup: NPA-GFP glioma cells were intracranially implanted in C57BL6 mice. Explant 1144 slice cultures of growing tumors were used for confocal time-lapse imaging of the tumor core. B) 1145 Single representative time-lapse confocal image of glioma cells within the tumor core (Movie #1). C) 1146 Tracking analysis of individual cell paths performed using the Track-Mate plugin from Image-J. D) 1147 Preferred directions of cells within three zones (A-C) superimposed onto a representative time lapse-1148 image. E) Speed distribution and mean speed $(\mu m/hr)$ in Zones A, B and C. F) Distribution of angle 1149 velocity for each zone. The Angle Velocity of each cell is denoted ϑ . The plot shows the proportion 1150 of cells moving in angle direction ϑ for each zone. **G-H)** Classification of collective motion patterns: stream, flock or swarm. The distribution is uni-modal for a flock (only one peak) and bi-model for a 1151 1152 stream (two peaks = 2 preferred angle velocity). For a *swarm*, the distribution is *flat* (no preferred 1153 angle velocity). In (G) Angle Velocity was transformed to a histogram; these data were then used to 1154 calculate the likelihood that a particular distribution of velocity angles corresponds to either a 1155 stream, flock, or swarm. The results are given in (H) for each zone. The frequency distribution of the 1156 data (shown in black) uses a non-parametric estimation (kernel density estimator). We tested three 1157 types of distributions, ρ , to describe the data-sets and give a likelihood for each case. The best fit 1158 was then determined by the Akaike weight (AW). I) Co-implantation of highly malignant GL26-citrine 1159 cells (green) and human MSP-12 glioma stem cells (ratio 1:30), and MSP-12 cells alone (control – left 1160 image). Immunohistochemistry of human nuclei (black) denote MSP-12 cells. Arrows show the 1161 distribution of MSP-12 cells within the brain or the tumor. Scale bar: 100 µm. J) Quantification of the distance of MSP-12 from the site of implantation. n=3 for control and n=5 for co-implantation (MSP-1162 1163 12+GL26). Error bars ± SEM; t-test, *p<0.05. K) Immunofluorescence images of human-nestin (red) 1164 labeling MSP-12 cells, and GL26-citrine cells. Note that MSP-12 cells have a multipolar morphological 1165 structure when alone, but a bipolar, elongated structure when aligned to GL26-citrine cells. Scale 1166 bar: 47.62 μm. L) Angle histogram plots quantify the alignment of MSP-12 within oncostreams, and 1167 the random alignment of MSP-12 cells when implanted alone (with dashed overlays of the other 1168 condition's alignment).

1169 Fig. 8. Collective invasion of COL1A1 enriched oncostreams contributes to malignant glioma 1170 behavior. A) Schematic representation of the experimental setup and location of imaging and 1171 quantification of tumor borders using td/mtTomato mice (Movie #5). B) Representative time-lapse 1172 scanning confocal image of glioma cells at the tumor border. This image was taken from border 1173 movie #1 and shows the subdivision into different zones. C) Preferred direction of cells within 1174 different zones superimposed onto a representative time lapse-image. D) Histogram of speed distribution and mean speed ($\mu m/h$) of Zones A, B, C and D. E) Angle Velocity distribution analysis 1175 1176 (ϑ) performed by zones. Plot shows overall direction and magnitude of cell movement. F) Likelihood 1177 analysis of the dynamic patterns at the tumor border. Graph of density estimation ρ flock (red), ρ 1178 stream (yellow) and p swarm (blue). The estimation of the black line (data) uses a non-parametric 1179 estimation. AW: 0 or AW:1. G) Immunofluorescence analysis of GFP expression in GEMM of glioma 1180 controls (NPA), and NPAshCOL1A1. Representative confocal images of the tumor borders. GFP 1181 expression is shown in green (Alexa 488) and nuclei in blue (DAPI). Dotted lines show tumor borders. 1182 Stars show tumor cell invasion patterns. Scale bar: 50 µm. H) The analysis of tumor borders was 1183 determined using the Allen-Cahn equation. Images were split into two values (-1 and +1)1184 representing the inside and outside of the tumor to analyze the sinuosity of the borders. Illustration 1185 of the sinuosity of a curve: it is defined as the ratio between the length of the curve L and the 1186 distance between the two extreme points. The sinuosity is close to 1 for a straight line. I) Sinuosity 1187 of the border for all experiments. 4-10 images of each tumor border were obtained. NPA: n=6 and 1188 NPAshCOL1A1: n=5 tumors for each experimental condition were used for the analysis. We detected 1189 a decrease of the sinuosity in COL1A1 knockdown tumors. t-test unequal variance, *p=0.0297.

1190

1191 Fig. 9. Intravital two-photon imaging reveals the collective patterns of glioma dynamics 1192 oncostream dynamics *in vivo*.

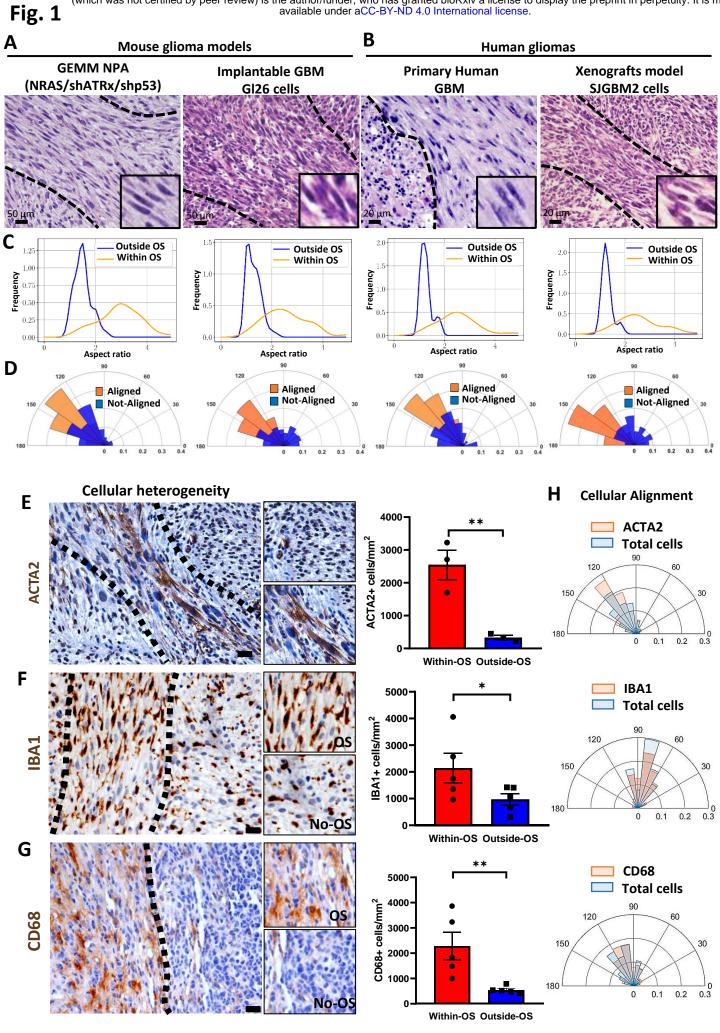
1193 A) Representative picture of the cranial window showing metallic head-bar positioned on the skull 1194 posterior to the cranial window and affixed using dental cement to stabilize the imaging plane and 1195 minimize motion artifacts during the time-lapse imaging. B) Single representative time-lapse two-1196 photon image of glioma cells within the tumor core in vivo (Movie #13). C) Tracking analysis of 1197 individual cell paths of the *in vivo* time-lapse. **D)** Speed distribution and mean speed in μ m/hr 1198 considering as a single zone named Zone A. E) Angle Velocity distribution for the *in vivo* time-lapse. 1199 The Angle Velocity of each cell is denoted θ . Plot shows overall direction and magnitude of cell 1200 movement in Zone A. F) Likelihood analysis of the dynamic patterns in *in vivo* intravital imaging. 1201 Graph of density estimation ρ flock (red), ρ stream (yellow) and ρ swarm (blue). The estimation of 1202 the black line (data) uses a non-parametric estimation. AW: 0 or AW:1. G) Single representative time-1203 lapse two-photon image of glioma cells invasion in the brain cortex in vivo (Movie #14). H) Individual 1204 cell paths trajectories of the in vivo time-lapse experiment. I) Preferred direction of cells within 1205 three zones (A-C) superimposed onto a representative in vivo time-lapse image determined by 1206 likelihood analysis. J) Speed distribution and mean speed (μ m/hr) in Zone A, Zone B, and Zone 1207 C. K) Angle Velocity distribution for the invasion *in vivo* time-lapse for each zone. The Angle Velocity 1208 of each cell is denoted θ . The plot shows the proportion of cells moving in the angle direction θ for 1209 each zone. L) Likelihood analysis of the dynamic patterns in *Movie #14 by* intravital imaging. Graph 1210 of density estimation ρ flock (red), ρ stream (yellow) and ρ swarm (blue). The estimation of the black 1211 line (data) uses a non-parametric estimation. AW: 0 or AW:1.

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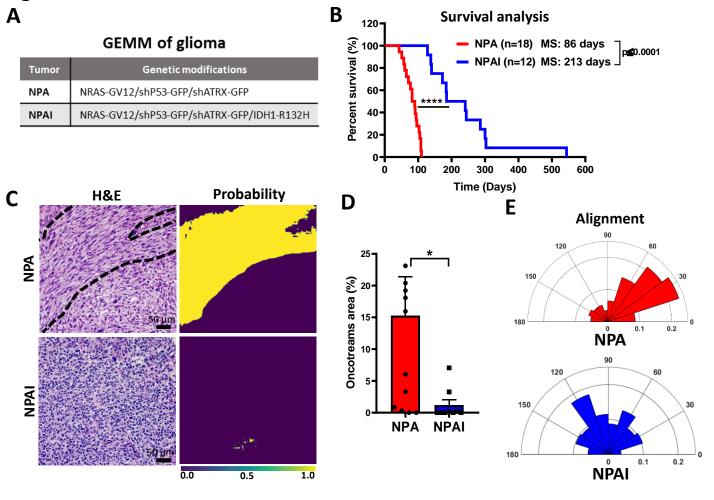
1214 Fig. 10. Oncostreams are COL1A1-rich multicellular dynamic mesenchymal structures that regulate 1215 glioma invasion and malignancy. Summary representation of mesenchymal dynamic fascicles 1216 (oncostreams) present in high grade gliomas. Our study reveals that oncostreams display directional 1217 collective motility patterns including streams and flocks. Non-directional collective motion (swarms) 1218 are represented by round cells that move do not have a preferred direction of motion. Directional 1219 dynamic patterns function as tumoral highways to facilitate the intra-tumoral spread of cells and 1220 participate in local invasion of normal brain. Oncostreams are areas of mesenchymal transformation 1221 defined by a molecular signature enriched in COL1A1. COL1A1 knockdown disrupts oncostream 1222 organization, decreases intratumoral heterogeneity and significantly increases animal survival. Our 1223 study reveals that oncostreams are anatomically and molecularly distinctive, are areas of 1224 mesenchymal transformation organized through interactions with the COL1A1 matrix, move by 1225 collective motion, and regulate glioma growth and invasion.

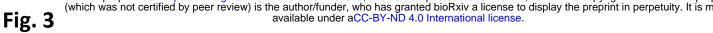
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- 1228 SUPPLEMENTARY INFORMATION
- 1229 Supplementary Material and Methods
- 1230 Supplementary Figures
- 1231 Supplementary Tables
- 1232 Supplementary Videos

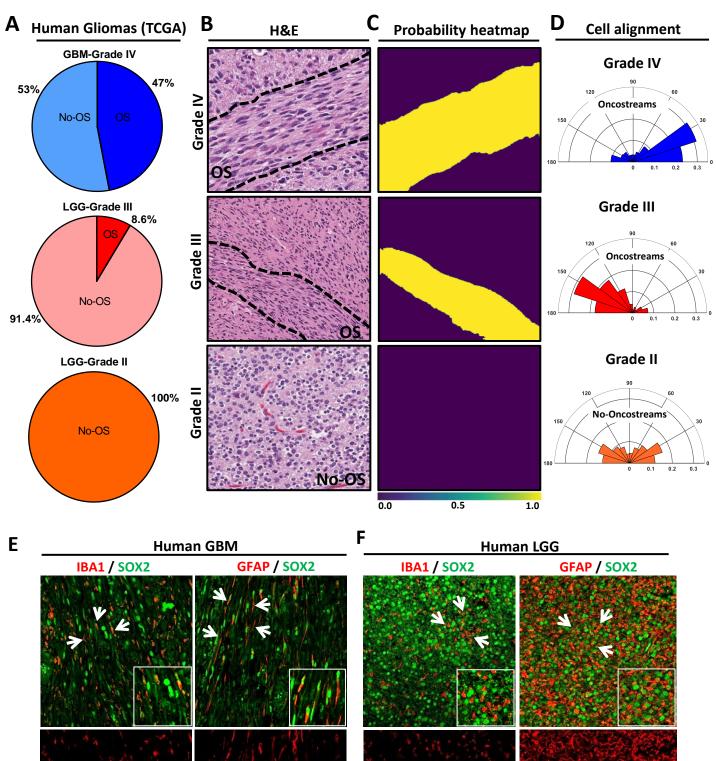
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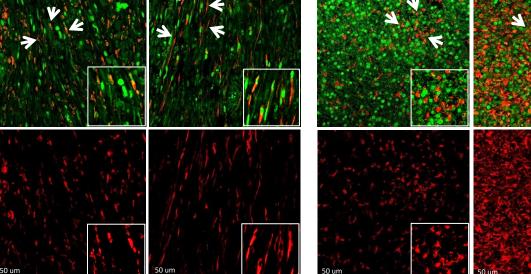


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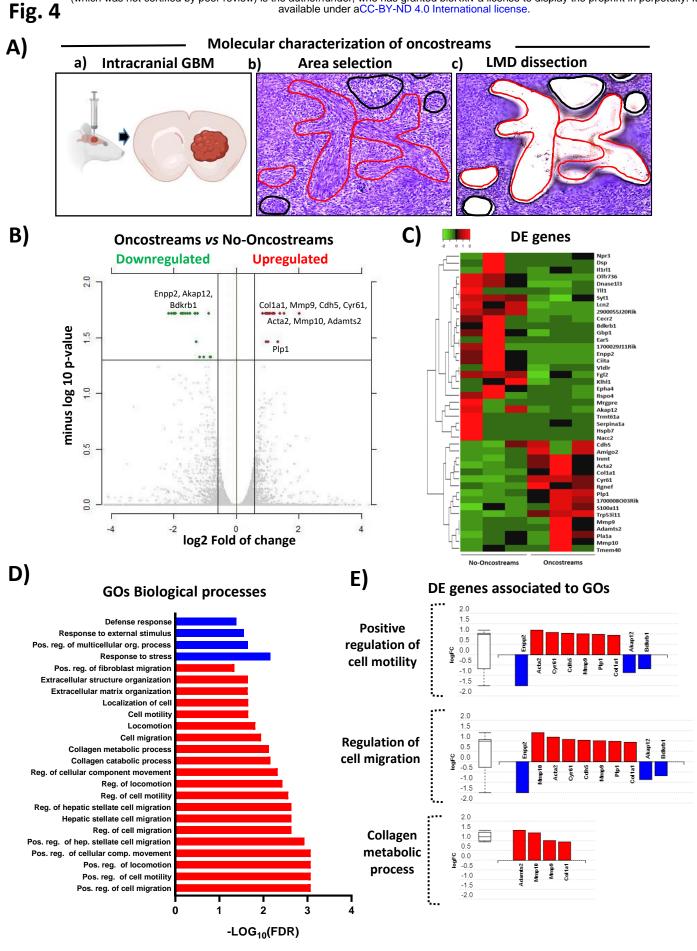


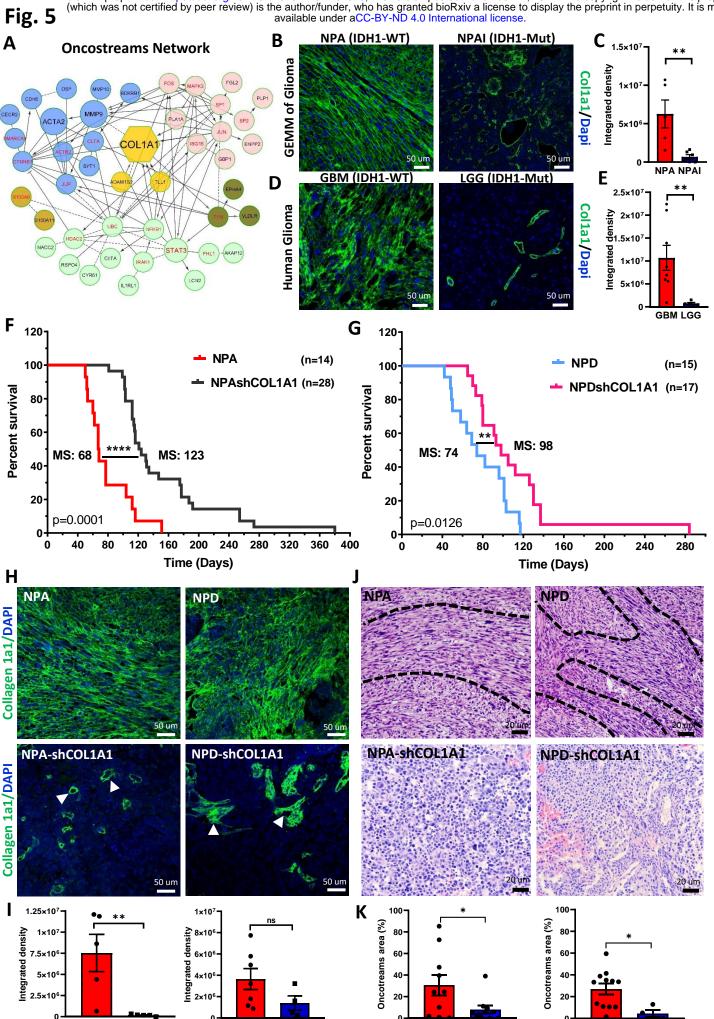






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NPA NPAshCOL1A1

0 NPD NPDshCOL1A1



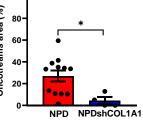
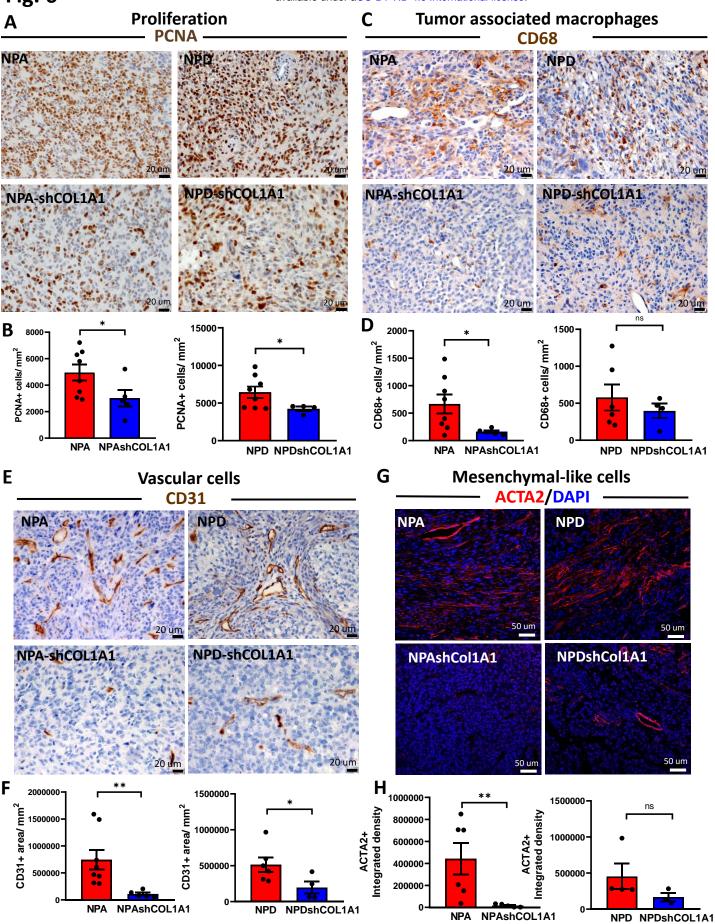
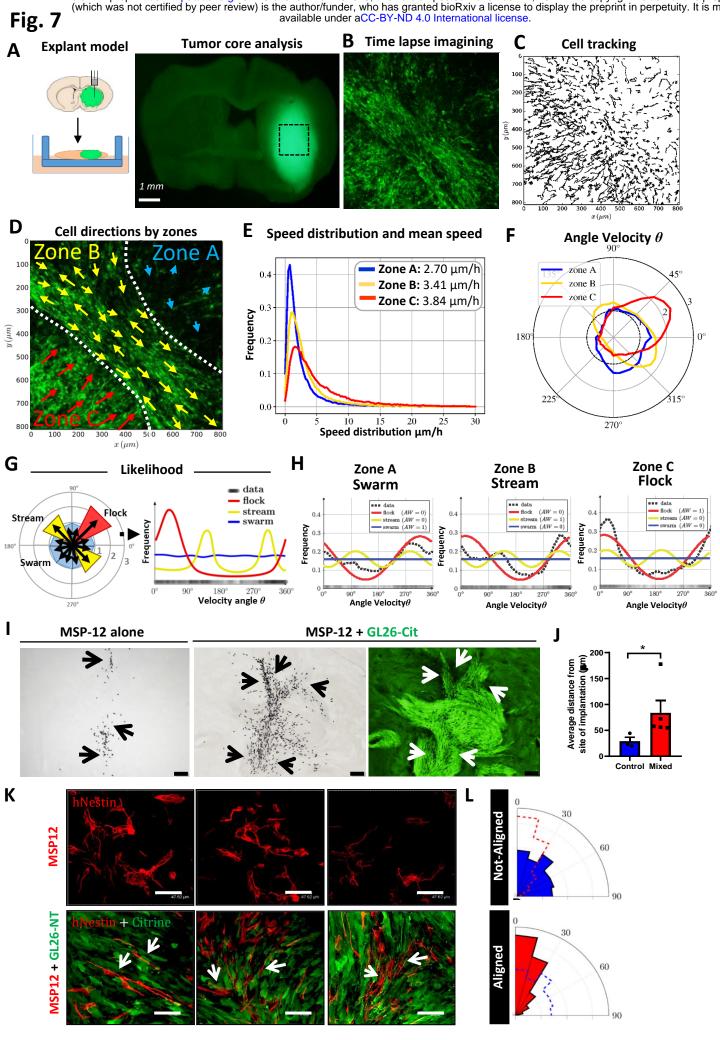
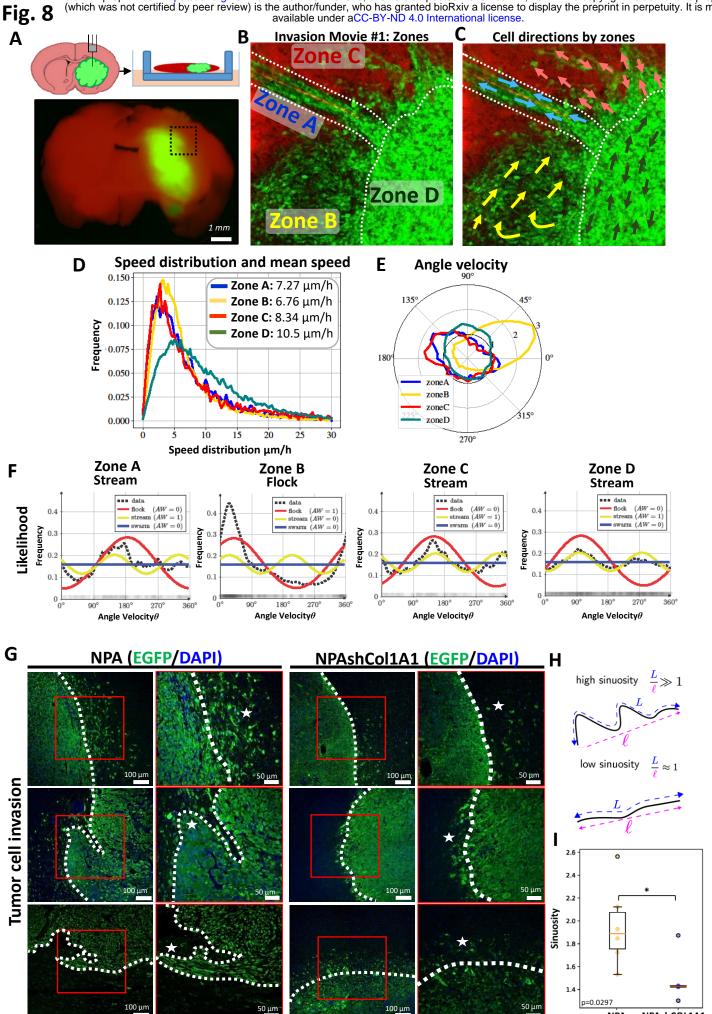


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NPA NPAshCOL1A1

