Specific pre-plasma cell states and local proliferation at the dark zone - medulla interface characterize germinal center-derived plasma cell differentiation in lymph node

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ABSTRACT

High affinity antibody-producing plasma cells (PC) generated in germinal centers (GC) are crucial for durable immunity after vaccination or infection. The selection of high affinity B cells in the GC light zone instructs PC differentiation in a subset of cells, but the cellular transitions and spatial organization of GC-to-PC differentiation remain poorly understood. Here, we have used a mouse model to track GC-derived B cells with integrative single-cell and spatial analyses in draining lymph node after immunization. We first identified putative PrePC cells in scRNA-seq datasets, then enriched those cells through their specific surface phenotype for further analysis of their gene expression trajectories and BCR repertoire. We found a continuum of actively proliferating transitional states bridging selected LZ GC B cells and recently exported PCs, with gradually increasing levels of endoplasmic reticulum stress-associated genes and Ig transcripts. Spatial analyses revealed that recently differentiated PC continued their maturation and affinity-restricted proliferation at a previously uncharacterized interface between the DZ and extensions of the lymph node medulla. Our findings provide insights into the intermediate stages and microenvironmental factors involved in the differentiation of GC B cells into PC, with implications for vaccine development and understanding antibody responses.

KEYWORDS

Germinal center, plasma cell, lymph node, differentiation trajectory, localization, transitional states

SHORT RUNNING TITLE

Mapping GC to PC differentiation

INTRODUCTION

B cell immune responses are important for long term protection against all kinds of pathogens, after natural infection or vaccination. Germinal Centers (GCs), micro-anatomical structures that form within B cell follicles in secondary lymphoid organs after T cell-dependent B cell activation, play a crucial part for long term antibody-based immunity. In GCs, the affinity maturation cyclic process enables the diversification and overall gain in affinity of antigen-specific antibodies expressed as surface B cell receptors (BCR) by GC B cells. In a maturation cycle, B cells first undergo cell division and somatic hypermutation (SHM) in the dark zone (DZ) in order to diversify their BCR before migrating to the light zone (LZ) and being selected based on the affinity of their B-cell receptor (BCR). Most studies have shown that selection broadly follows the affinity-dependent selection model[1]. In this model, the really low affinity B cells die by apoptosis, the low affinity ones differentiate in memory B cells (MBC), the intermediate ones recycle in the DZ and the high affinity ones differentiate in PC[1–4].

The differentiation of GC-derived MBC and plasma cells (PC) results in diversified and affinity-enhanced antibody specificities being expressed in long-lived cell types for long-term immune protection. In that regard, GC-derived PC have been shown to be the primary source of long-lived high-affinity antibody producing cells that home to the bone marrow[5]. The differentiation of GC B cells into long-lived PC results from the induction of gene expression modifications through different signals. First, GC B cells have to internalize the antigen contained in immune complexes at the FDC membrane before they present it to T follicular helper cells (T_{FH}) on MHC class II molecules and receive co-stimulatory signals. The BCR signal is also important in itself thanks to the repression of *Bcl6* transcription[6,7], responsible for the GC profile, and the induction of *Irf4* expression[8,9]. The BCR affinity allows to be more competitive and to extract more antigen at the FDC surface, which has been correlated with a better help from T_{FH} cells [10–13]. These cells allow the selection of GC B cells through a competitive access to co-stimulatory signals such as CD40L and IL21. CD40 signaling reinforces BCR signaling through the blockade of *Bcl6* expression and the induction of *Irf4* in a dose-dependent manner[14–16]. Finally, IL21 seems to have a dual role in the GC. It induces the expression of Bcl6, but when it synergizes with the BCR signal and the CD40 signal, it allows the expression of Irf4 and Blimp1 (encoded by *Prdm1*), two markers important for terminal PC differentiation[17–20].

The precise mapping of stepwise processes for PC differentiation has been mostly studied in *ex vivo* differentiation assays starting from naïve B cells or MBC. Although the situation may differ for the GC to PC differentiation *in vivo*, those assays have generated important findings at the genetic level. Notably, because some GC markers such as Bcl6 or Pax5 repress the PC phenotype[6,21], it is then mandatory for B cells to downregulate the expression of the GC profile in order to differentiate in GC-derived PC[9,22]. Once B cells have switched their gene expression profile towards a PC profile, they undergo several rounds of cell division that allow the hypomethylation of the chromatin parts encoding the proteins responsible for the PC profile in an irreversible way[23]. Thus, PC differentiation goes through a remodeling at the epigenetic, transcriptomic and phenotypic profile. These changes allow the functional remodeling of activated and proliferating GC B cells to antibody-secreting PC.

Several studies have tried to identify intermediate states bridging GC B cells to PC *in vivo*. Indeed, cells intermediate in their transcriptomic profile[24], or LZ GC B cells with PC features such as high expression of *Irf4* in mice[16,25], have been described in recent years. In humans, transcriptomic studies also highlighted intermediate states expressing PC and cell cycle markers[26–28]. Unfortunately, the scarcity of these states in lymphoid organs have most often precluded their in-depth study and the differentiation trajectory between selected GC B cells and differentiated antibody-producing cells remains uncharted. *In silico* modeling and imaging[29] and forced selection in *in vivo* models[4] have suggested that PC exit from the GC occurred through or proximal to the DZ. Several signals appear to be implicated in PC egress from GC (GPR183 signaling, S1P1 and CXCL12 gradients), but no final mechanism has been identified[30,31]. During early T-dependent responses, plasmablasts accumulate at the GC-T zone interface (GTI), but whether this also occurs for GC-derived PC is not known. PC that are found in the LN medulla, further away from the GC DZ, express less proliferation markers (Ki67$^+$) than those that are closer to the DZ, indicating a possible spatio-temporal axis of GC to PC differentiation[32]. In addition, it was recently shown that post-selection PC expand clonally without further SHM[33], but whether this expansion occurs inside or outside of the GC, or where in the dLN, is still unknown.

Here, we identified GC-to-PC intermediate differentiation states in mouse draining LN after model antigen immunization, that we characterized through integrative analyses of phenotypes, transcriptomes, BCR

repertoires, and microanatomic localization. We show that GC-derived PC differentiation implies local
proliferation at the interface between the GC DZ and the LN medulla.

RESULTS

Tracking GC B cells and recent GC emigrants after model immunization

We used a GC B cell fate mapping mouse model, *Aicda-Cre-ERT2 x Rosa26-lox-STOP-lox-eYFP*[34], to track
GC B cells and their recent progeny in draining lymph nodes after subcutaneaous immunization. After
immunization with the T-dependent model antigen 4-hydroxy-3-nitrophenylacetyl-conjugated keyhole lympet
hemocyanin (NP-KLH) in adjuvant, we gavaged the mice once with tamoxifen to induce the eYFP expression
in *Aicda* expressing cells, and analyzed cells in draining lymph nodes (dLN) by flow cytometry at day 13 after
immunization, 2 to 6 days after tamoxifen gavage (**Figure 1a**). In that model, gavaging with tamoxifen after
day 6 ensures that the eYFP expression is triggered almost exclusively in GC B cells. Thus, among activated
IgD^{neg} B cells, we detected eYFP expression in CD138$^+$ PC (**Figure 1b**), GL7$^+$CD38$^-$ GC B cells, and
CD38$^+$GL7$^-$ memory B cells (MBC) (**Figure 1c**). While tamoxifen-induced labeling was already maximal 2
days after gavage in GC B cells, GC-emigrant eYFP$^+$ MBC and PC were barely detectable at that time, but
increased gradually after 4 to 5 days (**Figure 1d**). We then assessed the proportion of NP-binding cells
among IgD^{neg} B cell subsets in dLN at day 14 after NP-KLH immunization and 4 days after tamoxifen gavage
(**Figure 1e-f**). Total GC B cells and PC included on average 20% and 30% NP-PE$^+$ cells, respectively, while
MBC included less than 5% NP-binding cells (**Figure 1e**). Restricting the analysis to eYFP$^+$ cells resulted in
higher proportions of NP-binding cells in all subsets, including MBC where 15% on average were NP-PE$^+$
(**Figure 1f**), indicating that PC and MBC exported from the GC from day 10 to day 14 were enriched in cells
that could detectably bind antigen. Thus, the *Aicda-Cre-ERT2 x Rosa26-lox-STOP-lox-eYFP* mouse model
enabled us to track GC output of PC and MBC in dLN after subcutaneous model antigen immunization.

Single-cell RNA-seq analysis of GC B cells and recent GC emigrants identifies putative PrePC

Using the GC B cell fate mapping mouse model in different immunization schemes with either the chicken
ovalbumin (OVA) or NP-KLH model antigens (**Figure 2a**), we generated 3 distinct single-cell RNA-seq
(scRNA-seq) datasets of IgD^{neg} eYFP$^+$ B cells from dLN. In dataset #1, mice were primed with OVA in Alum,
and we investigated cells 4 days after tamoxifen gavage, at day 10 and day 20 of either the primary response,
or the secondary response after boosting, using the FACS-based 5-prime-end sequencing (FB5P-seq)
method for integrative scRNA-seq analysis. In dataset #2, mice were primed with NP-KLH in Sigma Adjuvant
System, and we investigated cells 5 days after tamoxifen gavage, at day 16 of the primary response, using
droplet-based 5'-end scRNA-seq. In dataset #3, mice were primed with NP-KLH in Sigma Adjuvant System,
and we investigated cells 3 days after tamoxifen gavage, at day 10, day 16 and day 22 of the primary
response, using droplet-based 5'-end scRNA-seq. After standard quality controls of the resulting scRNA-seq
datasets, we integrated all experiments in a single low dimensional UMAP embedding (**Supplementary
Figure 2a-b**) and defined gene expression-based subsets by non-supervised clustering (**Figure 2b**). Based
on marker genes expression (**Figure 2c**), cluster 0 included mostly quiescent cells and corresponded to LZ
GC B cells (*Cd83, Il4i1*); clusters 1, 2, 3 and 7 expressed distinct combinations of cell cycle associated genes
and corresponded to DZ GC B cells; cluster 5 expressed the typical Myc-induced signature of positively
selected LZ GC B cells (*Npm1, C1qbp*); cluster 4 included mostly quiescent cells and expressed both LZ and
MBC markers (*Klf2, Serpinb1a*), suggesting those cells corresponded to preMBC or early MBC; and cluster
6 expressed high levels of Ig coding genes and corresponded to PC. Although there was a clear continuum
between quiescent LZ GC B cells and preMBC in the low dimensional embedding, the PC cluster was
completely separated and we failed to identify transitional states from GC to PC through non-supervised
analyses in that dataset.

We reasoned that a supervised approach may be better able to identify putative PrePC that would bridge the
GC and PC transitional states. We thus used a gene signature-based scoring approach to hierarchically
"gate" cells in the integrated scRNA-seq dataset (Methods, **Figure 2d** and **Supplementary Figure 2c**). In
particular, we identified rare cells expressing intermediate levels of a PC-specific gene signature that were
embedded in the GC "continent" of the UMAP embedding, which we named "PrePC" (**Figure 2e**). Other
supervised annotations, DZ, LZ, LZtoDZ, preMBC and PC were consistent with non-supervised clusters
(**Figure 2e** and **Supplementary Figure 2d**). Putative PrePC represented very low frequencies (average
0.2%) of sequenced IgD^{neg} eYFP cells in all time points of all datasets (**Figure 2f**). We computed marker
genes of the supervised annotated cell subsets in the 3 datasets (**Supplementary Figure 2e**). PreMBC, LZ,

DZ, LZtoDZ and PC expressed the expected gene expression programs that have already been described in other mouse GC B cell scRNA-seq datasets[35,36]. Putative PrePC expressed *Bst2*, *H2-Q7*, *Glipr1*, *Selplg*, *Itgb7*, *Plac8*, *Cd22*, *Usp8* and *Serpina3g* among other marker genes (**Supplementary Figure 2e**). In particular, high expression of *Psgl1* and *Bst2* (**Figure 2g**), encoding the surface markers PSGL1 and BST2, suggested that PrePC cells may be identified and enriched by flow cytometry.

Enrichment of putative PrePC by flow cytometry

We thus designed a 14-color flow cytometry panel targeting surface markers and transcription factors characteristic of GC, MBC and PC, and including the PSGL1 and BST2 markers, which we applied to analyze IgD[neg] B cells in dLN of *Aicda-Cre-ERT2 x Rosa26-lox-STOP-lox-eYFP* mice previously immunized with NP-KLH and gavaged with tamoxifen (**Figure 3a**). UMAP embedding, based on 10 surface and intracellular markers (Methods) identified clusters of phenotypically defined GC, MBC and PC, as well as cells located in intermediate areas of the low dimensional embedding (**Figure 3b**). We gated cells situated within GC and PC clusters as "GC-to-PC" and inspected their surface phenotype in comparison with the well-defined GC, MBC and PC clusters. GC-to-PC cells expressed high levels of the GC markers GL7, CD19 and B220 markers, high levels of the PC-specific transcription factor IRF4, intermediate levels of the PC marker CD138, and intermediate levels of PSGL1 (**Figure 3c**). Based on the specific surface phenotype of GC-to-PC cells, we thus reverse engineered a gating strategy (**Figure 3d**) that allowed for the identification and quantification of PrePC cells (**Figure 3e**). At days 14-16 after primary immunization with NP-KLH, PrePC accounted for approximately 0.08% of IgD[neg] B cells in dLN (**Figure 3f**), included significantly lower proportions of NP-binding cells (18% on average) compared to GC B cells (23%) and PC (24%), which was still much higher than the proportion of MBC detectably binding NP (2%) (**Figure 3g**). The proportions of PrePC were stable for all time points tested after primary immunization; we found a significant tendency for increased frequencies of PrePC during secondary responses as compared to primary responses (**Supplementary Figure 3**).

Characterization of PrePC cells in the GC-to-PC differentiation continuum

In order to gain more insight into the molecular features of PrePC cells, we next used our new gating strategy to enrich and sort phenotypically defined PrePC cells and compare them directly to DZ, LZ and PC in FACS-based 5-prime-end single-cell RNA-seq (FB5P-seq)[37] (**Figure 4a**). After quality control and cell cycle regression of the resulting dataset, low-dimensional UMAP embedding displayed a clear separation between DZ and LZ GC B cells on one side, and PC cells on the other side, with a subset of phenotypically defined PrePC cells bridging the two cell continents (**Figure 4b**). Consistent with our previous analysis on non-enriched cells (**Supplementary Figure 2e**), single PrePC combined the expression of GC B cell signature genes *Ms4a1*, *Cd19*, *Irf8*, with the expression of positive-selection induced gene *Myc*, and the expression of PC differentiation surface markers and transcription factors *Sdc1*, *Prdm1*, *Irf4*, *Selplg* and *Bst2* (**Figure 4c**). Those cells were also actively proliferating, either in S or G2/M phase of the cell cycle (**Figure 4d**).

PC differentiation is characterized by the production of high amounts of antibodies, requiring high level of Ig genes transcription and the induction of a specific endoplasmic reticulum (ER) stress response[31]. PrePC cells expressed intermediate levels of genes involved in the ER stress response (**Supplementary Figure 4a**), and intermediate levels of Ig transcript counts (**Supplementary Figure 4b**), when compared with GC B cells and PC. We thus defined a continuum of GC-to-PC differentiation based on single-cell gene expression of the ER stress module and Ig transcripts counts (**Figure 4e-g**), in which most phenotypically defined PrePC bridged the gap between GC B cells and PC. Another feature of PC differentiation, the gradual loss of antigen-presentation capacity on MHC-II[38], was also intermediate in phenotypically defined PrePC cells (**Figure 4h** and **Supplementary Figure 4c-d**). Functionally, FACS-sorted PrePC spontaneously secreted detectable amounts of soluble IgG in *ex vivo* cultures (**Figure 4i**). Thus, PrePC cells with the GL7+ B220+ CD138int PSGL1+ phenotype were engaged in the early stages of GC-to-PC differentiation and had already initiated the antibody production program.

The distribution of phenotypically defined PrePC cells along the GC-to-PC differentiation continuum showed that those cells were also heterogeneous, and could be further divided based on their degree of differentiation (**Figure 4j-k**). PrePC are very rare intermediates in non-enriched datasets (*e.g.* **Figure 2**), but our enriched dataset provided a unique opportunity to finely map the transcriptional changes that pave the GC-to-PC differentiation. We thus clustered genes according to the dynamic evolution of their expression as cells progress through the GC-to-PC continuum (**Figure 4l**), and computed the gene ontology enrichment of gene modules that are progressively lost (clusters 1-3) or induced (clusters 4-6) through differentiation (**Figure 4m**). Those analyses revealed that very early metabolic reprogramming (nucleotide biosynthesis process,

cellular respiration, oxidative phosphorylation, proton motive force−driven mitochondrial ATP synthesis)
preceded the gradual increase in antibody production-associated physiological responses (response to ER
stress, response to unfolded protein).

BCR analysis of PrePC cells

The affinity maturation trajectory of GC B cells and their progeny is in-part traceable through the analysis of
their Ig gene transcripts[39]. Our PrePC-enriched FB5P-seq analysis retrieved IgH and IgL sequences for the
vast majority of single cells (n=965/1367) For better resolution of the different gene expression programs and
their associated BCR sequence features, we re-annotated the cell types and states using the supervised
annotation method described in **Supplementary Figure 2c**, resulting in the identification of DZ, LZ, LZtoDZ,
PC, PreMBC and PrePC cells with specific gene expression programs and phenotypes (**Supplementary
Figure 4e**). While close to 90% of DZ, LZ and LZtoDZ subsets carried mutations in their IGH and IgL variable
genes, only 75% of preMBC and 40% of PrePC and PC were mutated (**Figure 5a-b**). Unmutated and mutated
PrePC and PC did not differ in their gene expression programs as testified by low-dimensional UMAP
embedding (**Figure 5c**), but were slightly different in isotype usage, with unmutated cells being more
frequently IgM-positive (**Figure 5d**). IgM-expressing cells represented approximately 25% of both DZ and LZ
cells, less than 5% of positively selected LZtoDZ cells, and about 60% of preMBC; in unmutated PC and
PrePC, 30% were IgM-positive, as compared to 5% and 10%, respectively, for their mutated counterparts.

The supervised annotation showed that PrePC were split into two subsets: one expressing intermediate
levels of ER stress response genes and the signature of recent positive selection; and the other, mostly
quiescent, expressing low levels of ER stress response genes and no sign of activation (**Figure 5e**). We re-
annotated those PrePC subsets as PrePC.Act and PrePC.ERstress[lo] (**Figure 5f-g**) and re-analyzed their
surface phenotype and BCR mutations. PrePC.Act cells had the typical PrePC phenotype identified in this
study (GL7[+] CD138[int] B220[+] CD19[+]), while PrePC.ERstress[lo] were phenotypically heterogeneous and
analogous to PC, suggesting that the latter were more advanced in their differentiation (**Figure 5h**). BCR
mutations did not differ between the two PrePC subsets (**Figure 5i**). Finally, we identified BCR clonotypes in
the dataset and analyzed the compositions in distinct cell types and states of clonal families (groups of 2 or
more cells sharing the same IgH and IgL clonotype). We first annotated clonal families depending on whether
they contained cells from one or several of the cell states defined by our supervised annotation; then we
analyzed for each cell state the proportion of cells being allocated to distinct clonal families (**Figure 5j**). About
half of DZ and LZ cells were in clonal families comprising either only GC B cells (DZ, LZ and/or LZtoDZ), GC
and PC, or GC, PC and PrePC cells. By contrast, LZtoDZ cells were more frequently clonally associated to
GC, PrePC and PC, and PreMBC were most often not clonally related to any other cell in the dataset. As
expected for transitional states in the GC-to-PC differentiation, PrePC.Act and PrePC.ERStress[lo] cells were
often clonally related to both GC and PC cells. The sizeable proportion of PrePC cells with no or very few
mutations in their IgH and IgL/K variable gene suggested that those had not been through affinity maturation
and had been generated in a largely GC-independent process. In support of that hypothesis, PrePC and PC
cells that were clonally related to contemporaneous GC B cells had significantly more somatic hypermutation
load than those that were not (**Figure 5k-l**).

Overall, the integration of gene expression, surface phenotype and BCR sequence measurements in single
cells enriched in putative PrePC enabled the fine characterization of the rare intermediate transitional states
in the GC-to-PC differentiation.

Recent GC-derived PC proliferate at the DZ-medulla interface in an affinity restricted manner

Some of the genes detected to be specifically induced from the PrePC state encoded chemotaxis (*Cxcr3*,
Gpr183) or adhesion (*Itga4*, *Itgb1*) receptors (**Figure 6a**), suggesting that PrePC cells may be localized in a
specific compartment in the dLN. To assess the location of GC-derived PC differentiation in our model, we
first performed whole transcriptome spot-based spatial transcriptomics (10x Genomics Visium) analysis of 4
sections of the same 2 dLN, separated by approximately 50μm in depth (**Figure 6b**). After quality control, we
annotated spots with a semi-supervised approach: first using non-supervised spot clustering and marker
genes to identify spots corresponding to B follicles, T zones, medulla, GC LZ and GC DZ; then defining three
zones at the interface of the GC (Methods), the GC-T zone interface (GTI), the GC-B follicle interface (GBI),
and the GC-medulla interface (GMI) (**Figure 6b-c**). We then scored a PC gene expression signature on every
spot as a proxy for detecting the presence of PC in the distinct dLN microenvironments and interfaces. The
PC score was high in medulla, as expected since PC are known to reside in medullary chords before dLN
exit[40]; among GC-proximal interfaces, GMI spots were the main areas with high PC scores (**Figure 6d**). This
result suggested that stromal components of the medulla that support PC survival extend to the direct

proximity of the GC, at the DZ side. Accordingly, several stromal cell subsets gene expression signatures[41] also scored highly in the medulla and GMI areas (**Supplementary Figure 5a-b**).

We thus turned to high-resolution microscopy to precisely map the location and interactions of eYFP$^+$ IRF4$^+$ cells, which we defined as recently differentiated PC. Consistent with our spatial transcriptomics results, we found clusters of eYFP$^+$ IRF4$^+$ cells accumulating in collagen-IV-rich areas in close proximity to the GC DZ (**Supplementary Figure 5c**). At the GMI, eYFP$^+$ IRF4$^+$ cells were surrounded by Lyve1$^+$ lymphatic endothelial cells and PDGFR1$^+$ stromal cells (**Figure 6e**). The density of eYFP+ IRF4+ cells in the GMI and medulla was similar (**Figure 6f**) Although on some sections, the eYFP$^+$ IRF4$^+$ cells in close proximity to the GC DZ seemed to express lower levels of CD138 than those more distal in the medullary areas (**Figure 6e**, panel *iv*), suggesting less advanced differentiation, that observation was not statistically significant when considering all GCs from multiple dLN sections (**Figure 6g**). Since we did not find a GMI area surrounding every GC in every tissue section, we reasoned that the medulla may connect with the GC DZ only at certain contact areas that are not always captured in a 5-10 µm tissue section. We thus run whole-LN 3D-imaging using light sheet microscopy, and could indeed show that every GC in a dLN included a GMI contact area, but that the area was not found at all depths (**Figure 6h** and **Supplementary Movie**).

Our scRNA-seq data and recently published observations[33] indicated that early differentiating PrePC, or recently differentiated PC proliferate extensively. *In situ*, we showed that a majority of eYFP$^+$ IRF4$^+$ at the GMI expressed Ki67, a hallmark of active proliferation (**Figure 6i**). In order to test whether there could be a positive selection checkpoint from the PrePC to PC stage, we measured the proliferative state (Ki67 expression measured by intracellular flow cytometry) of PrePC and PC cells, comparing NP-PE-positive and NP-PE-negative cells. We found that the vast majority of PrePC cells expressed high levels of Ki67, regardless of NP-PE binding capacity, similar to GC B cells; by contrast, the proportion of Ki67+ cells among NP-PE-positive PC was significantly higher than among NP-negative PC (82 ± 9 % vs. 47 ± 16 %, p = 0.0006 Mann Whitney two-tailed test) (**Figure 6j**). That difference in proliferation between NP-PE-positive and -negative cells was also significant when studying post-GC MBC. Overall, this suggests that an affinity-based positive selection checkpoint controls the proliferation of post-GC PC after the PrePC intermediate stage.

Overall, we have characterized the GMI, a specific dLN microenvironment defined as an extension of the medulla contacting the GC DZ at some level, as the main environment for exit and proliferation of GC-derived PC.

DISCUSSION

Our study provides a fine characterization of rare PrePC intermediate states in the GC-to-PC differentiation trajectory in draining LN after immunization. We showed that PrePC cells expressed a mixed phenotype expressing both GC and early PC markers in phenotypic and transcriptomic analyses; they were also actively cycling, expressed intermediate levels of MHC-II genes, had initiated metabolic reprogramming and started upregulating UPR programs, and produced antibodies *ex vivo* without any additional signal. Those data indicate that we have identified late intermediate states in the GC-to-PC differentiation trajectory. As PrePC spanned different unsupervised clusters in gene expression space, we analysed whether multiple PrePC cell states existed. PrePC could be subdivided based on either their BCR mutational profile, or their gene expression profile. In the latter, we separated ERstress[low] PrePC from activated PrePC which expressed a gene expression profile (Myc-induced genes) and phenotype (GL7[hi] CD138[lo] B220[int] CD19[hi]) related to selected LZ GC B cells. PrePC also expressed a distinctive repertoire of homing markers, suggesting they had already homed or were homing to extra-GC areas in draining LN. We found recently GC-exported PC proliferating just outside of the GC DZ, at the interface with medullary tissue microenvironments that connected every GC to the deep medulla. The GC-medulla interface contained Lyve1[+] and PDGFR1[+] stromal cells that may serve as a maturation niche for cells in the GC-to-PC differentiation trajectory.

Several studies have identified PrePC states as LZ GC B cells expressing high levels of *Irf4* in mice[42,43], and cells expressing both PC and cell cycle markers in humans[5,10,44]. Those studies lacked other information such as their surface phenotype, function, and *in situ* localization, making it difficult to compare them directly with our definition of PrePC. Recent studies in human tonsils has identified pre-plasmablasts that appeared integrated in the differentiation trajectory from GC to PC[26,27,45]. Indeed, the authors found cells that were intermediate in their transcriptomic profile, and shared clones with both GC B cells and PC. Even though the authors did not focus their analyses on these states, we could identify some common features with the PrePC we have identified in our study in mice (*e.g.* expression of PC lineage genes like *IRF4*, *PRDM1*, *XBP1*, *FKBP11*, and cell cycle progression), highlighting some evolutionary conserved modules activated at the GC-to-PC transition. In another high resolution human tonsil atlas[46], GC-derived PC branched both from LZ GCB cells and activated LZ-to-DZ B cells, the latter representing the likely counterparts of the activated PrePC we describe here in the mouse LN. Another study characterized human PrePC in an *ex vivo* MBC differentiation model using a combination of scRNAseq and ATACseq[47]. Although they originated from blood MBC, and not LN GC B cells, PrePC in that study were highly heterogenous and integrated in a continuum of differentiation with stepwise induction of the unfolded protein response, and with AP-1 family transcription factors like BATF playing a likely role at the onset of differentiation. In our study, PrePC branched from the LZ-to-DZ GC B cell state, where the transcription factor BATF has been shown to play a key role for metabolic refuelling induced by T cell help[48]. Altogether, our current results on PrePC are consistent with previous studies, refine the molecular, phenotypic and BCR repertoire characterization of those rare intermediate states, and introduce a gating strategy for enriching those cells from mouse lymph node by flow cytometry, which should be an asset for future studies.

In silico modeling experiments have suggested that PC exit the GC from the DZ[29]. Several homing receptors appear to be implicated in PC egress from GC (Gpr183, S1PR1, CXCR4), but the precise route that PrePC or PC use to exit GCs remained unidentified[30,31]. Based on gene expression, we found that PrePC were likely in the process of egress from the GC, but their scarcity precluded their robust identification by confocal microscopy in LN tissue sections. Instead, we analyzed recent GC emigrant PC (RGCE-PC) that were eYFP[+] IRF4[+] 3-4 days after tamoxifen induction of eYFP expression in GC B cells. RGCE-PC accumulated at the border of the GC. In previous studies, PC differentiated after early GC-independent B cell activation accumulated at the interface between the nascent GC and the T zone[41,42]. In our study, we showed that RGCE-PC were in fact accumulating and proliferating in DZ-proximal microenvironments characterized by a gene expression and cellular composition similar to the LN medulla. Notably, the GC-medulla interface was surrounded by Lyve1[+] cells resembling medullary sinus LEC[49], and contained numerous PDGFR1[+] stromal cells with high gene expression scores for previously described medullary stromal cell subsets (Nr4a1[+] and Inmt[+] stromal cells)[41]. Medullary fibroblastic reticular cells have been shown to provide essential survival and maturation cues to plasma cells in lymph nodes[50], but the fact that those cells may be in such a close proximity to the GC DZ has remained unnoticed in imaging studies focusing on B follicle stromal remodeling[51,52]. We attribute that to the fact that most 2D images of thin tissue sections failed to capture the GMI, which bridged medulla and GC DZ only at certain z-depth for a given GC structure. We did not detect GMI areas in non-immunized lymph nodes (unpublished observation), and other studies have shown a tight separation between medulla and B cell follicles in naïve LN[50,53]. However, adjuvanted immunization or infection induces

lymphangiogenesis and stromal remodeling[50,54] close to activated B cell follicles, which may initiate the
spontaneous organization of a GMI area that supports early PC proliferation, maturation and migration.

In our experiments, close to 25% of LZ-to-DZ cells were clonally related to contemporaneous PC or PrePC,
compared to less than 10% for LZ or DZ GC B cells (**Figure 5j**), suggesting that PrePC derive more closely
from GC B cells clones selected at the LZ-to-DZ stage. Thus, our study implies that, after being selected in
the LZ, a fraction of LZ-to-DZ state GC B cells commit to the PrePC state, proliferate at the external border
of the DZ and mature locally at the GC-medulla interface. For selected GC B cells recycling into the DZ
compartment, it has been shown that higher affinity induced more cycles of proliferation in recycled DZ B
cells[55,56]. Our analyses suggest that affinity-based selection also translates into a proliferative advantage of
high-affinity post-GC PC differentiating at the GMI, potentially explaining why loose affinity thresholds at the
LZ selection stage[33] may result in global affinity maturation of circulating antibody responses.

EXPERIMENTAL METHODS

Mouse models

Aicda-Cre-ERT2 x Rosa26-lox-STOP-lox-eYFP mice[34] were bred at the Centre d'Immuno-Phenomique, (Marseille, France), and transferred to the animal care facility of Centre d'Immunologie de Marseille-Luminy for experiments. C57BL/6 mice were purchased from Janvier Labs (Le Genest-Saint-Isle, France). All mice were maintained in the CIML mouse facility under specific pathogen-free conditions. Experimental procedures were conducted in agreement with French and European guidelines for animal care under the authorization number APAFIS #30945-2021040807508680, following review and approval by the local animal ethics committee in Marseille. Mice were used regardless of sex, at ages greater than 7 weeks and less than 3 months.

Mice were immunized with either 100µg of NP-KLH at 1µg/µL emulsified with Sigma Adjuvant System (SAS) at a 1:1 (v:v) ratio, or 100µg chicken ovalbumin (OVA) at 1µg/µl emulsified with Alum at a 1:1 (v:v) ratio, subcutaneously at the base of the tail, 50µl on each side. For induction of the Cre-ERT2-mediated labelling, we gavaged the mice once with 5mg of tamoxifen (TS648-1G, Sigma) in 200µL of peanut oil (P2144-250 ML, Sigma), at least 6 days after immunization. Mice were euthanized between 10 days and 21 days post-immunization (prime or boost) according to the experiments.

Flow cytometry

Single-cell suspensions from draining lymph nodes were washed and resuspended in FACS buffer (5% fetal calf serum, 2mM EDTA, 5% Brilliant Stain Buffer Plus in PBS 1X) at a concentration of 100 million of cells per ml. Cells were first incubated with FcBlock (Biolegend) for 10 min on ice. Then, cells were incubated with a mix of antibodies (see table1 below) conjugated with fluorochromes 30 min on ice. Cells were washed in PBS, and incubated with the Live/Dead Fixable Aqua Dead Cell Stain (Thermofisher) for 10 min on ice. Cells were then washed again in FACS buffer and resuspended in 2% paraformaldehyde during 50min to preserve the eYFP contained in the cytoplasm. Cells were washed and permeabilized using the FoxP3 permeabilization kit (eBioscience) during 30min, then washed again in the permeabilization buffer and incubated with intracellular antibodies for 45min at RT. Cells were finally washed in permeabilization buffer and resuspended in FACS buffer. Cell suspensions were analyzed on the LSRII UV cytometer (Becton Dickinson).

mAb	Clone	Manufacturer	Reference
Ki67 - BUV395	B56	BD	564071
PSGL1 (CD162) - BUV737	2PH1	BD	741796
GL7 - Pacific Blue	GL7	Biolegend	144614
CCR6 (CD196) - BV605	29-2L17	Biolegend	129819
CD86 - BV650	GL-1	Biolegend	105035
CD138 - BV711	281-2	BD	563193
CD19 - BV785	6D5	Biolegend	115543
anti-GFP (cross-react eYFP) – AF488		Thermofisher	A-21311
NP - PE		Home made	
B220 - PE-Cy5	RA3-6B2	Biolegend	103209
CXCR4 - PE-Cy7	2B11	Thermofisher	25-9991-80
IRF4 - AF647	IRF4.3E4	Biolegend	646409
CD38 - AF700	90	Thermofisher	56-0381-82
CD3 - APC-Cy7	17A2	BD	560590
Gr1 (Ly6G) - APC-Cy7	1A8	BD	560600
IgD - APC-Cy7	11-26c.2a	Biolegend	405716
Bst2 (CD317) - AF647	129c1	Biolegend	127106

For cell sorting, cells were pre-enriched using the "Pan B Cell Isolation Kit II mouse" enrichment kit from
Miltenyi Biotec (ref. 130-104-443) in which we added a biotinylated anti-IgD antibody for further enrichment
of IgD[neg] B cells (Biolegend ref. 405734). Cells were prepared as mentioned in the protocol provided and
passed through LS columns according to the manufacturer's instructions. We collected the negative fraction
and processed the cells according to the classical extracellular staining protocol, using antibodies described
in the table below. Cells were sorted on the BD Influx™ Cell Sorter, in 96-well plates, with index-sorting mode
for recording the fluorescence parameters associated to each sorted cell.

mAb	Clone	Manufacturer	Reference
B220 - FITC	RA3-6B2	Biolegend	103205
CXCR4 - PerCP-eF710	2B11	Thermofisher	46-9991-82
GL7 - Pacific Blue	GL7	Biolegend	144614
PSGL1 (CD162) - AF647	2PH1	BD	562806
CD3 - APC-Cy7	17A2	BD	560590
Gr1 (Ly6G) - APC-Cy7	1A8	BD	560600
IgD - APC-Cy7	11-26c.2a	Biolegend	405716
CD38 - PE	90	Invitrogen	12-0381-82
CD19 - PE-Dazzle594	6D5	Biolegend	115554
CD86 - PE-Cy7	GL-1	Thermofisher	12-0862-82

Data were analysed using FlowJo (v10.8.1). An example gating strategy for identifying different subsets of B
cells in draining lymph nodes is shown as **Supplementary Figure 1**.

FB5P-seq

The protocol was performed as previously described by Attaf et al.[37]. Individual cells were sorted into a 96-
well PCR plate, with each well containing 2 µL of lysis buffer. Index sort mode was activated to record the
fluorescence intensities of all markers for each individual cell. Flow cytometry standard (FCS) files from the
index sort were analyzed using FlowJo software, and compensation parameters were exported as CSV
tables for subsequent bioinformatic analysis. Immediately after sorting, plates containing individual cells were
stored at -80°C until further processing. Following thawing, reverse transcription was performed, and the
resulting cDNA was preamplified for 22 cycles. Libraries were then prepared according to the FB5P-seq
protocol. The FB5P-seq data were processed to generate both a single-cell gene count matrix and single-
cell B cell receptor (BCR) repertoire sequences for B cell analysis. Two separate bioinformatic pipelines were
employed for gene expression and repertoire analysis, as detailed in Attaf et al.[37].

10x 5' scRNA-seq

For experiments described in Figure 2a, cells from draining lymph nodes were washed and resuspended in
FACS buffer (PBS containing 5% FCS and 2 mM EDTA) at a concentration of 10^8 cells/ml. Samples from
different mice and different time points were processed separately using cell hashing as described[57]. Cells
were individually stained with distinct barcoded anti-mouse CD45 antibodies (in-house conjugated) in FACS
buffer for 30 minutes on ice. Subsequently, cells were washed and stained with a mix of primary antibodies,
then Live/Dead Fixable Aqua Dead Cell Stain (Thermofisher). Live cells of interest (IgD[neg] eYFP[+] PC and B
cells) from each sample were bulk-sorted using a BD Influx cell sorter. PC and non-PC cells were captured
in distinct wells for droplet-based scRNA-seq to maximize the recovery of BCR sequence information from
non-PC cells. Within each fraction, cells from different samples were pooled and loaded for the 10x Genomics
Single Cell 5' v2 workflow. Libraries were prepared according to the manufacturer's instructions with
modifications for generating B cell receptor (BCR) sequencing libraries. Following cDNA amplification, SPRI
select beads were used to separate the large cDNA fraction derived from cellular mRNAs (retained on beads)
from the hashtag oligonucleotide (HTO)-containing fraction (in supernatant). For the mRNA-derived cDNA
fraction, 50 ng were used to generate the transcriptome library, and 10-20 ng were used for BCR library
construction. Gene expression libraries were prepared according to the manufacturer's instructions. For BCR
libraries, heavy and light chain cDNAs were amplified by two rounds of PCR (6 cycles + 8 cycles) using

external primers recommended by 10x Genomics. Approximately 800 pg of purified amplified cDNA was tagmented using the Nextera XT DNA Sample Preparation kit (Illumina) and amplified for 12 cycles using the SI-PCR forward primer (10x Genomics) and a Nextera i7 reverse primer (Illumina). For the HTO-containing fraction, 5 ng were used to generate the HTO library. The resulting libraries were pooled and sequenced on an Illumina NextSeq2000 platform with single-indexed paired-end kits following the manufacturer's guidelines.

Ex vivo culture and ELISA

Single-cell suspensions FACS-sorted from draining lymph nodes were washed and resuspended in culture medium (10% FCS, 0.1% 2-Mercapto-ethanol, 1% non-essential amino acids, 1% Sodium Pyruvate, 1% HEPES buffer; 1% L-Glutamine, 1% Peniciline-Streptomycin in RPMI 1640) and placed in culture for different times at 37°C, 5% CO_2. Cell culture supernatants were collected and stored at -80°C until analysis. Mouse IgG concentrations were determined using the LSBio Mouse IgG ELISA Kit (Catalog No. LS-F10451) according to the manufacturer's instructions.

Spatial transcriptomics

Freshly dissected draining lymph nodes were dried on absorbent paper, embedded in OCT, snap frozen in isopentane over dry ice, and stored at -80°C until processing. On the day of the experiment, 10 µm-thick cryosections were prepared from the region of interest using a cryostat. Four sections separated by 50µm in depth were placed on a Visium slide within the designated capture areas. Hematoxylin and eosin (H&E) staining was performed according to the manufacturer's guidelines with the following modifications: hematoxylin incubation for 30 seconds, bluing buffer for 5 seconds, and eosin incubation for 1 minute 30 seconds, all at room temperature. Following washing, brightfield images of the stained sections were acquired before proceeding to subsequent steps of the 10x Genomics Visium Spatial Gene Expression protocol. The tissue was embedded in OCT medium and a 10 µm section was cut on a cryostat and deposited on the capture area of the Visium slide following the guidelines of the 10x Genomics Visium Spatial Gene Expression protocol. Briefly, we performed permeabilization (18 min), reverse transcription, second strand synthesis, denaturation, cDNA amplification (16 cycles of PCR), and library construction according to the manufacturer's instructions. The resulting libraries were sequenced on an Illumina NextSeq2000, generating an average of 120 million reads per library (Read 1: 28 cycles, Read i7: 10 cycles, Read i5: 10 cycles, Read 2: 79 cycles).

Confocal microscopy

Draining lymph nodes were harvested from immunized mice and fixed in antigen fix solution (DiaPath, ref. P0014) for 2 hours at 4°C. Samples were subsequently washed in phosphate-buffered saline (PBS) and cryoprotected in 30% sucrose overnight. The tissue was then embedded in optimal cutting temperature (OCT) compound and snap-frozen in isopentane. Cryosections (20 µm thickness) were prepared using a cryostat and stored at -20°C until staining. For immunofluorescence staining, sections were rehydrated in 1X PBS for 10 minutes. Non-specific binding was blocked by incubating sections for 30 minutes at room temperature in a blocking solution containing 0.1% Triton X-100, 1% fetal calf serum (FCS), 1% bovine serum albumin (BSA), and 1% serum from the host species of the secondary antibody in 1X PBS. Primary antibodies (see table below) were diluted in blocking solution, and sections were incubated overnight at 4°C. Following washing, slides were mounted using ProLong Gold antifade reagent (Invitrogen, ref. P36930). Confocal and spectral images were acquired using a Zeiss LSM 980 confocal microscope. Image processing for conventional analysis was performed using Zen software, while quantitative analysis was conducted using *QuPath* software (see "*Quantitative analysis of confocal microscopy images*" section for details).

mAb	Clone	Manufacturer	Reference
Lyve1 – eF570	ALY7	Thermofisher	41-0443-82
Ki67 – eF660	SolA15	Thermofisher	50-5698-82
IRF4 – AF594	IRF4.3E4	Biolegend	646409
CD3 – eF450	17A2	Thermofisher	48-0032-82
CD21/35 – AF700	7E9	Biolegend	123431
IgD – SNIR 685	11-26c.2a	Biolegend	405749
PDGFR1 - PE	18A2	Biolegend	323605
anti-GFP (cross-react eYFP) – AF488		Thermofisher	A-21311

mAb	Clone	Manufacturer	Reference
Bcl6 – AF647	IG191E/A8	Biolegend	648306
CD138 – PE	281-2	Biolegend	142504
Rabbit anti-mouse Col IV		Abcam	Ab19808
Donkey anti-rabbit IgG – Cy3		Jackson ImmunoResearch	711-165-152

Lightsheet microscopy

Draining lymph nodes were harvested and fixed overnight in 0.4% paraformaldehyde (Electron Microscopy
Science, ref. 15714) diluted in phosphate-buffered saline (PBS). Immunostaining was performed using the
Hubrecht protocol[58]. Organs were washed in PBS and incubated overnight at 4°C in blocking buffer (PBS
containing 5% donkey serum, 1% rehydrated milk, and 0.4% Triton X-100). Samples were then incubated for
5 days with primary antibodies (see table below) diluted in PBS containing 1% rehydrated milk, 0.4% Triton
X-100, 3% donkey serum, and 3% mouse serum. Following incubation, samples were washed in PBS with
0.4% Triton X-100 for 1 day. Secondary antibody incubation was performed similarly for 5 days, followed by
another 1-day wash in PBS. Samples were embedded in 1% low-melting agarose and subsequently
dehydrated through a graded methanol series (20%, 40%, 60%, 80%, and 2× 100% in PBS) for 1 hour per
concentration at room temperature. Lymph nodes were then incubated overnight in 100% dehydrated
methanol at room temperature. Clearing was initiated by incubating samples in a 1:1 mixture of methanol
and BABB (benzyl alcohol and benzyl benzoate at a 1:2 ratio, Sigma ref. 305197 and Fisher Scientific ref.
10654752), followed by overnight incubation in pure BABB. Finally, samples were transferred to ethyl
cinnamate for storage until imaging. Imaging was performed using a LaVision Ultramicroscope II (Miltenyi
Biotec). Image stacks were acquired with a step size of 6 μm at x2 magnification using an optic zoom with a
numerical aperture of 9 μm. Three-dimensional reconstruction and analysis of image stacks were conducted
using IMARIS software (Version 9.1.0, Bitplane).

mAb	Clone	Manufacturer	Reference
Chicken anti-GFP	-	AVES	GFP-1020
Rabbit anti-mouse Lyve1	-	Reliatech	103-PA50
Donkey anti-chicken IgY – Cy3	-	Jackson Immunoresearch	703-166-155
Donkey anti rabbit IgG – AF790	-	Jackson Immunoresearch	711-655-152
CD4 – AF488	GK1.5	eBioscience	53-0041-82
IRF4 – AF647	IRF4.3E4	Biolegend	646409

478
479
COMPUTATIONAL METHODS

Flow cytometry data analysis

The FlowJo UMAP extension was downloaded from FlowJo Exchange Website and used according to the
default parameters (euclidean distance, nearest neighbors = 15, minimum distance = 0.5, number of
components = 2). The UMAP presented in Figure 3b was computed based on the following compensated
parameters: IRF4, Bst2, CD38, GL7, CD138, CD19, PSGL1, Bcl6, B220, CCR6.

Pre-processing of scRNA-seq datasets

<u>10x Genomics 5'-end sequencing</u>

Raw fastq files from gene expression libraries were processed using *Cell Ranger* and *Cell Ranger VDJ*
(v3.0.1 for dataset #2, v6.1.2 for dataset #3), with alignment on the mm10 and vdj_GRCm38_alts_ensembl-
7.0.0 reference genomes, respectively. Quality control was performed on each dataset independently to
remove poor quality cells based on UMI counts, number of genes detected, percentage of transcripts from
mitochondrial genes, and percentage of transcripts from ribosomal protein coding genes. For each cell, gene
expression UMI count values were log-normalized with Seurat *NormalizeData* with a scale factor of 10,000
to generate normalized UMI count matrices. HTO barcodes for sample demultiplexing after hashing were
counted using CITE-seq-count and were normalized for each cell using a centered log ratio (CLR)
transformation across cells implemented in the Seurat function *NormalizeData*. Hashtags were demultiplexed

using *MULTIseqDemux* function and barcodes assigned as doublets or negative were excluded from further
analysis.

FB5P-seq

We used a custom bioinformatics pipeline to process fastq files and generate single-cell gene expression
matrices and BCR sequence files as previously described[37]. Detailed instructions for running the FB5P-seq
bioinformatics pipeline can be found at https://github.com/MilpiedLab/FB5P-seq. Quality control was
performed on each dataset independently to remove poor quality cells based on UMI counts, number of
genes detected, ERCC spike-in quantification accuracy, and percentage of transcripts from mitochondrial
genes. For each cell, gene expression UMI count values were log-normalized with Seurat *NormalizeData*
with a scale factor of 10,000 to generate normalized UMI count matrices.

Index-sorting FCS files were visualized in FlowJo software and compensated parameters values were
exported in CSV tables for further processing. For visualization on linear scales in the R programming
software, we applied the hyperbolic arcsine transformation on fluorescence parameters[59].

For BCR sequence reconstruction, the outputs of the FB5P-seq pipeline were further processed and filtered
with custom R scripts. For each cell, reconstructed contigs corresponding to the same V(D)J rearrangement
were merged, keeping the largest sequence for further analysis. We discarded contigs with no constant
region identified in-frame with the V(D)J rearrangement. In cases where several contigs corresponding to the
same BCR chain had passed the above filters, we retained the contig with the highest expression level. BCR
metadata from the *MigMap* and *Blastn* annotations were appended to the gene expression and index sorting
metadata for each cell.

Non-supervised analysis

For the analysis described in Figure 2, the three datasets after quality control and normalization were
integrated in *Seurat* using *SelectIntegrationFeatures* (2000 features, excluding BCR coding genes),
FindIntegrationAchors, and *IntegrateData* (30 components). After integration, we used *RunPCA* (30
components), *FindNeighbors* (30 components) and *FindClusters* (resolution 0.3) for non-supervised
clustering, and *RunUMAP* (30 components) for visualization. Marker genes were computed with
FindAllMarkers (Wilcoxon assay) and top markers visualized as a *Seurat* dot plot. For the analysis described
in Figure 4, we used a standard *Seurat* analysis workflow, including cell cycle regression.

Supervised annotation

Supervised annotation of scRNA-seq datasets were performed as described in Figure 2d and Supplementary
Figure 2c. We used the *AddModuleScore* function to compute gene expression scores for every cell in the
datasets for the indicated signatures (Supplementary Table 1). Thresholds for "gating" (Supplementary
Figure 2c) were defined empirically to optimize concordance between supervised annotation and non-
supervised clustering. The continuum score in Figure 4f was computed by fitting a linear regression on the
distribution of B cells in the scatter plot of ER stress score (x-axis) and Normalized Ig transcripts UMI counts
(y-axis), and projecting data points on the regression line.

Gene ontology

The evolution profile of genes along the differentiation continuum was computed by applying a kernel smoother on the distribution of data, for each gene expressed in more than 25% cells in the dataset (n=1834 genes after excluding mitochondrial and ribosomal genes). The derivatives of the resulting profiles were hierarchically clustered (pearson's correlation distance) to define groups of co-evolving genes in the continuum. Lists of genes grouped in distinct evolution clusters were then submitted to gene ontology analysis using *gprofiler2*, with default settings, and computing only the results for the "BP: biological processes" categories. The results of multiple gene clusters were appended in a single plot using *ggplot2*.

BCR-seq analysis

Single-cell BCR-seq data were further analyzed with Change-O[60] to compute clonotypes and annotate somatic hypermutation load. The intersection between cell clonotype identities and supervised annotations based on gene expression was used to compute the clonal compositions displayed in Figure 5j.

Spatial Transcriptomics analysis

Spatial transcriptomics FASTQ files were aligned to the mm10 reference genome using *SpaceRanger* v1.3.1. Downstream analysis was performed using the *Seurat* R package v4.2.1, employing log-normalization and the Louvain clustering algorithm. Annotation of the four merged slices was conducted iteratively, beginning with a clustering resolution of 0.1 (50 principal components) to identify main areas (B cell zone, T cell zone, and medulla), followed by a resolution of 0.2 (50 principal components) to reveal germinal centers. GC spots were further subclustered for light zone (LZ) and dark zone (DZ) segregation (resolution 0.2, 30 principal components). Border regions were annotated based on neighboring spots, defining GC-T zone interface (GTI), GC-medulla interface (GMI), and GC-B zone interface (GBI) based on neighbor spots annotation. Annotations were projected onto a UMAP calculated using 50 principal components from the complete dataset. Cell type-specific genes were identified using *FindAllMarkers* (default parameters) and filtered for adjusted p-value < 0.05. For stromal cell signature analysis, gene lists were derived from Rodda et al.[41] (log2FoldChange > 0.5, FDR < 0.05, pct_in > 0.10).

Quantitative analysis of confocal microscopy images

For cell detection and quantification in confocal microscopy images (Figure 6), we used the *QuPath* analysis platform. We annotated manually the GC and medulla areas based on the eYFP staining for the GC and on the IRF4 staining for the medulla. We defined the GC-medulla interface (GMI) as a 50µm-wide border surrounding the GC and intersecting the medulla annotation. We imported *Cellpose2coloc* in *QuPath* to segment the cells based on the different channels selected, which generated different cell segmentation masks. We determined the positivity thresholds for defining cell subsets, and used them in the colocalization analysis using a custom script. We exported the measurements and performed quantification analyses in GraphPad Prism. All scripts are publicly available here: https://github.com/Imagimm-CIML/Detection-of-B-cell-subtypes-in-a-draining-lymph-node-using-a-mask-colocalization-approach?tab=readme-ov-file#readme.

Statistical analyses

Statistical analyses were performed using Graphpad Prism or R softwares with tests and p-value significance criteria detailed in the figure legends.

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AUTHORSHIP CONTRIBUTIONS

L.B. designed experiments, performed experiments, analyzed the data, performed some of the bioinformatics analyses, prepared the figures and wrote the manuscript. C.D. performed most bioinformatics analyses and prepared the figures. N.A. designed and performed experiments at the initiation of the project. L.G. performed single-cell and spatial transcriptomics experiments. M.F. and T.B. designed and performed the quantitative analysis of confocal microscopy images. B.E. performed the bioinformatics analysis of the spatial transcriptomics data. L.C. performed the spatial transcriptomics experiment. C.S. and S.V.P. supervised the light-sheet microscopy staining, data acquisition, and data analyses. J.M.N. produced hashtag antibodies. P.M. designed experiments, performed experiments, analyzed the data, prepared the figures, wrote the manuscript, acquired funding and supervised the study. All authors revised and approved the manuscript.

DISCLOSURE OF CONFLICTS OF INTEREST

The authors declare no competing financial interests.

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FIGURES AND LEGENDS

Figure 1

Figure 1: Tracking GC B cells and recent GC emigrants after model immunization.

a, Experimental design. **b**, Flow cytometry gating of eYFP$^+$ and eYFP$^-$ PC (CD138^{hi}) and B cells among IgD^{neg} B cells from dLN in mice gavaged 4 days before sacrifice. **c**, Flow cytometry gating of GC B cells and MBC among eYFP$^+$ B cells. Numbers indicate % of parent gate. **d**, Percentage of the indicated eYFP$^+$ IgD^{neg} B cell types (GC, MBC or PC) among total live cells in dLN at the indicated times after tamoxifen gavage (2, 4, 5, 6 days). **e**, Flow cytometry gating of NP-specific cells among GC, PC and MBC (numbers indicate % of parent gate) (left). Cell counts of NP-specific GC, MBC and PC cells in dLN (middle). Percentage of NP-specific cells among the indicated cell types (right). Data compiled from 8 independent experiments. **f**, Flow cytometry gating of NP-specific cells among eYFP$^+$ GC, PC and MBC (numbers indicate % of parent gate) (left). Cell counts of eYFP$^+$ NP-specific GC, MBC and PC types in dLN (middle). Percentage of NP-specific cells among the indicated eYFP$^+$ cell types (right). Data compiled from 7 independent experiments. Symbols in panels **d**, **e** and **f** represent individual mice. * $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$, **** $p < 0.0001$ in Wilcoxon matched pairs signed rank test.

Figure 2

Figure 2: Single-cell RNA-seq analysis of GC B cells and recent GC emigrants identifies putative PrePC.

a, Experimental designs for the generation of three distinct scRNA-seq datasets of GC and GC-derived cells
in dLN. **b**, UMAP representation of single-cell gene expression profiles of GC and GC-derived cells integrated
from the three datasets, colored by unsupervised Louvain clusters (left) or by cell cycle status (right).　**c**, dot
plot of the expression of top 5 markers of each unsupervised cluster, as indicated. The percentage of cells
from a cluster expressing a marker is coded by the size of the circles, the average expression level by the
color. **d**, Hierarchical gating strategy for gene expression score-based supervised annotation of GC and GC-
derived cells. **e**, UMAP representation colored by supervised annotations. **f**, Total counts (left) and proportion
among total cells (right) of cells from the indicated supervised annotation. Each symbol represents a distinct
sample from the three datasets. **g**, Violin plots of gene expression levels of *Selplg* (left) and *Bst2* (right)
among cells from the indicated supervised annotations.

Figure 3

Figure 3: Enrichment of putative PrePC by flow cytometry.

a, Experimental design. **b**, UMAP representation of flow cytometry-based protein expression profiles of dLN IgD^{neg} B cells, pseudocolored by embedded cell density. Clusters of GC, PC, MBC and intermediate GC-to-PC cells are circled and annotated. **c**, Mean Fluorescence Intensity (MFI) histogram and quantification for the indicated GC (GL7, CD19, B220; top panels) and PC (IRF4, CD138, PSGL1; bottom panels) protein markers in the indicated cell subsets. **d**, Gating strategy to enrich for putative PrePC by flow cytometry. **e**, UMAP projection as in b, representing only the cells from the subsets gated in d (left) and then split by subset (right). **f**, Total counts (top) and proportion among dLN IgD^{neg} B cells (bottom) of cells from the indicated supervised gates defined in d and e. Data compiled from 6 experiments. **g**, Total counts (top) and proportion among dLN IgD^{neg} B cells (bottom) of NP-specific cells from the indicated supervised gates defined in d and e. Data compiled from 6 experiments. Each symbol in panels c, f and g represent a distinct mouse. *** p < 0.001, **** p < 0.0001 in Wilcoxon matched pairs signed rank test.

Figure 4

Figure 4: Characterization of PrePC cells in the GC-to-PC differentiation continuum.

a, Experimental design. **b**, UMAP projection of gene expression profiles of FACS-sorted DZ, LZ, PrePC and
PC cells, after cell cycle regression. Each dot is a cell, colored by the FACS sorting phenotype. **c**, Feature
plots showing the expression of the indicated GC (*Ms4a1*, *Cd19*, *Irf8*, *Myc*) and PC (*Sdc1*, *Prdm1*, *Irf4*, *Selplg*,
Bst2) marker genes in cells laid out as in b. **d**, UMAP projection as in b, with cells colored by cell cycle status.
e, Scatter plot representation of the ER stress score (x-axis) and the Ig transcripts UMI counts (y-axis) in
cells colored by the FACS sorting phenotype. **f**, Scatter plot presented in d, overlayed with the regression
line and colored by the GC-to-PC continuum score computed after projecting cells on the regression line and
ranking. **g**, UMAP projection as in b, colored by the continuum score. **h**, Scatter plot representation of the
continuum score (x-axis) and the MHC-II gene expression score (y-axis) in cells colored by the FACS sorting
phenotype. **i**, IgG concentration in culture supernatant of LZ B cells, PC and PrePC at the indicated times
after *ex vivo* cell culture. **j**, Cell density histogram distribution of continuum scores for cells in the dataset,
indicating the continuum thresholds used for defining low, intermediate and high values for the continuum
score in PrePC cells. **k**, Single-cell gene expression heatmap of the indicated GC (top 6 genes), PC (next 7
genes) and PrePC (bottom 2 genes) markers expression in the different subsets defined by their phenotype
and continuum score. **l**, Relative gene expression profile of cells ranked by continuum score, for genes
grouped according to their evolution profile (each gene a red line, the average cluster profile a black line). **m**,
Results of gene ontology (GO) enrichment analysis for genes in the different evolution clusters defined in l.

Figure 5

Figure 5: BCR analysis of PrePC cells.

a, Violin plot of BCR mutations (sum of IgH and IgL) in the indicated cell subsets from the dataset described in Figure 4. **b**, Proportion of mutated cells among total cells of the indicated subsets. Each dot represents a

distinct mouse, bar at median. **c**, UMAP projection as in Figure 4, colored by cell subsets, with PC and PrePC
divided into BCR-unmutated and BCR-mutated (left), and split to display only PC (middle) and PrePC (right).
d, Pie charts of the isotype repartition in the indicated cell subsets, with PC and PrePC divided into BCR-
unmutated and BCR-mutated. Numbers indicate the number of cells analyzed for each subset. **e**, UMAP
projection as in Figure 4, colored by the expression levels of the ER stress gene expression signature (left)
or the activation gene expression signature (right). **f**, UMAP projection as in Figure 4, colored by supervised
annotation 2 segragating PrePC into PrePC.Act and PrePC.ERstress^{lo}. **g**, Scatter plot representation of the
ER stress score (x-axis) and the Ig transcripts UMI counts (y-axis) in cells colored by supervised annotation
2. **h**, Flow cytometry signal intensity (index sorting) of the indicated surface markers in cells from the indicated
subsets (supervised annotation 2). **i**, Proportion of mutated cells among total cells grouped by supervised
annotation 2. Each dot represents a distinct mouse, bar at median. **j**, Pie charts of shared clonotype
composition within each subset of the supervised annotation 2. Numbers indicate the number of cells
analyzed for each subset. **k,** Pie charts of the repartition of the number of somatic mutations in BCR of PrePC
(top) or PC (bottom) grouped based on clonal overlap with GC B cells (left) or not clonally expanded (right).
The numbers in the center represent the number of cells considered in each pie chart. **l,** Violin plots
representing the number of mutations in PrePC (left) and PC (right) from the indicated clonal groups. **
p < 0.01, **** p < 0.0001 in Mann-Whitney non-parametric test.

Figure 6

Figure 6: Recent GC-derived PC proliferate at the DZ-medulla interface.

a, Feature plots showing the expression of the indicated genes (*Cxcr3*, *Gpr183*, *Itga4*, *Itgb1*) in cells laid out
as in Figure 4b. **b**, Experimental design of the spot-based spatial transcriptomics analysis of dLN, and semi-
supervised annotation of spots based on unsupervised spot clustering, gene expression scores, and
neighboring spot identity (GTI: GC-T zone interface; GMI: GC-medulla interface; GBI: GC-B zone interface).
c, Gene expression-based UMAP projection of spatial transcriptomics spots, colored by zonal identity. **d**,
Violin plot of PC gene expression score in spots from the indicated zones. **e**, Spectral confocal microscopy
images of serial dLN sections (14 days after primary NP-KLH immunization, 3 days after tamoxifen) stained
for the indicated markers (scale bar 200 μm). The images show a broad view around on GC (i). We zoomed
in this view to show the GTI (ii and iii, yellow arrows) and the GMI (ii and iii, white arrows). The white arrows
in iv represent the distinct populations of CD138^{lo} and CD138^{+} cells. **f**, Density of GC-derived PC (IRF4^{+} cells
expressing the GC-derived eYFP reporter) quantified from spectral confocal microscopy images after
segmentation in the indicated zones. Each dot represents a zone, combining analyses from 2 dLN in 2 distinct
experiments. **g**, Expression level of CD138 (geometric mean fluorescence intensity) in mature PC (IRF4^{+}
CD138^{+} cells) in the indicated zones. Each dot represents a zone, combining analyses from 2 dLN in 2 distinct
experiments. **h**, Lightsheet microscopy images of a whole dLN stained with the indicated markers, extracted
from different Z-positions as indicated in the top panel. **i**, Spectral confocal microscopy image of the indicated
markers in a whole dLN section (left) and zooming in on DZ proximal areas of two GCs (GC i: middle; GC ii:
right). Arrows in the zoomed-in panels indicate triple positive eYFP^{+} IRF4^{+} Ki67^{+} GC-derived proliferating PC.
j, Mean Fluorescence Intensity (MFI) histogram for Ki67 protein in NP-PE^{+} (left graph) and NP-PE^{-} (middle
graph) subsets among GC, PrePC, PC and MBC populations, and proportions of Ki67^{+} cells among the
indicated NP-PE^{+} and NP-PE^{-} subets (right panel). * $p < 0.05$, ** $p < 0.01$, **** $p < 0.0001$ in Wilcoxon matched
pairs signed rank test.

Supplementary Figures and Legends

Supplementary Figure 1

845   **Supplementary Figure 1**: flow cytometry gating strategy to analyze B cell subsets in dLN.

Supplementary Figure 2

Supplementary Figure 2: supervised annotation of integrated scRNA-seq datasets of GC B cells and recent
GC emigrants

a, UMAP representation of single-cell gene expression profiles of GC and GC-derived cells integrated from
the three datasets, colored by dataset identity (left) and split by dataset (right). **b**, UMAP representation of
single-cell gene expression profiles of GC and GC-derived cells integrated from the three datasets, colored
by time after immunization (d10p: day 10 of primary response, d10s: day 10 of secondary response). **c**,
Supervised gating strategy for the integrated scRNA-seq dataset based on gene expression scores for PC,
activated (ACT), memory (MBC), DZ and LZ scores. Dashed lines represent the thresholds used to gate the
cells in different annotations; the final supervised annotation is represented in the bottom right panel. **d**, Bar
graphs of the composition of the unsupervised Louvain clusters in the cell types or states identified by
supervised annotation. **e**, Single-cell gene expression heatmaps of the indicated marker genes in the different
subsets defined by supervised annotation, also indicating the cell cycle phase (G1, S, G2/M) (all datasets),
surface OVA binding (dataset #1) or time after primary immunization (dataset #3).

Supplementary Figure 3

Supplementary Figure 3: PrePC are generated in primary and secondary GC responses.
Proportions of cells among dLN IgD^{neg} B cells from the indicated gates defined in Figure 3d and 3e in mice analyzed at different time points after primary immunization (left panel), and at day 15 after primary or secondary immunization (right panel). * $p < 0.05$ in Mann-Whitney non-parametric test.

Supplementary Figure 4

Supplementary Figure 4: Characterization of PrePC cells in the GC-to-PC differentiation continuum.

a, UMAP projection as in Figure 4, colored by the expression levels of the ER stress gene expression signature (left) and violin plot of the expression of the ER stress signature in the cells grouped by phenotype (right). **b**, UMAP projection as in Figure 4, colored by the expression levels of Ig transcripts (left) and violin plot of the expression of Ig transcripts in the cells grouped by phenotype (right). **c**, UMAP projection as in Figure 4, colored by the expression levels of different MHC class II genes (*H2-Ob, H2-DMa, Ciita, H2-DMb1, H2-DMb2*). **d**, Violin plot of the expression of the MHC class II score in the cells grouped by phenotype. The MHC-II gene expression signature was built using the markers depicted in c. **e,** Flow cytometry signal intensity (index sorting) of the indicated surface markers in cells from the indicated subsets (supervised annotation).

Supplementary Figure 5

Supplementary Figure 5: Recent GC-derived PC localize at the DZ-medulla interface.

a, Spatial feature maps showing the expression levels of the stromal cell subset specific gene expression signatures, as indicated. **b**, Violin plots showing the expression of the stromal cell subset specific gene expression signatures indicated in the different zones identified. **c**, Confocal microscopy image of the indicated markers in a dLN section, focusing on two GC and the DZ-proximal area. **** p < 0.0001 in Wilcoxon matched pairs signed rank test.