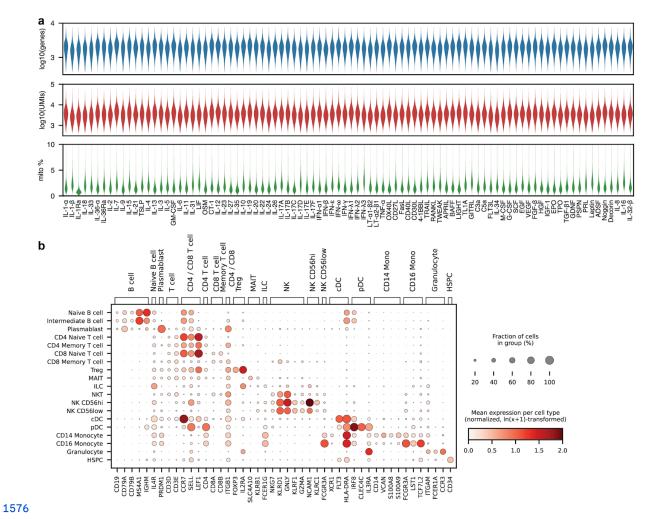
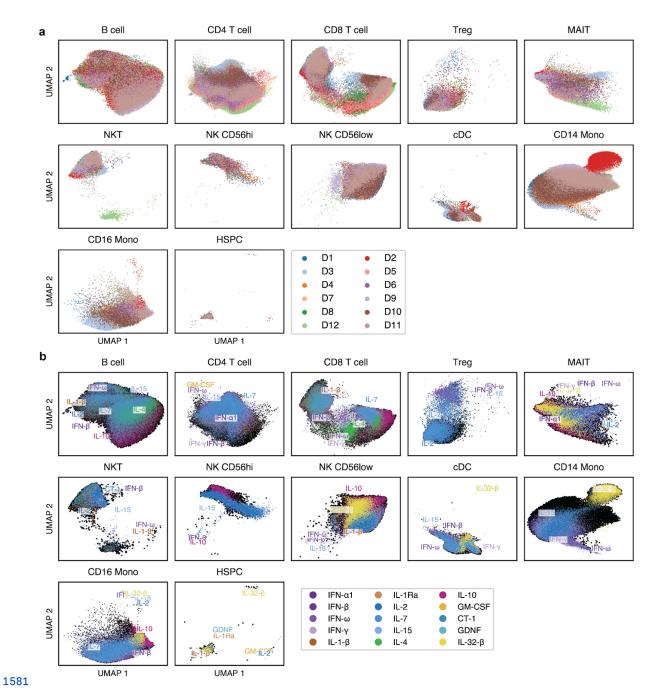
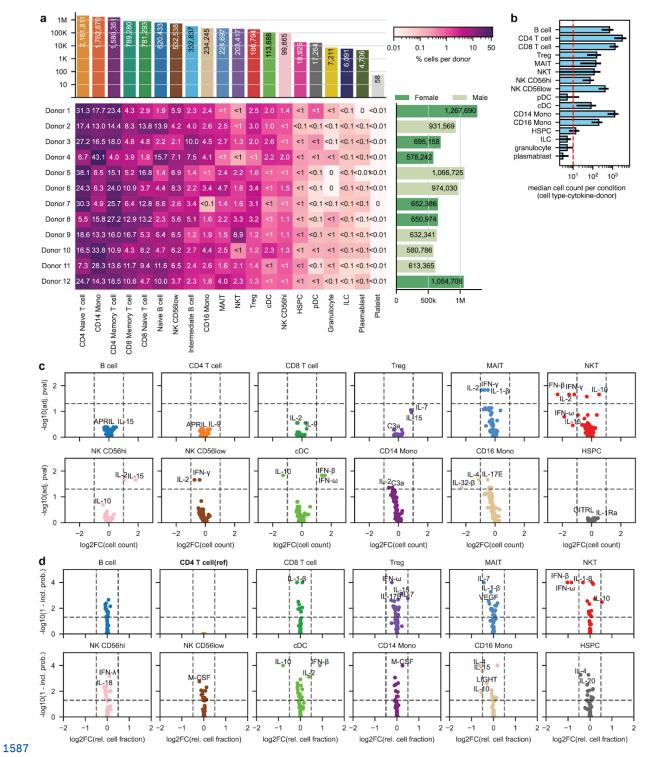
## **Supplementary Information** 1546 1547 A single-cell cytokine dictionary of human peripheral blood 1548 1549 1550 Lukas Oesinghaus<sup>1,2,\*</sup>, Sören Becker<sup>3,4,\*</sup>, Larsen Vornholz<sup>3,\*</sup>, Efthymia Papalexi<sup>5,\*</sup>, Joey 1551 Pangallo<sup>5</sup>, Amir Ali Moinfar<sup>3,4</sup>, Jenni Liu<sup>3</sup>, Alyssa La Fleur<sup>6</sup>, Maiia Shulman<sup>3,7</sup>, Simone 1552 Marrujo<sup>5</sup>, Bryan Hariadi<sup>5</sup>, Crina Curca<sup>5</sup>, Alexa Suyama<sup>5</sup>, Maria Nigos<sup>5</sup>, Oliver Sanderson<sup>5</sup>, 1553 Hoai Nguyen<sup>5</sup>, Vuong K. Tran<sup>5</sup>, Ajay A. Sapre<sup>5</sup>, Olivia Kaplan<sup>5</sup>, Sarah Schroeder<sup>5</sup>, Alec 1554 Salvino<sup>5</sup>, Guillermo Gallareta-Olivares<sup>5</sup>, Ryan Koehler<sup>5</sup>, Gary Geiss<sup>5</sup>, Alexander B. 1555 Rosenberg<sup>5</sup>, Charles M. Roco<sup>5,†</sup>, Georg Seelig<sup>1,2,6,†</sup> and Fabian J. Theis<sup>3,4,7,†</sup> 1556 1. Department of Electrical and Computer Engineering, University of Washington, 1557 Seattle, USA 1558 2. Botnar Institute for Immune Engineering, Basel, Switzerland 1559 3. Institute of Computational Biology, Helmholtz Center Munich, Munich, Germany 1560 4. School of Computing, Information and Technology, Technical University of Munich, 1561 Munich, Germany 1562 5. Parse Biosciences, Seattle, USA 1563 6. Paul G. Allen School of Computer Science and Engineering, University of Washington, Seattle, USA 1565 7. TUM School of Life Sciences Weihenstephan, Technical University of Munich, 1566 Munich, Germany 1567 1569 \*These authors contributed equally to this work. 1570 † correspondence: 1571 georg.seelig@immune.engineering 1572 fabian.theis@helmholtz-munich.de 1573 charlie@parsebiosciences.com 1574



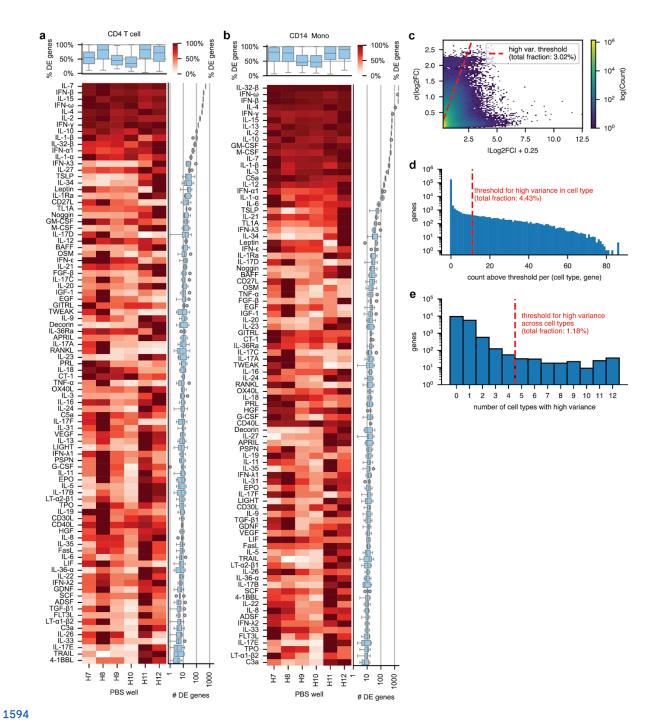
1577 Fig. S1. Quality control metrics and marker gene expression. a, Distribution of the number of detected genes, total UMIs, 1578 and percent mitochondrial genes across all cells for a given cytokine perturbation after QC steps. b, Expression of the marker 1579 genes used to annotate cell types.



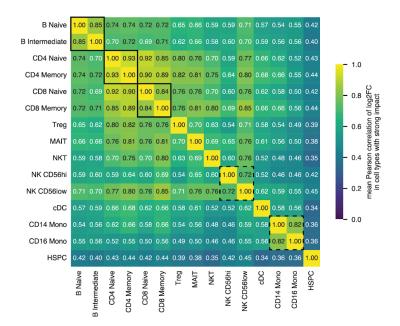
1582 Fig. S2. UMAP of cell type populations by donor and cytokine stimulation. a,b, UMAP of the cell type populations 1583 displaying (a) the PBS condition by donor and (b) the cytokines causing the strongest shift in the position of the population for 1584 different cell types. Although strong cytokine perturbations shift cells in the UMAP space, cells overall cluster more strongly by 1585 cell type rather than by donor or perturbation. 1586



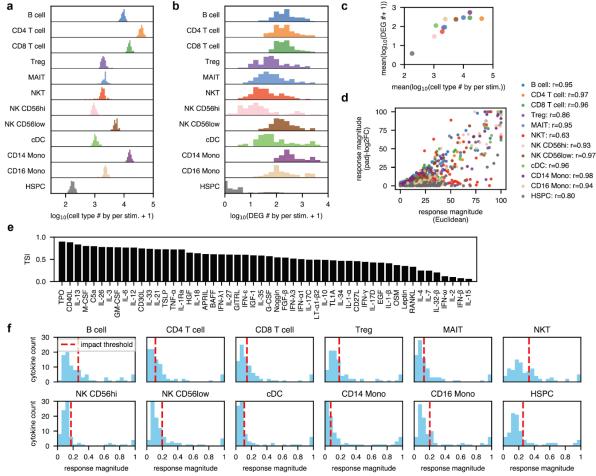
1588 Fig. S3. Cell type abundance analysis. a, Fraction of counts per cell type for each donor. b, Median cell count per condition. 1589 The dashed red line shows the cutoff of 10 cells per condition (cell type-cytokine-donor triplet) we used to exclude cell types 1590 from the analysis. c, Volcano plot of differential cell count analysis (absolute numbers) between cytokine perturbations and the 1591 PBS controls. d, Bayesian compositional modeling (scCODA) using CD4 T cells as reference. The y axis plots the posterior 1592 probability of no effect.



1595 Fig. S4. Rationale and process for additional DEG filtering based on consistency across wells. a, b, Relative number of 1596 DEGs (|Log2FC|>0.25, padj<0.05) obtained for different PBS wells in (a) CD4 T cells and (b) CD14 monocytes. The box plot on 1597 top shows the distribution of relative values by well and the box plot on the right shows the distribution of absolute numbers by 1598 cytokine. DEGs used in the main text are filtered for consistency across wells. c, Standard deviation of log2FCs for a particular 1599 gene, cell type, and cytokine perturbation across wells versus their mean. Some genes are highly variable across wells (dashed 1600 red lines) d, Number of times a certain gene is above the high variability threshold shown in (c) in a particular cell type. Those 1601 genes that are highly variable for at least 11 cytokine treatments are considered highly variable in that cell type. e, The number 1602 of times a gene is highly variable in a given cell type. Any gene that is highly variable in at least 5 cell types is considered highly 1603 variable across cell types and filtered from the DEG analysis.

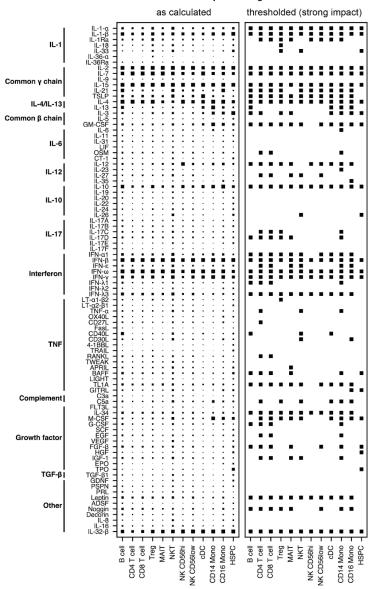


1606 Fig. S5. Merging of cell types. Mean Pearson r between log2FC values in cell types with a strong impact. Solid boxes show 1607 cell types that were merged, dashed boxes cell types that could be considered for merging but were kept separate. While these 1608 cell types are merged for the analysis in the main text, we provide pseudobulk data and DEGs both for the merged cell type and 1609 individual cell types for all boxes.



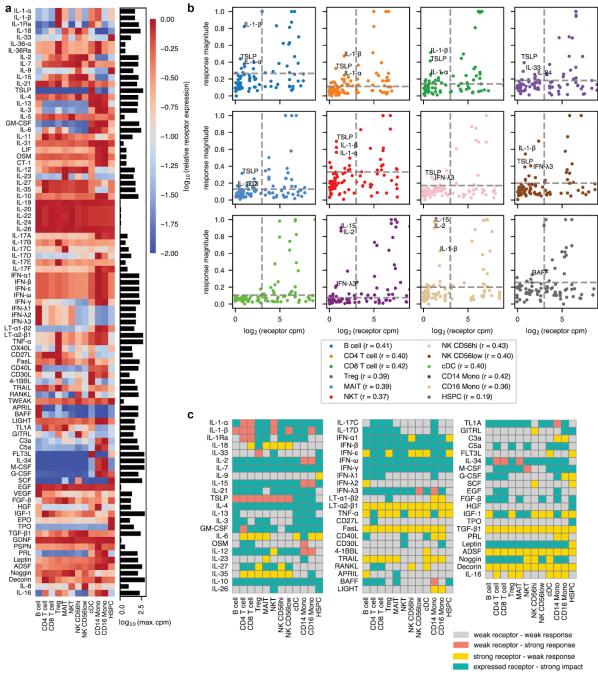
1613 Fig. S6. Cytokine DEG counts, response magnitudes, the tissue-specificity index, and strong impact classification. a, 1614 b, Distribution of the total number of cells (a) and DEGs by cell type (b) across all cytokine perturbations before log2FC filtering. 1615 c, Association between the mean of the values in (a) and (b). d, Two measures of the response magnitude (Euclidean distance 1616 and padj-log2FC-based) are highly correlated in all cell types. e, Tissue-specificity index of the response magnitude for 1617 cytokines with at least one strong response. f, Distribution of the overall response magnitude (mean of Euclidean and 1618 padj-log2FC-based) across the cell types. The dashed red line shows the response magnitude threshold used to define strong 1619 impact.

## response magnitudes

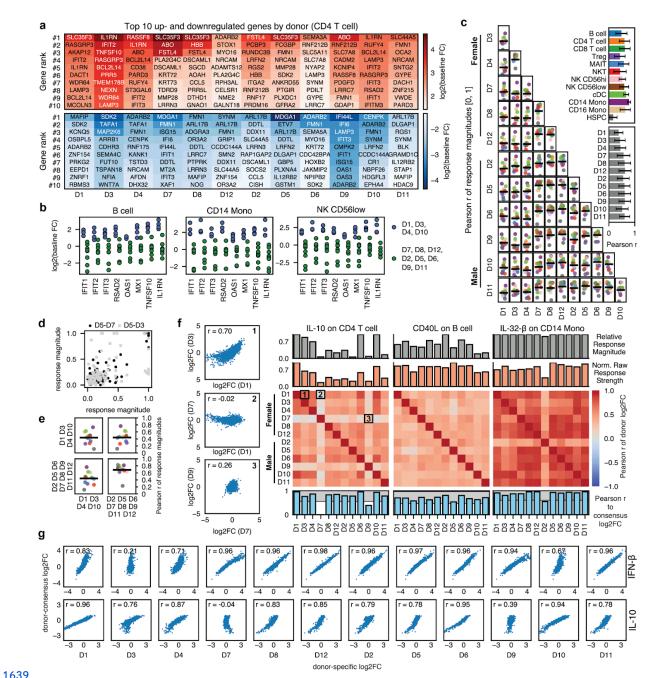


1622 Fig. S7. Response magnitude by individual cell types and cytokines. Left: The calculated response magnitude is linearly 1623 proportional to the box size. Right: The presence of a box indicates a strong response.

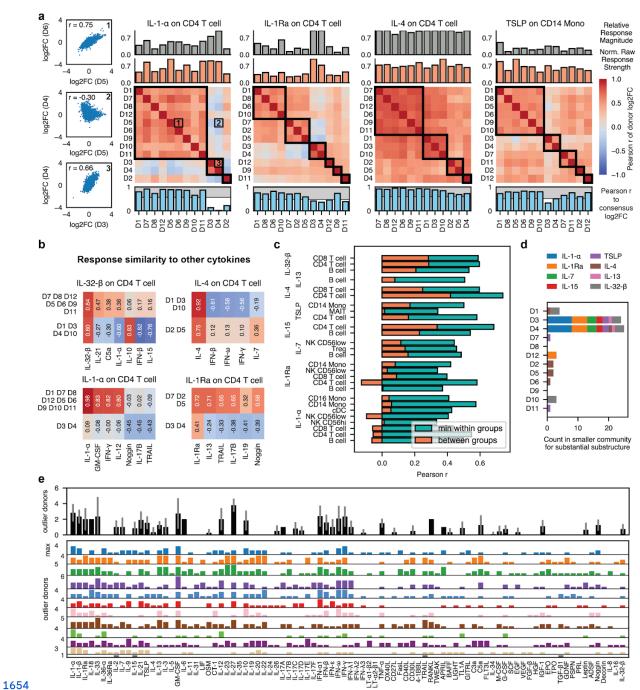
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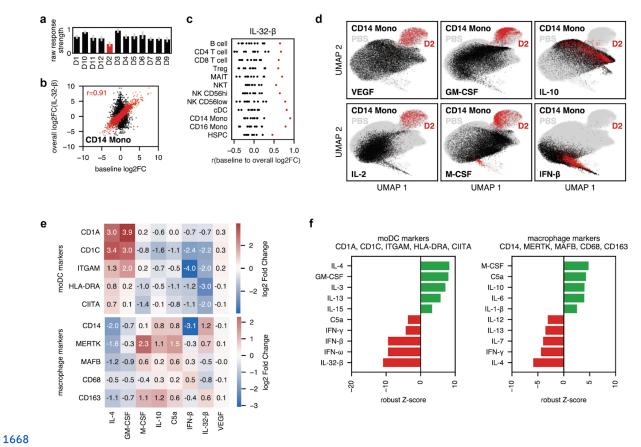
1627 Fig. S8. Cytokine receptor expression and its association with the response magnitude. a, Relative receptor expression 1628 (cpm divided by the maximum across cell types) for all cytokines with known receptors (i.e., every tested cytokine except 1629 IL-32-β). The y-axis label shows the name of the cytokine for which the receptor expression is shown. The bar plot on the right 1630 shows the maximum cpm for each receptor. b, Response magnitude versus absolute receptor expression for all cytokines and 1631 cell types. The response magnitude threshold for strong activity is shown as a horizontal dashed line, and a receptor 1632 expression of 8 cpm is shown as a vertical dashed line and acts as the approximate threshold for functional receptor 1633 expression. For each cell type, a number of cytokines with a strong response despite low receptor expression are highlighted 1634 by name. c, Some cytokines show weak responses despite strong receptor expression (>32 cpm, response magnitude < 1635 threshold). Others show strong responses despite low receptor expression (<5 cpm, response magnitude > 1.5x threshold). We 1636 additionally annotate those with expressed receptors and a strong response (>8 cpm, response magnitude > threshold) and 1637 those with weak receptor expression and a weak response (<8 cpm, response magnitude < threshold).



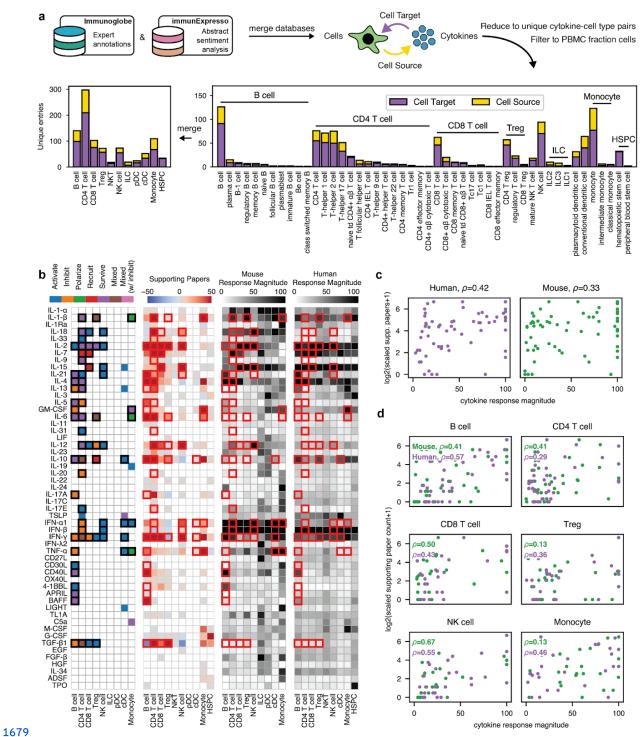
1640 Fig. S9. Additional analysis of donor variability in cytokine responses. a, Top up- and downregulated genes in the 1641 baseline FC for each donor. Genes were pre-filtered to exclude ribosomal and mitochondrial genes, sex chromosome-linked 1642 genes, and non-protein coding genes. b, Baseline log2FC of interferon response genes for two groups of donors and three cell 1643 types. c, Pearson r of response magnitudes for all 90 cytokines within a given cell type between two donors. The solid line 1644 shows the mean across all cell types. Bar plots on the right show the marginal means across cell types and donors. d, 1645 Correlation of the response magnitudes for two sets of donors. e, As in (a) but averaged over the two indicated groups of 1646 donors. f, Correlation of donor-specific log2FC scores for different cell types and cytokines. The topmost barplot shows a 1647 response magnitude calculated using the Euclidean distance of vectors. The second barplot from the top shows a noise-robust 1648 measurement of the total response strength, normalized to the strongest response across donors for the shown condition. The 1649 bottom barplot shows the correlation to the donor-consensus log2FCs. The shaded gray area denotes a region around the 1650 median which we interpret to mean that the response is consistent, whereas responses outside this region are considered 1651 outliers. g, Donor-specific log2FC versus donor-consensus log2FC for two cytokines (IL-10 and IFN-β) and all donors. The 1652 Pearson r is shown on the top left of each plot.



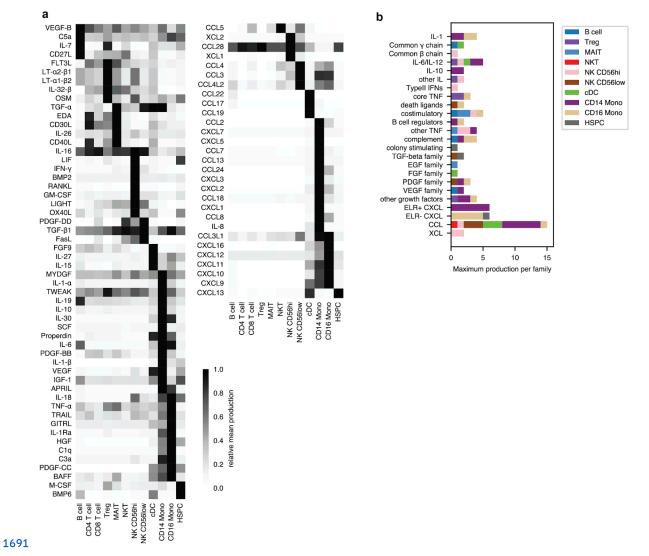
1655 Fig. S10. Detection of substantial sub-structure in donor responses. a, Correlation of donor-specific log2FC scores for 1656 different cell types and cytokines. The topmost barplot shows a response magnitude calculated using the Euclidean distance of 1657 vectors. The second barplot from the top shows a noise-robust measurement of the total response strength, normalized to the 1658 strongest response across donors for the shown condition. The bottom barplot shows the correlation to the donor-consensus 1659 log2FCs. The shaded gray area denotes a region around the median which we interpret to mean that the response is 1660 consistent, whereas responses outside this region are considered outliers. Boxes indicate groupings of correlation patterns by 1661 Leiden clustering. b, Pearson r of the donor-group averaged log2FC for different cytokines and cell types (indicated in individual 1662 titles) compared to the most similar donor-consensus log2FC vectors across cytokines in the same cell type. c, Pearson r of 1663 donor group responses for cytokine-cell type pairs where substantial sub-structure was detected. We show the minimum of the 1664 two within-group comparisons and the between-group comparison. d, Presence of different donors in the minority group of 1665 responses with substantial substructure by cytokine. e, Count of donors with outlier responses for each cell type and cytokine. 1666 The maximum for each cell type is indicated on the left. The barplot on the top shows the mean across cell types.



1669 Fig. S11. Donor 2 shows evidence of baseline IL-32-β signaling. a, Raw response strength to IL-32-β by donor averaged 1670 over cell types. b, Baseline log2FC versus donor-consensus IL-32-β log2FC in CD4 T cells for either donor 2 or all other 1671 donors. c, Correlation between the baseline log2FC and the donor-consensus log2FC for IL-32-β by donor in all cell types. 1672 Donor 2 is highlighted in red. d, UMAP of CD14 monocytes in response to different cytokine perturbations. e, Up- and 1673 downregulation of macrophage (top) and moDC (bottom) identity markers by several relevant cytokines. Log2 Fold Changes 1674 were calculated using across-donor cpm means for the perturbation condition divided by across-donor cpm means for the PBS 1675 condition because filtering criteria for the consensus (edgeR-derived) log2FC impede the calculation of robust z scores across 1676 perturbation conditions. f, Sum of robust z-scores (calculated across perturbation conditions) for the markers shown in (e) by 1677 cytokine.

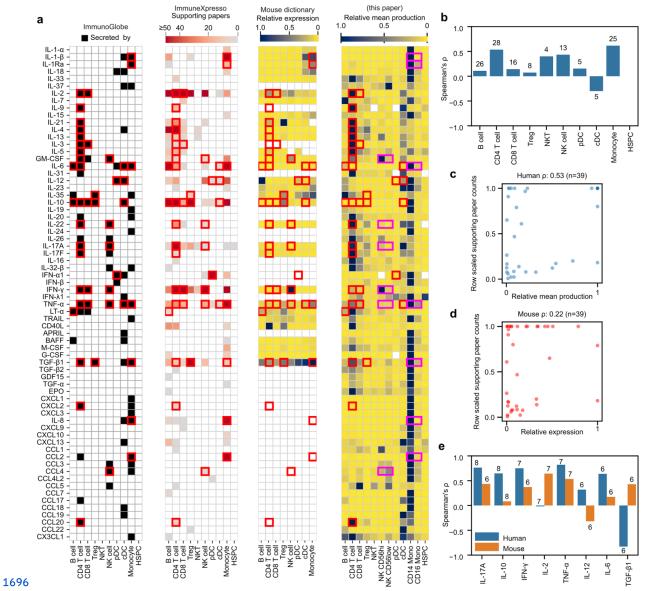


1680 Fig. S12. Database comparison with cytokine response magnitudes. a, The construction of the reference dataset from the 1681 two input databases by merging on cell ontology identifiers. b, All relevant annotations for cytokine effects on specific cell types. 1682 From left to right, these include ImmunoGlobe (the colors in the legend represent the authors' annotation, with the two mixed 1683 categories being any combination of the other labels with or without inhibition included), ImmuneXpresso (where paper counts 1684 were considered positive if there were more positive labels per the relationship than negative, or negative otherwise), mouse 1685 cytokine response magnitude data for relevant cytokines, and our response magnitude values. Boxed cells are high-quality 1686 annotations. For the human data, we used NK CD56low to represent NK cells and CD14 Mono to represent monocytes as 1687 these are by far the majority populations. c, Correlation between mouse or human response magnitudes and scaled paper 1688 counts across all cell types by species. d, Correlation between mouse or human response magnitudes and scaled paper counts 1689 for each cell type.

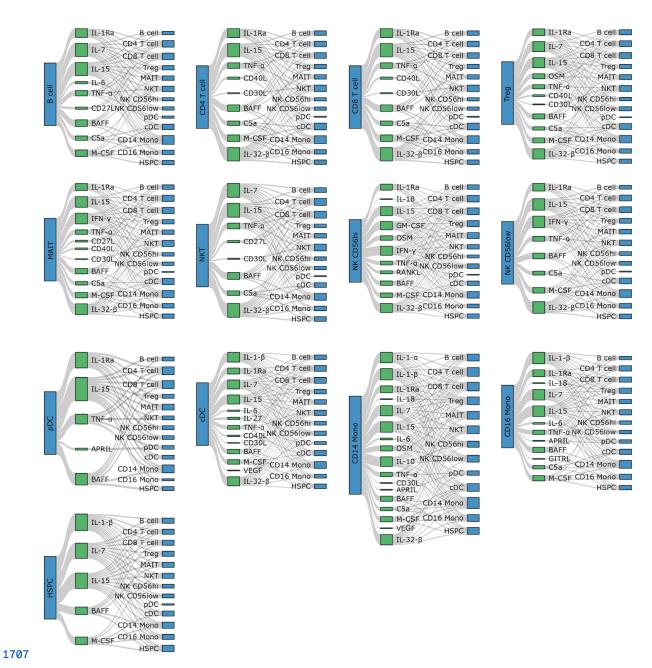


1692 Fig. S13. Expression of individual cytokines by cell type across stimulation conditions. a, Mean production of cytokines 1693 across stimulation conditions for all cytokines expressed in at least one cell type. b, Number of times a given cell type has the 1694 strongest production of cytokines and chemokines of different families.

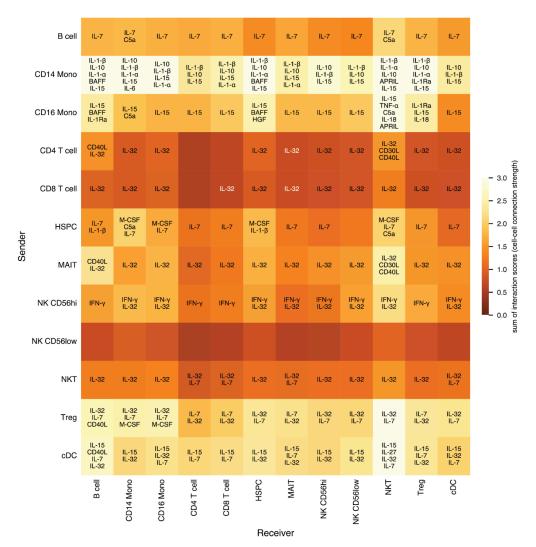
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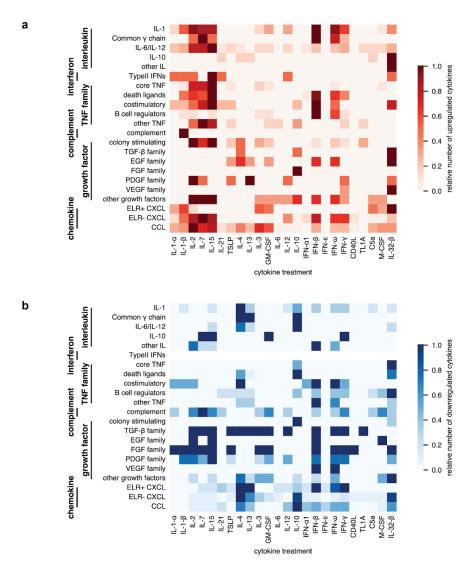
1697 Fig. S14. Cytokine secretion reference dataset comparisons. a, All cytokine secretion values for immunoGlobe, 1698 ImmuneXpresso, relative expression mouse cytokine dictionary information, and relative mean expression (this publication) 1699 from left to right. Note that the boxed values are those shared between databases, with the purple two by one boxes denoting 1700 cell types which were simplified for the end analysis. b, Correlation of relative mean expression values with row-scaled paper 1701 counts per cell type for all annotations with paper count information. Numbers are n for cells with secretion data. c, Human 1702 relative mean production vs row scaled paper counts (see methods) for only high quality annotations shared between the 1703 mouse and human datasets. d, Mouse relative expression vs row scaled paper counts for only high quality annotations shared 1704 between the mouse and human datasets. e, Per-cytokine correlations of relative expression and production with row scaled 1705 supporting paper counts for well-studied cytokines.



1708 Fig. S15. Cytokines secreted by a given cell type cause a strong impact in other cell types. Each connection shows 1709 expression of a cytokine by a cell type that fulfills the strong impact criterion in the target cell type. 1710



1712 Fig. S16. Cell-cell communication map by specific cytokines. For each sender cell-cytokine-target cell combination, an 1713 interaction score was calculated as the product of the relative production of that cytokine in the sender cell type compared to 1714 other cell types times the response magnitude of the cytokine in the target cell type. An overall cell type-cell type connection 1715 strength is calculated as the sum of their interaction scores. Cytokines with an interaction score above 0.25 are annotated.



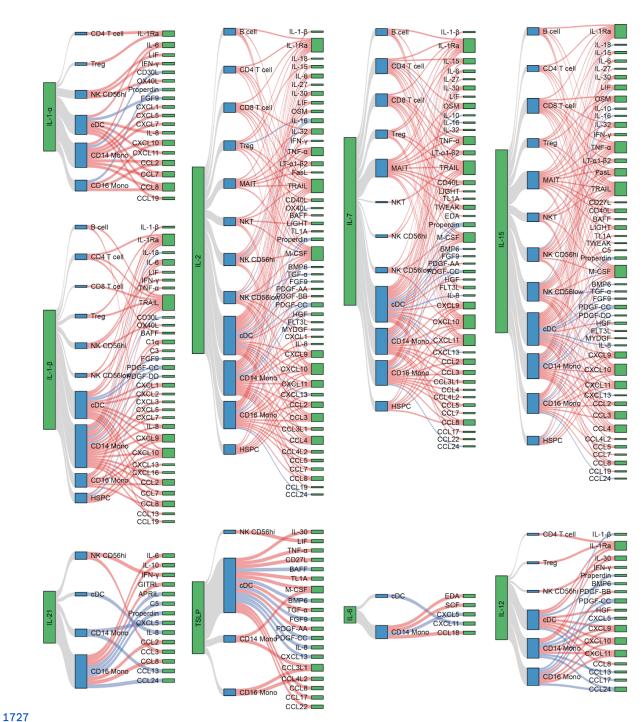
1718 Fig. S17. Cytokine-cytokine family crosstalk. a, Up- and b, downregulation of different cytokine families by cytokine 1719 treatments. The figure sums regulation in the different cell types and then divides by the maximum value for a given family to 1720 derive a relative number.

1721

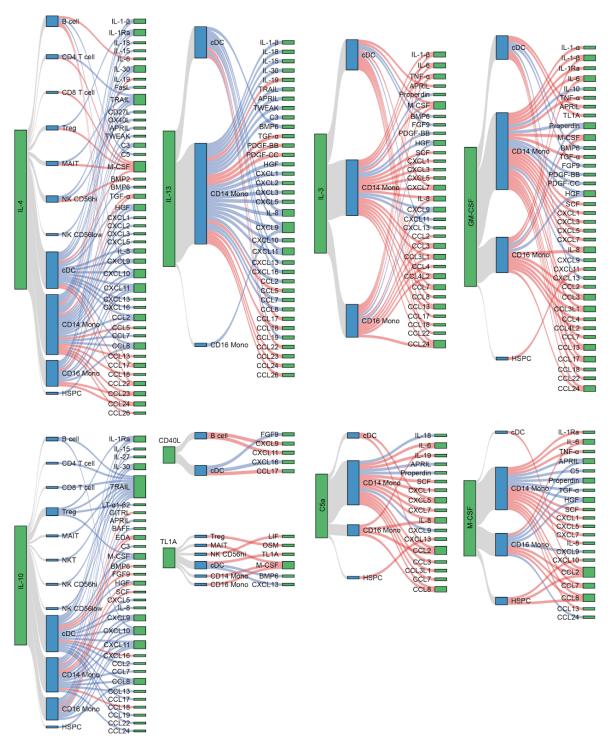
produced cytokine	cell type	treatment	mean cpm	FC to PBS
APRIL	CD16 Mono	IL-3	87	1.8
BAFF	CD14 Mono	IFN-β	718	3.2
BMP6	cDC	IL-32-β	170	8.9
C1q	CD16 Mono	IFN-γ	105	1.7
C3a	CD14 Mono	IL-32-β	75	2.9
C5a	B cell	IL-12	42	1.2
CCL13	CD14 Mono	IL-4	135	5.0
CCL17	cDC	IL-4	570	28.2
CCL18	CD14 Mono	IL-4	225	15.5
CCL19	cDC	IL-1-β	172	3.8
CCL2	CD14 Mono	M-CSF	2115	7.5
CCL22	cDC	TSLP	2508	4.7
CCL23	CD14 Mono	IL-4	71	53.3
CCL24	CD14 Mono	IL-3	609	5.7
CCL3	NK CD56low	IL-15	570	11.8
CCL3L1	CD16 Mono	IL-32-β	154	20.0
CCL4	NK CD56low	IL-15	317	5.9
CCL4L2	NK CD56low	IL-2	51	3.3
CCL5	NKT	IFN-ω	731	1.3
CCL7	CD14 Mono	M-CSF	363	32.4
CCL8	CD14 Mono	IL-2	967	7.4
CD30L	MAIT	IL-7	127	1.6
CD40L	MAIT	IL-15	150	4.8
CXCL1	CD14 Mono	C5a	498	7.1
CXCL10	CD16 Mono	IL-15	4069	4.3
CXCL11	CD14 Mono	IFN-β	1725	28.0
CXCL13	HSPC	IL-32-β	978	30.3
CXCL16	CD16 Mono	IL-32-β	236	1.7
CXCL2	CD14 Mono	IL-32-β	134	5.7
CXCL3	CD14 Mono	C5a	450	3.0
CXCL5	CD14 Mono	C5a	2569	41.1
CXCL7	CD14 Mono	C5a	474	37.2
CXCL9	CD16 Mono	IL-2	1972	10.1
EDA	CD4 T cell	IL-7	240	1.9
FGF9	cDC	IL-10	45	4.0
FLT3L	Treg	IL-24	136	1.4
FasL	NK CD56low	IL-15	299	2.3
GITRL	CD16 Mono	IL-21	61	9.4
GM-CSF	NK CD56hi	IL-1-β	53	9.0
HGF	CD16 Mono	CD27L	64	1.2

produced cytokine	cell type	treatment	mean cpm	FC to PBS
IFN-y	NK CD56hi	IL-1-β	367	19.7
IL-1-α	CD14 Mono	IL-32-β	122	26.5
IL-1-β	CD14 Mono	IL-32-β	3287	46.7
IL-10	CD14 Mono	IL-32-β	60	4.9
IL-15	cDC	CD40L	1082	1.3
IL-16	NK CD56hi	IL-34	177	1.3
IL-18	CD16 Mono	IFN-ε	56	1.3
IL-1Ra	CD16 Mono	IFN-β	1038	5.9
IL-24	CD14 Mono	IL-32-β	52	71.2
IL-27	cDC	IFN-y	43	5.3
IL-30	CD14 Mono	IFN-β	131	7.3
IL-32	MAIT	IL-15	351	2.9
IL-6	CD14 Mono	C5a	133	7.8
IL-7	B cell	CD40L	244	1.6
IL-8	CD14 Mono	IL-32-β	2299	35.1
LIF	NK CD56hi	IL-15	104	5.8
LIGHT	NK CD56hi	IL-1-β	100	1.4
LT-α1-β2	MAIT	IL-7	125	5.5
LT-α2-β1	MAIT	IL-7	125	12.6
M-CSF	HSPC	IL-24	206	1.0
MYDGF	CD14 Mono	IL-8	66	1.4
OSM	MAIT	IL-7	41	21.6
OX40L	NK CD56hi	IL-32-β	114	3.9
PDGF-CC	CD16 Mono	IL-4	527	1.8
PDGF-DD	NK CD56low	IL-4	462	1.2
Properdin	CD16 Mono	GITRL	95	0.9
RANKL	NK CD56hi	CD30L	59	1.5
SCF	CD14 Mono	M-CSF	89	10.8
TGF-α	CD14 Mono	IL-4	190	5.6
TGF-β1	CD16 Mono	Leptin	217	1.2
TGF-β2	HSPC	FGF-β	40	23.6
TNF-α	MAIT	IL-7	106	6.6
TRAIL	CD16 Mono	IFN-β	2193	3.1
VEGF	CD14 Mono	IL-32-β	114	7.9
XCL1	NK CD56hi	IFN-λ3	75	1.5
XCL2	NK CD56hi	TSLP	45	1.6

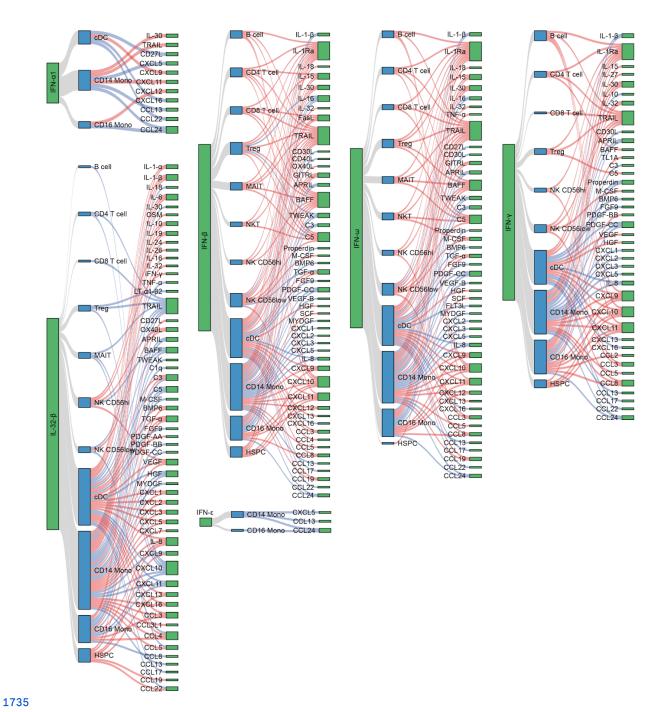
1723 Fig. S18. Table of maximum expression values per expressed cytokine by cell type and stimulation condition. We show 1724 the condition with the largest cpm value for a given cytokine, which doesn't necessarily imply a strong upregulation for that 1725 cytokine relative to baseline, as it might be expressed constitutively (cf. the FC to PBS column). 1726



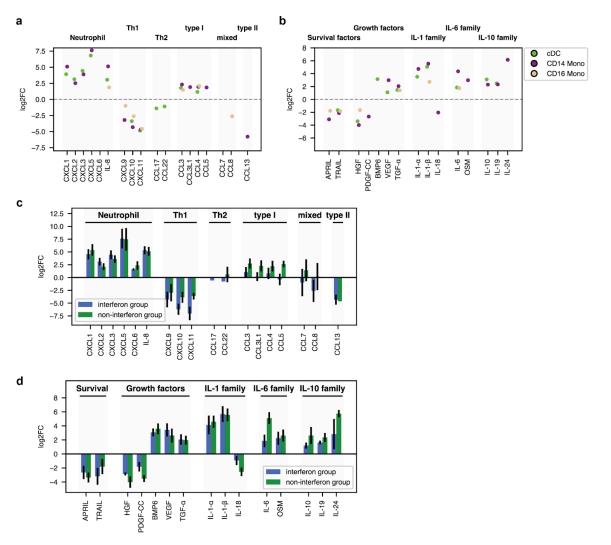
1728 Fig. S19. Cytokine-to-cytokine crosstalk (IL-1- $\beta$ , IL-2, IL-7, IL-15, IL-21, TSLP, IL-6, IL-12). Each connection shows 1729 regulation (FC > 2 in blue, FC <0.5 in blue) of a given cytokine by another cytokine in a given cell type. 1730



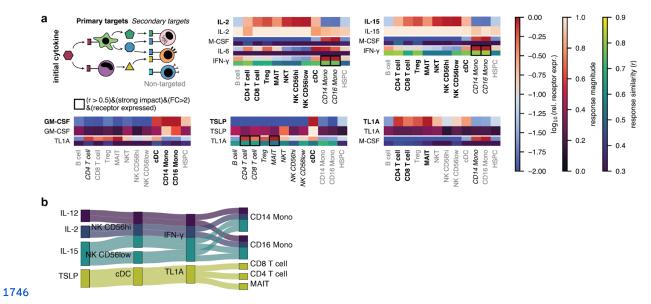
1732 Fig. S20. Cytokine-to-cytokine crosstalk (IL-4, IL-13, IL-3, GM-CSF, IL-10, CD40L, TL1A, C5a, M-CSF). Each connection 1733 shows regulation (FC > 2 in blue, FC <0.5 in blue) of a given cytokine by another cytokine in a given cell type.



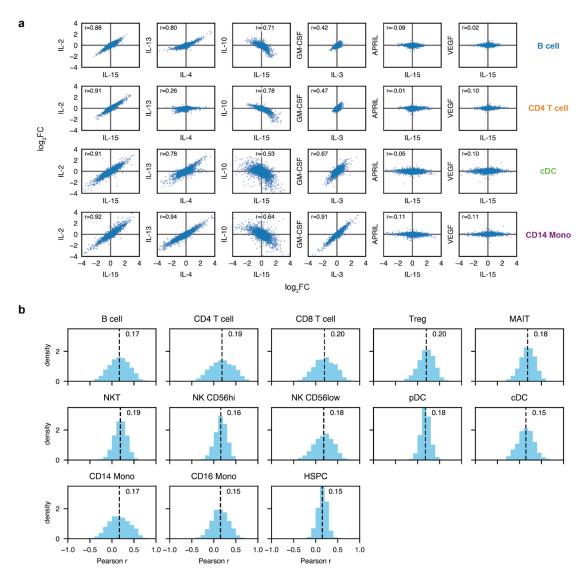
1736 Fig. S21. Cytokine-to-cytokine crosstalk (IFN- $\alpha$ 1, IFN- $\beta$ , IFN- $\omega$ , IFN- $\gamma$ , IFN- $\epsilon$ , IL-32- $\beta$ ). Each connection shows regulation 1737 (FC > 2 in blue, FC <0.5 in blue) of a given cytokine by another cytokine in a given cell type. 1738



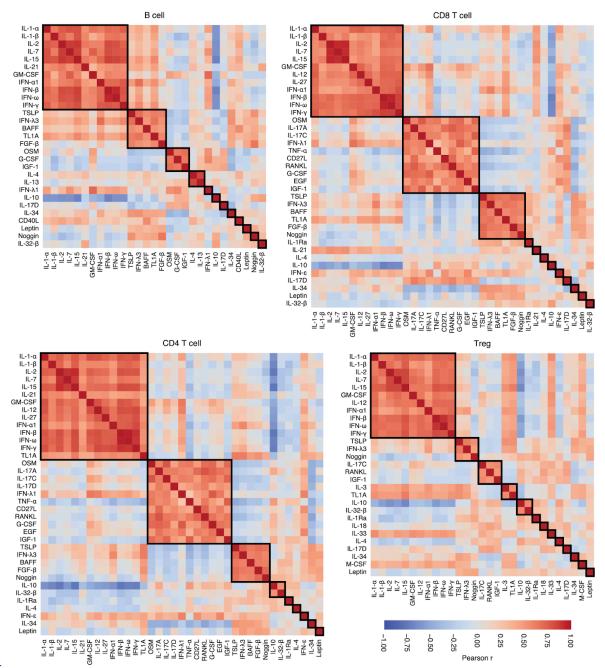
1740 Fig. S22. Cytokine signaling induced by IL-32-β in myeloid cells and by donor group. a,b, Fold changes for chemokines 1741 (a) and other cytokines (b) grouped by cytokine family in myeloid cell types. Missing values indicate that the cytokine did not 1742 pass the filters used for DEG calculation. c,d, Fold changes for chemokines (a) and other cytokines (b) grouped by cytokine 1743 family in CD14 monocytes for two different groups of donors. See Fig. 2 for the list of donors associated with each group. Donor 1744 2 was excluded from this calculation to the strong pre-existing IL-32-β signaling.



1747 Fig. S23. Exploration of potential secondary responses for different cytokines. a, Potential secondary responses were 1748 detected by a combination of low receptor expression and high response magnitude in at least one cell type. We show all 1749 potential secondary cytokines, i.e., cytokines significantly upregulated (FC>2, padj<0.05) in at least one primary target cell type 1750 with receptor expression and a strong response in at least one secondary cell type, are shown. High response similarity (r>0.5) 1751 is used as an additional criterion to narrow down potential secondary signaling mechanisms (black box). b, IL-2, IL-12, and 1752 IL-15 all likely signal to secondary cells via IFN-γ released by (CD56-positive) natural killer cells. TSLP might use TL1A 1753 signaling via cDCs as a mechanism to affect T cells.



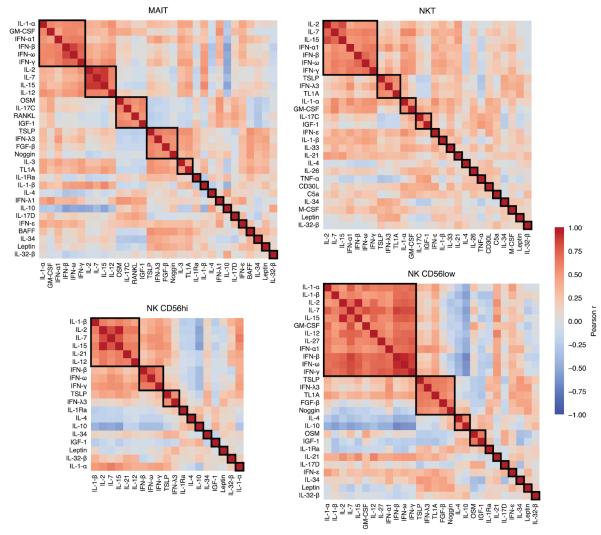
1756 Fig. S24. Cytokine-cytokine log2FCs correlations quantify response similarity. a, Scatter plots of log2FC value 1757 correlations for different chosen cytokines in B cells, CD4 T cells, cDCs, and CD14 Monocytes. b, Distribution of Pearson r 1758 values between all cytokine pairs in different cell types. The dashed line and number show the mean value. There is a slight 1759 bias towards positive correlation.



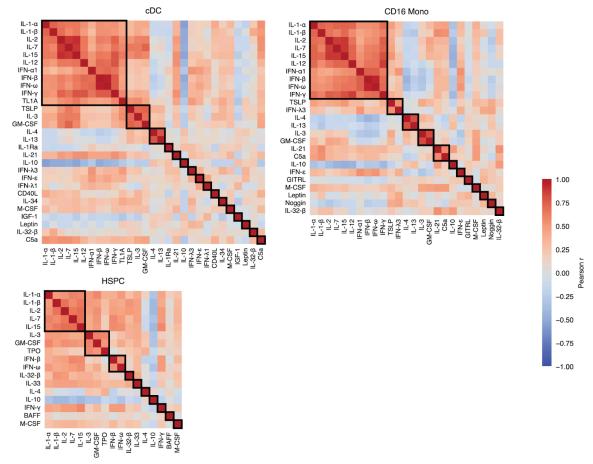
1761 Pearson r

1762 Fig. S25. Correlation heatmaps between log2FC values for different cytokines in B cells, CD4 and CD8 T cells, and 1763 Tregs. Cytokines were grouped based on their correlation patterns using the Leiden algorithm (black boxes).

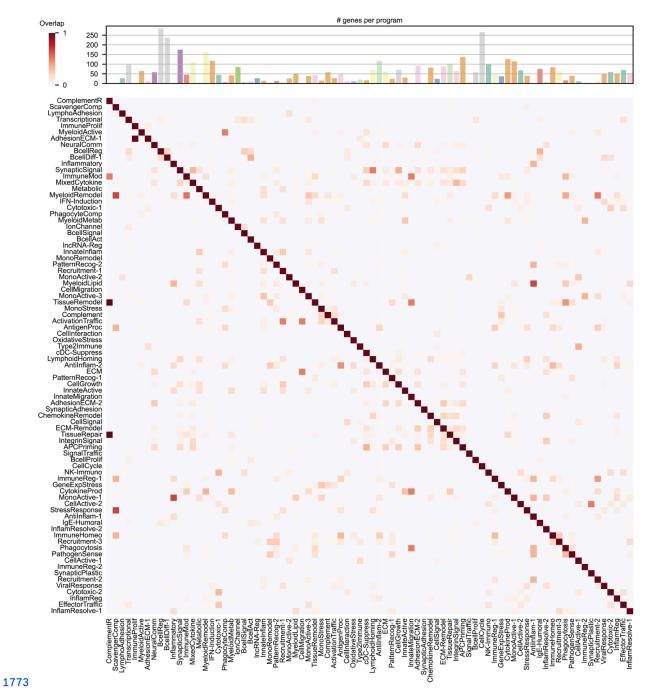
1764



1766 Fig. S26. Correlation heatmaps between log2FC values for different cytokines in MAIT, NK CD56hi and CD56low, and 1767 NKT cells. Cytokines were grouped based on their correlation patterns using the Leiden algorithm (black boxes). 1768

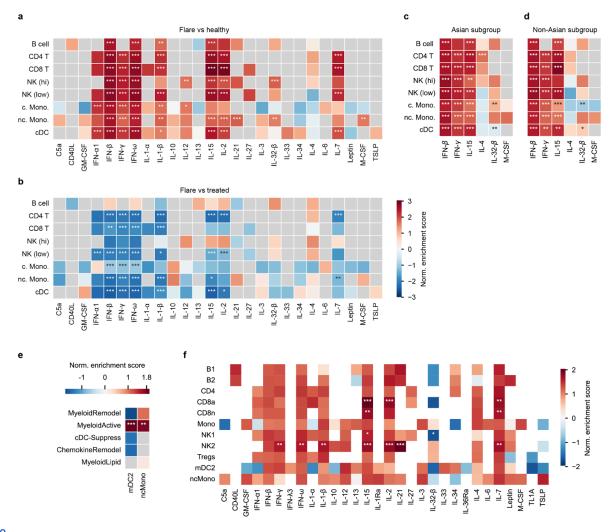


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1770 Fig. S27. Correlation heatmaps between log2FC values for different cytokines in cDCs, CD16 Monocytes, and HSPCs.
1771 Cytokines were grouped based on their correlation patterns using the Leiden algorithm (black boxes).
1772



1774 Fig. S28. Overlap of CIP-associated genes. The barplot above shows the total number of genes per program. The main plot 1775 shows that there is little overlap in genes between programs.

1776



1779
1780 Fig. S29. Additional results of the enrichment analysis. a, Enrichment results for all tested cytokines when comparing Flare 1781 group vs. Healthy group. b, Enrichment results for all tested cytokines when comparing Flare group vs. treated group. c, 1782 Enrichment results for the Asian subgroup, when comparing Flare vs. Healthy. d, Enrichment results for the Asian subgroup, 1783 when comparing Flare vs. Healthy. Results in (c) and (d) do not show qualitative differences between both subgroups. e, CIP 1784 enrichment results for the myeloid compartment, comparing MS vs healthy. f, Enrichment results for all tested cytokines when 1785 comparing MS vs healthy.