# 1 Appendix

- 2 Comprehensive innate immune profiling of chikungunya virus infection in
- 3 pediatric cases

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# 28 Appendix Table of Contents

29	Appendix Figures	3
30	Appendix Figure S1	4
31	Appendix Figure S2	5
32	Appendix Figure S3	6
33	Appendix Figure S4	7
34	Appendix Figure S5	8
35	Appendix Figure S6	10
36	Appendix Figure S7	11
37	Appendix Figure S8	12
38	Appendix Figure S9	13
39	Appendix Figure S10	14
40	Appendix Figure S11	15
41	Appendix Figure S12	16
42	Appendix Figure S13	17
43	Appendix Figure S14	18
44	Appendix Figure S16	20
45	Appendix Figure S17	21
46	Appendix Figure S18	22
47	Appendix Figure S19	23
48	Appendix Figure S20	24
49	Appendix Figure S21	25
50	Appendix Figure S22	26
51	Appendix Figure S23	27
52	Appendix Figure S24	
53	Appendix Figure S25	
54	Appendix Figure S26	
55	Appendix Figure S27	31
56	Appendix Figure S28	32
57	Appendix Figure S29	33
58	Appendix Figure S30	34

#### **Appendix Figures** 60



62

continued on next page 63





# 65 Appendix Figure S1.

66	Example output of MetaHybridLouvain for the representative sample used in Figure 2A, Figure
67	3B, and 3D. At left, frequencies for each sub-community at each timepoint, and at right, a
68	heatmap of the distribution of channel values is plotted for each sub-community (within each
69	row). As expected, sub-communities generally show similar values among all channels for their
70	constituent events, with some exceptions (heatmap cells with vertical gradients).



# 72 Appendix Figure S2.

73 Heatmap of differences in marker protein expression between four naïve B cell sub-

74 communities identified by MetaHybridLouvain. All channels listed here differentiate sub-

community 4 from the median value for the other three sub-communities at a threshold of fold

change > 1.5 and FDR < 0.05 (using a moderated paired t-test under the mixed effects model).

77 Colors represent log<sub>2</sub> fold change in channel intensity from the median for all communities (red

means higher intensity). Notably, sub-community 4 is characterized by a much higher

repression of CXCR5 as compared to all other sub-communities of naïve B cells. Otherwise, its

80 marker expression is very similar to sub-community 1 of naïve B cells, albeit with much lower

81 expression of CCR4, CXCR3 and CD80.



# 85 Appendix Figure S3.

Correlations between acute-phase cell sub-community frequencies and log<sub>10</sub> CHIKV antibody 86 titer at 15d post-symptom onset (p.s.o.). A, Scatterplot of log<sub>10</sub> cell sub-community frequencies 87 at 1d p.s.o. against log<sub>10</sub> CHIKV antibody titer at the 15d timepoint for CD14<sup>+</sup>CD16<sup>+</sup> monocyte 88 89 sub-community 1. This is the only correlation that is significant after multiple hypothesis 90 correction (FDR P = 0.0050, Spearman's  $\rho$  = 0.60). B, Scatterplot as in A but for two sub-91 communities that have significant correlations at the 15d p.s.o. timepoint: again, CD14<sup>+</sup>CD16<sup>+</sup> 92 monocyte sub-community 1 (FDR *P* = 0.035,  $\rho$  = 0.51), and central memory CD4<sup>+</sup> T cell sub-93 community 2 (FDR *P* = 0.035,  $\rho$  = -0.52). 94



## 96 Appendix Figure S4.

97 Differences in per-sample protein expression between two CD14<sup>+</sup>CD16<sup>+</sup> sub-communities

98 identified by MetaHybridLouvain on CyTOF data. 1 corresponds with the "intermediate"

99 CD14<sup>++</sup>CD16<sup>+</sup> phenotype, while 2 corresponds with the "nonclassical" CD14<sup>+</sup>CD16<sup>++</sup> phenotype.

100 The X axis is filtered to channels with differences significant at FDR < 0.05, and ordered from

101 differences where sub-community 1 > 2 on the left to sub-community 1 < 2 on the right; i.e., red

- bars are "intermediate"-associated markers while gray bars are "nonclassical"-associated
- 103 markers. Error bars correspond to 95% confidence intervals. \*, FDR-adjusted *P* (FDR *P*) < 0.05;

104 \*\*, FDR *P* < 0.01, \*\*\*, FDR *P* < 0.001.

- 105
- 106



MetaHybridLouvain: cd14\_cd16\_monocyte\_1

### 108 Appendix Figure S5.

- 109 Summary of frequencies (per timepoint) and mean channel values for sub-community 1 of
- 110 CD14<sup>+</sup>CD16<sup>+</sup> monocytes, across all samples. At left, frequencies in each sample split by

- timepoint; at right, within each row, heatmaps of the distribution of channel values for all events
- 112 within this sub-community for each sample. A gray row indicates this sub-community was not
- identified in this sample.



MetaHybridLouvain: cd14\_cd16\_monocyte\_2

- 115 Appendix Figure S6.
- 116 Summary of population frequencies (per timepoint) and mean channel values for sub-
- 117 community 2 of CD14<sup>+</sup>CD16<sup>+</sup> monocytes, across all samples. Layout as in Appendix Figure S5.



# CD14+ monocytes

119

## 120 Appendix Figure S7.

- 121 Manual gating analysis of a representative sample demonstrates the presence of a small but
- 122 distinct CD14<sup>+</sup> monocyte subpopulation that displays non-canonical markers CXCR3 and
- 123 CCR4, among others (see sub-community 3 in Appendix Figure S8).



126 Appendix Figure S8.

Heatmap of differences in marker protein expression between three CD14<sup>+</sup> sub-communities
identified by MetaHybridLouvain. All channels listed here showed differences significant at FDR
< 0.05 (moderated paired t-test under the mixed-effects model). The patterns of fold change</li>
differences in channel intensities between sub-communities are ordered vertically with
hierarchical clustering (red indicates increased relative to the other sub-communities). The
mean intensity of each channel across *all* sub-communities is shown by the adjacent purple
annotation column.



MetaHybridLouvain: cd14\_monocyte\_1

- 135 Appendix Figure S9.
- 136 Summary of population frequencies (per timepoint) and mean channel values for sub-
- 137 community 1 of CD14<sup>+</sup> monocytes, across all samples. Layout as in Appendix Figure S5.



MetaHybridLouvain: cd14\_monocyte\_2

- 139 Appendix Figure S10.
- 140 Summary of population frequencies (per timepoint) and mean channel values for sub-
- 141 community 2 of CD14<sup>+</sup> monocytes, across all samples. Layout as in Appendix Figure S5.



- 143 Appendix Figure S11.
- 144 Summary of population frequencies (per timepoint) and mean channel values for sub-
- 145 community 3 of CD14<sup>+</sup> monocytes, across all samples. Layout as in Appendix Figure S5.



# 147 Appendix Figure S12.

148 Differences in serum growth factor levels between the acute and convalescent timepoints, as

149 measured by multiplex ELISA (Luminex). Only one of the differences depicted here achieved

150 statistical significance at FDR < 0.05, indicated by the asterisk.



# 152 Appendix Figure S13.

- 153 Clustered heatmap of Spearman correlations between log-scaled serum cytokine concentration
- 154 (Luminex) and log-scaled cell subphenotype frequencies (CyTOF) of changes from the acute to
- 155 *convalescent timepoints*. CyTOF values are indicated by an asterisk following the label.



# 158 Appendix Figure S14.

159 Clustered heatmap of Spearman correlations between log-scaled serum cytokine concentration

160 (Luminex) and log-scaled cell subphenotype frequencies (CyTOF) within the acute timepoint.

161 CyTOF values are indicated by an asterisk following the label.



- 164 Appendix Figure S15.
- 165 Clustered heatmap of Spearman correlations between log-scaled serum cytokine concentration
- 166 (Luminex) and log-scaled cell subphenotype frequencies (CyTOF) within the convalescent
- 167 *timepoint*. CyTOF values are indicated with an asterisk following the label.



## 170 Appendix Figure S16.

171 Clustered heatmap of Spearman correlations between log-scaled serum cytokine concentration 172 and log-scaled monocyte subphenotype frequencies. *A*, within *acute*-phase samples. *B*, within 173 *convalescent*-phase samples. CCL2 within the convalescent timepoint (highlighted) displayed 174 the only set of Spearman correlations that differed significantly from those of the other cytokines 175 at an FDR threshold of < 0.05 (Mann-Whitney *U* test).



### 178 Appendix Figure S17.

179 Pathview plot of log<sub>2</sub> fold change in gene expression between acute and convalescent

180 timepoints for the cytokine-cytokine receptor interaction pathway in KEGG (accession

181 hsa04060). Positive values indicate upregulation during the acute phase of infection.

182



# 184 Appendix Figure S18.

- 185 Pathview plot of log<sub>2</sub> fold change in gene expression between acute and convalescent
- timepoints for the RIG-I-like receptor signaling pathway in KEGG (accession hsa04622).
- 187 Positive values indicate upregulation during the acute phase of infection.

188



# 190 Appendix Figure S19.

191 Pathview plot of log<sub>2</sub> fold change in gene expression between acute and convalescent

timepoints for the JAK-STAT signaling pathway in KEGG (accession hsa04630). Positive

193 values indicate upregulation during the acute phase of infection.



# 196 Appendix Figure S20.

197 Pathview plot of log<sub>2</sub> fold change in gene expression between acute and convalescent

timepoints for the chemokine signaling pathway in KEGG (accession hsa04062). Positive values

199 indicate upregulation during the acute phase of infection.



# 202 Appendix Figure S21.

203 Pathview plot of log<sub>2</sub> fold change in gene expression between acute and convalescent

timepoints for the toll-like receptor pathway in KEGG (accession hsa04620). Positive values

205 indicate upregulation during the acute phase of infection.

206



207

# 208 Appendix Figure S22.

- 209 Pathview plot of log<sub>2</sub> fold change in gene expression between acute and convalescent
- 210 timepoints for the TNF signaling pathway in KEGG (accession hsa04668). Positive values
- 211 indicate upregulation during the acute phase of infection.



- 213 05164 10/23 (c) Kanehise
- 214 Appendix Figure S23.
- 215 Pathview plot of log<sub>2</sub> fold change in gene expression between acute and convalescent
- timepoints for the Influenza A pathway in KEGG (hsa05164). Positive values indicate
- 217 upregulation during the acute phase of infection.



# 226 Appendix Figure S24.

- 227 Heatmap of gene expression across different types of blood cell lines for genes expressed
- higher in acute phase (A) or higher in convalescent phase (B). Rows represent genes. Columns
- represent blood cell lines which are grouped according to the lineage (column legend). HSC,
- Hematopoietic stem cell; MYP, Myeloid progenitor; ERY, Erythroid cell; MEGA, Megakaryocyte;
- 231 GM, Granulocyte/monocyte; EOS, Eosinophil, BASO, Basophil; DEND, Dendritic cell.



- 234 Appendix Figure S25.
- 235 Correlation between cell frequency derived from CyTOF and that estimated from gene
- 236 expression using expression of cell-specific markers (*A*) or the CIBERSORT algorithm (*B*).





# 239 Appendix Figure S26.

240 Q-Q plot of the distribution of observed  $-\log_{10} P$  values for differentially expressed transcripts 241 between severe and non-severe cases against the distribution of  $-\log_{10} P$  values expected 242 under the null hypothesis (that no transcripts are differentially expressed). Gray shaded band 243 indicates the 95% confidence interval for the expected *P* value distribution under the null 244 hypothesis.



## 246 Appendix Figure S27.

- 247 Weighted multiscale interaction network depicting all Pearson's correlations between
- subpopulation frequencies (gold nodes), serum cytokine concentrations (small black nodes),
- 249 coexpression modules (ME prefix, gray nodes), and clinical variables (large black nodes).
- 250 Serum cytokine concentrations cluster into two highly interconnected components (upper right),
- 251 farther from the clinical variables than the other node types. Edges are filtered to correlations
- significant at *P* < 0.001, and thickness is scaled to the magnitude of the correlation.
- 253





255 Appendix Figure S28.

Receiver operator characteristic (ROC) curves measuring the performance of elastic net logistic 256 regression models predicting the phase of infection (acute vs. convalescent) for each sample, 257 258 using progressively dimensionality-reduced versions of the dataset. Thin grey lines show the 259 ROC curves for 100 bootstrap replicates. The area under the curve (AUC) along with its 95% 260 confidence interval are shown underneath each plot; a perfect classifier would achieve AUC=1 261 while a random classifier is expected to achieve AUC=0.5 (dashed diagonal line). A, model 262 trained using all CyTOF sub-community frequencies, all quantified RNA-seq transcripts, and all 263 Luminex cytokine measurements achieves near-perfect performance. B, model trained as in A 264 but with eigengene values for 92 coexpression modules replacing the RNA-seg transcript-level 265 quantification; performance is still near-perfect. C, model trained as in B but without the Luminex 266 cytokine measurements; performance is still essentially equivalent to the other two models.



#### 270 Appendix Figure S29.

271 Principal variance component analysis of CyTOF data for evaluation of potential batch effect. 272 Since the 3 potential batch effect variables were largely collinear (Spearman correlation > 0.93 273 for thaw, staining and acquisition batch variables), we included only the acquisition date in our 274 PVCA analysis. A, Barplot of contributions for each variable to overall variance. Acquisition date 275 explains 3.8% of the overall variance. B, Scatterplot of all samples in principal component space 276 for the first two principal components. Acquisition date is used to color the samples, and grey 277 lines connect pairs of samples across the two timepoints. As expected, the first principal 278 component (explaining 78% of the overall variance) roughly parallels the timepoint contrast, 279 while the acquisition date variable does not correlate with either of the first two principal 280 components.



#### 283 Appendix Figure S30.

284 High precision, recall scores and F1 score show that the classifier is returning accurate results 285 with low false negative rates. A, Precision versus recall for each cell subset. The NOD classifier 286 was trained over all samples except one, then applied to the remaining sample. This process 287 was repeated for all samples ("jackknifing"). For each cell subset, precision (TP / TP + FP) and recall (TP / TP + FN) values were calculated, and the mean over all samples is presented here. 288 289 B. The classifier is an accurate classifier as shown by high F1 score values for all large cell 290 subsets. The F1 score for each cell subset is shown. The F1 score was calculated as the 291 harmonic mean of precision and recall (2 x precision x recall / (precision + recall)). The line denotes the mean F1 score over all subsets. TP, true positive; FP, false positive; FN, false 292 293 negative.