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34 Abstract

35 Plasmacytoid dendritic cells (pDCs) are a major source of type I interferon (IFN-I). What other functions pDCs exert in vivo during viral infections is controversial and 36 37 more studies are needed to understand their orchestration. Here, we characterize in-38 depth and link pDC activation states in animals infected by mouse cytomegalovirus, by combining *lfnb1* reporter mice with flow cytometry, single-cell RNA sequencing, 39 40 confocal microscopy and a cognate CD4 T cell activation assay. We show that IFN-I production and T cell activation were performed by the same pDC, but sequentially in 41 42 time and in different micro-anatomical locations. In addition, we show that pDC 43 commitment to IFN-I production was marked early on by their downregulation of LIFR and promoted by cell-intrinsic TNF signaling. We propose a novel model of how 44 45 individual pDCs are endowed to exert different functions in vivo during a viral 46 infection in a manner tightly orchestrated in time and space.

48 Introduction

Plasmacytoid dendritic cells (pDCs) produce massive amounts of type I interferons (IFN-I) in response to virus-type stimuli¹, including in vivo during mouse cytomegalovirus (MCMV) infection^{2, 3}. pDCs exert this function by engulfing material derived from viruses or infected cells, and routing it to dedicated endosomes for tolllike receptors 7/9 triggering and activation of the MyD88-to-IRF7 signaling pathway¹.

54 Soon after their discovery, pDCs were proposed to first produce IFN-I and then acquire features of conventional dendritic cells (cDCs)⁴, including dendrites and 55 the ability to present antigens⁴⁻⁷, upon specific stimulation conditions. However, 56 57 adequate pDC purification methods and single-cell analysis tools were lacking to prove this hypothesis. Indeed, when defined by few surface markers, pDCs can be 58 contaminated with cDC precursors⁸⁻¹⁰, pDC-like cells¹¹, non-canonical CD8α⁺ cDCs¹², 59 ¹³ or transitional DCs (tDCs)^{14, 15}. These cells harbor mixed phenotypic and functional 60 characteristics of pDCs and cDCs and are more powerful than resting pDCs for T cell 61 activation. Hence, T cell activation by pDCs was claimed to result from this 62 63 contamination. However, discrepancies in the stimuli used to activate pDCs likely contributed to fuel the controversy on their T cell activation capacity^{16, 17}. It is not 64 clear to which extent pDC phenotypic and functional heterogeneity reflects their true 65 developmental or stimulation-induced plasticity versus a lineage contamination by 66 other cells¹⁸⁻²¹. pDC activation differs in response to synthetic TLR ligands, viral 67 particles or infected cells²²⁻²⁴. Rigorously characterizing pDC cell fate and functional 68 plasticity in vivo during viral infections thus remains a major challenge^{15, 19, 21}. 69

Here, we characterized and linked pDC activation states in MCMV-infected animals, by combining *lfnb1*-reporter mice²⁵ with flow cytometry, single-cell RNA sequencing (scRNAseq), confocal microscopy and a CD4 T cell priming assay.

Contrary to currently prevailing models of pDC plasticity^{19, 20}, we demonstrated that
IFN-I production and T cell activation were performed by the same pDC, but
sequentially in time and in different microenvironments.

77 Results

78 pDCs specifically express IFN-I during MCMV infection.

We analyzed IFN-I production kinetics in splenic pDCs during MCMV infection 79 (Fig. 1a-b). IFN-I-producing pDCs were detected at 33h post-infection, their 80 percentage increased at 36h and 40h, diminished at 44h and became undetectable 81 at 48h. Since pDC-like cells share Zbtb46 expression with cDCs^{10, 11, 26}, we 82 harnessed *Zbtb46^{Egfp}* mice to identify them and show that they do not express IFN-I at 83 84 36h post-infection (Extended Data Fig. 1a). Thus, IFN-I expression occurred 85 specifically in a minor fraction of splenic pDCs around 36h post-infection, consistent with previous work ²⁷. 86

87

88 Microarray analysis of pDCs isolated from MCMV-infected *lfnb1^{Eyfp}* mice 89 suggests their heterogeneity.

We harnessed the *lfnb1^{Eyfp}* reporter mice²⁵ to characterize lfnb1-producing 90 pDCs and follow their fate in vivo during MCMV infection. The half-life of YFP in live 91 cells exceeds 24h^{28, 29}. Consistently, YFP⁺ pDCs sorted from 36h infected Ifnb1^{Eyfp} 92 mice and cultured ex vivo for 8h harbored stable YFP expression (Extended Data Fig. 93 1b). At 36h, 44h and 48h post-infection, both the percentage and MFI of YFP⁺ cDCs 94 were very low (Extended Data Fig. 1c-e), whereas around 80% of YFP⁺ splenocytes 95 were XCR1⁻ CD11b⁻ CD11c^{int} Bst2^{hi} (Extended Data Fig. 1f-g), therefore being bona 96 fide CCR9⁺ pDCs (Extended Data Fig. 1c-e). Thus, *lfnb1^{Eyfp}* mice allow reliable 97 identification and fate-mapping of Ifnb1-producing pDCs in vivo. 98

99 YFP and CD86 co-expression at 36h post-infection identified three pDC 100 subpopulations (Fig. 1c,d). In uninfected (UN) animals, most pDCs were CD86⁻YFP⁻

(DN cells), with the remainder being CD86⁺YFP⁻ (SP cells). In infected (IN) mice, half 101 of the pDCs expressed CD86 (IN_SP cells), with a minority being YFP⁺ (IN_DP 102 cells). pDC subpopulations were sorted by flow cytometry for gene expression 103 profiling (Fig. 1e). Interferon-stimulated genes (ISG) were induced in all pDC 104 subpopulations from infected mice (Fig. 1e, red names). Cd86 induction was stronger 105 in IN SP and IN DP pDCs (Fig. 1e), consistent with their purification as CD86⁺. 106 Unexpectedly, IN_DN and IN_SP pDCs expressed IFN-I genes (Fig. 1e, green 107 names), including *lfnb1* (green arrow), and other MyD88-dependent genes (MSG)²² 108 (Fig. 1e, blue names), but at lower levels than IN DP cells. The IN DP cells 109 specifically up-regulated genes known to be higher in steady state pDC-like cells¹¹ or 110 tDCs^{13, 15} (Fig. 1e, violet names), including genes involved in T cell activation by 111 cDCs. The IN DP cells expressed lower Tcf4 and higher Id2 levels (Fig. 1e, black 112 arrows), two transcription factors driving pDC versus cDC1 differentiation and identity 113 maintenance respectively^{19, 30}. All pDC subpopulations expressed high and 114 comparable levels of genes known to be specific of steady state pDCs as compared 115 to cDC-like cells, tDCs and cDCs^{10, 11, 13, 15}, including Ccr9, Ly6d, Klk1 and Klk1b27 116 117 (Fig. 1e, bold black names). Thus, YFP⁻ pDCs isolated from infected mice unexpectedly expressed IFN-I genes, while YFP⁺ pDCs acquired a transcriptomic 118 program shared with tDCs and cDCs, suggesting further heterogeneity within each 119 pDC subpopulation. 120

121

122 IFN-I and YFP co-expression reveals three consecutive pDC activation states.

We analyzed YFP and IFN-I co-expression in splenic pDCs isolated from 36h MCMV-infected *Ifnb1^{Eyfp}* mice (Fig. 1f). A small fraction of YFP⁻ pDCs expressed IFN-I, albeit with a lower MFI than YFP⁺ pDCs. Thus, IFN-I expression preceded YFP

detection in pDCs, likely explaining the moderate induction of IFN-I genes in IN DN 126 and IN_SP pDCs in microarray analyses. Half of YFP⁺ pDCs did not express IFN-I, 127 128 suggesting that they were in a late activation state, where they had already secreted IFN-I but retained YFP expression. Therefore, we examined IFN-I and YFP co-129 expression kinetically during infection (Fig. 1g-i). Whereas IFN-I expression peaked 130 between 36h and 40h post-infection (Fig. 1g-h), the frequency of YFP⁺ pDCs 131 continued increasing to reach a plateau between 44h and 48h (Fig. 1g-h). Hence, 132 within pDCs expressing IFN-I or YFP, the proportions of IFN-I⁺YFP⁻ cells decreased 133 after 33h, to become undetectable at 48h, whereas the proportions of IFN-I⁺YFP⁺ 134 cells peaked at 36h, and the proportions of IFN-I⁻YFP⁺ cells increased steadily (Fig. 135 136 1i). Thus, pDCs undergo a more complex activation than anticipated, encompassing at least three consecutive stages for IFN-I production. At an early stage, cytokine 137 production starts at relatively low level as revealed by lack of YFP expression (IFN-138 I⁺YFP⁻). Then, high expression of both YFP and IFN-I marks pDCs at peak IFN-I 139 production (IFN-I⁺YFP⁺). Finally, pDCs undergo a late activation stage, previously 140 141 overlooked, where they have secreted IFN-I but remain fate-mapped for their 142 expression (IFN-I⁻YFP⁺). Thus, pDCs survive more than 12h after ceasing IFN-I production, raising the question of their terminal activation state and functions. 143

144

scRNAseq confirms the existence of different pDC activation states during infection.

To better characterize pDC activation states without the confounding effect of cross-contamination between populations as can happen in bulk transcriptomics, we performed scRNAseq on splenic pDCs from one uninfected and one 36h MCMVinfected *lfnb1*^{Eyfp} mice (Fig. 2a), using Smart-seq2 (SS2)³¹. For the infected mouse,

94 pDCs were index sorted, with enrichment of YFP⁺ pDCs to one third of all cells
(SS2 dataset#1).

Principal component analysis (PCA) did not show heterogeneity in steady-153 154 state pDCs (Fig. 2b). In contrast, the pDCs from the infected mouse (Fig. 2c) clearly segregated along the first PCA axis (PC1) into YFP⁺ (red dots) and YFP⁻ (blue dots) 155 subpopulations. Consistently, IFN-I genes and other MSG strongly contributed to 156 157 PC1. Based on top PC1 genes, hierarchical clustering not only separated most YFP⁻ from YFP⁺ pDCs, but also split YFP⁺ pDCs in two clusters (Fig. 2d). The first YFP⁺ 158 159 pDC cluster co-expressed *Eyfp* and IFN-I genes (Fig. 2d, green arrow and names), therefore corresponding to the peak IFN-I⁺YFP⁺ activation state identified by flow 160 cvtometry. Most cells of the second YFP⁺ pDC cluster expressed neither Eyfp nor 161 IFN-I genes, or at very low levels (Fig. 2d). However, Eyfp⁻YFP⁺ and Eyfp⁺YFP⁺ 162 pDCs shared the expression of other MSG, including *II12b*, *Ccl5*, *Ccl3* and *Tnf* (Fig. 163 2d, bold blue names). As compared to YFP⁻ pDCs, YFP⁺ pDCs expressed pDC-like 164 165 cell/tDC/cDC genes, including Cd83, Cd40 and Ccr7 (Fig. 2d, bold violet names). Thus, the *Eyfp*⁻YFP⁺ pDC cluster matched with the late IFN-I⁻YFP⁺ activation state 166 identified by flow cytometry, corresponding to pDCs that had secreted IFN-I and 167 168 stopped transcribing IFN-I genes.

Out of the 65 YFP⁻ pDCs, 4 segregated with YFP⁺ pDCs, both in PCA and hierarchical clustering (Fig. 2c-d). These YFP⁻ pDCs expressed high levels of *Eyfp* and IFN-I genes (Fig. 2d,e), confirming that detectable induction of IFN-I precedes that of YFP. Thus, *Eyfp*⁺YFP⁻ pDCs corresponded to the early IFN-I⁺YFP⁻ activation state identified by flow cytometry.

Eyfp and IFN-I gene expression were tightly correlated in single pDCs (Fig. 2de and Supplementary Table 1), consistent with the known synchronous induction of

all IFN-I³² in a manner dependent on IRF7 but not IFN-I receptor signaling, specfically in pDCs^{22, 33-35}. Thus, in *lfnb1^{Eyfp}* mice, *Eyfp* reliably reports expression of all IFN-I genes in pDCs.

In summary, combined expression of the *Eyfp* gene and of the YFP protein in *Ifnb1^{Eyfp}* mice enabled accurate tracking in scRNAseq analyses of three consecutive activation states of the pDCs engaged in IFN-I production, corresponding to early *Eyfp*⁺YFP⁻, peak *Eyfp*⁺YFP⁺ and late *Eyfp*⁻YFP⁺ pDC clusters.

183

184 LIFR is selectively downregulated on IFN-I-expressing pDCs.

The small number of *Eyfp*⁺YFP⁻ pDCs limited the accuracy of the analysis of 185 their gene expression program. Therefore, we sought for a cell surface marker to 186 enrich them. Cd83, Ccr7 and Lifr were differentially expressed between YFP⁺ and 187 YFP⁻ pDCs. Cd83 and Ccr7 were induced on pDCs expressing lfnb1 or YFP. 188 Reciprocally, Lifr was selectively downregulated in the vast majority of $lfnb1^+$ pDCs 189 190 regardless of their YFP expression (Fig. 2f-g). Flow cytometry analysis confirmed a stronger downregulation of LIFR on YFP⁺ or IFN-I⁺ pDCs (Fig. 2h-i), whereas CD83 191 and CCR7 were induced only on a fraction of YFP⁺ pDCs. Since IFN-I⁺ pDCs express 192 the highest levels of BST2 (Fig. 1a), we combined LIFR downregulation with YFP and 193 BST2 expression for enriching $Eyfp^+YFP^-$ pDCs (Extended Data Fig. 2a). 194

pDCs from one uninfected and one 36h MCMV-infected *lfnb1^{Eyfp}* mice were index-sorted with enrichment of LIFR^{Io} cells (SS2 dataset#2). This led to a contamination of pDCs by limited numbers of pDC-like or B cells, none of which was YFP⁺ (Supplementary Data 1, Extended Data Fig 2b-f). The expression of *Eyfp* and IFN-I genes were again tightly correlated (Supplementary Table 2). The remaining

cells were bona fide pDCs expressing high levels of key pDC signature genes,
including *Klk1*, *Siglech*, *Pacsin1*, *Dntt* and *Irf8* (Extended Data Fig 2f, black arrows),
but low levels of the genes selective of lymphocytes³⁶, B cells, plasma cells³⁷ or pDClike cells¹¹ (Extended Data Fig 2f).

Upon UMAP dimensionality reduction and graph-based cell clustering (Fig. 204 3a), the pDCs from the uninfected mouse regrouped together (C0) remotely from the 205 206 7 clusters derived from the pDCs of the infected mouse. Projection on the UMAP of 207 the expression levels of YFP (Fig. 3b), Eyfp (Fig. 3c) and Ifnb1 (Fig. 3d) identified the 208 clusters corresponding to the three pDC activation states uncovered in our previous experiments: *Eyfp*⁺YFP⁻ (C5), *Eyfp*⁺YFP⁺ (C6) and *Eyfp*⁻YFP⁺ (C7). *Eyfp*⁺YFP⁻ C5 209 cells expressed Eyfp and most IFN-I genes, although at lower levels than the 210 Evfp⁺YFP⁺ C6 cells (Fig. 3e; green names on Fig. 3f). C5 represented 16% of the 211 LIFR^{lo}BST2^{hi} pDCs from the infected mouse (Supplementary Table 3), and 212 encompassed 88% of LIFR¹⁰BST2^{hi} sorted pDCs. LIFR expression was 213 214 downregulated most strongly in clusters C5, C6 and C7 (Fig. 3g-h), thus helping to discriminate C5 Eyfp⁺YFP⁻ cells from all other YFP⁻ cells. In contrast to steady state 215 pDCs (C0), all pDC clusters from the infected mouse (C1-C7) harbored high 216 expression of Oasl1 (Fig. 3i) and other ISG (Fig. 3f, red names), confirming that all 217 individual pDCs had responded to IFN-I at 36h post-infection. The expression of 218 219 IL12b, Ccr7 and several other genes occurred in C6 rather than C5 and was 220 maintained in C7, thus being delayed and prolonged as compared to IFN-I genes (Fig. 3f, violet names; Fig. 3i). 221

In summary, our enrichment strategy of $LIFR^{Io}$ cells improved recovery of the rare $Eyfp^+YFP^-$ pDCs, enabling the generation of a more robust scRNAseq dataset to characterize these cells including their relationships with other pDC activation states.

226 **Reconstruction of the pDC activation trajectory.**

We used Monocle to infer a pseudo-temporal sequence of pDC activation 227 states during MCMV infection (Fig. 4a), which yielded a distinctive linear branch 228 ordering pDCs clusters along pseudo-time in the following sequence: Eyfp⁻YFP⁻ (C2) 229 \rightarrow Eyfp⁺YFP⁻ (C5) \rightarrow Eyfp⁺YFP⁺ (C6) \rightarrow Eyfp⁻YFP⁺ (C7) (Fig. 3a and Fig. 4a). We 230 defined an IFN-I meta-gene and plotted its expression along pseudo-time in 231 232 comparison with that of other genes or of phenotypic markers. Transcription of the 233 IFN-I meta-gene started around pseudo-time 10, preceding detectable expression of 234 YFP around pseudo-time 12 (Fig. 4b, upper graph). At later pseudo-time, expression of the IFN-I meta-gene stopped whereas YFP expression remained high. LIFR 235 expression decreased to reach its minimal plateau around pseudo-time 10, 236 simultaneously to the initiation of IFN-I meta-gene expression (Fig. 4b, lower graph). 237 Hence, pseudo-time analysis fits with delayed but prolonged expression of YFP, as 238 239 compared to IFN-I transcription, and with LIFR downregulation upon pDC engagement into IFN-I production. 240

241 For independent trajectory inference, we used the Velocyto software, which computes for each cell a vector predicting its future state by precisely defining and 242 integrating the transcriptional states of all the cells from the dataset, based on their 243 unspliced/spliced mRNA ratio for many genes³⁸. The projection of Velocyto vectors 244 245 onto the UMAP space revealed the predicted pDC activation trajectory (Fig. 4c). YFP 246 expression was not taken into account in this model, and neither were Eyfp and IFN-I 247 genes since they are intron-less. Yet, Velocyto predicted the same activation trajectory (Fig. 4c) as Monocle (Fig. 4a) and YFP/IFN-I co-expression analysis (Fig. 248 249 1g-i).

RNA Velocity unraveled a new pDC activation state among C2, immediately preceding the initiation of IFN-I production corresponding to C5. We manually assigned these pDCs to "sub-cluster 8" (Fig. 4c, dotted black ellipse). Consistent with their location in the Velocyto activation trajectory, sub-cluster 8 pDCs presented a distribution of Monocle pseudo-time values higher than those of other C2 pDCs and just below those of C5 *Eyfp*⁺YFP⁻ pDCs (Fig. 4d), further illustrating the convergence and complementarity between Velocyto and Monocle for trajectory inference.

257 Both Monocle (Fig. 4e, Extended Data Fig. 2g) and Velocyto (Fig. 4f, 258 Extended Data Fig. 2h) predicted similar waves of gene expression along the pDC 259 activation trajectory. Tnf and Ccl3 induction preceded that of IFN-I genes, or of their 260 most correlated intron-bearing genes such as Kdm6b. II12b transcription started later, 261 followed by that of *Ccr7*, both reaching their maximum at the latest pseudo-times. As compared to the other C2 pDCs, sub-cluster 8 pDCs specifically expressed Eqr1. 262 also detected in the C5 *Eyfp*⁺YFP⁻ pDCs. Consistently, the pseudo-temporal analysis 263 264 showed that Egr1 expression pattern resembled and closely preceded that of the IFN-I meta-gene (Fig. 4g, top graph). Egr1 is a transcription factor induced early in 265 response to TLR triggering^{39, 40}. Hence, *Eqr1* expression is likely an early and 266 267 transient marker of the initial sensing of the viral infection by individual pDCs, and might promote their IFN-I production. Finally, Monocle showed a late induction of Id2 268 (Fig. 4g, bottom graph), which might have contributed to cDC gene induction in Eyfp⁻ 269 YFP⁺ pDCs. 270

We analyzed the kinetic of IFN-I, YFP and IL-12 expression in pDCs during infection (Extended Data Fig. 3). IL-12 peak production was reached later than that of IFN-I (Extended Data Fig. 3a-b), and, at each time point analyzed, most IL-12⁺ pDCs were YFP⁺ (Extended Data Fig. 3c). Thus, these results confirm at the protein level

that pDC IL-12 production during infection occurred mostly in the cells that expressed
IFN-I, but with a delay and for a longer time window.

277 We attempted characterizing by scRNAseg the activation trajectory of splenic 278 pDCs from 36h MCMV-infected WT C57BL/6 mice, upon harnessing LIFR 279 downregulation to enrich the pDCs engaged in IFN-I production. We used a different method for library preparation, FB5P-seq⁴¹, that overcomes the throughput limitations 280 281 of SS2, while keeping a better sensitivity for low copy number gene detection than 282 high-throughput droplet-based methods. We index-sorted and analyzed 188 pDCs from one uninfected mouse, and 468 pDCs from a 36h infected animal split in 2/3 283 LIFR^{hi} and 1/3 LIFR^{lo} pDCs (FB5P C57BL/6 dataset, Extended Data Fig. 4). Monocle 284 285 analysis resulted in a pDC activation trajectory bifurcating in two main branches, Y53 and Y50. The Y53 branch ended with IFN-I-expressing pDCs and harbored a 286 stronger downregulation of LIFR shortly after its bifurcation from the Y50 branch 287 (Extended Data Fig. 4a-c). The gene expression pattern of the IFN-I-producing pDCs 288 from WT C57BL/6 mice was very similar to that observed in *Ifnb1^{Eytp}* mice, with an 289 early induction of Cc/3 preceding that of the IFN-I genes, whereas II12b and Ccr7 290 expression were maximal at late pseudo-times (Extended Data Fig. 4b.d). 291

In conclusion, three independent methods based on distinct information and different methodologies, flow cytometry, Monocle and Velocyto, converged onto the same pDC activation trajectory during infection, both in *lfnb1^{Eyfp}* reporter and WT mice, with *Ccl3* and *Tnf* initial induction preceding the transient expression of IFN-I genes that was followed by prolonged *ll12* expression and late *Ccr7* induction.

297

An independent kinetic scRNAseq dataset confirms and extends the pDC activation trajectory.

300 To reinforce our results, we performed another scRNAseq experiment using 301 pDCs harvested at different times after infection: 0h, 33h, 36h and 48h (FB5P kinetic 302 dataset). We surmised that using LIFR downregulation on pDCs from WT animals may not guarantee sufficiently unbiased and efficient sampling of the different pDC 303 304 activation states for robust inference of their activation trajectories. Therefore, we used *lfnb1^{Eyfp}* mice. Total pDCs were sorted at 0h. For the other time points, we 305 sorted LIFR¹⁰ pDCs for enriching cells engaged in IFN-I production, as well as YFP⁺ 306 and bulk pDCs (Extended Data Fig. 5a-b). We removed contaminating pDC-like and 307 B cells (Extended Data Fig. 5b-d). 308

Monocle identified 15 pDC clusters (Fig. 4h) and inferred an activation 309 trajectory encompassing 4 branches (Fig. 4i). The pDCs from the uninfected mouse 310 and most of those from the 48h infected animal were at opposite ends of the longest 311 branch, whereas the pDCs from 33h and 36h infected animals were spread in-312 between (Fig. 4i-j). The pseudo-time analysis for this dataset confirmed the 313 progression of pDC activation from "early" (cluster 15, Eyfp⁺YFP⁻) to "peak" (clusters 314 4, *Evfp*^{hi}YFP⁺ and 7, *Evfp*^{hi}YFP^{hi}) to "late" (clusters 8, *Evfp*^{lo}YFP^{hi} and 9, *Evfp*⁻YFP^{hi}) 315 316 states (Fig. 4k). IFN-I transcription started around pseudo-time 25, preceding detectable YFP expression, and stopped around pseudo-time 35 (Fig. 4I) at the 317 transition from C7 to C8 (Fig. 4k), whereas YFP remained expressed at high levels 318 319 until late pseudo-times in C8 and C9 (Fig. 4k-l). The activation state inferred as the initiation phase of IFN-I production (C15) was more represented in pDCs harvested 320 at the earliest time points after infection (33h and 36h) and had disappeared at 48h 321 (Fig. 4m). Conversely, the late activation states corresponding to post-IFN-I 322

production (C8 and C9) were absent at 33h and increased at the latest time points 323 (Fig. 4m). 324

325 Velocyto confirmed in part the trajectory inferred from Monocle (C15, $Eyfp^+YFP^- \rightarrow C4$, $Eyfp^{hi}YFP^+ \rightarrow C7$, $Eyfp^{hi}YFP^{hi}$), as well as C9 as one endpoint of 326 pDC activation (Extended Data Fig. 5e). Results were less clear for the relationships 327 between C7, C8 and C9. However, computing Jaccard indexes to determine 328 329 equivalencies between the pDC clusters of the FB5P kinetic dataset and SS2 330 dataset#2 showed good consistency (Fig. 4n). Consistent with its late pseudo-time and detection only at 48h post-infection, the FB5P C9 *Evfp*⁻YFP^{hi} pDC cluster did not 331 strongly overlap with a specific 36h SS2 cluster (Fig. 4m). 332

333 In summary, this kinetic analysis confirmed and extended the interpretation drawn from the 36h dataset, showing that, upon commitment to IFN-I production, 334 individual pDCs sequentially underwent 5 distinct activate states: *Eyfp^{-//o}YFP⁻* (FB5P) 335 C15, SS2 C8) $\rightarrow Eyfp^{hi}YFP^{-/int}$ (FB5P C4, SS2 C5) $\rightarrow Eyfp^{hi}YFP^{hi}$ (FB5P C7, SS2 C6) 336 $\rightarrow Evfp^{-}VFP^{hi}$ (FB5P C8, SS2 C7) $\rightarrow Evfp^{-}VFP^{hi}$ (FB5P C9). 337

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IFN⁻YFP⁺ end-matured pDCs transcriptionally converge towards tDCs.

We examined in the FB5P kinetic dataset the expression along pseudo-time of 340 genes known to be expressed at steady state to higher levels in pDCs over tDCs 341 342 ("pDC-versus-tDC UP" meta-gene) or conversely ("pDC-versus-tDC DN" metagene). The latest IFN⁻YFP⁺ pDC activation state (FB5P C9, "end-matured") 343 344 transcriptionally converged towards tDCs (Fig. 5a). However, all pDCs expressed the pDC meta-gene at a higher level than the tDC meta-gene (Fig. 5a), and maintained a 345 346 distinctive expression of pDC signature genes including Ly6d, Ly6a, Klk1, Klk1b27,

and *Ccr9* to some extent (Fig. 5b). Some pDC-like cell or tDC genes were induced to high levels in a sizeable fraction of end-matured pDCs (Fig. 5c), which also harbored a decreased *Tcf4* and increased *Id2* expression (Fig. 5d). Hence, the end-matured pDCs did not convert into pDC-like cells/tDCs. Rather, the pDCs engaged into IFN-I production progressively converged transcriptionally toward tDCs/cDCs, already starting at the time of their peak IFN-I production.

353

354 GSEA-based inference of the molecular regulation and functional 355 specialization of pDC activation states.

We performed a Gene Set Enrichment Analysis (GSEA) using BubbleGUM⁴², to 356 357 identify candidate biological processes or upstream regulators associated to gene modules induced or repressed in the FB5P kinetic dataset along the pDC activation 358 trajectory (Fig. 6a). The expression of selected meta-genes (Fig 6b) and of one of 359 360 their representative genes (Fig. 6c, e) were plotted along pseudo-time. Egr1 was selectively expressed in C15, with a transient induction around pseudo-time 25 (Fig. 361 6c), consistent with its pattern of expression in the SS2 dataset#2 (Fig. 4g). ISG were 362 363 induced strongly early on, around pseudo-time 5 (cluster C1), followed by a further increase centered around pseudo-time 20 (C3), a decrease most pronounced around 364 pseudo-time 35 (C7), and a late re-expression (C8 and C9), as illustrated with Oasl2 365 (Fig. 6c). A similar expression pattern was observed for the genes induced by IFN-y 366 367 (Fig. 6a-c, e). The genes associated with TNF and NF-κB signaling also followed an induction pattern in three waves, but with a delayed central wave reaching its peak 368 369 immediately before maximal expression of IFN-I genes, around pseudo-time 25 (C15) (Fig. 6a-c, e). This suggested that TNF signaling through NF-κB might be one of the 370 371 early signals that pDCs need to receive together with TLR9 triggering to become high

IFN-I producers. The last wave of TNF response encompassed genes associated
with antigen processing and presentation, including *Cd83* (Fig. 6c), *Cd80*, *II12b*, *Relb*, *Rel*, *Tap1* and *Tapbp* (Fig. 6e, bold violet names). This implied that TNF could
be involved later in shutting off IFN-I production by pDCs and promoting their end
maturation, as previously reported for human pDCs^{4, 17, 43}.

377 The expression of gene modules associated with antigen processing and 378 presentation decreased progressively from early pseudo-time, until minimal 379 expression coinciding with peak expression of the IFN-I genes, followed by a sharp 380 increase at late pseudo-time (Fig. 6a, d-e). This is consistent with the transcriptomic 381 convergence of end-matured pDCs with tDCs/cDCs (Fig. 5). Genes contributing to oxidative phosphorylation followed the same expression pattern (Fig. 6a, e), 382 suggesting different metabolic regulation of IFN-I production versus cognate 383 interaction with T cells in pDCs. 384

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386 Cell-intrinsic TNF signaling promotes IFN-I production by pDCs.

Because *Tnf* expression and transcriptomic responses to TNF were induced in 387 pDCs earlier than IFN-I genes, we tested whether TNF activity promoted pDC IFN-I 388 production. In vivo antibody blockade of the TNF receptor did significantly reduce the 389 390 frequency of IFN-I⁺ pDCs at 36h post-infection (Fig. 7a-b). To test whether this process was due to cell-intrinsic TNF responses in pDCs, we examined IFN-I 391 392 production in CD45.1 versus CD45.2 pDCs from CD45.1⁺ WT: CD45.2⁺ WT versus CD45.1⁺ WT: CD45.2⁺ Tnfr1^{-/-}Tnfr2^{-/-} (TNFRDKO) mixed bone marrow chimera 393 mice infected with MCMV (Fig. 7c-d). We observed a significant decrease in the 394 CD45.2⁺ to CD45.1⁺ ratio within IFN-I⁺ pDCs in the TNFRDKO MBMC mice as 395

compared to their CTRL counterparts (Fig. 7d). Thus, cell-intrinsic TNF signaling is
 necessary for optimal IFN-I production by pDCs.

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399 YFP⁺CCR7⁺ pDCs harbor phenotypic and functional features of mature DCs.

Our scRNAseq data suggested sequential induction in single pDCs first of IFN-I genes, transiently, and then of genes modules associated with antigen processing and presentation or transcriptional convergence towards tDCs. We thus analyzed the phenotype, functions and anatomical location of the early, peak and late IFN-I-producing pDC activation states, to investigate their abilities to activate T lymphocytes.

A fraction of YFP⁺ pDCs transiently upregulated CCR7 between 36h and 44h 406 after MCMV infection (Fig. 8a-b), corresponding mostly to cells that were terminating 407 408 or had stopped IFN-I production (Fig. 8c). Since CCR7 in DCs promotes their migration to the T cell zones of secondary lymphoid organs for T cell activation, we 409 410 compared the expression of T cell regulatory signals by steady state pDCs and by the three final activation states of the pDCs engaged in IFN-I production in 36h 411 infected mice (Fig. 8d). CCR7 expression was significantly higher on IFN-I⁻YFP⁺ 412 pDCs than on IFN-I⁻YFP⁻, IFN-I⁺YFP⁻ and IFN-I⁺YFP⁺ pDCs. We sorted YFP⁻, YFP⁺ 413 CCR7⁻, and YFP⁺ CCR7⁺ pDCs from 36h infected *lfnb1^{Eyfp}* mice, cultured them for 8h 414 415 ex vivo in feeder CD45.1 bone marrow FLT3-L cultures and assessed whether their phenotype changed (Extended Data Fig. 6). All of the YFP⁺ CCR7⁻ pDCs increased 416 417 their expression of CCR7, with 37% becoming strongly positive, whereas YFP⁺ 418 CCR7⁺ pDCs maintained a stable phenotype. A very small proportion of the YFP⁻ pDCs became YFP⁺. Thus, this experiment confirmed the transition from YFP⁺ 419

420 CCR7⁻ to YFP⁺ CCR7⁺ pDC activation states, and the lack of reverse transition. The 421 YFP⁺ CCR7⁺ pDCs also expressed higher levels of MHC-II, CD80, CD86, CD40 and 422 PD-L1 (Fig. 8d). Hence, after termination of their IFN-I secretion, pDCs likely acquire 423 the ability to engage CD4 T cells in an antigen-specific manner, by up-regulating the 424 three signals required for this function.

We assessed the ability of different pDC activation states to induce the 425 426 proliferation of CD4 T cells in an antigen-specific manner, as compared to pDCs from 427 uninfected mice that are quite limited in this function. As CCR7 was upregulated preferentially on IFN-I⁻YFP⁺ pDCs (Fig. 8c), as a surrogate purification strategy for 428 this activation state we sorted YFP⁺CCR7⁺ pDCs from 38h MCMV-infected mice. We 429 sorted in parallel YFP⁻CD86⁻, YFP⁻CD86⁺ and YFP⁺CCR7⁻ pDCs (Fig. 8e). 430 YFP⁺CCR7⁺ pDCs were the only population harboring a significant higher ability for 431 CD4 T cell activation than pDCs from uninfected mice (Fig. 8f). We evaluated by 432 confocal microscopy whether and when pDCs migrated in the T cell area of the 433 spleen of MCMV-infected mice. At 36h post-infection, splenic YFP⁺ cells were in a 434 large part IFN-I⁺ (Fig. 8g and Extended Data Fig. 7a-c), and mainly located in the 435 marginal zone (Fig. 8g-j and Extended Data Fig. 7d) as previously published^{3, 22}. In 436 contrast, at 48h, splenic YFP⁺ cells were mostly IFN⁻ and located in the T cell zone 437 (Fig. 8g-k and Extended Data Fig. 7d). Hence, our scRNASeq, flow cytometry, 438 functional and microscopy data converge towards the conclusion: during MCMV 439 440 infection, after having produced IFN-I, pDCs undergo further activation promoting 441 their cognate interactions with CD4 T cells. Our study therefore supports a novel 442 model of the molecular and spatiotemporal regulation of pDC functions during viral 443 infections (Extended Data Fig. 8).

444 **Discussion**

It has been claimed that pDCs only produce IFN-I, because the pDC-like cells 445 or tDCs that can contaminate them fully accounted for their antigen presentation⁸⁻¹⁵. 446 However, under proper stimulation conditions, bona fide pDCs can exert both 447 functions^{16, 17}. Different models have attempted to explain how. First, pDC functional 448 plasticity could be developmental, with prominent IFN-I production by bone marrow 449 450 pDC precursors, before their terminal differentiation inducing their egress to the periphery and switch to antigen presentation¹⁸. Alternately, pDC functional plasticity 451 452 could be instructed by stimulation. IFN-I production versus antigen presentation could be induced in pDCs by distinct, eventually antagonist, stimuli⁴⁴, or reflect the 453 induction of a division of labor by a single stimulus. Indeed, in vitro exposure to the 454 influenza virus can instruct initially identical human pDCs to diversify into distinct and 455 non-interconvertible activation states, with some pDCs producing IFN-I and others 456 presenting antigens ¹⁷. Finally, antigen presentation by pDCs was proposed to result 457 from their cell fate conversion into cDC-like cells/tDCs^{19, 20, 45}, due to an activation-458 459 induced inversion of their Tcf4/ld2 expression ratio. However, the studies supporting these models were performed in vitro or upon genetic inactivation of *Tcf4*, and under 460 conditions unable to assess dynamically the fate of individual pDCs. Rigorously 461 characterizing pDC functional plasticity and cell fate during viral infections in vivo thus 462 remains an important challenge^{19, 20}, which we addressed by combining fate-mapping 463 of individual pDCs for IFN-I production with flow cytometry, scRNAseq, 464 465 immunohistological imaging and an ex vivo assay for CD4 T cell priming.

We identified different pDC activation states during MCMV infection and deciphered their relationships, showing that the same individual pDCs first produce

IFN-I and then activate T cells in the spleen of infected mice, contrary to what was
 proposed in the previous models of pDC plasticity discussed above.

Steady state pDCs were transcriptionally homogeneous, as previously reported^{9, 11, 14, 46}. Thus, it is unlikely that the minor pDC fraction producing IFN-I upon activation corresponds to a distinct subset already primed for this function at steady state, at least not in a manner detectable transcriptionally.

474 A recent study reported the gene expression profiles and anatomical locations of YFP⁺ versus YFP⁻ pDCs from CpG-injected *Ifnb1^{Eyfp}* mice⁴⁷. YFP⁺ pDCs 475 476 upregulated CCR7, were located in the T cell zone and strongly expressed genes 477 coding for T cell-activating cytokines. Hence, professional IFN-I-producing cells were proposed to be a distinct and polyfunctional pDC subset, specialized in 478 simultaneously producing many cytokines, in lymphoid organ T cell zones, to 479 coordinate innate and adaptive cellular immunity⁴⁷. However, this study was 480 performed at bulk level and only at one time point, preventing any definite conclusion 481 on gene co-expression in single pDCs and on their ability to simultaneously perform 482 different functions. In contrast, here we show that the YFP⁻ and YFP⁺ pDC 483 populations are both heterogeneous, encompassing distinct activation states, and 484 that, in vivo during MCMV infection, IFN-I production by pDCs is dissociated in time 485 486 and space from their acquisition of the ability to interact with T cells in a cognate 487 manner.

We identified LIFR downregulation as a novel phenotypic marker of pDC commitment to IFN-I production. LIFR is specifically expressed by pDCs and can inhibit their IFN-I production⁴⁸. Hence, the failure of most pDCs to strongly downregulate LIFR during infection may contribute inhibiting their IFN-I production.

Pseudo-time analysis identified gene modules dynamically regulated along the 492 pDC activation trajectory. Genes in the TNF-to-NFKB signaling pathway were 493 induced early during pDC commitment to IFN-I production, consistent with the cell-494 intrinsic role of *lkbkb* in promoting this function²² and suggesting that it occurs in part 495 downstream to TNF stimulation which we proved experimentally. Yet, IFN-I and TNF 496 play antagonistic roles in health and disease⁴³. Autocrine TNF production by human 497 pDCs in vitro does not impede their simultaneous production of IFN-I, whereas 498 exposure to exogenous TNF prior to TLR triggering does⁴³. TNF promotes pDC 499 maturation^{4, 17, 43}, which is tightly associated with IFN-I production shutdown, as 500 reported on bulk pDCs ⁴⁴ and here at the single-cell level. We observed three 501 502 consecutive waves of TNF responses in pDCs during infection. The last wave 503 correlated with the induction in pDCs of genes associated to antigen processing and 504 presentation. Hence, TNF may play different roles on pDCs depending on its origin, concentration and timing of delivery as compared to that of other signals. We 505 506 identified transient Egr1 induction, and shutdown of oxidative phosphorylation, as 507 candidate mechanisms for the induction and termination of pDC IFN-I production, 508 respectively. Further studies are warranted to test these hypotheses.

Unexpectedly^{19, 49}, pDCs did not die early after termination of their IFN-I 509 production. They survived for over 12 hours, producing IL-12, upregulating CCR7 and 510 511 acquiring the ability to activate naïve CD4⁺ T cells. We identified a novel end-mature 512 state of activated pDCs, resulting from a progressive transcriptional convergence 513 towards tDCs, which started at the state of peak IFN-I production and occurred in 514 parallel to an inversion of the Tcf4/Id2 expression ratio. This physiological observation is reminiscent of the acquisition of cDC features caused in pDCs by their 515 genetic inactivation for Tcf4⁵⁰. However, end-matured pDCs maintained a clear 516

517 imprint of their lineage identity, rather than undergoing the proposed cell fate 518 conversion into tDCs⁴⁵.

519 In conclusion, our study uncovered the dynamic activation trajectory of pDCs 520 during MCMV infection in vivo, leading us to propose a novel model of the functional 521 plasticity of pDCs, explaining how individual pDCs exert different functions in a manner tightly orchestrated in time and space. IFN-I production is exerted first, 522 523 transiently, at the site of virus replication in the marginal zone of the spleen, whereas 524 antigen presentation occurs later, after a further activation of pDCs leading to their 525 transcriptional, phenotypic and functional convergence towards tDCs, including their 526 migration to the splenic T cell zone. Future work will examine whether similar pDC 527 activation trajectories occur during other systemic viral infections, including in primates. Once suitable mouse models have been generated²¹, it will be possible to 528 529 assess the importance of antigen presentation by pDCs for shaping antiviral adaptive 530 immunity.

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557 Author contributions

558	Studies were designed by E.T. , M.D. , A.A. and TP.V.M., with help from A-C.V and
559	P.M. Experiments were performed by A.A., E.T., M.V., N.C. and N.A., with help from
560	K.N. and G.Be. Data were analyzed by A.A., TP.V.M., M.V., N.C., M.D. and E.T.,
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563	P.VM., M.D. and E.T. All authors contributed to discussions and manuscript edition.

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565 **Competing interests**

566 The authors declare no competing interests.

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568 **Online content**

569 Any methods, additional references, Nature Research reporting summaries, 570 source data, extended data, supplementary information, acknowledgements, peer 571 review information; details of author contributions and competing interests; and 572 statements of data and code availability are available at XXX.

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Fig. 1. Bulk transcriptional profiling suggests the induction of distinct pDC 699 activation states in vivo during MCMV infection. a, pDC gating strategy within live 700 single-cells. **b**, Percentages of IFN-I⁺ cells within pDCs at indicated time points 701 702 afterMCMV infection. *p<0.05, **p<0.01 (One-way ANOVA). c, YFP and CD86 coexpression in pDCs isolated from uninfected (UN) or 36h MCMV-infected (IN) 703 Ifnb1^{EYFP} mice. d, Percentages of subpopulations within splenic pDCs of 36h MCMV-704 infected *lfnb1^{EYFP}* mice. The data are shown for n=9 individual mice pooled from 3 705 independent experiments. e, Heatmap showing mRNA expression levels of selected 706 genes (rows) across pDC subpopulations (columns), with hierarchical clustering 707 using City block distance for cells and Euclidian distance for genes. f, Co-expression 708 of IFN- α/β and YFP in pDCs isolated from one representative 36h MCMV-infected 709 710 If $nb1^{Eytp}$ mouse. **g**, Co-expression of IFN- α/β and YFP in pDCs isolated from If $nb1^{Eytp}$ 711 mice, at 0h (UN), 33h, 36h, 40h, 44h and 48h after MCMV infection. For each time point, one representative mouse is shown. **h**, Proportions of IFN- α/β^+ YFP⁻, IFN-712 α/β^+YFP^+ and IFN- α/β^-YFP^+ cells amongst pDCs, at indicated time points. i. Pie 713 charts recapitulating the mean proportions of cells expressing IFN-I and/or YFP (see 714 color key) amongst pDC positive for either molecule at different time points during the 715 course of MCMV infection in *Ifnb1^{Eyfp}* mice. For panels b, d and h, data are presented 716 717 as mean±s.e.m, and for panel i as mean percentage. Panels b and f-i show data from individual mice, with n=5 at 0h, 7 at 33h, 10 at 36h, 5 at 40h, and 3 at 44h and 48h, 718 pooled from 2 (resp. 3) independent experiments for 33h and 40h (resp. 0h and 36h). 719 One experiment was performed for 44h and 48h. 720

Fig. 2. scRNAseg analysis of pDCs from 36h MCMV-infected mice confirms 722 their heterogeneity and pinpoints to LIFR downregulation as a selective marker 723 724 of IFN-I-producing pDCs. a, Flow cytometry gating strategy for index sorting of pDC from one uninfected mouse (top) and from one 36h MCMV-infected mouse (bottom). 725 726 Numbers in parentheses correspond to the number of cells sorted for each population. b-c, Principal component analysis (PCA) on 1,016 highly variable genes 727 of single pDCs isolated from uninfected, b, or from one MCMV-infected mice (SS2 728 dataset#1), c, encompassing 29 YFP⁺ cells (red) and 65 YFP⁻ cells (blue). d, 729 Heatmap showing mRNA expression of selected genes, and YFP protein 730 731 fluorescence intensity obtained from index sorting data, with hierarchical clustering 732 using one minus Pearson correlation as distance metric for both cells and genes. e, Scatter plots showing *lfnb1* expression vs YFP protein fluorescence intensity (left), or 733 Ifnb1 vs Eyfp mRNA expression (right). f, Violin plot showing Lifr expression in YFP⁻ 734 vs YFP⁺ pDCs. **g**, Scatter plot showing *lfnb1* vs *Lifr* mRNA expression. **h**, Flow 735 cytometry analysis showing the downregulation of LIFR protein expression on IFN-736 α/β^+ (middle) or YFP⁺ (right) pDC isolated from 36h MCMV-infected *lfnb1^{EYFP}* mice. 737 738 as compared to pDC isolated from uninfected animals (left). The dot plots shown are from one mouse representative of 10 animals. i, Relative median fluorescence 739 intensity (MFI) of LIFR on IFN- α/β^-YFP^- (black), IFN- α/β^+YFP^- (blue), IFN- α/β^+YFP^+ 740 741 (green) and IFN- α/β^- YFP⁺ (pink) populations. The data are shown for n=10 individual 742 animals pooled from 2 independent experiments, with overlay of mean±s.e.m. values. 743 ****p<0.0001 (One-way ANOVA with Tukey's post hoc test).

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Fig. 3. scRNAseq analysis identifies 8 different pDC activation states in vivo
 during MCMV infection. a, Dimensional reduction performed using the UMAP

algorithm, and graph-based cell clustering, for 264 bona fide pDCs isolated from one 747 control and one 36h MCMV-infected mice (SS2 dataset#2). b, Inverse hyperbolic 748 749 arcsine (asinh) fluorescence intensity of YFP projected on the UMAP space. c-d, Expression of Yfp and Ifnb1 on the UMAP space. e, Violin plots showing mRNA 750 751 expression profiles of selected genes across all individual cells and in comparison 752 between the clusters identified in (a), cluster 0= UN pDCs, clusters 1-4= MCMV Eyfp⁻ YFP⁻ pDCs, cluster 5= MCMV Eyfp⁺YFP⁻ pDCs, cluster 6= MCMV Eyfp⁺YFP⁺ pDCs, 753 cluster 7= MCMV Evfp⁻YFP⁺ pDCs. f, Heatmap showing mRNA expression levels of 754 755 selected genes (rows) across individual pDCs (columns), with hierarchical clustering 756 using one minus Spearman rank correlation for cells and Kendall's tau distance for 757 genes. The top differentially expressed genes between Seurat clusters are shown, as well as representative ISG and pDC-specific genes. B2m was included as an 758 invariant control housekeeping gene. YFP and LIFR protein fluorescence intensities, 759 obtained from index sorting data, are also shown on the top of the heatmap, as well 760 761 as the infection status of the mice from which the pDCs were isolated, and the belonging of individual pDCs to the Seurat cell clusters. **g**, LIFR expression intensity 762 763 projected on the UMAP space. h, Violin plots showing asinh fluorescence intensity of EYFP and LIFR across Seurat clusters. i, Violin plots showing mRNA expression 764 765 profiles of selected genes across Seurat clusters.

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Fig. 4. Inference of the pDC activation trajectory during MCMV infection. a, Monocle pseudo-temporal inference of the pDC activation trajectory for the SS2 dataset#2. **b**, Expression of the YFP or LIFR proteins vs the IFN-I meta-gene along pseudo-time. Dots correspond to individual cells and a 6-order polynomial curve was fit to the data. **c**, RNA Velocity reconstruction of the pDC activation trajectory.

Projections of the velocity vector of each pDC are represented as arrows in the 772 773 UMAP space. The black dotted ellipse corresponds to the new subcluster 8. d, Violin 774 plots showing pseudo-time distribution across Seurat clusters, including subcluster 8. e, Expression along pseudo-time of the indicated genes vs the IFN-I meta-gene, 775 776 each normalized to their maximal value. f, Predicted induction (red) vs termination (blue) of the transcription of selected genes as obtained using Velocyto. g, 777 Normalized expression along pseudo-time of Egr1 (top) and Id2 (bottom) vs the IFN-I 778 meta-gene. h. UMAP and clustering analysis of the FB5P kinetics dataset after 779 780 removal of contaminants. i, Pseudo-temporal inference of the pDC activation 781 trajectory obtained using Monocle. j, Distribution in the UMAP space of the cells 782 isolated at different time points after infection. The cell numbers analyzed for each time point (black dots) are indicated in parentheses. k, Violin plots showing the 783 values for pseudo-time (top), Eyfp (middle) and YFP (bottom) expression, across cell 784 clusters. I, Eyfp and YFP expression across pseudo-time. Individual cells are shown 785 786 as dots and a polynomial curve was fit to the data. \mathbf{m} , Calculus of the frequency over real-time of pDCs being in the different activation states (cell clusters) associated to 787 788 IFN-I production. n, Heatmap showing the equivalences between the pDC clusters from the SS2 dataset#2 and the FB5P kinetics dataset based on the calculation of 789 790 Jaccard Indexes reflecting their marker gene content overlap. One minus Pearson 791 correlation was used as distance metric for hierarchical clustering.

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Fig. 5. Transcriptional convergence of pDCs towards tDCs over pseudo-time. a, Meta-gene expression levels (y-axis) along pseudo-time (x-axis), in the FB5P kinetic dataset, for genes higher in steady state pDCs over tDCs (pDC_versus_tDC_UP gene set, dark red dots and upper black curve) and for the genes lower in steady

state pDCs as compared to tDCs (pDC versus tDC DN gene set, violet dots and 797 798 lower black curve). Individual cells are shown as dots and a polynomial curve was fit 799 to the data. The gene sets were generated by reanalyzing public data (GEO Series GSE76132, see Supplementary Data 2). b, Expression level along pseudo-time of 800 801 representative individual genes of the pDC_versus_tDC_UP gene set. c, Expression level 802 along pseudo-time of representative individual genes of the 803 pDC_versus_tDC_DN gene set. d, Expression level along pseudo-time of the pDC master transcription factor Tcf4 (blue dots and black curve) and of its counter-804 805 regulated cDC1 master transcription factor Id2 (pink dots and black curve). The 806 expression of the IFN-I meta-gene is shown on each graph for comparison (pale red 807 dots and gray curve).

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Fig. 6. Identification of annotated gene modules regulated along pseudo-time 809 during pDC activation. a, BubbleMap showing the gene sets with the most relevant 810 811 and significant enrichment patterns along the pseudo-time of pDC activation 812 trajectory. Dark red bubbles represent significant increased expression of the gene 813 set (columns) across two consecutive pDC clusters along pseudo-time (rows). Dark blue bubbles represent significant decreased expression. Empty bubbles correspond 814 815 to lack of significant differences in gene set expression between the pDC clusters compared. Hence, the expression pattern of each gene set along pseudo-time is 816 visualized as a vertical succession of red or blue bubbles. b,c Normalized expression 817 818 level (y-axis) along pseudo-time (x-axis) of the meta-genes corresponding to the 819 selected gene sets (left) and of one of their representative genes (right) (pale orange 820 dots and black curves). The expression of the IFN-I meta-gene is shown on each graph for comparison (pale red dots and gray curve). d, Normalized expression level 821

822 (y-axis) along pseudo-time (x-axis) of the metagene corresponding to the antigen 823 presentation gene set, vs the IFN-I meta-gene. e, Heatmap showing the expression 824 patterns along pseudo-time of individual genes selected from the analysis illustrated on panel (a), with hierarchical clustering of the genes using one minus Pearson 825 826 correlation as distance metric. Belonging of the genes to the gene sets illustrated in 827 a-c is shown in the grid on the right of the heatmap, with gene set 1, ISG; 2, IFN-y 828 responsive genes; 3, TNF signaling pathway and responsive genes; 4, pDC Myd88dependent genes UP: 5, KEGG TLR signaling pathway: 6, Reactome Traf6-mediated 829 830 IRF7 activation; 7, NFKB signaling pathways and responsive genes; 8, Zwang class 831 2 transiently induced by EGF; 9, CHO NR4A1 targets; 10, KEGG Oxidative 832 phosphorylation; 11. GO antigen processing and presentation; 12. pDC_versus_tDC_DN; 13, pDC_versus_tDC_UP. 833

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Fig. 7. Cell-intrinsic TNF signaling promotes pDC IFN-I production. a, 835 836 Experimental protocol followed to evaluate the role of TNF signaling in pDC activation during MCMV infection. **b**, Percentages of IFN⁺ cells in splenic pDCs isolated from 837 isotype control (IC)-treated animals vs anti-TNFR2-treated mice. The data are shown 838 for 10 individual animals pooled from 2 independent experiments, with overlay of 839 840 mean±s.e.m. values. c, Generation of mixed bone marrow chimera (MBMC) mice; CTR, control. Recipient C57BL/6 CD45.1 (5.1) mice were lethally irradiated and then 841 842 reconstituted with equal proportions of bone marrow (BM) cells isolated from WT 5.1 843 mice and from indicated CD45.2 (5.2) mice, either WT for CTR MBMC mice or 844 TNFR1/2-KO (TNFRDKO) for TEST MBMC animals. Reconstituted MBMC mice were 845 infected by MCMV and their splenic pDCs examined ex vivo for IFN-I production. d, Measuring the cell-intrinsic role of TNF signaling in promoting pDC IFN-I production. 846

Results are expressed as 5.2/5.1 ratio of the percentages of IFN-I-producing pDCs obtained from MCMV-infected CTR WT versus TEST TNFRDKO MBMC mice. The data are shown for 5 individual animals for each type of MBMC mice, pooled from 2 independent experiments, with overlay of mean±s.e.m. values. For b and d, the statistical analysis was performed using one-sided Mann-Whitney U test. **e**, Dot plots show representative data of intracellular IFN-I staining in 5.2⁺ versus 5.1⁺ pDCs isolated from the indicated MCMV-infected MBMC mice.

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855 Fig. 8. Characterization of the phenotype, function and micro-anatomical location of pDC activation states. a, YFP vs CCR7 expression in splenic pDCs 856 during MCMV infection. Data from one representative *lfnb1^{Eyfp}* mouse is shown for 857 each time points. b, Frequency of YFP⁺CCR7⁺ cells within splenic pDCs during 858 MCMV infection. The data represent individual mice with n = 5 at 0h, 7 at 33h, 10 at 859 36h, 5 at 40h, and 3 at 44h and 48h, from one experiment for 44h and 48h, and 860 861 pooled from 2 (resp. 3) independent experiments for 33h and 40h (resp. 0h and 36h). 862 c, Mean fluorescence intensity of IFN- α/β on YFP⁺CCR7⁺ vs YFP⁺CCR7⁻ pDCs isolated from 36h MCMV-infected mice. d, Relative median fluorescence intensity 863 (MFI) of indicated markers on pDC subpopulations isolated from 36h MCMV-infected 864 *lfnb1^{Eyfp}* mice. The data in c and d are from n = 10 (resp. 5) mice from two 865 independent experiments (c) or one experiment representative of two (d). e, Flow 866 cytometry sorting strategy for splenic pDC subpopulations from uninfected (UN) or 867 38h MCMV-infected *lfnb1^{Eyfp}* mice, starting from live single CD11b^{neg} Lin^{neg} cells. **f**, 868 Expansion of naïve CD4 OT-II cells upon co-culture with the indicated OVA peptide-869 pulsed pDC subpopulations. The graph shows individual data points pooled from 3 870 independent experiments, each with 2-4 replicate co-cultures for each pDC 871

subpopulation. g, Immunohistological analysis of spleen sections from uninfected 872 (UN) or MCMV-infected *lfnb1^{Eyfp}* mice harvested at the indicated time points. Top 873 right image, 5x zoom of the region delimited in the previous image. h, Definition of 874 spleen zones for cell quantification (see online methods). i, Distribution of YFP⁺ cells 875 across the different spleen zones during the course of MCMV infection. j, Detail of 876 877 the individual data collected and used for generating the graph of panel I. k, A similar analysis was performed as in panel (j) for the percentages of WP YFP⁺ pDCs 878 residing in the TCZ. For g-k, data were analyzed from 3 different mice, with 5 879 different whole splenic sections per mouse (i.e. 15 images per time point). In b-d,f,i-k, 880 881 data are shown as mean±s.e.m. and P values are from One-way ANOVA with Tukey's post hoc test, with *p<0.05, **p<0.01, ***p<0.001, ***p<0.0001. 882

884 **On-line methods**

885 **Mice**

886 C57BL/6 and BALB/c mice were purchased from Janvier, France. Congenic CD45.1 C57BL/6 mice were purchased from Charles River. Italy. All other mouse strains 887 were bred at the Centre d'ImmunoPhénomique (CIPHE) or the Centre d'Immunologie 888 de Marseille-Luminy (CIML), Marseille, France, under specific pathogen free-889 conditions and in accordance with animal care and use regulations. *Ifnb1^{Eyfp} (B6.129-*890 Ifnb1^{tm1Lky}) mice²⁵ were purchased from Jackson Laboratories, USA. Tnfr1^{-/-}Tnfr2^{-/-} 891 (B6.129S-Tnfrsf1a^{tm1lmx};Tnfrsf1b^{tm1lmx}/J, TNFRDKO) were provided by J-.L.D. All 892 893 animals used were age-matched females (8 to 12 weeks old unless specified otherwise). The animal care and use protocols were designed in accordance with 894 national and international laws for laboratory animal welfare and experimentation 895 (EEC Council Directive 2010/63/EU, September 2010). They were approved by the 896 Marseille Ethical Committee for Animal Experimentation (registered by the Comité 897 898 National de Réflexion Ethique sur l'Expérimentation Animale under no. 14; authorization #11-09/09/2011 and APAFIS#1212-2015072117438525 v5). 899

900 **Virus**

Virus stocks were prepared from salivary gland extracts of 3-weeks old MCMVinfected BALB/c mice. All mice used in the experiments were infected i.p. with 10^5 pfu Smith MCMV and sacrificed at indicated time points.

904 In vivo treatment with blocking antibodies

To block TNFR2, mice were injected i.p. 24 hrs before sacrifice with 0.5 mg of Armenian Hamster IgG (isotype control) or of anti-TNFR2 mAbs (both from BioXCell).

907 Generation of Mixed Bone Marrow Chimera (MBMC) mice

Congenic CD45.1 mice were 7 Gy irradiated, then reconstituted with equal proportions of BM cells derived from CD45.1 animals and from CD45.2 mice either wild type (WT) or deficient for the indicated genes. Mice were used at least 8 weeks after BM reconstitution.

912 Cell preparation, flow cytometry analysis and cell sorting

Spleens were harvested and submitted to enzymatic digestion for 25 minutes at 37°C 913 with Collagenase IV (Wortington biochemicals) and DNAse I (Roche Diagnostics), in 914 915 the presence of Brefeldin A (10 µg/ml, Sigma-Aldrich) when intracellular cytokine 916 staining was required. Red blood cells were then lysed by using RBC lysis buffer (Life 917 Technologies), and pDCs were enriched with the mouse plasmacytoid dendritic cell isolation kit (Miltenyi Biotech, France). Extracellular staining with BV650 anti-BST2 918 (or alternatively with PerCP-eF710 anti-BST2), Alexa700 anti-CD3e, APC-Cy7 anti-919 920 CD11b, Alexa700 anti-Ly-6G, Alexa700 anti-NK1.1, Alexa700 anti-CD19 and BV421 921 anti-CD11c (or alternatively with BV785 anti-CD11c) was performed in PBS 1X supplemented with 2 mM EDTA (Sigma-Aldrich) and 0.5% bovine serum albumin 922 923 (BSA, H2B, Limoges, France). A complete list of the antibodies with clones and dilutions used for flow cytometry is provided in the Nature Research Reporting 924 925 Summary. All extracellular stainings were performed for 20 min, at 4°C with the 926 exception of anti-mouse CCR7 mAb which was performed at 37°C. Dead cell staining 927 (LIVE/DEAD[™] Fixable Aqua Dead Cell Stain, Life Technologies) was performed in 928 PBS 1X as per the manufacturer's recommendations. For preservation of the YFP signal in pDC from *Ifnb1^{Eyfp}* mice treated for intracellular staining, cells were first fixed 929 with PBS 1x containing 2% of formaldehyde solution (Sigma-Aldrich) for 40 min, 930 931 followed by incubation with Cytofix/Cytoperm (BD Biosciences) for 20 min. Cells were

then stained with specific anti-cytokine mAbs diluted in 1x Perm/Wash (BD 932 Biosciences) following the manufacturer's recommendations. Purified anti-mouse 933 934 IFN- α (RMMA1) and IFN- β (RMMB1) were purchased from PBL Interferon Source; purified rat IgG was used as isotype control (Jackson Immuno Research); these 935 mAbs were coupled to Alexa647 by using a monoclonal labelling kit (Southern 936 Biotech)²². Samples were acquired on a LSR Fortessa X-20 (BD Biosciences, 937 France) and analyzed with FlowJo software (FlowJo, LLC). For microarray studies, 938 pDCs were sorted to high purity by flow cytometry on a FACS Aria II cell sorter (BD 939 Biosciences, France) and directly collected in RLT buffer (Qiagen) supplemented with 940 941 10% β -mercaptoethanol. For SS2 scRNAseq, cells were index sorted on a BD FACS Aria III in 96-well plates, with each well filled with 10µL of lysis buffer (TCL buffer, 942 943 Qiagen, supplemented with 1% β-mercaptoethanol). For FB5P single-cell RNA sequencing, cells were index sorted on a BD FACS Influx in 96-well plates, with each 944 well filled with 2µl FB5P lysis buffer, as described⁴¹. 945

946 Generation of FLT3-L bone marrow cultures

Bone marrow cells were isolated from CD45.1 C57BL/6 mice and cultured for 6-7 days at 2 10^6 cells/ml in 24-well plates in RPMI supplemented with 10% FCS, 1% glutamine (Gibco), 100U/mL Penicillin streptomycin, 1% non-essential amino acids, 1% sodium pyruvate and 0.05 mM β -mercaptoethanol, with a 1/10 dilution of B16-FLT3-L culture supernatant.

952 Immunohistological analysis

Spleen fragments isolated from uninfected or MCMV-infected *lfnb1^{EyfP}* reporter mice were fixed for 2 hours at 4°C with Antigenfix (DiaPath), then washed in phosphate buffer (PB1X: 0.025 M NaH2PO4 and 0.1 M Na2HPO4) for 1 hour, dehydrated in

30% sucrose overnight at 4°C, and embedded in OCT freezing media (Sakura 956 Fineteck). 20µm-thick tissue sections were blocked in PB1X containing 0.5 % 957 958 saponin, 2% BSA, 2% 2.4G2 supernatant and streptavidin/biotin blocking kit (Vector Laboratories), and first stained in PB1X, 0.5% saponin, containing the following 959 960 antibodies: rat anti-mouse B220 (BD Bioscience) followed by Alexa546-conjugated 961 goat anti-rat mAb (Molecular probes) and hamster anti-mouse CD3 (BD Bioscience) followed by Alexa594-conjugated goat anti-hamster mAb (Jackson Immunoresearch). 962 Tissue sections were then blocked with 5% rat serum (Jackson Immunoresearch) in 963 PB1X and stained in PB1X, rat serum 5%, 0.5% saponin with the following 964 965 antibodies: Alexa488-conjugated rabbit anti-GFP/YFP (Thermofisher), biotin rat anti-966 mouse CD169 (MOMA1, AbCam) followed by eFluor450-conjugated streptavidin (Thermofisher), and Alexa647-conjugated rat anti-mouse IFN α/β (Thermofisher)²². 967 968 Stained sections were mounted in ProLong Gold antifade reagent (Thermofisher) 969 mixed with SYTOX Blue nucleic acid stain (Thermofisher), acquired on a LSM 880 970 confocal microscope (Zeiss) and analyzed with ImageJ software. The red pulp (RP) 971 was identified as the zone outside of the external boundary of CD169 staining. The 972 marginal zone (MZ) was delimited between the external boundary of CD169 staining and the external boundary of B220 staining. The white pulp (WP) was identified as 973 974 the zone included within the external boundary of B220 staining. The T cell zone 975 (TCZ) corresponds to the WP area encompassing CD3 staining, whereas the B cell 976 zone (BCZ) was identified within WP as selectively including B220 staining.

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978 Microarray data generation and analysis

Pooled spleens of uninfected or MCMV-infected mice were used in order to obtain10,000 cells for each pDC subpopulation, with two independent duplicate samples

generated per condition. Total RNA was prepared as described²² and converted to
biotinylated double-stranded cDNA targets which were hybridized on GeneChip
Mouse Gene 1.0 ST arrays (Affymetrix) that were scanned to extract raw data (.CEL
Intensity files). Bioinformatics analyses were performed as previously described^{51, 52}
and specified in Figure legends. Heatmaps were generated using Gene-E
(http://www.broadinstitute.org/cancer/software/GENE-E/).

987 Single-cell RNA sequencing

988 For SS2 scRNASeq, single cells were sorted into 96-well full-skirted Eppendorf plates 989 chilled to 4°C, prepared with lysis buffer. Single-cell lysates were sealed, vortexed, 990 spun down at 300g at 4°C for 1 min, immediately placed on dry ice and transferred for storage at -80° C. The SS2 protocol³¹ was performed on single-cells as previously 991 described ⁵³. Briefly, plates containing cell lysates were thawed on ice, followed by 992 RNA purification and first strand synthesis. cDNA was then amplified by PCR for 22 993 cycles and libraries were generated using ¼ of Nextera XT library preparation kit 994 (Illumina) with custom barcodes as described¹⁴. Libraries were then pooled 995 996 equimolarly and sequenced on an Illumina HiSeq 4000 sequencer (paired-end, 50-bp reads). FACS-based 5-prime end single-cell RNAseg (FB5P-seg) was performed as 997 described⁴¹. 998

999 Single-cell RNA sequencing data analysis

For SS2 scRNAseq, raw FASTQ files were aligned with STAR ⁵⁴ against the GRCM38.94 genome to which the sequence of *Eyfp* was added as an extra chromosome (see Supplementary text for the *Eyfp* sequence). Gene-specific read counts were calculated using HTSeq-count⁵⁵. FB5P sequencing data were processed to generate single-cell UMI counts matrices as described⁴¹. The counts matrix was

loaded to R, and Seurat ⁵⁶ was used for downstream analyses such as normalization 1005 (LogNormalize), shared nearest neighbor graph-based clustering, differential 1006 expression analysis and visualization, eventually followed by Monocle^{57, 58} or 1007 Velocvto³⁸ for the inference of the pDC activation trajectory, as summarized in 1008 Supplementary Data 1 and described below. Unless specified otherwise, gene 1009 expression is shown as log-normalized values and protein expression as inverse 1010 hyperbolic arcsine (asinh) of fluorescence intensity raw or scaled values, namely 1011 asinh(fluorescence) for SS2 dataset#1, asinh(fluorescence/100) for SS2 dataset#2, 1012 and asinh(fluorescence/10) for FB5P datasets. For dimensionality reduction, we 1013 1014 performed a uniform manifold approximation and projection (UMAP), using the RunUMAP function. The different clusters were projected on the UMAP plot for 1015 visualization. The differentially expressed genes were determined using the 1016 FindAllMarkers function. The heatmaps were plotted using the Gene-E program. For 1017 some of the datasets, a first analysis was performed to identify and exclude 1018 contaminating cells, as summarized in the corresponding Figure legends and in 1019 Supplementary Data 1. 1020

Alignment of clusters from the SS2 dataset#2 and the FB5P kinetic dataset

In order to compare the clusters obtained in the SS2 dataset#2 and the FB5P kinetic dataset, we first extracted the marker genes of each cluster for each dataset separately, using the FindMarker function of Seurat, using the "bimod" statistical test and an adjusted p-value cut-off of 0.25. We then computed the Jaccard Indexes (JI) between each cluster of one dataset with any other cluster of the other dataset. The JI measures similarity between finite sample sets, and is defined as the size of the intersection divided by the size of the union of the sample sets: $J(A,B) = |A \cap B| / |A|$

U B|. The JI matrix was then used to generate a heatmap with hierarchical clustering,using Gene-E.

1031 Monocle analysis

We used Monocle3 to investigate cell trajectories and the dynamics of gene 1032 expression within these trajectories. The cells from the uninfected mouse were set as 1033 1034 the root of the trajectory and the input genes were selected using Seurat. We first clustered and projected the cells onto a low-dimensional space (UMAP) calculated 1035 1036 either by Monocle or by Seurat. Monocle then resolved the activation trajectories and 1037 calculated along them the pseudo-time for each cell. In some cases where several 1038 trajectories were found, we exported the branches identified, using the centroid-Monocle 1039 based principal graph generated by (https://gist.github.com/mochar/70e67bfba87ceedbd1435b83e6464ba2). 1040 Each individual path from the starting point (cells from non-infected mice) to each of the 1041 end points of the constructed graphs were retrieved. The expression profile of genes 1042 1043 of interest were then projected on the cells of chosen branches. For certain gene 1044 modules or gene sets, a meta-gene was defined as the average of log-normalized 1045 values of all of its member genes that were included in the Leading Edge of at least one of the GSEA results illustrated in Fig. 6a. The expression of this meta-gene was 1046 then plotted, with each cell shown as a dot and a polynomial curve computed to 1047 illustrate the trend in dynamical expression change of the gene set along the pseudo-1048 time. When specified, the expression levels of each meta-gene or gene was 1049 1050 normalized, to its maximal value, to ease comparison of the temporal evolution of 1051 their expression pattern.

1052 RNA velocity analysis

RNA velocity was performed using the Velocyto software. Velocyto predicts the future 1053 direction in which cells are moving in the transcriptional space by estimating, for each 1054 cell, the time derivative of RNA abundance³⁸. Specifically, it computes the predicted 1055 fate of each single cell based on the integration of its precise transcriptional state for 1056 hundreds of intron-bearing genes, as assessed by their ratio of unspliced to spliced 1057 RNA, in comparison with the other cells. The directionality of the transitions is 1058 represented as vectors, whose length is proportional to the velocity of these 1059 transitions (longer vectors meaning faster transitions). Annotation of the spliced and 1060 unspliced reads was performed using velocyto.py on the SS2 dataset#2 and using 1061 1062 DropEst on the FB5P datasets. Only the cells and genes that had been kept in the 1063 previous analysis using the Seurat R package were considered in the Velocyto analysis. PCA was performed with pagoda2 (https://github.com/hms-dbmi/pagoda2), 1064 with spliced expression matrix as input, and the cell-to-cell distance matrix was 1065 calculated using Euclidean distance. RNA velocity was estimated using gene-relative 1066 model with k-nearest neighbor cell pooling. The overall pattern generated by 1067 projecting the RNA velocities of each individual cell onto the UMAP space enables 1068 1069 visualizing the corresponding model of the trajectory.

1070 BubbleGUM analysis

BubbleGUM, a high throughput gene set enrichment analysis (GSEA) software ⁴² was used to assess the enrichment of selected transcriptomic signatures across populations of the microarray data or clusters of the scRNAseq data. BubbleGUM is composed of two modules; i) GeneSign, which generates statistically significant gene signatures and ii) BubbleMap, which automatically assesses the enrichment of input gene signatures between all possible pairs of conditions from independent datasets, based on gene set enrichment analysis (GSEA) methodology⁵⁹, and which generates

an integrated graphical display. To run the final BubbleMap analysis, we used a total 1078 of 24 selected gene sets (Supplementary Data 2). These gene sets had been 1079 selected from an initial pool of 27 homemade GeneSets and of all of the 17,810 gene 1080 sets from the MSigDB (msigdb.v6.2.symbols.gmt), upon iterative runs of GSEA and 1081 BubbleGUM analyses to focus the final analysis on the gene sets that yielded 1082 significant and meaningful enrichments and were not too redundant with one another. 1083 BubbleMap was used with 1,000 gene set permutations, and with "difference of 1084 classes" as a metric for ranking the genes since the data were expressed in log 1085 1086 scale. The results are displayed as a BubbleMap, where each bubble is a GSEA 1087 result and summarizes the information from the corresponding enrichment plot. The 1088 color of the Bubble corresponds to the condition from the pairwise comparison in which the signature is enriched. The bubble area is proportional to the GSEA 1089 normalized enrichment score (NES). The intensity of the color corresponds to the 1090 statistical significance of the enrichment, derived by computing the multiple testing-1091 adjusted permutation-based p-value using the Benjamini-Yekutieli correction. 1092

1093 In vitro T cell activation assay

For each experiment, spleens were isolated from 10 *lfnb1^{Eyfp}* mice infected with 1094 MCMV 38h before sacrifice. Splenocytes were pooled and submitted to magnetic 1095 depletion of Lineage⁺ cells. pDCs in the negative fraction were identified as live 1096 single-cells, CD11b⁻ Lin(CD3,CD19,Ly6G,NK1.1)⁻ CD11c^{int} SiglecH⁺ BST2^{hi} cells, 1097 and sorted as YFP⁻CD86⁺, YFP⁻CD86⁻, YFP⁺CCR7⁻ or YFP⁺CCR7⁺. Cells were 1098 1099 collected into complete RPMI supplemented with 10% FCS, 1% glutamine (Gibco), 100U/mL Penicillin streptomycin, 1% non-essential amino acids, 1% sodium pyruvate 1100 and 0.05 mM β-mercaptoethanol. pDCs sorted from uninfected mice were used as 1101 controls. Cells were spun down, resuspended in complete RPMI with 5 µg/mL OVA 1102

peptide (323-339; H-Ile-Ser-GIn-Ala-Val-His-Ala-Ala-His-Ala-Glu-Ile-Asn-Glu-Ala-Gly-1103 Arg-OH) (Sigma-Aldrich) and plated at 2,000 cells in 100 µL per well in a V-bottom 1104 plate for 20 min at 37°C. Naïve CD4⁺ T cells from Rag2KO x Tg OT-II x B6.Ly5.1 1105 mice were isolated using a naïve CD4⁺ T cell enrichment kit (Miltenyi Biotec), labeled 1106 1107 with 2.5 µM of carboxyfluorescein diacetate succinimidyl ester (CFSE) for 10 min, and blocked for 5 min at RT with 5x volumes of complete RPMI. Cells were spun 1108 down and washed once again. 10,000 T cells/50 µL were added to each well 1109 containing sorted pDCs. Wells without pDCs were used as negative controls. After 4 1110 1111 days of co-culture, cells were washed in PBS, stained with Live Dead Agua, APC-1112 Cy7 anti-CD4 and Alexa700 anti-CD3e mAbs. Cells were then resuspended in 75 µL 1113 of PBS supplemented with 1 mM EDTA (Sigma-Aldrich) and 0.5 % Bovine Serum Albumin (BSA, H2B). 25 µL of absolute counting beads (Beckman Coulter) were 1114 added to each well before acquisition by FACS. The proliferation in the negative 1115 control wells (background) was substracted from each condition. For each 1116 experiment, the fold expansion was calculated as the number of proliferating cells in 1117 each condition divided by the mean number of proliferating cells (3 replicates per 1118 1119 experiment) in the wells with pDCs isolated from uninfected mice.

1120 Statistical analyses

All quantifications were performed unblinded. Statistical parameters including the definitions and exact value of n (number of biological replicates and total number of experiments), and the types of the statistical tests are reported in the figures and corresponding figure legends. Statistical analyses were performed using Prism (GraphPad Software) or R statistical programming language. Statistical analysis was conducted on data with at least three biological replicates. Comparisons between groups were planned before statistical testing and target effect sizes were not

predetermined. Error bars displayed on graphs represent the mean ± SEM. Statistical

significance was defined as *, P<0.05; **, P<0.01; ***, P<0.001; ****, P<0.001.

1130

1131 Data availability

1132 Microarray and scRNAseq data have been deposited in the GEO repository under

accession codes GSE150664 and GSE151248, respectively. All the other data that

support the findings of this study will be available from the corresponding authors

1135 upon reasonable request.

1136

1137 Methods only References

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1159 Additional information

- 1160 Extended data is available for this paper at XXX.
- 1161 **Supplementary information** is available for this paper at XXX.
- 1162 **Correspondence and requests for materials** should be addressed to M.D. or E.T.
- **Peer review information** loana Visan was the primary editor on this article and managed its editorial process and peer review in collaboration with the rest of the editorial team.
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	total	70	24	94	188			
48	QC1	64	22	84	170			
	pDC ²	59	21	77	157			
	total	304	72	1034	1410			
total	QC1	260	57	856	1173			
	pDC ²	251	53	790	1094			
¹ # cells that passed technical quality controls; ² #cells that were identified as bona fide pDCs, as opposed to								
contaminating pDC-like pre-								
cDC/tDC or B cells.								

JMAP2

P2

pDC enrichment by MACS (negative selection)

Index sorting

FB5P scRNAseq library generation

YFP

Whole spleen section

ΜZ

Whole spleen section

RP

Whole spleen section

WP

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