Spatiotemporal orchestration of multicellular transcriptional programs and communications in the early stage of spinal cord injury

Zeqing Wang[1,5,6], Zhuxia Li[2,5,6], Tianle Luan[1,5], Guizhong Cui[4], Shunpan Shu[1,5], Yiyao Liang[3], Jingshu Xiao[4], Kai Zhang[1,5], Wei Yu[3], Jihong Cui[1], Ang Li[3,4,*], Guangdun Peng[2,4,5,*], and Yanshan Fang[1,5,*]

[1]Interdisciplinary Research Center on Biology and Chemistry, Shanghai Institute of Organic Chemistry, Chinese Academy of Sciences, Shanghai, 201210, China

[2]Center for Cell Lineage and Development, CAS Key Laboratory of Regenerative Biology, Guangdong Provincial Key Laboratory of Stem Cell and Regenerative Medicine, GIBH-HKU Guangdong-Hong Kong Stem Cell and Regenerative Medicine Research Centre, Guangzhou Institutes of Biomedicine and Health, Chinese Academy of Sciences, Guangzhou, 510530, China

[3]Guangdong-Hong Kong-Macau Institute of CNS Regeneration, Key Laboratory of CNS Regeneration of Ministry of Education, Jinan University, Guangzhou, 510632, China

[4]Bioland Laboratory (Guangzhou Regenerative Medicine and Health Guangdong Laboratory), Guangzhou, 510005, China

[5]University of Chinese Academy of Sciences, Beijing, 100049, China

[6]These authors contributed equally to this work

*Co-corresponding authors:

 Ang Li: anglijnu@jnu.edu.cn, ORCID: 0000-0002-9886-4880

 Guangdun Peng: peng_guangdun@gibh.ac.cn, ORCID: 0000-0002-8586-0637

 Yanshan Fang: fangys@sioc.ac.cn, ORCID: 0000-0002-4123-0174

Running title: Spatial transcriptomics of SCI

Keywords: RNA-seq; spatial transcriptomics; spinal cord injury; neuroimmune; neuroinflammation; astrocytes.

Word counts: ~7,200 words (*excluding the title page, methods, acknowledgements and references*)

ABSTRACT (*≤200 words*):

While spinal cord injury (SCI) involves a complex cascade of cellular and pathological changes

that last for months to years, the most dramatic and comprehensive molecular rewiring and

multicellular re-organization occur in the first few days, which determine the overall

progression and prognosis of SCI, yet remain poorly understood. Here, we resolved the

spatiotemporal architecture of multicellular gene expression in a mouse model of acute SCI,

and revealed the coordinated gene co-expression networks, the upstream regulatory

programs, and *in situ* cell-cell interactions that underlay the anatomic disorganization as well

as the immune and inflammatory responses conferring the secondary injury. The spatial

transcriptomic analysis highlights that the genes and cell types in the white matter (WM) play

a more active and predominant role in the early stage of SCI. In particular, we identified a

distinct population of WM-originated, *Igfbp2*-expressing reactive astrocytes, which migrated

to the grey matter and expressed multiple axon/synapse-supporting molecules that may foster

neuron survival and spinal cord recovery in the acute phase. Together, our dataset and

analyses not only showcase the spatially-defined molecular features endowing the cell

(sub)types with new biological significance but also provide a molecular atlas for disentangling

the spatiotemporal organization of the mammalian SCI and advancing the injury management.

MAIN (*≤2,500 words*):

Spinal cord injury (SCI) causes neuronal death, synapse loss, axon retraction, demyelination,

inflammation and immune activation, macrophage and immune cell infiltration, gliosis and so

forth, which may result in permanent motor, sensory and autonomic dysfunctions

accompanied by various local or systemic complications. Both the primary injury and the

secondary injury (caused by the initial injury-induced responses) contribute to the overall

pathology and disease progression[1-4]. While the chronic phase of SCI can last for months to

years, the most dramatic and intensive changes in gene expression and molecular

programming take place in the immediate phase (within hours) and the acute phase (a couple

of days), followed by the subacute phase (the first two weeks) when secondary injury is

triggered and glial scar forms, which leads to further damage to the spinal cord and sets a

barrier to axonal regeneration in the chronic phase[1,5,6]. Thus, dissecting the multicellular

transcriptional programs and cell-cell communications in the early stage of SCI, in particular

the immediate and the acute phases that foretell the injury site re-organization and the overall

prognosis, is instrumental to the understanding of SCI pathology and development of new

therapeutic strategies.

Significant advancements and valuable insights into cell type-specific regulatory

programs of gene expression in SCI have been achieved with the advent of single-cell (sc)

and single-nucleus (sn) RNA-sequencing (RNA-seq)[7-9]. However, a spatially-resolved view of

the changes in the transcriptional landscape and the molecular anatomy of the different spinal

cell types as well as their context-dependent injury responses are largely missing, especially

for the "enigmatic" early stage of SCI. The recent innovation in spatial transcriptomics, which

allows high-throughput measurement of gene expression while retaining the anatomic

information[10,11], has been applied to build cell atlases and characterize different biological

processes, including normal tissue and organ development, pathological and

neurodegenerative diseases[12-21]. Given its value in molecular profiling of physically intact

tissue samples in chronic conditions, we made an attempt to employ spatial transcriptomics

to investigate traumatically damaged mouse spinal cords in acute injury, aiming to gain a

holistic understanding of the intertangled early-phase transcriptional programs and regulatory

networks that determine and underlie the SCI-induced molecular and cellular changes in their

morphological context.

The mouse SCI model and the workflow of the spatial transcriptomic analyses

For the spatial transcriptomic analysis of mammalian SCI in this study, the spinal cord of

C57BL/6J mice was fully transected at the tenth thoracic vertebra (T10) (Fig. 1a and Extended

Data Fig. 1a). The motor function and the morphology of the spinal cord after injury were

examined at 0 (uninjured), 3, 24 and 72 hours post injury (hpi) to confirm a complete

transection (Extended Data Fig. 1b-d). We performed hematoxylin and eosin (H&E) staining

on transverse sections of various distances from the injury epicenter on both the rostral (-)

and caudal (+) segments (Extended Data Fig. 1e). As shown in Extended Data Fig. 1d and e,

the spinal cords were severely deformed within ±0.25 mm, precluding an informative spatial

transcriptomic analysis. Therefore, the rostral (-) and the caudal (+) sections at ±0.5 and ±1.0

mm from the injury epicenter were subjected to the 10x Visium spatial RNA-seq analysis. Four

consecutive cryosections each at ±0.5 and ±1.0 mm were captured, and the spinal sections

of the same distance at different time points (0, 3, 24 and 72 hpi) were attached to the same

Visium Gene Expression slide and processed simultaneously in the subsequent procedure

(Fig. 1b). Together, 64 spinal cord sections from 9 different mice were examined in this spatial

transcriptomic analysis, representing an inclusive coverage of the dynamic injury responses

in a high spatial and temporal resolution.

Overall, 22,820 spatial spots were obtained and profiled after filtering, with a mean of

12,570 RNA reads and 3,996 genes per spot (Extended Data Fig. 2a-d and Supplemental

Table 1). The discernible grey matter (GM) regions containing condensed neuronal cells

showed higher numbers of expressed genes (Extended Data Fig. 2d). We compared the intra-

and inter-correlations of the sample groups and found a high data reproducibility of the

replicate sections within each group at the bulk level (Extended Data Fig. 2e). We therefore

analyzed all the sampled spots of the four replicates but visualized the results on each

representative sample section in all the spatial view charts in this study.

Further spatiotemporal evaluation of the expression of known injury-related genes (e.g.,

Atf3, *Hmox1* and *Timp1*), aka, the injury score (see Methods), displayed remarkable patterns

in association with the injury time, the distance to injury epicenter and the anatomy of the

spinal cord (Extended Data Fig. 2f, g). For example, the ±0.5 mm spinal sections at 72 hpi

displayed the highest injury scores (Extended Data Fig. 2f, g) and the lowest transcriptomic

similarity to others (Extended Data Fig. 2e), which were in accordance with the most severe

damage among all the 64 sections (Extended Data Fig. 1e). Therefore, in the subsequent data

analyses, we focused on revealing: (1) the spatiotemporal molecular annotation of the spinal

cord anatomy, (2) the gene co-expression modules and regulatory networks, and (3) the spinal

cell type composition and cell-cell communications in response to SCI (Fig. 1c). For readers

with broader research interests, an interactive data exploration portal is available at: *(will*

release to the public upon acceptance of the paper).

Cluster analysis of the spatial spots reveals anatomic domain- and injury time-featured

expression profiles of SCI

We first performed dimension reduction and cluster visualization on the collective spatial spots

using uniform manifold approximation and projection (UMAP) (see Methods). Seven distinct

gene expression clusters were identified (Fig. 1d) and Clusters 1, 2, 6 and 7 showed a bona

fide anatomic and molecular annotation of the intact spinal cord at 0 and 3 hpi (Fig. 1e). This

allowed us to categorize the seven clusters into four spatial domains according to the four

major anatomic regions of the spinal cord: white matter (WM), middle grey (MG), dorsal horn

(DH) and ventral horn (VH) (Fig. 1f, g). With the injury score increased along the time

(Extended Data Fig. 2f, g), the physical boundary of the anatomic regions became less clear

at 72 hpi especially with the ±0.5 mm sections (Fig. 1e). However, the cluster analysis could

still classify the spatial spots to the four distinct spatial domains following the coherence in the

UMAP plot (Fig. 1g), which showcased the advantage of spatial transcriptomics in providing

the molecular annotation of tissue architecture to better understand the SCI pathology.

Time after injury, along with the anatomic domain, comprised the major transcriptomic

differences amongst the spatial spots of the spinal cord examined in this study (Fig. 1h). In

contrast, the rostral/caudal directionality and the distance to the injury site showed mild

impacts on distinguishing the clusters (Fig. 1i, j). Consistently, analysis of differentially

expressed genes (DEGs) revealed remarkable domain-dependent alterations, whereas only

a small number of directionality- or distance-specific DEGs were identified during SCI (except

for the ±0.5 mm sections at 72 hpi, where more severe damage was associated) (Extended

Data Fig. 3a-c). As such, we focused on the factors of spatial domain and injury time in the

subsequent analyses (also see Extended Data Fig. 3d-g).

A remarkable shift of the compositions of spatiotemporal clusters in the WM and MG

domains was elicited by SCI (Fig. 1k). To understand the molecular changes and functional

relevance of these changes, we identified the cluster-enriched genes (CEGs) for each cluster

and analyzed their spatiotemporal expression profiles (Fig. 1l). The CEGs included not only

the well-known markers of the four anatomic regions of the intact spinal cord, such as *Mbp*

(Cluster 1) for WM, *Snap25* (Cluster 2) for MG, *Tac2* (Cluster 6) for DH and *Prph* (Cluster 7)

for VH, with their spatial expression validated by *in situ* hybridization (ISH) (Fig. 1m), but also

uncharted signature genes for the different spatial domains especially after injury (Extended

Data Fig. 4). Gene ontology (GO) term enrichment analyses of the CEGs revealed the

biological processes that not only reflect the unique physiological function of the different

anatomic regions, but also underlie the pathological progression of SCI. For example, Cluster

1 was predominantly localized to the WM at 0 and 3 hpi (Fig. 1k), and its CEGs, such as the

myelin-related genes *Plp1*, *Mog* and *Mbp*[22], were enriched in the biological functions of

myelination and oligodendrocyte differentiation (Fig. 1l). At 24 and 72 hpi, the main clusters of

the WM switched to Clusters 4 and 5 (Fig. 1k), whose CEGs, such as the reactive astrocyte

markers *Vim* and *Tm4sf1*[23] in Cluster 4 and the activated macrophage markers *Lyz2* and

Lgals3[24] in Cluster 5, were enriched with the terms of protein synthesis, cell migration,

responses to wounding, *etc*. (Fig. 1l). Thus, the switch of the spatiotemporal clusters in the

WM between 3 and 24 hpi suggested a reduction of myelination-related functions along with

the activation of astrocytes and infiltration of myeloid cells in the WM regions between the

immediate phase and the acute phase of SCI.

Spatiotemporal analysis of gene co-expression patterns and transcriptional programs

in SCI

Next, to depict the intricate and dynamic transcriptional changes in SCI and unveil the

commonalities in the molecular events and regulations, we performed the weighted gene co-

expression network analysis (WGCNA)[25] (Fig. 2a). Thirty-three spatially and temporally co-

expressed gene modules (CGMs) were identified (see Methods and Supplemental Table 2),

which displayed unique and diverse spatiotemporal expression patterns, cellular functions,

gene connectivity networks and transcriptional regulations (Extended Data Fig. 5 and 6),

together providing an unbiased molecular annotation of the injury-induced expression

programs in SCI. For example, CGM13 showed an injury-induced expression in the WM

specifically at 3 hpi, after which the expression was sharply decreased (Fig. 2b). This module

was highly associated with the cellular functions of regulating the ERK1/2 signaling cascade,

and the node genes of the connectivity network were *Csrnp1* and *Maff*, which were known

players in stress, immune and inflammatory responses[26,27]. Thus, CGM13 likely represented

a group of MAPK/ERK-mediated "early response genes" regulating immunity and

inflammation in the acute phase of SCI.

In addition, we noticed that some modules, such as CGM9 and CGM11, exhibited an

interesting spatiotemporally-complementary expression pattern: CGM9 was expressed at a

low level in the WM of the intact spinal cord (0 hpi) and at 3 hpi, but was robustly increased at

24 hpi, which expanded into the MG in the ±0.5 mm sections at 72 hpi (Fig. 2c). In contrast,

CGM11 was highly expressed in the MG at 0 and 3 hpi, which started to decrease at 24 hpi

and only retained weak expression at 72 hpi, with the expression in the ±0.5 mm sections

reduced the most (Fig. 2d). Scrutinizing the biological functions and the connectivity network

of the genes enriched in these two CGMs, we found that CGM9 was mainly involved in stress,

wound and immune responses, whereas CGM11 was mostly associated with the function and

maintenance of neuronal synapse. Thus, this complementary expression pattern of the two

CGMs likely underscored the interplay of SCI-induced immunity and inflammation with the

loss of neurons and synapses during SCI.

To further unravel the regulatory mechanisms that determine the spatiotemporal changes

in gene expression in SCI, we performed single-cell regulatory network inference and

clustering (SCENIC) analysis[28] to identify transcriptional regulons (consisting of the

transcription factors (TFs) and their target genes) that could coordinately regulate gene

expression in response to SCI. Overall, 147 regulons were identified and classified into seven

regulon modules (RMs) (Fig. 2e and Supplemental Table 3). The RMs showed remarkable

anatomic domain- and/or injury time-specificity (Fig. 2f and Extended Data Fig. 7). For

example, RM4 and RM7 functioned predominantly in the DH and the VH, respectively; RM6

was enriched in the GM regions including MG, DH and VH and its activity was gradually

reduced along the time in SCI; RM5 was highly expressed in the DH and the WM at 3 hpi,

which may be associated with the induction of the early response genes.

The RMs of the WM were more complicated and dynamic: RM2 was the main RM at 0

and 3 hpi, and it was superseded by RM3 at 3 and 24 hpi and finally by RM1 at 72 hpi (Fig.

2f, g and Extended Data Fig. 7a, b). Along with these changes, the TFs of the main regulons

in the WM switched from Sox10, Oligo2 and Srebf2 (RM2), to Fos, Jun and Nfkb1/2 (RM3),

and ultimately to Stat3, Stat6 and Irf5 (RM1) (Fig. 2h). Sox10 and Oligo2 are important TFs in

oligodendrocyte differentiation and maturation[29] and Srebf2 regulates genes involved in

cholesterol biosynthesis[30]; Fos and Jun are key TFs in TGFβ-mediated signaling and regulate

cell proliferation, differentiation and death[31] and Nfkb1/2 is a central activator of genes related

to inflammation and immune functions[32]; Stat3, Stat6 and Irf5 mediate gene expression in

response to cell stimuli and are widely involved in immunity, inflammation and cell death[33].

Together, the shifts of the main RMs in the WM (Fig. 2f) were in agreement with the changes

of the spatiotemporal clusters in the WM during SCI (Fig. 1k), which underscored the injury-

induced oligodendrocyte dysfunction and immune and inflammation activation in the WM. Of

note, the immune and inflammatory responses mediated by RM3 at 3 and 24 hpi were largely

restricted to the WM, whereas those by RM1 at 72 hpi were comparably activated in all the

four anatomic domains (Fig. 2f and Extended Data Fig. 7a), suggesting a widespread

upregulation of immune and inflammatory genes by the end of the acute phase of SCI.

Integration of the SCI scRNA-seq data for an *in situ* multicellular gene expression and

cell communication atlas

Because the Visium spot is not single-cell resolution, to allocate the cell type information to

the SCI spatial spots in a comparable manner, we performed deconvolution analysis by robust

cell type decomposition (RCTD[34]; also see Methods) using the combined normal and injured

spinal cord scRNA-seq datasets as reference[9,35]. The proportions of different cell types

detected within a single spatial spot were displayed as a pie chart and the entire spinal cord

section was displayed as an assembly of the spatial pie charts (Fig. 3a, b). The overall

composition of the main cell types and their anatomic localizations were in good accordance

with the spatial architecture of the spinal cord. For example, neurons and oligodendrocytes

were predominantly localized to the GM and the WM, respectively (Fig. 3b). Interestingly, we

found that the two molecularly-defined astrocyte subtypes, astrocyte-*Gfap* and astrocyte-

Slc7a10[35] (Supplemental Table 4), showed distinct and mutually exclusive distributions in the

white (namely, astrocyte-WM) and grey matter (namely, astrocyte-GM), respectively (Fig. 3b),

which represented the general categories of astrocyte heterogeneity in location of white

(fibrous astrocyte) versus grey matter (protoplasmic astrocyte)[36]. Furthermore, deciphering *in*

situ spinal cell type compositions and their spatiotemporal dynamics during SCI revealed

sophisticated spatial domain- and injury time-associated alterations (Extended Data Fig. 8),

including the emergence of the cell types that participate in the scar formation such as

fibroblasts, ependymal and vascular cells as well as the invasion of peripheral immune and

inflammatory cell types such as myeloid cells, macrophages, monocytes and neutrophils (Fig.

3c and Extended Data Fig. 9).

Cell-cell interaction plays important roles in maintaining tissue integrity and normal

functions. To denote faithful *in situ* cell-cell interactions from our spatial transcriptomic data,

we performed a spatial-aware cell-cell communication (SA-CCC) analysis. In brief, spatial

colocalization of cell-type pairs within spots and spot-enriched ligand_receptor (L_R) pairs

deduced from CellChat package[37] were subjected to co-occurrence test (Fig. 3d and see

Methods). Upon injury, both the number of interacting L_R pairs per cell-type pair and the

complexity of the network of co-localized cell types were markedly increased, especially at 72

hpi (Fig. 3e). Among all the cell types, oligodendrocytes and astrocytes had the most extensive

connections and served as the nodes to interact with other cell types in both intact and injured

spinal cords, especially at 72 hpi. In addition, the connectivity of macrophages, reactive

microglia, OPCs and fibroblasts was substantially increased, indicating much escalated

diversity of cell-cell communications in this stage of SCI.

To pinpoint the specific L_R pairs that mediate the spatiotemporal dynamics of cell-cell

communications, the top L_R pairs in each cell-type pair at different time points in SCI were

identified and examined (Extended Data Fig. 10a) and three main classes of L_R pairs were

discerned (Extended Data Fig. 10b). Unlike the cell-cell communication analysis based solely

on scRNA-seq datasets, the SA-CCC analysis is featured by identifying both the expression

levels and the *in situ* interacting cell types for a particular L_R pair in a context-dependent

manner. For example, the L_R pair of *Spp1_Cd44* was expressed predominantly in the WM

at moderate levels at 0 and 3 hpi, which was increased remarkably at 24 hpi (primarily in the

WM and scattered in some GM regions) and spread throughout the entire spinal cord section

at 72 hpi (Extended Data Fig. 10c). Accompanying this profound induction, the interactions of

the cell-type pair(s) mediated by the L_R pair of *Spp1_Cd44* expanded from

"oligodendrocyte_astrocyte-WM" to "astrocyte-WM_OPC" in the WM at 24 hpi and finally to

"oligodendrocyte_macrophage" and "OPC_macrophage" in the GM at 72 hpi (Extended Data

Fig. 10c), suggesting a multifaceted role of this L_R pair along the progression of SCI.

266   **Injury induces redistribution and repurposing of astrocytes-WM into the GM of the**

267   **spinal cord**

268   Astrocytes are morphologically and functionally diverse and play vital roles in the nervous

269   system in both physiological and pathological conditions[38]. Being a significant portion in the

270   spinal cord, not only the proportions of the two main astrocyte subtypes (WM and GM) altered

271   dramatically (Fig. 3c), but also their weights in coordinating cell-cell communications in SCI

272   differed a lot (Fig. 3e). Indeed, the gene expression features, GO term analysis of the DEGs

273   and their upstream transcriptional regulons as well as cross-reference with two single-cell SCI

274   datasets all pointed to the molecular basis underlying their spatial and functional divergences

275   (Extended Data Fig. 11). Thus, our spatial classification of astrocytes-WM and astrocytes-GM

276   corroborates the different physiological functions of the known astrocyte subtypes in the

277   mammalian nervous system[39,40].

278   Strikingly, the two astrocyte subtypes displayed quite different injury-induced changes in

279   their spatiotemporal distributions in response to SCI: astrocytes-WM were largely unchanged

280   at 3 hpi but were drastically increased at 24 and 72 hpi, at which they were even present in

281   the GM region; astrocytes-GM were increased at 3 hpi but then drastically decreased at 24

282   hpi and became almost undetectable at 72 hpi (Fig. 3f, h). The dynamic changes of distribution

283   were validated by examining the spatiotemporal expression patterns of two representative

284   marker genes (*Gfap* and *Pla2g7*; see Extended Data Fig. 11e) by ISH (Fig. 3g, i). Furthermore,

285   by the SA-CCC analysis of cell-cell interactions in the GM and the L_R pairs mediating these

286   interactions (e.g., the *Nrxn_Nlgn* families) (Fig. 3j), we revealed a clear switch of neuron-

287   interacting cells from "astrocytes-GM" at 0 and 3 hpi to "astrocytes-WM" at 24 and 72 hpi,

288   accompanied by the synchronized switch of the spatial L_R pairs (e.g., *Nrxn3_Nlgn3*) (Fig.

289   3k). The *Nrxn3_Nlgn3* is well known for regulating astrocyte maturation as well as synapse

290   development and function[41]. Thus, these findings strongly suggested a functional relay of the

291   original astrocytes-GM to GM-relocated astrocytes-WM at the end of the acute phase of SCI.

292   To further confirm the injury-induced relocation of astrocytes-WM to the GM at 72 hpi, we

293   performed immunostaining of the spinal astrocytes with anti-GFAP. The results not only

verified the predominant localization and fibrous appearance of GFAP$^+$-astrocytes in the WM

of the intact and injured spinal cord but also showed a remarkable emergence of GFAP$^+$-

astrocytes in the GM after injury, and these GM-relocated GFAP$^+$-astrocytes displayed a

hypertrophic morphology (Fig. 3l). The loss of astrocytes-GM at 24 and 72 hpi and the striking

occupation of massive GFAP$^+$-astrocytes in the GM prompted us to further investigate the

molecular origin and the potential functions of these GM-relocated astrocytes. To avoid

confusion due to the discrepancy in the anatomic localization and the transcriptional features,

we designated the injury-induced, GM-relocated astrocytes as "astrocytes-GMii" in the

subsequent study.

The astrocytes-GMii are molecularly distinct from the original astrocytes-WM/-GM and

may contribute to spinal cord rescue/repair

We were keen to know where these astrocytes-GMii came from and what function(s) they

played in the GM. Toward this end, we performed a series of exploratory data analyses and

also cross-tested with the published scRNA-seq datasets of mouse SCI (Fig. 4a). As an

attempt to address the question whether astrocytes-GMii were derived from local cell division

or remote cell migration, we calculated and compared the proliferation score and the migration

score (see Methods) at the spatial and cell type levels (Fig. 4b-e and Extended Data Fig. 12a,

b). The results showed that significant cell proliferation was detected in the spots enriched by

astrocytes-WM, reactive microglia and other cell types only at 72 hpi, much later than the time

when a large number of astrocytes-GMii emerged. In contrast, the migration score started to

increase in astrocytes-WM as early as 3 hpi, which kept rising at 24 hpi and expanded into the

GM at 72 hpi. Notably, astrocytes-GM persistently showed the low proliferation and migration

scores with a minimal increase of migration at 3 hpi. In addition, our analysis of the SCI

scRNA-seq dataset[9] also indicated a markedly higher migration ability associated with

astrocytes-WM than astrocytes-GM (Extended Data Fig. 12b). Together, astrocytes-GMii

likely originated from astrocytes-WM, which migrated into the GM and proliferated to

compensate for the loss of original astrocytes-GM in the acute phase of SCI.

To reveal what unique molecular features astrocytes-GMii possess that made these cells

distinct from the other astrocytes-WM and drove them into the GM upon injury, we assessed

all the spots containing astrocytes-WM (astrocyte-WM spots) in our spatial transcriptomic

dataset (Fig. 4f) using the anatomical definition by aforementioned UMAP clustering (Fig. 1g,

k). In addition, to exclude potential confounding influences from other cell types in the same

spot, only the 317 astrocyte marker genes used for cell type deconvolution were included.

Despite considerable overlapping in their expression features that contribute to the general

identity of astrocytes, these 317 marker genes could be clustered into four spatiotemporally-

distinct astrocyte (SA) groups (Fig. 4g and Supplemental Table 5). Among them, the SA-1

group showed specifically high expression in the GM region at 24 and 72 hpi, while the SA-4

group also showed increased expression in the GM after injury, albeit to a lesser extent. We

speculated that a SA score computed by these two SA gene sets would serve to spatially

define the molecular features of astrocyte-GMii. Interestingly, the SA score was positively

correlated with the proportion of remaining neurons in the spatial spots at 72 hpi (Fig. 4h, i).

Together with the earlier SA-CCC analysis indicating a functional relay of astrocytes-GM to

astrocytes-GMii at the end of the acute phase of SCI (Fig. 3j, k), these data suggested that

the emergence of astrocytes-GMii and their interaction with neurons likely compensated for

the loss of original astrocytes-GM and provided structural and nutritional support to the

neurons spared from the primary injury.

To test the utility of the SA score for distinguishing astrocytes-GMii from the general

population of astrocytes-WM in other spinal cord studies, we re-clustered the astrocytes-*Gfap*

in the SCI scRNA-seq dataset[9]. Indeed, these cells were gated into two distinct populations

based on the SA scores: astrocytes-GMii and the rest astrocytes-WM (-WMrs) (Extended Data

Fig. 12c), which were also separated from each other in the UMAP plot (Fig. 4j). Next, to

expand the spatially-defined molecular features of astrocytes-GMii, we identified the signature

genes of astrocytes-GMii and -WMrs in the whole injured single-cell dataset (Extended Data

Fig. 12d). Also, to facilitate precise sorting and genetic manipulation of astrocytes-GMii in the

future, we further examined the potential surface markers and the upstream TFs of astrocytes-

WMrs, -GMii and -GM (Extended Data Fig. 12e, f). It was worth noting that astrocytes-GMii

were strongly associated with the increased activity of the Irf and Stat families of TFs, pointing

to an immune-triggered transition of astrocytes-WM to astrocytes-GMii. Thus, the findings

based on the spatially defined molecular features together the expanded signature genes of

astrocytes-GMii demonstrated that our spatial transcriptomic analysis could provide valuable

metrics to annotate independent cell states that are often hidden in common single-cell

approaches.

To elucidate the molecular programming driving the transition of the original astrocytes-

WM to the SCI-induced astrocytes-GMii, we performed pseudo-trajectory analysis of the SCI

single-cell dataset by slingshot[42]. We identified continuous transitive structures between

associated clusters in the low dimensional data (Fig. 4j, k) and highlighted the most

dramatically altered genes along this transition (Fig. 4l). One of such genes was *Igfbp2*

(*insulin-like growth factor binding protein 2*), and fluorescence ISH (FISH) of *Igfbp2* together

with immunostaining of GFAP confirmed that astrocytes-GMii, represented by the *Igfbp2*$^+$-

GFAP$^+$ double-positive cells, were specifically induced in the GM of injured spinal cord

especially at 72 hpi (Fig. 4m). *Igfbp2* is a developmentally regulated gene that is highly

expressed in the developing brain and markedly decreases after birth[43,44], while its aberrant

expression in cancer has been proposed to act as a hub of the oncogenic network that

promotes tumor growth and metastasis[44,45]. Here, we showed that overexpression (OE) of

Igfbp2 in primary cultures of mouse astrocytes by lentiviral infection significantly promoted

astrocyte migration in the *in vitro* scratch assay (Fig. 4n). Thus, the expression of *Igfbp2* not

only can serve as a molecular marker for astrocytes-GMii but may also underlie the migration

of astrocytes-WM to the GM in response to SCI.

Finally, to explicate the biological significance of astrocytes-GMii, we compared the

DEGs and their enriched GO terms of astrocytes-WMrs, -GMii and -GM in the SCI single-cell

dataset (Fig. 4o and Supplemental Table 6). The results indicated that astrocytes-GMii shared

several neurotrophic functions in common with astrocytes-GM, which once again suggested

a functional replacement of astrocytes-GM by astrocytes-GMii. In addition, the astrocyte-GMii

signature genes such as *Vegfa*, *Gpr37l1*, *Fgfr3* and *Fgfr1* were associated with a protective

role in neurological disorders[46], which was consistent with the outcome of better neuronal

survival observed in the higher SA scores of astrocytes-GMii (Fig. 4h, i). Together, as both the

previous[47] and this study showed that migrating astrocytes together with other cells in the SCI

lesions expressed multiple axon/synapse-growth-supporting molecules, we anticipate that the

surge of *Igfbp2*+-astrocytes to the GM of the injured spinal cord may provide neurotrophic

support for spare neuron survival and spinal cord rescue in the acute phase of SCI (Fig. 4p).

Future investigation is warranted for the finer classification of astrocytes-GMii and

understanding of their spatial features and functional implications.

Discussion

The spinal cord is a highly structured organ composed of various cell types that are well

organized to perform complicated yet interdependent physiological functions. Two earlier

spatial transcriptomic studies of the mammalian spinal cord identified pathway dynamics

during the disease progression of ALS[13] and described various cell transcriptional states over

the course of injury-induced scar formation[48]. Meanwhile, several single-cell studies have

demonstrated complex cellular heterogeneity and interactions in the mouse spinal cord[9,35,49-

52]. How the diverse cell types respond and crosstalk to one another in a spatial- and context-

dependent manner, especially in the early stage of SCI, drives the subsequent pathological

process and is central to the pathophysiology of SCI. Here, we implemented the spatial

transcriptomics to characterize the spatiotemporal programming of gene expression and cell

organization in the immediate and the acute phase of SCI. We characterized and analyzed

the spatiotemporal changes in gene expression, TF activities, regulon networks, cell

compositions as well as cell-cell communications in SCI. To demonstrate the utility of the rich

information generated in this study, we have set up an interactive web-portal for other

researchers to explore *in situ* gene expression dynamics in SCI *(will release to the public upon*

acceptance of the paper).

Glial cells in the WM are generally considered to provide support and maintain

extracellular homeostasis to neurons, forming myelin and clearing debris. Interestingly, our

spatiotemporal transcriptomic data indicate that genes and cells in the WM respond to SCI

more rapidly, robustly and sophisticatedly than those in the GM. The main molecular features

of the WM shifted twice from Cluster 1 to 4 and then to 5, whereas the GM changed only once

from Cluster 2 to 3 (Fig. 1). Moreover, the early response gene modules (e.g., CGM13 and

CGM22), involving the MAPK/ERK signaling cascade, cytokine production and other immune

and inflammatory responses, were activated in the immediate phase at 3 hpi in the WM but

not the GM (Fig. 2). Thus, the glial cells in the WM do not merely insulate axons or respond

subordinately to neuronal damage. Rather, they behave as the "first responders" in SCI and

play a more active, leading and probably decisive role than previously thought.

SCI triggers multicellular gene expression changes in a temporally and spatially

coordinated manner. When characterizing cell type-specific responses and cell-cell

communications, single cell-based analyses are limited by the sampling bias in the tissue

dissociation procedure and low-coverage sequencing[53]. To amend this, we took the

advantage of spatial transcriptomics and developed a new computation strategy (SA-CCC) to

unfold the cell communication by simultaneous evaluation of the physical distance between

any cell-type pair and the locally aggregated expression level of a L_R pair that is supposed

to express in this cell-type pair (Fig. 3). Using this approach, we not only systematically

illustrate an overview of *in situ* cell-cell interactions of SCI, but also provide a useful toolset for

evaluation of intercellular communications in a comprehensive multicellular context in

biological processes and diseases.

Astrocytes are dynamic, diverse cells whose molecular and functional heterogeneity is

closely related with the local environment[54,55]. Despite their being a barrier to axonal

regeneration in the later phases, migrating reactive astrocytes that form the glial scar play a

beneficial role by the closure of the wound and segregation of immune/inflammatory cells,

which contribute to the spontaneous, partial functional recovery in the subacute phase[56,57].

And, cell-based treatments including the induction of differentiation and migration of reactive

astrocytes have emerged as a potential therapeutic strategy to improve the functional recovery

in patients[58-61]. However, the specific dynamics and the regulatory programs of astrocyte

migration in SCI are still poorly understood, largely due to the difficulty in exploring these

mechanisms *in vivo*. In this study, by leveraging the spatiotemporal transcriptomic allocation

and profiling different cell types and subtypes, we uncover a distinct population of injury-

induced reactive astrocytes (astrocytes-GMii) and explicitly show that these astrocytes

originate from the WM and migrate to the GM, in which they interact with spare neurons and

other cell types at 24 and 72 hpi. Thus, unlike the double-edged scar-forming astrocytes and

those promote detrimental secondary inflammation and immune responses in the subacute

and the intermediate phases, this population of astrocytes-GMii respond much rapidly in SCI

and function as a hub of the multicellular communication and re-organization in the acute

phase, which may support neuronal survival and spinal cord rescue after the primary injury.

In addition, the pseudotime trajectory and the DEG analysis strongly suggested that the local

environment of the GM triggered the transition of the relocated astrocytes-WM to astrocytes-

GMii, and the latter shared certain gene expression similarity with the original astrocytes-GM,

especially regarding the neurotrophic function in maintaining neuronal synapses.

Nevertheless, these astrocytes-GMii do not necessarily migrate from the same segmental

level of the injured spinal cord[62] and are unlikely through tangential astrocyte migration that is

absent in adult spinal cord even after injury[63].

Astrocytic scars have been studied for several decades; however, the exact molecular

mechanisms regulating their formation are not fully understood. This is in part due to the lack

of individual markers for the diverse range of subtype- and context-specific astrocytes. For

example, in this study, GFAP is strongly expressed in both the original astrocytes-WM and the

SCI-induced astrocytes-GMii, despite that their spatial distributions, morphologies, molecular

features and cellular functions are quite different. By a series of integrated spatial and single-

cell transcriptomic analyses, we not only demonstrate that astrocytes-GMii are molecularly

distinct from the original astrocytes-WM and -GM but also identify that these astrocytes-GMii

express both the reactive astrocyte marker GFAP and the IGF signaling factor IGFBP2, and

the latter may underlie the SCI-induced migration of astrocytes-WM to the GM. Hence, our

work exemplifies how the spatiotemporal transcriptomic signatures combined with single-cell

resolution may facilitate subtype annotation and functional interrogation in SCI as well as other

physiological and pathological conditions. In addition, although the secreted protein Igfbp2

was believed to play an essential role in promoting neurite outgrowth in developing and injured

brains[64,65], a recent study showed that blocking Igfbp2 in the astrocyte-conditioned medium

derived from a Rett syndrome model induced a significant increase in neurite outgrowth[66].

Thus, future efforts are warranted to understand the specific role of the increased expression

Igfbp2 in these astrocytes-GMii and to induce the astrocytes-GMii to adopt a neuroprotective

and pro-regenerative state that may aid spinal cord repair.

Taken together, our work here provides the most comprehensive gene expression

profiling of SCI (to the best of our knowledge), involving injury time, distance, directionality,

anatomic domain, cell type and other dimensions. As summarized in Extended Data Fig. 13,

this study disentangles the SCI-induced complex transcriptional programs and multicellular

responses in the tissue context and establishes a spatiotemporal transcriptomic landscape for

in-depth research in the future, providing an open data resource and a platform for identifying

potential targets for therapeutic intervention of SCI.

FIGURE LEGENDS *(< 250 words each)*

Fig. 1. Spatiotemporal transcriptomics and molecular clusters of mouse SCI using

Visium

(**a**) A carton of the mouse SCI model and the spinal cord sections obtained in this spatial

transcriptomic study. (**b**) A schematic of the Visium Gene Expression slide and the acquisition

of the mouse SCI sections for RNA-sequencing. (**c**) The main workflow of the spatial

transcriptomic data analyses. (**d**) The UMAP plot of the 22,820 transcriptomic profiles (one

profile for each spatial spot) of all the SCI samples identified 7 spatiotemporal clusters. (**e**)

The clusters are color coded and shown in the spatial view of the spinal cord sections at the

indicated locations and time points after injury. (**f**) A cartoon showing the main anatomic

regions of the spinal cord: WM, white matter; MG, middle grey; DH, dorsal horn; VH, ventral

horn. (**g-j**) The UMAP plots of the spatial spots were colored according to anatomic region (**g**),

time after injury (**h**), directionality (**i**) and distance to the injury site (**j**). (**k**) Quantification of the

numbers of the spatial spots in each cluster at the indicated time points. (**l**) The heatmap

showing the expression levels of the top 10 CEGs (left) and the dot-plot showing the top 5 GO

terms (right) for each of the 7 spatiotemporal clusters. (**m**) The spatial view of the gene

expression levels in the spatiotemporal transcriptomics (ST, left) and *in situ* hybridization (ISH,

right) of four representative anatomic markers as indicated. Scale bar: 500 μm. SCI, spinal

cord injury; hpi: hours post injury; UMI, unique molecular identifier; UMAP, uniform manifold

approximation and projection; CEGs, cluster-enriched genes; GO terms, gene ontology terms;

p. adjust, adjusted *p* value.

Fig. 2. Spatiotemporal analyses of gene co-expression dynamics and regulatory

networks in SCI

(**a**) Heatmap showing the weighted-correlation network of the genes constituting 33 SCI

spatiotemporal CGMs. (**b-d**) The molecular features of three representative CGMs: CGM13

(**b**), CGM9 (**c**) and CGM11 (**d**). From left to right: Left, the spatial view of average levels of the

genes expressed in the indicated CGM during the process of SCI; middle, the ridge plot

quantitatively illustrating the density distribution of the CGM expression within the 4 anatomic

domains at the indicated time points of SCI; right, the representative GO terms (top) and the

gene network showing the highest 50 connectivity weights (bottom) of the indicated CGM. The

number of genes in each CGM is shown in the parentheses. (**e**) The CSI heatmap of 147

transcriptional regulons grouped into 7 RMs by hierarchical clustering. (**f**) The heatmap of the

overall activity of each RM enriched in the four anatomic domains at the indicated time points

of SCI. (**g**) The spatial view of the average regulon activity of RM2 at the indicated time points

of SCI. The number of regulons in RM2 is shown in the parentheses. (**h**) The Pearson

correlation networks and the CSI connectivity degree of the transcriptional regulons within and

across the 7 RMs. CGM, co-expressed gene module; CSI, connectivity specificity index; RM,

regulon module.

Fig. 3. Analyses of multicellular responses and cell-cell communications in SCI identify

injury-induced, GM-relocated astrocytes-WM

(**a, b**) The spatial views of the spots showing the proportions of all cell types in each spot (as

a pie chart) of the spinal sections at the indicated time points in SCI (**a**), and the zoom-in of

the yellow-boxed area with the spatial distributions of the representative spinal cord cell types

is shown in (**b**); also see Extended Data Fig. 7-8. (**c**) The fractions (%) of different cell types

at indicated time points in SCI. (**d**) A schematic cartoon illustrating the main procedure for the

spatial-aware cell-cell communication (SA-CCC) analysis that evaluates *in situ* intercellular

communications by testing the statistical significance of the co-occurrence of a cell-type pair

and an L_R pair. (**e**) The SA-CCC networks at each indicated time point in SCI. The size of

the orange circles represents the number of the interacting L_R pairs expressed in each cell-

type pair. For the connecting lines, the thickness denotes the GeoM of the significance of the

interaction and the shade indicates the average expression levels of the L_R pairs for each

cell-type pair. (**f-i**) The spatiotemporal distributions of astrocytes-WM (**f**) and astrocytes-GM

(**h**), validated with ISH of *Gfap* for astrocytes-WM (**g**) and *Pla2g7* for astrocytes-GM (**i**). Scale

bars: 500 µm. (**j**) Dot plots showing the major interacting L_R pairs of the main cell-type pairs

in the GM at the indicated time points in SCI (from Extended Data Fig. 10a). (**k**) From left to

right: left, the geometric proportions of the cell-type pairs of "astrocyte-GM_neuron" and

"astrocyte-WM_neuron"; middle, the expression levels of the individual ligand, receptor, and

their geometric mean of the L_R pair of *Nrxn3_Nlgn3*; right, the average expression of the

Nrxn3_Nlgn3 pair within the indicated co-localized cell-type pairs along the time in SCI. (**l**)

Immunostaining of GFAP in the mouse spinal cord (-1.0 mm section) at 0 and 72 hpi. The

morphologies of GFAP$^+$-astrocytes in the GM (yellow box) and the WM (cyan box) are shown

in the zoom-ins on the right. Scale bars: 200 µm (left) and 20 µm (right). Astro, astrocytes;

WM, white matter; GM, grey matter; H-Microglia, homeostatic microglia; R-Microglia, reactive

microglia; GeoM, geometric mean; EM, assembly mean; L_R pair, ligand_receptor pair; ISH,

in situ hybridization.

Fig. 4. Characterization of the molecular features of SCI-induced astrocytes-GMii with

spatial transcriptomics and scRNA-seq integration

(**a**) An overview of the comprehensive spatial and single-cell transcriptomic analyses (in Fig.

4 and Extended Data Fig. 12) for the identification and characterization of the SCI-induced

astrocytes-GMii. (**b, c**) The spatiotemporal dynamics of the proliferation (**b**) and migration (**c**)

scores in SCI. (**d, e**) The jaccard similarity (JC index) of the proliferation score (**d**) or migration

score (**e**) for the spatial spots containing the specified individual cell types at the indicated time

points in SCI. (**f**) An example of the spinal cord section showing the H&E staining, the spatially-

defined (sd) WM and sdGM domains (annotated using the spatial clusters in Fig. 1g, k), and

the spatial spots containing astrocytes-WM. (**g**) The 317 astrocyte deconvolution marker

genes in the spatial spots containing astrocytes-WM are clustered into four SA groups based

on their bulk expression levels according to the sdWM or sdGM domain and the injury time.

(**h, i**) The spatial distributions of the sdWM and sdGM domains, the SA score and neuron

proportion in each spot at 72 hpi (**h**) and their scatterplot along the distance (-1.0 to +1.0 mm)

as well as the corresponding locally weighted regression fitting curves (**i**) are shown. The

normalized values are the z-score [0, 1] of the SA scores and neuron proportions. The

Spearman rank correlation coefficient of the SA score and neuron proportion is shown in the

paratheses. (**j, k**) The UMAP re-clustering (**j**) and pseudotime trajectory (**k**) analyses of the

astrocytes-*Gfap* in the SCI single-cell dataset[9] identify distinct populations of astrocytes-WMrs

and astrocytes-GMii sharing a continuous cell lineage. (**l**) The expression heatmap of the top

50 most dramatically-altered genes along the trajectory in (k). Among them, an example of

Igfbp2 is examined by FISH in (m). (**m**) Representative confocal images of immunostaining of

GFAP combined with FISH of *Igfbp2* of the mouse spinal cord (-1.0 mm section) at the

indicated time points in SCI. The white-boxed areas are shown in a higher magnification in the

zoom-ins and the 3D rendering of a "GFAP+-*Igfbp2*+" cell at 72 hpi (yellow box) is shown, with

DAPI staining to indicate the location of the nucleus. The average densities of the indicated

immunophenotypes of cells in the mouse spinal cord area and the percentage of "GFAP+-

Igfbp2+" cells to total cell counts (indicated by DAPI) at the specified time points in SCI are

quantified. Scale bars: 10 μm in 3D rendering, 25 μm in zoom-ins, and 200 μm in the rest. (**n**)

Representative brightfield images of the scratch assay of *in vitro* cultured mouse primary

astrocytes infected with pCDH-vector (control) or pCDH-*Igfbp2*. The gap (cell-free area)

caused by the scratch in multiple random visions of two pooled repeats were traced along time

and the average size of the gap was calculated and shown on the right. Scale bar, 100 μm.

Mean ± SEM; n = 6; *$p < 0.05$ and ***$p < 0.001$; one-way ANOVA in (m) and Student's t-test

in (n). (**o**) GO term analysis of the unique and shared DEGs (left) and the heatmap showing

the average normalized expression levels of the DEGs (right) associated with astrocytes-

WMrs, -GMii and -GM. The DEGs that are cross-referenced as pro-inflammatory (red),

neurotrophic (green) or neuroprotective (in bold) genes by Kaufmann et al.[46] are highlighted.

(**p**) A schematic of the SCI-induced, *Igfbp2*-expressing astrocytes-GMii, which originate from

reactive astrocytes in the WM and migrate into the GM to provide support to spare neurons.

Astro, astrocytes; WM, white matter; GM, grey matter; Astro-GMii, injury-induced GM-

relocated astrocytes; Astro-WMrs, the rest astrocytes-WM; SA, spatiotemporally-distinct

astrocyte; FISH, fluorescence ISH; hps, hour post scratch.

METHODS *(< 3,000 words)*

Animal Care

All practices on mice in this study were performed in compliance with the institutional

guidelines on the scientific use of living animals at the Shanghai Interdisciplinary Research

Center on Biology and Chemistry, the Chinese Academy of Sciences (CAS). Animal distress

and conditions requiring euthanasia were addressed and the number of animals used was

minimized. All mice were housed in a pathogen-free barrier facility with free access to food

and water.

Mouse spinal cord injury (SCI)

The C57BL/6J mice (Jackson Laboratory, #000664) used in this study were purchased from

the Shanghai Laboratory Animal Center (SLAC). Adult female mice of 8-9 weeks old were

used in the SCI. The spinal cord transection was performed as previously described[67]. Briefly,

after exposure and removal of the lamina of the tenth thoracic vertebra (T10) spinal segment,

the spinal cord was transected with Micro Scissors, and the wound was then cleaned and

sutured. For sham controls, only laminectomy at T10 was performed. For post-operative care,

the mice received analgesics at the wound every 12 h and were subjected to manual bladder-

emptying at least twice daily.

Behavioral evaluation of SCI

All animals were accommodated to the testing room or apparatus for at least 1 h before

behavioral assessments. Blind scoring was performed to ensure that observers were unaware

of treatments. The animals were evaluated for the hindlimb locomotion performance at 0

(uninjured), 3, 24 and 72 hours post injury (hpi) by the Basso mouse scale (BMS)[68]. Briefly,

mice were placed in an open field for 4 min. Two observers assessed the performance

independently with the reference to the motion attributes of the hindlimbs, including joint

movement, weight support, plantar stepping, coordination, paw position, and trunk and tail

Spatial transcriptomics of SCI_ZWang & ZLi

control. The BMS scores range from 0 to 9 points (0 means complete paralysis and 9 means

normal mobility). For the mechanical sensitivity (von Frey) test[69], the mice were placed on the

bottom of a metal grid and the mid-plantar surface of the hindpaw of the mice was stimulated

with 2 g von Frey filaments vertically. The filaments were applied to the hindpaw surface until

the filament bent, and paw withdrawal or flinching was considered as a positive response. The

test of tactile withdrawal threshold was repeated 10 times in each mouse. The mean value of

the performance of the left and right hindlimbs was calculated and used in the BMS and von

Frey tests.

Cryosection and immunohistochemistry

Anesthetized mice were fixed via transcardial perfusion with diethylpyrocarbonate-treated

phosphate-buffered saline (PBS) followed by 4% (w/v) paraformaldehyde (PFA, RNase-free;

Bioss, C2055). Spinal cord tissues were then dissected and post-fixed in 4% PFA at 4°C

overnight. The fixed spinal cord samples were cryoprotected in 30% (w/v) sucrose (Sangon,

A0498) in PBS, embedded in Tissue-Tek O.C.T. (Sakura, 4583), snap-frozen in liquid nitrogen,

and stored at -80°C until cryosection. Before cryosection, samples were equilibrated to -20°C

in the cryostat for at least 30 min. Cross sections (20 μm) of the spinal cord were collected on

the slides and could be stored at -80°C up to 2 weeks.

For hematoxylin and eosin (H&E) staining, the spinal cord sections on the slides were

incubated at 37 °C for 1 min, and then completely immersed in the prechilled methanol at -

20 °C for 30 min. After three washes with deionized water (diH$_2$O), isopropanol was added to

fix the tissue slides at room temperature (RT) for 1 min. Thereafter, the slides were air-dried

and hematoxylin (Agilent, S330930-2) was added to the slides at RT for 7 min. After three

washes with diH$_2$O, slides were treated by Bluing Buffer (Agilent, CS70230-2) at RT for 2 min.

Finally, eosin (Sigma, HT110216-500ML) was added at RT for 1 min and the slides were

washed in diH$_2$O for three times, followed by imaging with Aperio LV1 slide scanner (Leica).

For immunostaining, the spinal cord sections on the slides were incubated at 37 °C for

1 min, post-fixed once more in 4% PFA at RT for 15 min, and then permeabilized and blocked

with the blocking buffer (10% normal goat serum and 0.5% Triton X-100 in PBS) at RT for 1

h. The primary rat anti-GFAP antibody (Invitrogen, 13-0300) was subsequently included in the

blocking buffer at 4 °C overnight. The sections were washed with 0.3% Triton X-100 in PBS

for three times and the secondary goat anti-rat-Alexa Fluor 568 antibody (Life Technologies,

A11077) was added to the slide for 1 h at RT. After washed by 0.3% Triton X-100 in PBS for

three times, the spinal cord sections on the slides were mounted with coverslips using the

VECTASHIELD Antifade Mounting Medium with DAPI (Vector Laboratories, H1200) and

imaged by Dragonfly Spinning Disk Confocal Microscope (Andor) with a 20x/0.75 NA objective.

***In situ* hybridization (ISH) and fluorescence *in situ* hybridization (FISH)**

ISH was performed as previously described[70]. Briefly, the mouse spinal cord sections were

treated with 10 µg/mL proteinase K (QIAGEN, 19131) at 37°C for 5 min and then post-fixed in

4% PFA at RT for 10 min. After washing and dehydration, DIG-labeled RNA probes at

approximately 1 µg/mL in hybridization buffer were incubated with the sections on the slides

at 65°C overnight. Afterwards, the slides were washed with 50% formamide (Sangon,

A600212) in 2x saline sodium citrate (SSC) (Invitrogen, AM9770) at 65°C for 30 min, followed

by incubation with 1 mg/mL Rnase A (Roche, 10109142001) at 37°C for 30 min. The slides

were then washed sequentially in 2x SSC at 60°C for 20 min, 0.2x SSC at 60°C for 20 min,

and 0.1x SSC at RT for 20 min. For ISH, the slides were incubated with anti-DIG-AP (Roche,

11093274910) at 4°C overnight, followed by washing and incubation with BM-purple (Roche,

11442074001) until dark purple color became visible.

The composition of the hybridization buffer: 50% (v/v) formamide; 10 mM Tris-HCl solution,

pH 8.0 (Sangon, B548127); 200 µg/mL yeast tRNA (ThermoFisher, AM7119); 10% (v/v)

dextran sulfate (Millipore, S4030); 1x Denhardt's solution (Sigma, D9905); 600 mM NaCl,

RNase-free (ThermoFisher, AM9760G); 0.25% SDS (ThermoFisher, AM9820); 1 mM EDTA

(0.5 M), pH 8.0, RNase-free (ThermoFisher, AM9261); Nuclease-free water (ThermoFisher,

AM9937).

For FISH, the procedure was similar to the ISH but the anti-DIG-POD antibody (Roche,

11207733910) was used for signal development. The slides were finally incubated with

TSA_Plus_Fluorescence Kits (Akoya, NEL745001KT) according to the manufacturer's

instruction and imaged.

For simultaneous detection of the immunofluorescence signals of GFAP with FISH, the

immunostaining was conducted after FISH images had been obtained. Antigen retrieval was

performed with Antigen Retrieval Solution (Solarbio, C1035) before the conventional

immunostaining procedure.

The RNA probes used in this study were adapted from the Allen Spinal Cord Atlas

(https://mousespinal.brain-map.org/), and synthesized with the DIG RNA labeling kit (Roche,

11277073910) following the procedure described in (Peng et al., 2016)[70]. The following

primers were used:

Gfap forward: 5'---GTGGATTTGGAGAGAAAGGTTG---3'

Gfap reverse: 5'---CTGGAGGTTGGAGAAAGTCTGT---3'

Pla2g7 forward: 5'---AAGGTCGCCTCGACACTG---3'

Pla2g7 reverse: 5'---TCAAAGGGTGACCCAGGA---3'

Igfbp2 forward: 5'---ACAGTGATGACGACCACTCTGA---3'

Igfbp2 reverse: 5'---CCTCTCTAACAGAAGCAAGGGA---3'

Culture primary astrocytes from mouse pup brains

Primary astrocytes were isolated from C57BL/6J newborn mice at postnatal day 2 (P2) as

previously described[71]. In brief, after removal of the meninges, mouse cortical tissues were

minced and digested sequentially with 0.05% DNase I (Sigma, DN-25) for 5 min and 0.05%

trypsin (Gibco, 15090-046) for 20 min at RT, followed by filtration through a 70-µm cell strainer

(FALCON, 352350) to remove undissociated tissues and debris. The filtrate was centrifuged

at 150 *g* for 10 min, and the cell pellet was resuspended after decanting the supernatant. The

harvested cells were plated into T-75 flasks coated with 50 µg/mL poly-L-Lysine, and

incubated at 37 °C with 5% CO_2 in an incubator. The growth medium (10% fetal bovine serum

and 2% penicillin-streptomycin in DMEM) was changed every other day. Upon the confluence

of the cells, oligodendrocyte progenitor cells (OPCs) could be detached by tapping the flask

and DNase I digestion (37 °C for 5 min). After completely removing the supernatant, the

remaining adherent cells were subsequently digested with 0.25% trypsin for 5 min. The cell

suspension was mixed with an equal volume of the growth medium to deactivate trypsin, and

centrifuged for collecting the pellet. Thereafter, the resuspended cells were subjected to two

rounds of incubation in bacterial grade plates for 20 and 90 min, respectively. This helped to

effectively remove the microglia as they are more prone to attach to the plate surface

compared with astrocytes. The supernatant was then transferred to a new T-75 flask, and

primary astrocytes were allowed to grow to confluence before use.

Lentivirus production

To generate lentivirus for infecting primary astrocytes, 293T cells were co-transfected with

target plasmid (pCDH-vector-flag or pCDH-*Igfbp2*-flag), psPAX2 and pMD2.G with a ratio of

4:3:1 in Opti-MEM (Gibco, 31985070) using Lipofectamine 2000 (Invitrogen, 11668). The

culture medium was collected at 48 h after transfection and passed through a 0.45-µm filter.

Viral particles were concentrated using the Lenti-X Concentrator (Clontech, 631232) and the

viral pellets were resuspended in PBS for astrocyte infection.

The *in vitro* cell migration assay

Primary cultures of mouse astrocytes were grown *in vitro* in the plates coated with 5 µg/mL

fibronectin (Sigma, F2006) for 4-5 days before they were infected with lentivirus. Three days

after infection, the astrocytes were treated with 10 µg/mL mitomycin C (Sigma, M5353) for 2

h to suppress cell proliferation and then the middle of the monolayered astrocytes was

scratched with a pipette tip to create a gap (cell-free area), which was traced and imaged at

0, 24 and 48 hours post scratch (hps) using phase contrast microscopy.

Image acquisition, processing and quantification of light microscopy

Brightfield images of H&E staining and ISH were acquired using the Aperio LV1 slide scanner (Leica) and processed in ImageScope. FISH and immunofluorescence images were acquired using the Dragonfly Spinning Disk Confocal Microscope (Andor) with a 20x/0.75 NA objective. The 3D rendering of the GFAP$^+$-*Igfbp2*$^+$ astrocytes was performed in Imaris using the Surface function. The images were processed and assembled into figures using Adobe Photoshop 2022 and Adobe Illustrator 2022.

For quantification of astrocytes, two regions of interest (ROIs) were drawn to outline the left and right grey matter (GM) of each spinal cord section from three different animals (n = 6 ROIs for each time point). The number of DAPI, GFAP$^+$ or *Igfbp2*$^+$ cells within each ROI was counted by ImageJ using the Analyze Particles function. And the GFAP$^+$/*Igfbp2*$^+$ double-positive cells over DAPI were visually examined and manually counted. Statistical analysis was determined by one-way analysis of variance (ANOVA) with Tukey's HSD post-hoc test at *$p < 0.05$, **$p < 0.01$, and ***$p < 0.001$. Error bars represent the standard error of the mean (SEM).

For quantification of the gap size in the cell migration assay, the cell-free area was delineated, with its size measured by ImageJ at 0, 24 and 48 hps. Six randomly-selected, non-overlapping regions of the samples prepared independently in each group were included at each time point for calculating the percentage of the gap area, using the following equation: Relative gap size (%) = S/S_{0hps} x 100%, where S stands for the number of the pixels of the cell-free area of a sample, and S_{0hps} is that of the same sample at 0 hps.

Acquisition of mouse spinal cord sections for 10x Genomics Visium spatial RNA-seq

Dissected spinal cords were washed with 1x PBS and subsequently embedded in Tissue-Tek O.C.T. through a bath of dry ice and prechilled ethanol. Before sectioning, Visium slides and samples were equilibrated to -20°C in the cryostat for at least 30 min. Four consecutive spinal cord cryosections were cut at 20-µm thickness each and attached to one of the four capture

areas of the Visium Spatial Gene Expression slides (10x Genomics, 1000184). The slides

were stored individually in sealed containers at -80°C before library construction.

Spatial transcriptomics library construction and sequencing

The spatial transcriptomics library was constructed according to the Visium Spatial Gene

Expression Reagent Kits User Guide (10x Genomics, CG000239 Rev D). In brief, after H&E

staining and imaging, the spinal cord samples on the slides were permeabilized for 18 min,

which was determined based on the pre-tests using the tissue optimization kit (10x Genomics,

1000193). Thereafter, the tissue slides were reverse transcribed and amplified, producing

cDNAs containing the spatial barcodes. The amplified cDNAs were subjected to library

construction and sequenced with the Illumina NovaSeq 6000 system using a 150-bp paired-

end setting at a depth of 100k reads per spot.

<u>**Data analyses of spatial transcriptomics**</u>

Pre-processing and quality control

The raw Visium spatial RNA-seq data and histological H&E images were processed using the

Space Ranger pipeline (v.1.0.0, 10x Genomics) to align and summarize unique molecular

identifier (UMI) counts according to the reference genome "Mouse Genome mm10".

To assess the quality of the spatial RNA-seq data, we grouped the SCI samples based

on the replicate batch and estimate the sample correlation coefficients. In brief, average

expression of each spinal cord section replicate was calculated and the Pearson correlation

coefficient (PCC) analysis was performed across all samples. The expressed genes, UMI

counts and mitochondrial levels were computed by Seurat (version 4.1.1)[72]. Gene expression

levels of all spatial spots were normalized and scaled using the 'SCTransform' function[73] and

merged by regressing out 'section' factors to create a single Seurat object for further analyses.

All the hemoglobin-related genes (due to infiltration of red blood cells from the damaged blood-

spinal cord barrier) were removed from the matrix before merging. Other potential confounding

effects including cell cycle, sequencing library size and percentage of mitochondria were

regressed out as well.

Spatial cluster analysis

The top 3,000 highly variable genes during SCI were identified and used for the subsequent

analysis. We computed 20 principal components (PCs) with 'runPCA' and constructed the

shared nearest neighborhood graph with 50 local neighbors of each pixel using the

'FindNeighbors' function. We used the Louvain clustering algorithm to optimize the modularity

of neighbors and the UMAP (Uniform Manifold Approximation and Projection) visualization at

the resolution of 0.25. Seven spatiotemporal clusters were recognized and classified into four

anatomic domains according to their spatial distributions and the identified signature genes:

white matter (WM), middle grey (MG), dorsal horn (DH) and ventral horn (VH).

Evaluation of activity scores of gene sets

Injury, proliferation and migration scores of the indicated gene sets were computed using the

function of 'AddModuleScore' in Seurat with gene sets extracted from published literatures or

defined by the enrichment analysis. Specifically, the injury score was calculated based on the

expression of the following 22 genes: *Adamts1*, *Atf3*, *Ccl2*, *Ccnd1*, *Cd68*, *Cebpd*, *Cyba*, *Fn1*,

Gal, *Gap43*, *Hmox1*, *Hspb1*, *Igfbp2*, *Jun*, *Junb*, *Fos*, *Lgals1*, *Neat1*, *Socs3*, *Tnc*, *S100a10*,

and *Timp1*; the proliferation score was calculated based on the expression of *Mki67*; and the

migration score was calculated based on the expression of the genes in the "positive

regulation of cell migration" gene ontology (GO) term.

Differential expression gene (DEG) and GO term analyses

For DEG analysis, 'FindAllMarkers' and 'FindMarkers' of Seurat were run with the following

parameters: logfc.threshold = 0.5, min.pct = 0.25 and method = "wilcoxon". GO term analysis

was done using the "clusterProfiler" R package[74]. The 'compareCluster' function was used for

gene list enrichment against all expressed genes by setting *p*. adjust < 0.01 or 0.05. The top

5 or 10 (*p*. adjust) terms were displayed by DotPlot.

Gene co-expression analysis

The WGCNA (weighted gene co-expression network analysis) package in R[75] was used to

measure co-expression networks across all the spatial spots. Genes expressed in over 100

spatial spots were selected and their average expression levels in each group of injury time

and anatomic domains were calculated. To calculate the adjacent matrix, the soft power of 18

was chosen with the WGCNA function 'pickSoftThreshold'. The module identification and

topological connectivity network were computed using the 'DynamicTreeCut' function with

min.module.size = 30, deep = 2 and mergeHeight = 0.15. The most connected genes in the

topological network of each module were defined as the hub genes and were visualized with

Cytoscape[76]. Ribosome-related genes were filtered manually to avoid overdominance in co-

expression analysis when displaying connectivity network of Co-expressed Gene Module

(CGM) 9. Enriched regulons of CGMs were determined using the hypergeometric test against

previously identified regulons of transcription factors (TFs) and the cutoff *p*. adjust (which was

calculated with the bonferroni multiple comparison correction of *p* value in hypergeometric test

by the number of modules) was set to < 0.05. Indices of significance were calculated by (0.05

- *p*. adjust)/0.05. The top 3 (ranked by significance index) regulons of gene modules were

displayed.

Analysis of transcriptional regulons and regulon modules (RMs) with SCENIC (single-

cell regulatory network inference and clustering)

The Python package "pyscenic"[28] was used to detect active transcription RMs. Briefly, the

spatial gene expression matrix was first filtered to exclude genes expressed in fewer than 0.1%

of all the spots. The gene-gene correlation matrix for module identification was computed with

the Random Forest based on the 'GRNBoost2' algorithm. Each module was further pruned to

only include genes that were present in the RcisTarget database with the 'prune2df' function.

Therefore, regulons consisting of the TFs and their transcriptional targets in the co-expression

module were detected finally. Overall, 515 regulons were found in this study and the regulon

scores were calculated by the 'AUCell' function. To evaluate the co-expression relationship of

the regulons, the top 20 regulons (based on the regulon specificity score, the RSS[77] for each

combination of anatomic domain and injury time was picked (147 regulons in total; less than

predicted due to redundant regulons), which were then used to compute their connectivity

specificity index (CSI) for each pair of regulons following the instruction in (Bass et al., 2013)[78].

Based on the CSI, 7 RMs were identified via hierarchical clustering. Activity score of each CSI

module was calculated by averaging the regulon scores within each module. Regulon modules

with PCC > 0.65 were visualized with Cytoscape[76].

For regulon analysis of the single-cell data of astrocytes, the same procedure was

followed and 433 regulons were identified. The scaled average activity was computed across

three spatial associated states and the top 10 regulons of each state (based on RSS) were

shown.

Cell type deconvolution and identification of cell type-enriched DEGs

To estimate the proportions of the cell types in each spatial spot, all the spots were

deconvoluted with the "Spatial eXpression R (spacexr)" package (version 2.0, formerly

RCTD[34]). For a better inclusiveness of diverse cell types, we imported and combined two

mouse spinal cord single-cell datasets[9,35] in the RCTD analysis. We further defined astrocytes

in the SCI single-cell data[9] into three subtypes (astrocyte-*Gfap*, -*Slc7a10* and -*Svep1*) with the

annotation by Rosenberg et al. (2018)[35] as the reference by label transfer[79]. The cell types in

the sham and injured spinal cords (up to 72 hpi) identified by Milich et al. (2021)[9] were used

to construct the spacexr reference. Finally, the percentage of different cell types for each spot

was calculated using 'run.RCTD' with "multi" mode (CELL_MIN_INSTANCE = 10, UMI_min =

(min(puck@UMI)-1), UMI_max = (max(puck@UMI)+1), fc_cutoff_reg = 1). For robust cell type

decomposition, confident sub_weight was used to evaluate the final proportions of different

cell types in all spots. Astrocytes-*Svep1* and dendritic cells were merged into the class of

"other cells" due to their extremely low presence in the overall spatial transcriptomic data and

were not considered in the subsequent analysis. In addition, excessive subtype classifications

showing scattered spatial distributions were merged and analyzed at the cell type level.

The DEGs of the indicated cell type were calculated with the 'C-SIDE' function[80]. For

analysis of astrocyte subtype-enriched DEGs, astrocytes-WM or -GM in the four replicates of

the +1.0 mm sections at 0 hpi were passed to C-SIDE with the parameters of

"cell_type_threshold = 1, weight_threshold = 0.9 and doublet_mode = F" and their spatial

localization was used as a function of the covariates.

Spatial-aware cell-cell communication (SA-CCC) analysis

The ligand_receptor (L_R) pair interaction information was extracted from CellChatDB.mouse

of CellChat[37]. To further pinpoint the L_R pair interactions between different cell-type pairs in

the spatial context, we binarized the L_R pair consensus matrix of each spatial spot based on

whether its value exceeded its assembly mean across all spots. Similarly, the cell-type pairs

were also binarized depending on their deconvolution proportions. To minimize the

interference from ubiquitous and/or ambiguous interactions from other cell types, the

proportions of neurons and oligodendrocytes were filtered by their mean values when

assessing the spatial interactions. Additionally, to exclude "false interaction" between spatially

nonadjacent cells, the cell-type pairs were eliminated if the Spearman rank correlation

coefficient of their deconvolution proportions was < -0.4. The spatial correlation between each

L_R pair and cell-type pair was calculated using the one-side Fisher's exact test. When the p

value was < 0.001, the cell-type pair was considered to have a bona fide spatial cell-cell

interaction. The expression level of a specific L_R pair for an associated cell-type pair was

calculated by averaging the values of L_R pair geometric mean (GeoM) across their

colocalized pixels. The aggregated types of cell-cell interactions, the average expression

levels of all L_R pairs between indicated cell-type pairs and the -log10(GeoM of p values) at

each time point of SCI were visualized by igraph[81]. Only interactions between the cell-type

pairs that have over 10 shared positive spatial spots were shown. The SA-CCC results were

further applied to assess the cell-cell interactions identified in the SCI single-cell study[9] using

the CellChat with the default parameters. The SA-CCC analysis is accessible and searchable

in our interactive data exploration portal: *(will release to the public upon acceptance of the*

paper).

Re-analysis of the SCI single-cell datasets of the spatially-defined astrocytes

To re-cluster the astrocytes in the SCI single-cell dataset[9], the single-cell matrix was

normalized and scaled using the 'SCTransform' function (var.to.regress =

c("orig.ident","S.Score","G2M.Score","percent_mt","percent_rp"), do.scale = TRUE). Cells

with low counts (< 2,500) were filtered. Ribosome- and mitochondria-related genes were

regressed out. The top 16 PCs, 40 k.param and resolution = 0.2 were used in the UMAP

analysis (Extended Data Fig. 9d). Two molecularly-defined subtypes of astrocytes-WM (*Gfap*[+])

and astrocytes-GM (*Slc7a10*[+]) were identified, and the subsets of the RNA matrix of the

astrocytes were used in the subsequent analyses.

The spatially scaled data were used to extract the expression profiles of astrocytes-WM

in all the spots that contained astrocytes-WM based on the deconvolution result. The 317

astrocyte marker genes used in the cell type deconvolution analysis were retained. Scaled

expression was binarized into 1 and -1 based on the threshold = 0, which was used to estimate

the expression of each marker gene at the bulk level in the WM and the GM at all time points

of SCI. K-means clustering (k = 4) was employed to identify spatially-defined marker genes

for the SA groups of astrocytes in Fig. 4g.

To evaluate whether the SA-1 and SA-4 gene groups were able to distinguish astrocytes-

GMii from the rest astrocytes-WM, the astrocytes-*Gfap* (WM) defined in the SCI single-cell

study[9] were re-clustered using the maximum distance and the 'cutree' function based on the

scores of the SA-1 and SA-4 gene sets. The DEGs of the original astrocytes-WM and -GM as

well as the injury-induced astrocytes-GMii were identified and compared pairwise to reveal

their overlapping relationships (logfc.threshold = 0.5, *p*. adjust < 0.01). Surface marker

analysis was done by overlapping the surface genes downloaded from Cell Surface Protein

Atlas with astrocyte-GMii signature genes.

Pseudotime analysis

Trajectory inference was performed in low dimension data of the astrocytes-*Gfap* (WM),

including the astrocytes in the uninjured WM and the GM-relocated, WM-featured astrocytes

after injury, in the SCI single-cell study[9] using the slingshot package (v2.5.5)[42].

Statistical information

The exact statistical information of the comprehensive spatial transcriptomic analyses in this

study is specified in each related section of the Methods above or as indicated in the legends

of the relevant figures.

Data and Materials Availability

Raw data generated by SpaceRanger pipelines are available through the NODE project under

the accession number OEP003508. Our SCI spatial transcriptomic resources can be explored

at the SCI-ST web portal *(will release to the public upon acceptance of the paper)*. All other

data are available in the main text and the supplementary materials. All unique and stable

reagents generated in this study are available from the corresponding authors with a

completed Material Transfer Agreement (MTA).

Supplemental Information

Supplemental information includes Extended Data Fig. 1-13, Supplemental References 82-86,

and Supplemental Tables 1-6.

Acknowledgements

We thank Drs. J. Hu, Z. He, C. Zhang, and Misses. J Zhu and Z. Sun for assistance in mouse operation, Dr. Q. Wang for help with cryosection, Mr. G. Zhang for technical support with 10x Genomics Visium spatial RNA-seq, Dr. X. Sun for help with ISH, and Drs. W. Li, Y. Song, B. Zhou, P. Yu, C. Liu, Z. He, G. Bai, H. Liu and J. Yuan for comments and critical reading of the manuscript.

Fundings

This study was supported by grants from National Key R&D Program of China (2018YFA0801402 to G.P.); the National Natural Science Foundation of China (NSFC) (31970697 and 32270812 to Y.F., 32270854 to G.P., and 82071372 to A. L.); the Science and Technology Commission of Shanghai Municipality (201409003300, 20490712600 and 2019SHZDZX02 to Y.F.); the "Strategic Priority Research Program" of the Chinese Academy of Science (XDA16020404) to G. P.; Natural Science Foundation of Guangdong Province (2021A1515011231 to A.L. and 2019B151502054 to G.P.), the Outstanding Scholar Program of Bioland Laboratory (2018GZR110102002) and the Science and Technology Program of Guangzhou (202007030012) to A. L.

Author Contributions

AL, GP and YF conceived the research; ZW, SS, and YF designed the experiments; ZW, SS and TL performed the experiments and analyzed the data; AL, WY and YL contributed important reagents; ZL, JX and GP performed the bioinformatic analyses; ZW, ZL, TL, GC, KZ, GP, AL and YF interpreted the results; ZW, ZL, TL, GP and YF prepared the figures; ZW, ZL, TL, AL, GP and YF wrote the paper. All authors read and approved the final manuscript.

Competing Interests

The authors declare no competing interests.

REFERENCES *(≤50 references)*

1. Ahuja, C. S. *et al.* Traumatic spinal cord injury. *Nat. Rev. Dis. Primers.* **3**, 17018 (2017).

2. Courtine, G. & Sofroniew, M. V. Spinal cord repair: advances in biology and technology. *Nat. Med.* **25**, 898–908 (2019).

3. Anjum, A. *et al.* Spinal Cord Injury: Pathophysiology, Multimolecular Interactions, and Underlying Recovery Mechanisms. *Int. J. Mol. Sci.* **21**, 7533 (2020).

4. Zhang, K., Jiang, M. & Fang, Y. The Drama of Wallerian Degeneration: The Cast, Crew, and Script. *Annu. Rev. Genet.* **55**, 93–113 (2021).

5. Okada, S. *et al.* Conditional ablation of Stat3 or Socs3 discloses a dual role for reactive astrocytes after spinal cord injury. *Nat. Med.* **12**, 829–834 (2006).

6. Adams, K. L. & Gallo, V. The diversity and disparity of the glial scar. *Nat. Neurosci.* **21**, 9–15 (2018).

7. Noristani, H. N. *et al.* Spinal cord injury induces astroglial conversion towards neuronal lineage. *Mol. Neurodegener.* **11**, 68 (2016).

8. Wang, J.-J. *et al.* Molecular Expression Profile of Changes in Rat Acute Spinal Cord Injury. *Front. Cell. Neurosci.* **15**, 720271 (2021).

9. Milich, L. M. *et al.* Single-cell analysis of the cellular heterogeneity and interactions in the injured mouse spinal cord. *J. Exp. Med.* **218**, e20210040 (2021).

10. Moor, A. E. & Itzkovitz, S. Spatial transcriptomics: paving the way for tissue-level systems biology. *Curr. Opin. Biotechnol.* **46**, 126–133 (2017).

11. Williams, C. G., Lee, H. J., Asatsuma, T., Vento-Tormo, R. & Haque, A. An introduction to spatial transcriptomics for biomedical research. *Genome Med.* **14**, 68 (2022).

12. Asp, M. *et al.* A Spatiotemporal Organ-Wide Gene Expression and Cell Atlas of the Developing Human Heart. *Cell* **179**, 1647-1660.e19 (2019).

13. Maniatis, S. *et al.* Spatiotemporal dynamics of molecular pathology in amyotrophic lateral sclerosis. *Science* **364**, 89–93 (2019).

14. Peng, G. *et al.* Molecular architecture of lineage allocation and tissue organization in early mouse embryo. *Nature* **572**, 528–532 (2019).

15. Chen, W.-T. *et al.* Spatial Transcriptomics and In Situ Sequencing to Study Alzheimer's Disease. *Cell* **182**, 976-991.e19 (2020).

16. Ortiz, C. *et al.* Molecular atlas of the adult mouse brain. *Sci. Adv.* **6**, eabb3446 (2020).

17. Fawkner-Corbett, D. *et al.* Spatiotemporal analysis of human intestinal development at single-cell resolution. *Cell* **184**, 810-826.e23 (2021).

18. Rao, A., Barkley, D., França, G. S. & Yanai, I. Exploring tissue architecture using spatial transcriptomics. *Nature* **596**, 211–220 (2021).

19. Li, Z. & Peng, G. Spatial transcriptomics: new dimension of understanding biological complexity. *Biophysics Rep.* (2021) doi:10.52601/bpr.2021.210037.

20. Ratz, M. *et al.* Clonal relations in the mouse brain revealed by single-cell and spatial transcriptomics. *Nat. Neurosci.* **25**, 285–294 (2022).

21. Tavares-Ferreira, D. *et al.* Spatial transcriptomics of dorsal root ganglia identifies molecular signatures of human nociceptors. *Sci. Transl. Med.* **14**, eabj8186 (2022).

22. Monje, M. Myelin Plasticity and Nervous System Function. *Annu. Rev Neurosci.* **41**, 61–76 (2018).

23. Li, L. *et al.* The Specific Role of Reactive Astrocytes in Stroke. *Front. Cell Neurosci* **16**, 850866 (2022).

24. Willemsen, L. & Winther, M. P. Macrophage subsets in atherosclerosis as defined by single‐cell technologies. *J. Pathol.* **250**, 705–714 (2020).

25. Langfelder, P. & Horvath, S. WGCNA: an R package for weighted correlation network analysis. *BMC Bioinformatics* **9**, 559 (2008).

26. Macdonald, C. D. *et al.* Cytokine-induced cysteine- serine-rich nuclear protein-1 (CSRNP1) selectively contributes to MMP1 expression in human chondrocytes. *PLOS ONE* **13**, e0207240 (2018).

27. von Scheidt, M. *et al.* Transcription Factor MAFF (MAF Basic Leucine Zipper Transcription Factor F) Regulates an Atherosclerosis Relevant Network Connecting Inflammation and Cholesterol Metabolism. *Circulation* **143**, 1809–1823 (2021).

28. Aibar, S. *et al.* SCENIC: single-cell regulatory network inference and clustering. *Nat. Methods* **14**, 1083–1086 (2017).

29. Zhou, Q., Wang, S. & Anderson, D. J. Identification of a Novel Family of Oligodendrocyte Lineage-Specific Basic Helix–Loop–Helix Transcription Factors. *Neuron* **25**, 331–343 (2000).

30. Pierrot, N. Amyloid precursor protein regulates SREBP-mediated neuronal lipid homeostasis in mice and humans. *Alzheimers Dement.* **7**, S572 (2011).

31. Shaulian, E. & Karin, M. AP-1 in cell proliferation and survival. *Oncogene* **20**, 2390–2400 (2001).

32. Liu, T., Zhang, L., Joo, D. & Sun, S.-C. NF-κB signaling in inflammation. *Sig. Transduct. Target. Ther.* **2**, 17023 (2017).

33. Mogensen, T. H. IRF and STAT Transcription Factors - From Basic Biology to Roles in Infection, Protective Immunity, and Primary Immunodeficiencies. *Front. Immunol.* **9**, 3047 (2019).

34. Cable, D. M. *et al.* Robust decomposition of cell type mixtures in spatial transcriptomics. *Nat. Biotechnol.* **40**, 517–526 (2022).

35. Rosenberg, A. B. *et al.* Single-cell profiling of the developing mouse brain and spinal cord with split-pool barcoding. *Science* **360**, 176–182 (2018).

36. Köhler, S., Winkler, U. & Hirrlinger, J. Heterogeneity of Astrocytes in Grey and White Matter. *Neurochem. Res.* **46**, 3–14 (2019).

37. Jin, S. *et al.* Inference and analysis of cell-cell communication using CellChat. *Nat. Commun.* **12**, 1088 (2021).

38. Escartin, C. *et al.* Reactive astrocyte nomenclature, definitions, and future directions. *Nat. Neurosci.* **24**, 1–14 (2021).

39. Sidoryk-wegrzynowicz, M., Wegrzynowicz, M., Lee, E., Bowman, A. B. & Aschner, M. Role of Astrocytes in Brain Function and Disease. *Toxicol. Pathol.* **39**, 115–123 (2011).

40. Molofsky, A. V. *et al.* Astrocytes and disease: a neurodevelopmental perspective. *Genes Dev.* **26**, 891–907 (2012).

41. Stogsdill, J. A. *et al.* Astrocytic neuroligins control astrocyte morphogenesis and synaptogenesis. *Nature* **551**, 192–197 (2017).

42. Street, K. *et al.* Slingshot: cell lineage and pseudotime inference for single-cell transcriptomics. *BMC Genomics* **19**, 477 (2018).

43. Khan, S. IGFBP-2 Signaling in the Brain: From Brain Development to Higher Order Brain Functions. *Front. Endocrinol. (Lausanne)* **10**, 822 (2019).

44. Li, T. *et al.* IGFBP2: integrative hub of developmental and oncogenic signaling network. *Oncogene* **39**, 2243–2257 (2020).

45. Yau, S. W., Azar, W. J., Sabin, M. A., Werther, G. A. & Russo, V. C. IGFBP-2 - taking the lead in growth, metabolism and cancer. *J. Cell Commun. Signal.* **9**, 125–142 (2015).

46. Kaufmann, M. *et al.* Identification of early neurodegenerative pathways in progressive multiple sclerosis. *Nat. Neurosci.* **25**, 944–955 (2022).

47. Anderson, M. A. *et al.* Astrocyte scar formation aids central nervous system axon regeneration. *Nature* **532**, 195–200 (2016).

48. Gong, L. *et al.* Spatiotemporal Dynamics of the Molecular Expression Pattern and Intercellular Interactions in the Glial Scar Response to Spinal Cord Injury. *Neurosci. Bull.* (2022) doi:10.1007/s12264-022-00897-8.

49. Sathyamurthy, A. *et al.* Massively Parallel Single Nucleus Transcriptional Profiling Defines Spinal Cord Neurons and Their Activity during Behavior. *Cell Rep.* **22**, 2216–2225 (2018).

50. Zeisel, A. *et al.* Molecular Architecture of the Mouse Nervous System. *Cell* **174**, 999-1014.e22 (2018).

51. Li, C. *et al.* Temporal and spatial cellular and molecular pathological alterations with single-cell resolution in the adult spinal cord after injury. *Signal Transduct. Target. Ther.* **7**, 65 (2022).

52. Matson, K. J. E. *et al.* Single cell atlas of spinal cord injury in mice reveals a pro-regenerative signature in spinocerebellar neurons. *Nat. Commun.* **13**, 5628 (2022).

53. Denisenko, E. *et al.* Systematic assessment of tissue dissociation and storage biases in single-cell and single-nucleus RNA-seq workflows. *Genome Biol.* **21**, 130 (2020).

54. Yoon, H., Walters, G., Paulsen, A. R. & Scarisbrick, I. A. Astrocyte heterogeneity across the brain and spinal cord occurs developmentally, in adulthood and in response to demyelination. *PLOS ONE* **12**, e0180697 (2017).

55. Hasel, P., Rose, I. V. L., Sadick, J. S., Kim, R. D. & Liddelow, S. A. Neuroinflammatory astrocyte subtypes in the mouse brain. *Nat. Neurosci.* **24**, 1475–1487 (2021).

56. Renault-Mihara, F. *et al.* Spinal cord injury: Emerging beneficial role of reactive astrocytes' migration. *Int. J. Biochem. Cell Biol.* **40**, 1649–1653 (2008).

57. White, R. E. & Jakeman, L. B. Don't fence me in: harnessing the beneficial roles of astrocytes for spinal cord repair. *Restor. Neurol. Neurosci.* **26**, 197–214 (2008).

58. Lukovic, D. *et al.* Concise Review: Reactive Astrocytes and Stem Cells in Spinal Cord Injury: Good Guys or Bad Guys? *Stem Cells* **33**, 1036–1041 (2015).

59. Bradbury, E. J. & Burnside, E. R. Moving beyond the glial scar for spinal cord repair. *Nat. Commun.* **10**, 3879 (2019).

60. Fischer, I., Dulin, J. N. & Lane, M. A. Transplanting neural progenitor cells to restore connectivity after spinal cord injury. *Nat. Rev. Neurosci.* **21**, 366–383 (2020).

61. Zipser, C. M. *et al.* Cell-based and stem-cell-based treatments for spinal cord injury: evidence from clinical trials. *Lancet Neurol.* **21**, 659–670 (2022).

62. Okada, S., Hara, M., Kobayakawa, K., Matsumoto, Y. & Nakashima, Y. Astrocyte reactivity and astrogliosis after spinal cord injury. *Neurosci. Res.* **126**, 39–43 (2018).

63. Tsai, H.-H. *et al.* Regional Astrocyte Allocation Regulates CNS Synaptogenesis and Repair. *Science* **337**, 358–362 (2012).

64. Jeong, E. Y. *et al.* Enhancement of IGF-2-induced neurite outgrowth by IGF-binding protein-2 and osteoglycin in SH-SY5Y human neuroblastoma cells. *Neurosci. Lett.* **548**, 249–254 (2013).

65. Khan, S. *et al.* IGFBP2 Plays an Essential Role in Cognitive Development during Early Life. *Adv. Sci. (Weinh)* **6**, 1901152 (2019).

66. Caldwell, A. L. M. *et al.* Aberrant astrocyte protein secretion contributes to altered neuronal development in multiple models of neurodevelopmental disorders. *Nat. Neurosci.* **25**, 1163–1178 (2022).

67. Reshamwala, R. *et al.* Induction of Complete Transection-Type Spinal Cord Injury in Mice. *J. Vis. Exp.* **159**, 10.3791/61131 (2020).

68. Basso, D. M. *et al.* Basso Mouse Scale for Locomotion Detects Differences in Recovery after Spinal Cord Injury in Five Common Mouse Strains. *J. Neurotrauma.* **23**, 635–659 (2006).

69. Martinov, T., Mack, M., Sykes, A. & Chatterjea, D. Measuring Changes in Tactile Sensitivity in the Hind Paw of Mice Using an Electronic von Frey Apparatus. *J. Vis. Exp.* **82**, e512512 (2013).

70. Peng, G. *et al.* Spatial Transcriptome for the Molecular Annotation of Lineage Fates and Cell Identity in Mid-gastrula Mouse Embryo. *Dev. Cell* **36**, 681–697 (2016).

71. Güler, B. E., Krzysko, J. & Wolfrum, U. Isolation and culturing of primary mouse astrocytes for the analysis of focal adhesion dynamics. *STAR Protoc.* **2**, 100954 (2021).

72. Hao, Y. *et al.* Integrated analysis of multimodal single-cell data. *Cell* **184**, (2021).

73. Hafemeister, C. & Satija, R. Normalization and variance stabilization of single-cell RNA-seq data using regularized negative binomial regression. *Genome Biol.* **20**, 296 (2019).

74. Yu, G., Wang, L.-G., Han, Y. & He, Q.-Y. clusterProfiler: an R Package for Comparing
Biological Themes Among Gene Clusters. *OMICS.* **16**, 284–287 (2012).

75. Langfelder, P. & Horvath, S. WGCNA: an R package for weighted correlation network
analysis. *BMC Bioinformatics* **9**, 559 (2008).

76. Shannon, P. Cytoscape: A Software Environment for Integrated Models of Biomolecular
Interaction Networks. *Genome Res.* **13**, 2498–2504 (2003).

77. Suo, S. *et al.* Revealing the Critical Regulators of Cell Identity in the Mouse Cell Atlas.
Cell Rep. **25**, 1436-1445.e3 (2018).

78. Bass, J. I. F. *et al.* Using networks to measure similarity between genes: association
index selection. *Nat. Methods* **10**, 1169–1176 (2013).

79. Stuart, T. *et al.* Comprehensive Integration of Single-Cell Data. *Cell* **177**, 1888-1902.e21
(2019).

80. Cable, D. M. *et al.* Cell type-specific inference of differential expression in spatial
transcriptomics. *Nat. Methods* **19**, 1076–1087 (2022).

81. Csárdi, G. & Nepusz, T. The igraph software package for complex network research.
InterJournal, complex systems **1695**, 1–9 (2006).

Figure 4 - SCI ST