

1 **The shifting of dominating roles between structural cells and immune**
2 **cells are key regulators of human adipose tissues aging**

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20

21 **Abstract**

22 Adipose tissue is a highly dynamic organ with complex cellular composition. Aging
23 induces adipose tissue function decline and relocation of peripheral adipose tissue to
24 abdominal compartment, which often associated with inflammation and metabolic
25 disorders. Here we performed single-cell RNA sequencing to comprehensively and
26 unbiasedly deconvolve how subcutaneous adipose tissue (SAT) responses to aging. We
27 collected >25,000 stromal vascular cells from abdominal and gluteofemoral SAT of
28 young and old donors. Analyses of transcription signatures and cell networks uncovered
29 impaired adipogenesis and extracellular matrix synthesis capacity of APC, altered
30 metabolic phenotype of immune cells and shifted tissue-dominating cells that can be
31 used to predict adipose tissue aging. We also reported aging-associated distinct
32 transcriptional program between gluteofemoral SAT and abdominal SAT. Our work
33 thus reveals unanticipated cellular, immunological, metabolic and site-specific aspects
34 of human adipose tissues aging process, providing valuable resource for better
35 understanding of aging-associated adipose tissue dysfunction.

36

37 **Introduction**

38 Adipose tissue distributes throughout the body. It is an extremely complex organ with
39 profound effects on physiology and pathophysiology of human body, and plays critical
40 roles in mechanical support, temperature regulation, metabolic and immune
41 homeostasis¹. Aging is increasingly recognized as a kind of systemic chronic disease,
42 leading to the functional decline of vital tissues. Age-dependent adipose tissue
43 dysfunction is associated with metabolic declines, heterotopic fat accumulation and
44 chronic systemic inflammation, increasing the risk of diseases like diabetes,
45 cardiovascular disease, and even cancer².

46 Aging changes the mass and distribution of adipose tissue throughout the body.
47 Body weight increases with age and fat mass peaks occur in middle or early old age³⁻⁵. With increasing age, peripheral subcutaneous adipose tissue tends to be lost, while
48 visceral adipose tissue (VAT) tends to be preserved^{6,7}. Similar to the VAT, the fat mass
49 of the abdominal subcutaneous adipose tissue (ASAT) increases with advancing age in
50 both male and female^{8,9}, which positively correlates with cardiovascular disease and
51 type 2 diabetes mellitus^{10,11}. Recent studies have shown that adipose progenitor
52 differentiation, adipose tissue immune and metabolic tenors are tightly regulated by the
53 crosstalk between adipose progenitor cells and non-adipocyte fraction of human
54 adipose tissue, including immune cells and structural cells^{12,13}. However, how aging
55 influences cell subpopulations and their crosstalk in human adipose tissue across depots
56 is still unclear.

58 In this study, we present high-throughput single-cell transcriptome analyses of

59 young and old human ASAT and GSAT. Our results reveal the aging-dependent
60 alterations in cell subpopulations, cell-cell interactions, as well as site-specific response
61 to aging of human SAT.

62

63 **Results**

64 **Single-cell atlas of human ASAT**

65 We performed single-cell RNA sequencing on stromal vascular fraction (SVF) cells of
66 ASAT from young and aged participants ([Figure. 1A](#)). In total, 14,073 single cells
67 derived from the ASAT of 3 young (26.33 ± 3.79 years old) and 3 old (74.33 ± 3.51 years
68 old) donors were analyzed ([Supplementary Figure. 1](#)), unsupervised clustering of the
69 gene expression profiles identified 11 cell types, each containing cells from both young
70 and old samples ([Figure. 1B](#), [Supplementary Figure. 2](#)). A list of differentially
71 expressed genes (DEGs) that define the clusters are presented in Supplementary Table
72 1. Analysis of DEGs identified six major clusters of adipose progenitor cells (APC),
73 three clusters of immune cells (IC), a population of vascular endothelial cells (VEC)
74 and a small population of smooth muscle cells (SMC) ([Figure. 1B,C](#)).

75 Our data showed that APC3, APC5, APC6 expressed higher level of stem cell
76 markers (*CD55*, *PII6*, *SEMA3C*, *DPP4*) ([Figure. 2A](#)), while APC1, APC2 and APC4
77 expressed higher level of early adipogenic markers (*APOE*, *FABP4*, *CD36*, *VCAMI*)¹⁴⁻
78¹⁶ ([Figure. 2B](#)). The three stem-like APC clusters could be distinguished by high
79 expression of genes corresponding to temperature response in APC3, detoxification in
80 APC5, and extracellular matrix (ECM) organization in APC6 ([Figure. 2C](#)). Within the

81 three committed preadipocyte population, APC2 and APC4 exhibited the similar
82 phenotype with stem-like population APC3 and APC6 respectively, which represents
83 the temperature-effector differentiation trajectory (APC3-APC2) and ECM-effector
84 differentiation trajectory (APC6-APC4). APC1 was featured by high expression of
85 genes associated with leukocyte migration (Figure. 2D). These results reveal the
86 coordination of cell type and maturation stage within human APC populations.

87 It was suggested that aging increased cellular transcriptional instability, which led
88 to cell fate drift and function loss ¹⁷. Therefore, we analyzed transcriptional noise
89 following previous work, and identified that transcriptional noise was increased with
90 aging in major cell populations inhabited in human ASAT (Supplementary Figure. 3),
91 which was consistent with the previous findings in mice ^{17,18}.

92

93 **Aging causes the accumulation of a dysfunctional APC subpopulation**

94 Segregation of the aggregated t-SNE plot of APC demonstrated that committed
95 preadipocyte APC4 was mainly composed of cells from the young ASAT (YASAT),
96 whereas stem-like population APC5 included cells mostly from the old ASAT (OASAT)
97 (Figure. 3A,B). Besides the stem cell property, APC5 expressed high level of
98 metallothionein genes (*MT1A*, *MT1E*, *MT1M*, *MT2A*), which are associated with cell
99 dysfunction ¹⁹(Figure. 3C). These alterations indicate that aging impeded the
100 differentiation of ECM-trajectory (APC6-APC4), result in the accumulation of
101 dysfunctional stem-like APC5. Although the changes of cellular quantity were minor,
102 the split t-SNE plot further indicated that aging shifted the expression profile of APC1,

103 APC2, APC3, and APC6.

104 To characterize the common features caused by aging, we analyzed differentially
105 expressed genes for APC1, APC2, APC3, and APC6 across young and aged ASAT, and
106 identified a series of genes commonly changed by aging in these clusters. Only those
107 genes consistently changed (up-regulated or down-regulated) across all these four
108 clusters were defined as common DEGs ([Supplementary Table 2](#)). Gene ontology (GO)
109 analysis of the common DEGs of these four APC populations during aging showed that
110 genes related to extracellular structure organization were downregulated, while genes
111 involved in apoptotic signaling pathway, blood coagulation, and neutrophil activation
112 were upregulated ([Figure. 3D](#)). These results show that aging induced the expansion of
113 dysfunctional aged adipose progenitor cells and partly impedes the adipogenic
114 differentiation ([Figure. 3E](#)), but the ECM synthesis ability was broadly hindered across
115 all APC populations.

116

117 **Aging alters the immune and metabolic phenotype of human ASAT**

118 Previous studies have highlighted immune cells alteration responding to adipose tissue
119 aging. To learn more details about immune cell profile of ASAT, we re-clustered cells
120 of IC1-IC3 and identified 8 specific immune cell subpopulations, labelled as ICS1-
121 ICS8 ([Figure. 4A](#)).

122 ICS1-ICS4 showed gene expression signatures of T cell and natural killer (NK)
123 cell ([Figure. 4B,C](#)). Unsupervised cell type annotation using human primary cell atlas
124 reference (Methods) showed that both of ICS1 and ICS4 were consisted of a mixture

125 of naïve CD8⁺ T cell and memory CD4⁺ T cell ([Supplementary Table 3](#)). High
126 expression level of ribosomal protein-related genes distinguished ICS1 from ICS4,
127 while ICS4 expressed higher level of activated T cell related genes *CXCR4* and *IL32*
128 ^{20,21} ([Figure. 4C, Supplementary Table 4](#)). ICS2 was annotated as NK cell for the unique
129 marker *NKG7* ([Figure. 4C](#)). ICS3 possessed 56% tissue stem cells according to our
130 unsupervised cell annotation ([Supplementary Table 3](#)), and highly expressed *CD34* and
131 genes related to ECM organization, epithelial cell proliferation, vasculature
132 development ([Figure. 4C,D](#)). Cell quantification demonstrated the most obvious
133 difference between young and old ASAT was the expansion of ICS1 in old tissues
134 ([Figure. 5A,B](#)). These results indicate the expansion of a specific T cell subpopulation
135 with active ribosome biogenesis activity during ASAT aging.

136 Recent study showed that aging up-regulates specific T cell markers in old VAT ²².
137 To investigate whether the expression of these aging-related T cell markers was also
138 upregulated in the expanded ICS1, we analyzed the DEGs between young and old T
139 cell subpopulations. Surprisingly, ICS4 but not ICS1 expressed higher level of aging-
140 related T cell markers *CD44* and *PDCD1* ([Supplementary Table 5](#)). Besides, the DEGs
141 analysis revealed that ICS1, ICS2, and ICS4 possessed similar top 10 DEGs during
142 aging ([Supplementary Table 5](#)), and in total we found 289 genes were similarly changed
143 by aging in these three subpopulations ([Supplementary Table 6](#)). The common
144 upregulated genes were related to antimicrobial humoral response and detoxification of
145 copper ion ([Figure. 5C](#)), while genes involved in interferon signaling pathway were
146 downregulated ([Figure. 5D](#)). Taken together, these results suggest that T cells in

147 subcutaneous adipose tissue may take a different way responding to aging from visceral
148 adipose tissue, although the cell number was both increased.

149 ICS5-7 highly expressed *CD86* and antigen presentation-related genes, the
150 markers of monocyte/macrophage. The expression of *CD163* distinguished ICS6 as M2
151^{12,23}, and the high expression level of *S100A9* indicated that ICS7 was M1 macrophage
152²⁴ (Figure. 4B). ICS5 could be identified as dendritic cell according to the unsupervised
153 cell type annotation (Supplementary Table 3). Aging reduced the total proportion of
154 ISC5-7 (Figure. 5A,B) and significantly shifted their transcriptional profile. Genes
155 involved in chemokine activity of ISC5-7 were significantly upregulated upon aging
156 (Figure. 5E), whereas genes related to NADH dehydrogenase was downregulated,
157 suggesting the downregulated oxidative phosphorylation activity, especially in ISC7
158 (Figure. 5F). These results indicate that aging increases inflammatory level and alters
159 metabolic phenotype of monocyte/macrophage in human ASAT (Figure. 5G).

160 The smallest cluster ICS8 was designated to B cell for the expression of *CD79A*,
161 *JCHAIN*, and *MZB1* (Figure. 4B, Supplementary Table 4). Compared to young ASAT,
162 B cells of old ASAT expressed higher level of *IGHA2*, *IGHG1*, *IGHG4* and *IGHA1*,
163 which encoded the heavy chain of IgG and IgA (Supplementary Table 5). IgG produced
164 by B cells is known to be associated with insulin resistance in obese mice²⁵. Thus, our
165 data indicate that B cells may contribute to the glucose intolerance and insulin
166 resistance in aged ASAT.

167

168 **Aging resets the cell subpopulation interaction pattern in ASAT**

169 Having defined the alterations of progenitor cell and immune cell populations during
170 aging, we utilized iTALK to perform an unbiased ligand-receptor interaction analysis
171 between these populations ²⁶.

172 We inspected the top 20 chemokines and growth factors mediated ligand-receptor
173 interactions. In young ASAT, APC showed strong secretion activity of chemokine
174 CXCL12, an important adipose environmental factor and growth factor CTGF ²⁷.
175 CXCL12 acts on the receptors of lymphocyte subpopulations (ICS1-4, and ICS8),
176 whereas CTGF mostly acts within APC subpopulations ([Figure. 6A,C](#)). In keeping with
177 our finding that aging increased the inflammatory level in monocyte/macrophage
178 subpopulations, aging dramatically enhanced the secretion of pro-inflammatory factor
179 IL1B by M1 macrophage (ICS7) and the expression of IL1B receptor IL1R1 by APC
180 subpopulations ([Figure. 6B](#)). However, the secretion activity of APC was suppressed in
181 aged ASAT ([Figure. 6B](#)). During aging, growth factors-mediated interactions were
182 decreased in immune cell subpopulations, and maintained in APC populations ([Figure.](#)
183 [6D](#)). Collectively, our analyses demonstrate that the ligand-receptor interaction activity
184 of ICS1 was decreased despite of its increased cell proportion during ASAT aging, and
185 M1 exhibited the most obvious variation in cell interactome, although its cell proportion
186 was slightly decreased during aging.

187 To investigate whether gluteofemoral subcutaneous adipose tissues show the same
188 phenomenon during aging, we performed single cell RNA sequencing on gluteofemoral
189 subcutaneous adipose tissues (GSAT) from young and aged participants. Unsupervised
190 cell clustering and marker-based cell type annotation identified 6 APC populations, 3

191 IC populations, 1 VEC and 1 SMC population ([Supplementary Figure. 4](#)), indicating
192 that overall cell population constitution of GSAT was similar to that of ASAT.

193 We next asked how GSAT-derived APC respond to aging. Similar to ASAT, aging
194 shifts the distribution of APC populations in GSAT. APC1 mainly composed of cells
195 from old GSAT, and highly expressed cell dysfunctional-related genes (MT2A, MT1E,
196 MT1A and MT1M), which means aging causes the accumulation of a dysfunctional
197 APC subpopulation in GSAT as in ASAT ([Supplementary Figure. 5](#)).

198 Re-clustering of immune cells identified 8 cell subpopulations ([Supplementary](#)
199 [Figure. 6](#)). Unlike ASAT, the proportion of T cell decreased in aging GSAT, while the
200 proportion of DC cell population was increased. Ligand-receptor interaction analysis
201 showed that APC from YGSAT showed strong secretion activity of chemokine
202 CXCL12 acting on the receptors of lymphocyte subpopulations (ICS1, ICS3, ICS7). In
203 addition, aging dramatically enhanced the secretion of pro-inflammatory factor IL6 by
204 the dysfunctional APC (APC1). These results indicated that APC dominated cell-cell
205 interactions in both young and aged GSAT but mediated by different cytokine
206 ([Supplementary Figure. 7](#)).

207

208 **The change of dominating cell population determined the aging of human adipose**
209 **tissues**

210 In this research, we found aging changes the cell population ratio and crosstalk in
211 adipose tissues ([Figure. 7](#)). In both human ASAT and GSAT, aging induced the
212 expansion of dysfunctional aged cells. In young ASAT and GSAT, APC populations

213 dominated the cell-cell interactions through paracrine or autocrine of CXCL12. In old
214 ASAT, immune cell populations secret inflammatory factor IL1 to act on APC
215 populations, indicating an inflammatory tissue microenvironment. In contrast to ASAT,
216 aging didn't shift the dominating cell population in GSAT. It seems that inflammatory
217 factor IL6 secreted by dysfunctional APC1 of GSAT is the key regulator of the old
218 GSAT tissue microenvironment (Figure. 7).

219

220 **Discussion**

221 In this study, we constructed a single-cell aging atlas of human abdominal subcutaneous
222 adipose tissue. The overall cellular component and proportion were similar to a recent
223 published research demonstrating the obese subcutaneous adipose tissue at single-cell
224 level ²⁸, which verified the reliability of our data.

225 Multiple studies have reported the heterogeneity in adipose tissue stem progenitor
226 cells ¹⁴⁻¹⁶. In addition to the degree of differentiation, there are also different trajectories
227 during stem cell differentiation ²⁹. Our results showed the coordinating of maturation
228 stage and differentiation path of APC, and identified at least two trajectories exhibited
229 in ASAT, the ECM secreting APC, and temperature responding APC. Collagen is an
230 important kind of ECM in adipose tissue, it surrounds adipocytes limiting adipocyte
231 hypertrophy and promoting APC adipogenesis ³⁰. We found that the ECM-secreting
232 APC (APC4) sharply decreased during aging. The decrease in collagen levels may
233 cause adipose tissue to lose its ability to maintain homeostasis, leaving fat cell
234 hypertrophy in an uncontrolled way, leading to increased hypoxia in adipose tissue and

235 enhanced apoptotic signal, finally suppress adipogenesis of APC. Our data also showed
236 a cluster of stem-like but dysfunctional APC accumulated in elderly adipose tissue,
237 again demonstrating that aging tissue is not always accompanied by stem cells
238 exhaustion, but by arrest of dysfunctional stem cells.

239 Previous studies demonstrated that the phenotype and function of macrophages
240 are closely related to their metabolic patterns. Pro-inflammatory M1 macrophages
241 mainly rely on glycolysis and present breaks on the tricarboxylic acid cycle. On the
242 contrary, M2 cells are more dependent on oxidative phosphorylation (OXPHOS)^{31,32}.
243 Our data showed that aging enhanced the inflammatory level of adipose tissue
244 macrophages as well as disturbed its OXPHOS process. The synergic effect of
245 inflammation and metabolism deteriorates the elderly macrophages in ASAT.

246 ASAT aging atlas confirmed the earlier findings that aging is associated with
247 inflammatory macrophages in adipose tissue³³. However, our cell subpopulations
248 interaction analysis provided additional support to the idea that M1 macrophages
249 dominated the inflammatory microenvironment of old ASAT. The secretary activity of
250 APC was suppressed by the inflammatory environment during aging, however, the
251 expression level of IL1 receptor responding to the inflammatory signals of M1 was up-
252 regulated. These results also indicate that APC from elderly people may not be suitable
253 for using as stem cell source to treat inflammatory diseases, because it's susceptible to
254 the inflammatory environment.

255 In a recently research, the immune functions of structural cells across multiple
256 mice organs were identified³⁴. APC is the structural cells of adipose tissue. Oure data

257 verified the important immune functions of structural cells both in human abdominal
258 and gluteofemoral adipose tissues. Besides, we found that aging could change the
259 immune functions of APC by weakening the strength in abdominal adipose tissue or
260 shifting towards to a proinflammatory status in gluteofemoral adipose tissue. These
261 phenomena suggest that targeting the immune functions of structural cells may be a
262 feasible strategy to treat diseases related to tissue aging.

263 In conclusion, our study provides a comprehensive characterization of structural
264 cells, immune cells and their crosstalk in human adipose tissues during aging. We
265 emphasize the importance of crosstalk between dominating cell populations and other
266 cell populations in the regulation of tissue microenvironment upon aging. We see
267 our study and large-scale dataset as a starting point and reference atlas for
268 mechanistic explorations in structural cells-immune cells crosstalk mediated tissue
269 homeostasis maintenance.

270

271 **Methods**

272 **Study subjects**

273 Human adipose tissues were obtained from patients undergoing a specific surgical
274 procedure with the approval of the Second Affiliated Hospital, Zhejiang University.
275 Specifically, man and woman who had a body mass index between 18.5 kgm^{-2} and 30
276 kg m^{-2} and required surgery which could harvest abdominal or gluteofemoral
277 subcutaneous adipose tissues were included. For scRNA-seq, the young group included
278 patients between the ages of 16 to 29 years old, the old group included patients between

279 the ages of 68 to 87 years old. The exclusion criteria included pregnancy, liver cirrhosis,
280 inflammatory bowel disease. The characteristics of included patients were in
281 supplementary table 7.

282

283 **SVF isolation**

284 To collect SVF cells, adipose tissues were digested at 37°C using collagenase type I
285 (Gibco). SVF cells were collected every 30 mins, and the residual adipose tissues were
286 added to fresh collagenase type I until 95% tissues were digested. Then the SVF cells
287 well treated with RIPA lysis buffer to remove red blood cells. Cells were resuspended
288 with full culture medium (L-DMEM (Gibco)+10% FBS (Gibco)), and 2× freezing
289 medium was added. Finally, cells were placed in programmed cooling box at -80°C for
290 24 hours, and then stored in liquid nitrogen until analysis.

291

292 **scRNA-seq**

293 The scRNA-seq experiment was performed using the Chromium Single Cell 3' Solution
294 v2 platform (10x Genomics), following the manufacturer's protocol. SVF cells were
295 thawed. After calculating, the same number of young or old SVF cells were mixed in
296 one tube respectively, and dead cell removal kit (Miltenyi Biotec) was used to improve
297 cell quality. scRNA-seq libraries were prepared using the Chromium Single Cell 3'
298 Reagent Kit v2. Libraries were sequenced by HiSeq X Ten (Illumina) system.

299

300 **Quality control and analysis of single-cell data**

301 The 10x Genomics Inc. software package CellRanger (v2.1.0) and the GRCh38
302 reference genome were used to perform sample de-multiplexing, alignment and
303 quantification of unique molecular identifiers (UMI). The UMI matrices produced by
304 CellRanger were used for downstream analysis by using R (version 3.6.0) and the
305 Seurat package (version 3.1.3)³⁵. Gene expression matrices were filtered to remove
306 cells with >10% mitochondrial genes and < 500 genes. After quality control filtering,
307 data were normalized to remove the effects of the number of genes detected per cell,
308 the number of counts, and the percentage of mitochondrial reads. Normalized data of
309 young subcutaneous adipose tissue were combined into one object and integrated with
310 data of old subcutaneous adipose tissue. Variable genes were discovered using the
311 *SelectIntegrationFeatures* function with nfeatures = 1000. Integration anchors across
312 all samples were discovered using the *FindIntegrationAnchors* function command with
313 default parameters. The *IntegrateData* function was run on the anchor set to integrate
314 all samples with default arguments. Dimensionality reduction was performed with
315 *RunPCA*. Then t-stochastic neighboring embedding method (tSNE) dimensionality
316 reduction was carried out and Shared Nearest Neighbour (SNN) graph constructed
317 using dimensions 1-18 (for ASAT) or 1-15 (for GSAT) as input features and default
318 PCA reduction. Cell clustering was performed on the integrated assay at a resolution of
319 0.5. SingleR with default parameters was used to generate annotation for each cell
320 cluster³⁶. Differentially expressed genes among clusters was identified by using the
321 function *FindAllMarkers* and examined by Wilcox test, only test genes that are detected
322 in a minimum fraction of 25% cells in either of the two populations. Violin plots, dot

323 plots, and tSNE plots for the given genes and cell populations were generated by using
324 the *VlnPlot*, *DotPlot*, and *FeaturePlot* functions, respectively.

325

326 **Multi-lineage interactome analysis of single-cell data**

327 Cell-cell interaction analysis was performed based on the scRNA-seq data by using
328 iTALK²⁶. Top 50% highly expressed genes were used for further analyses. The
329 software built-in database containing a total of 2,648 unique ligand-receptor interacting
330 pairs were used to identify significant interactions. The top 20 significant interactions
331 were visualized based on the R package circlize³⁷.

332

333 **Transcriptional noise analysis**

334 Transcriptional noise between each ASAT cell type derived from old and young donors
335 was quantified based on previous work using the gene expression profiles^{18,38}. Firstly,
336 to remove the effects of the differences in total UMI counts all cells were downsampled
337 to have equivalent number of total counts. Downsampling was also performed on the
338 cell numbers so that equal numbers of young and old ASAT cells in each cell type were
339 used. Then, all the genes were assigned to ten equally sized bins by their mean
340 expression, and the top and bottom 10% of genes were excluded for further analyses.
341 Next, the 10% of genes with the lowest coefficient of variation within each bin were
342 selected. The down-sampled UMI count matrix was reduced to these genes and
343 transformed by square root. Then, the Euclidean distance between each cell and the
344 mean expression level of each cell-type within young and old group was determined.

345 Spearman's correlation coefficients were calculated on the subsampled UMI count
346 matrix between all pairwise cell comparisons within each cell type and age group.
347 Wilcoxon's rank sum test was used to evaluate the association between transcriptional
348 noise and age within each cell type. The *p* values were adjusted for multiple testing
349 using the FDR procedure Bonferroni–Hochberg.

350

351 **Data availability**

352 The datasets generated during the study are not publicly available for now but will be
353 available before publication.

354

355 **Code availability**

356 The code to reproduce the analyses and figures described in this study is available upon
357 reasonable request.

358

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454

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461

462 **Author contributions**

463 W.Z. and J.L. designed and performed experiments and analyzed the data. W.Z. and J.L.

464 wrote the manuscript. X.H. collected human adipose tissues. X.Y. assisted with

465 manuscript writing. H.W.O conceived ideas and oversaw the research program.

466

467 **Declaration of interests**

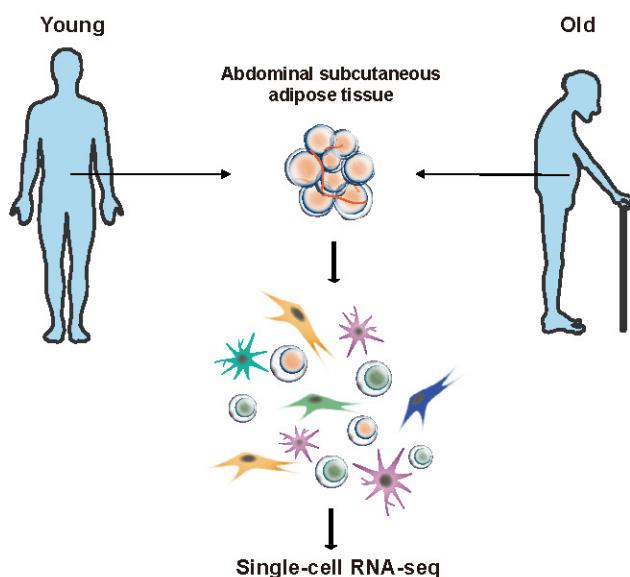
468 The authors declare no competing interests.

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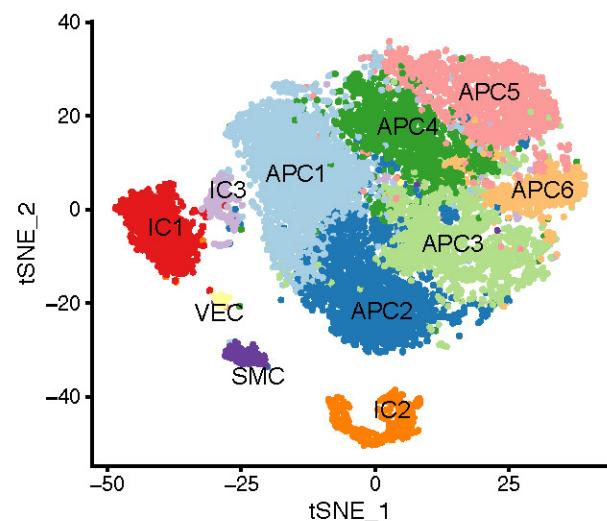
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Figures and legends

A



B



C

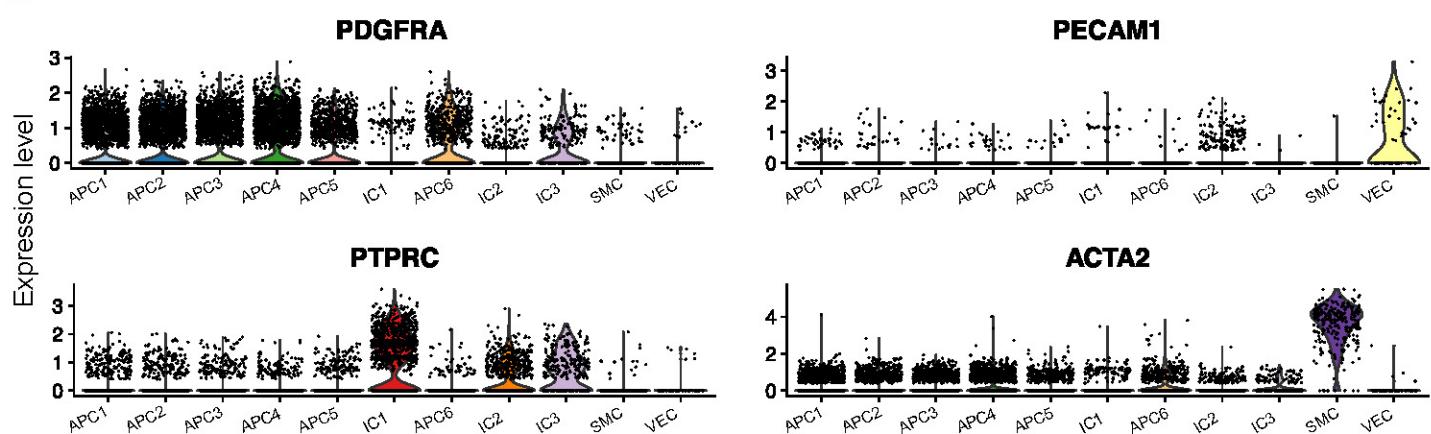
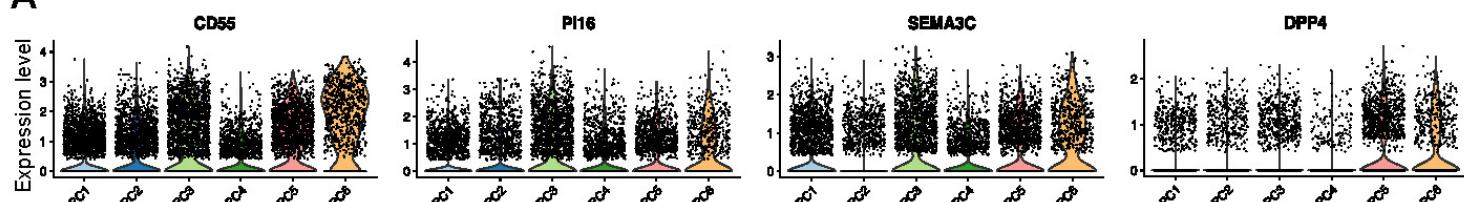
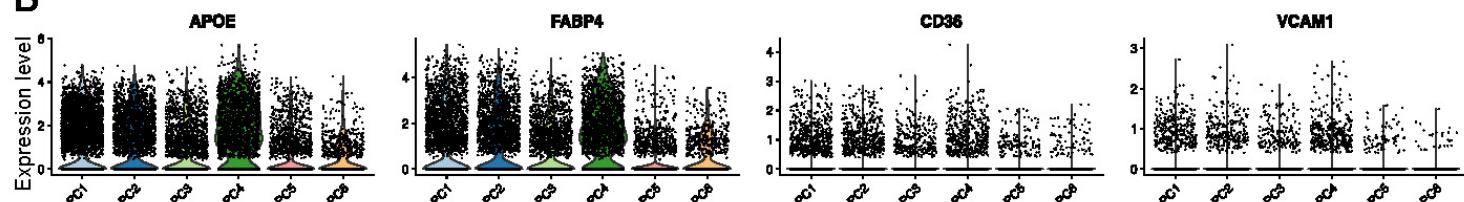


Figure 1 Single-cell RNA-sequencing of human ASAT. (A) Schematic diagram of the experimental workflow. (B) The t-Distributed stochastic neighbor (t-SNE) plot shows unsupervised clustering of 14,073 single-cell transcriptomes. (C) Violin plots show the expression levels of representative cell-type-specific marker genes across all 11 cell types.

A



B



C

APC3	APC5	APC6	response to temperature stimulus
13	0	0	
11.3	0	0	response to heat
0	9.93	0	detoxification of copper ion
0	7.88	0	detoxification
0	0	15.1	extracellular matrix organization
0	0	6.74	collagen metabolic process

D

APC1	APC2	APC4	myeloid leukocyte migration
8.471	0	0	
8.359	0	0	leukocyte migration
0	15.71	0	response to temperature stimulus
0	12.34	0	response to heat
0	0	9.447	extracellular structure organization
0	0	6.615	extracellular matrix organization

Figure 2 APC heterogeneity of human ASAT. (A) Violin plots show the expression levels of stem cell markers and (B) early adipogenic markers of human APC. (C) Heatmap shows the gene ontology term of differentially expressed genes of stem-like populations APC3, APC5, APC6 and (D) committed preadipocyte populations APC1, APC2, APC4.

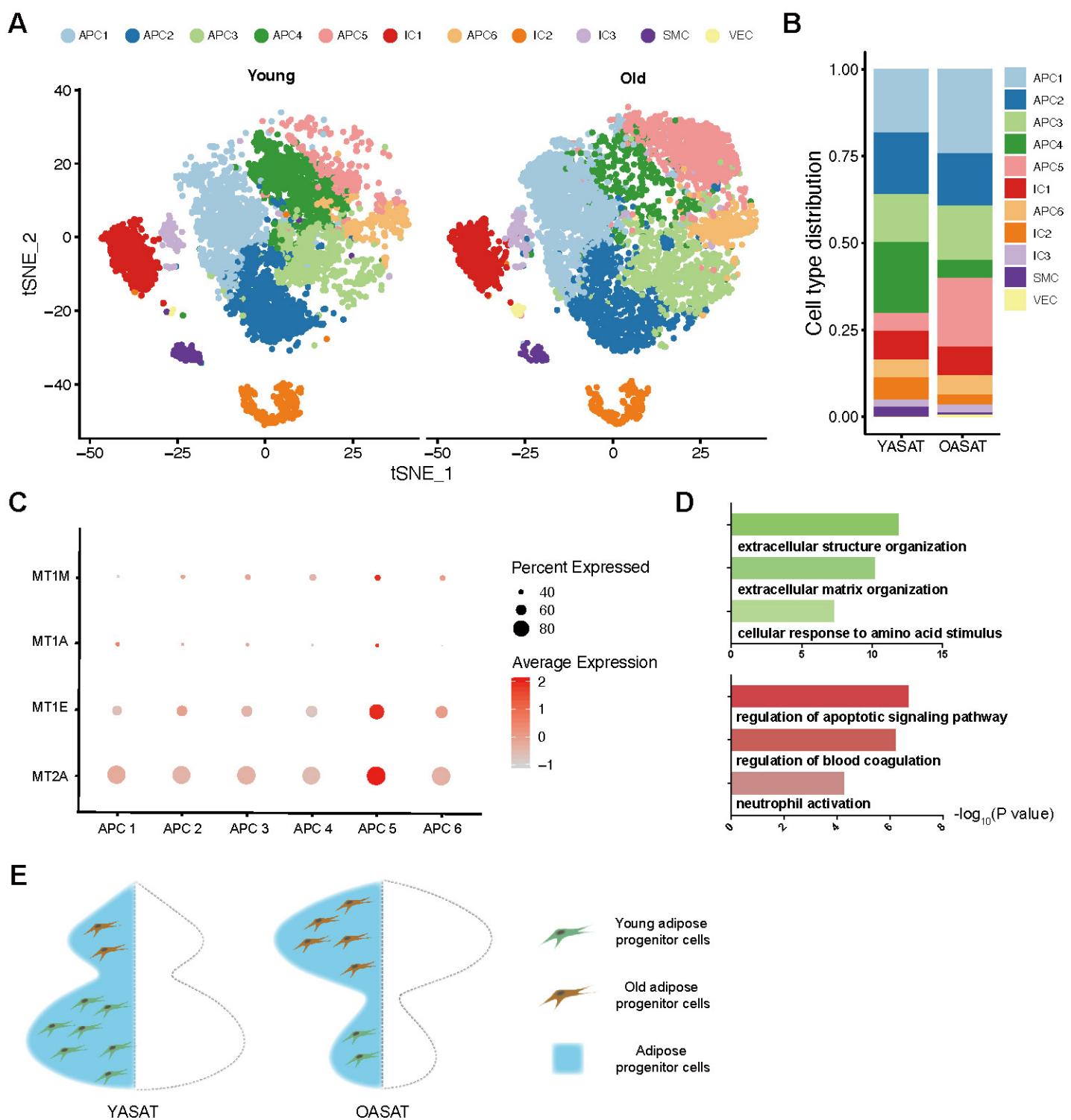


Figure 3 Alterations of APC subpopulations during ASAT aging. (A) The t-SNE plots of cell clusters in young and old ASAT split from Figure 1B. (B) Cell type distribution of young and old ASAT. (C) Dot plot of the expression of metallothionein genes across APC populations. (D) Representative GO terms of common downregulated (upper plot) and upregulated (lower plot) genes during aging of APC1, APC2, APC3, and APC6. YASAT: young abdominal subcutaneous adipose tissue; OASAT: old abdominal subcutaneous adipose tissue. (E) A schematic representation of the changes of adipose progenitor cell subpopulations in young and old ASAT.

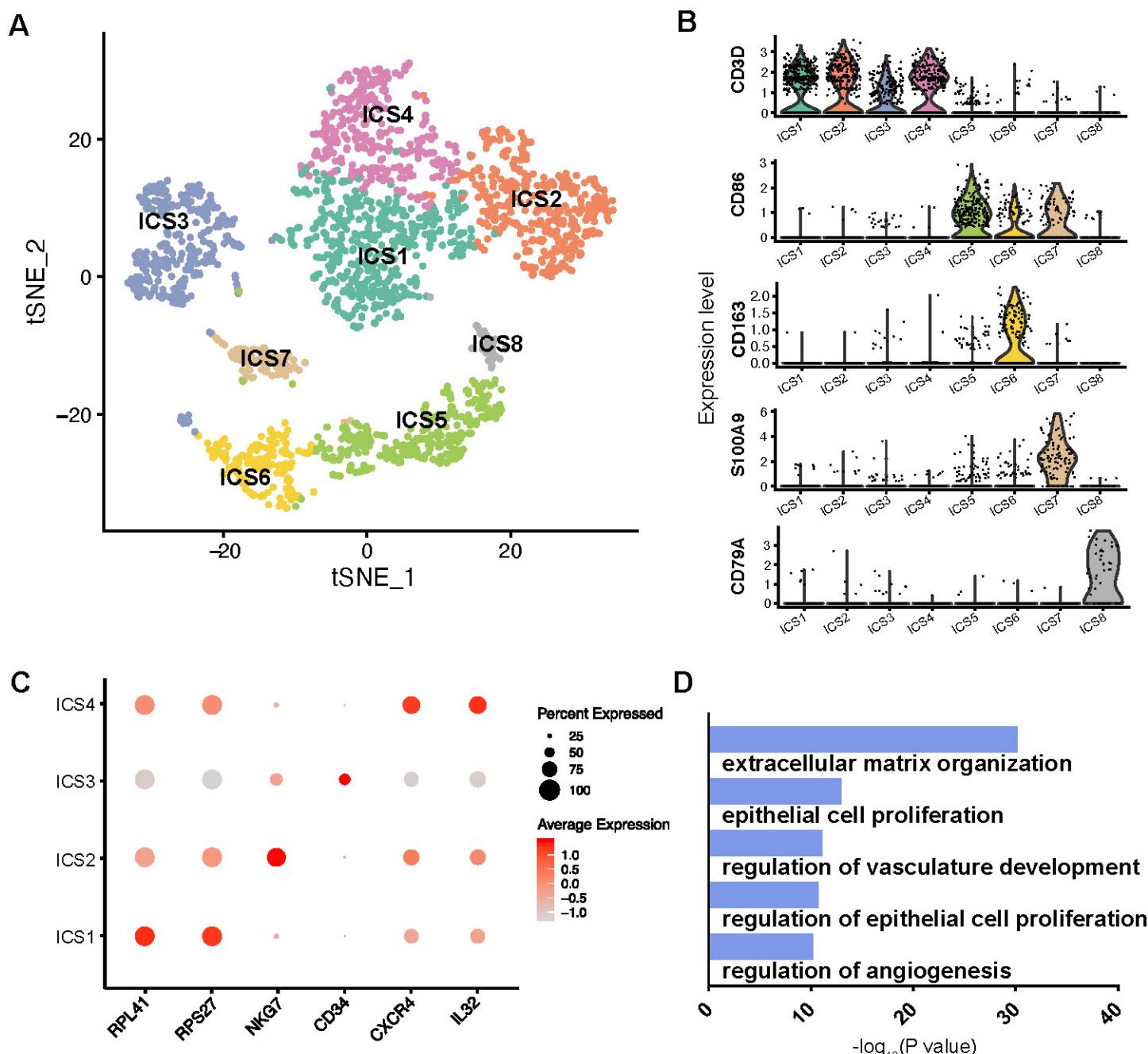
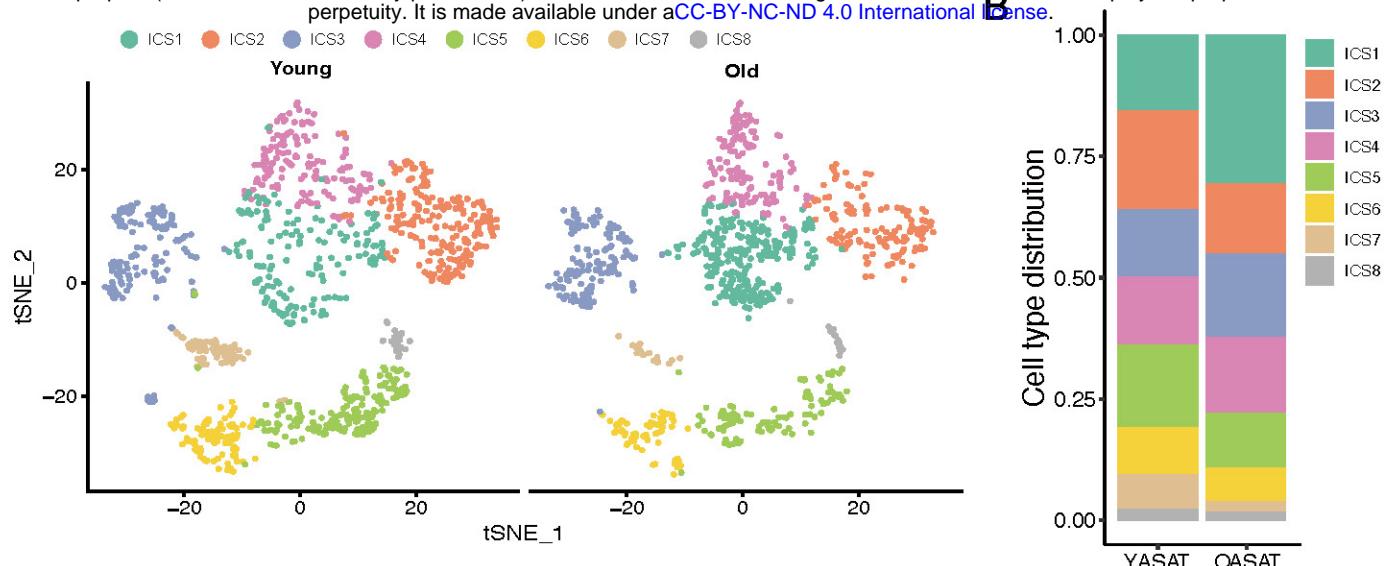
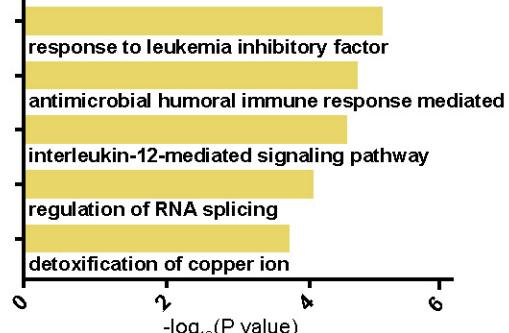


Figure.4 Immune cell heterogeneity of human ASAT. (A) Re-clustering of IC1-IC3 in Figure.1B identified 8 specific immune cell subpopulations. (B) Violin plots show the expression levels of representative cell-type-specific marker genes across all these 8 immune cell subpopulations. (C) Dot plot of the expression of representative genes across ICS1-ICS4. (D) GO analysis of specific expressed genes of ICS3.

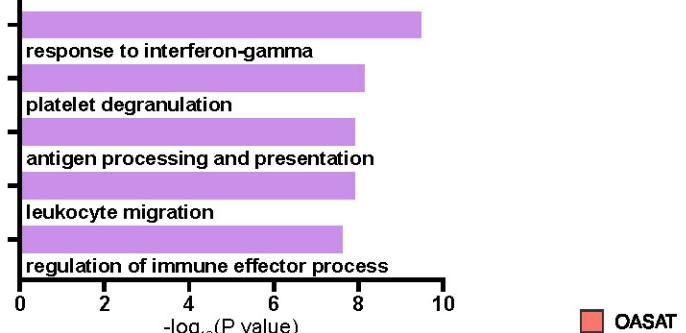
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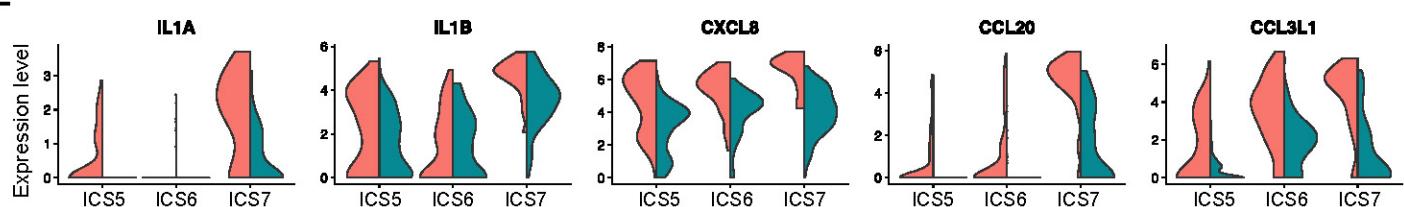
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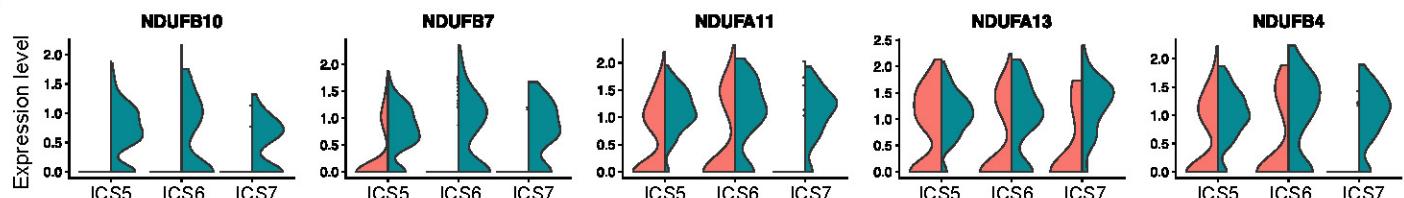
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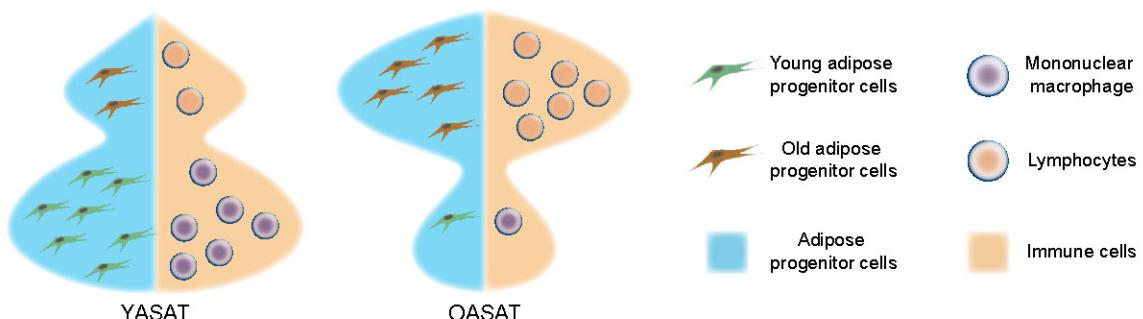


Figure 5 Alterations of immune cell subpopulations during ASAT aging. (A) The t-SNE plots of cell clusters in young and old ASAT split from Figure 4A. (A) Immune cell type distribution of young and old ASAT. (C) GO analysis of the common upregulated genes in ICS1, ICS2 and ICS4. (D) GO analysis of the common downregulated genes in ICS1, ICS2 and ICS4. (E) Violin plots show the expression levels of chemokine activity-related genes and (F) NADH dehydrogenase-related genes in ICS5-7. YASAT: young abdominal subcutaneous adipose tissue; OASAT: old abdominal subcutaneous adipose tissue. (G) A schematic representation of the changes of immune cell subpopulations in young and old ASAT.

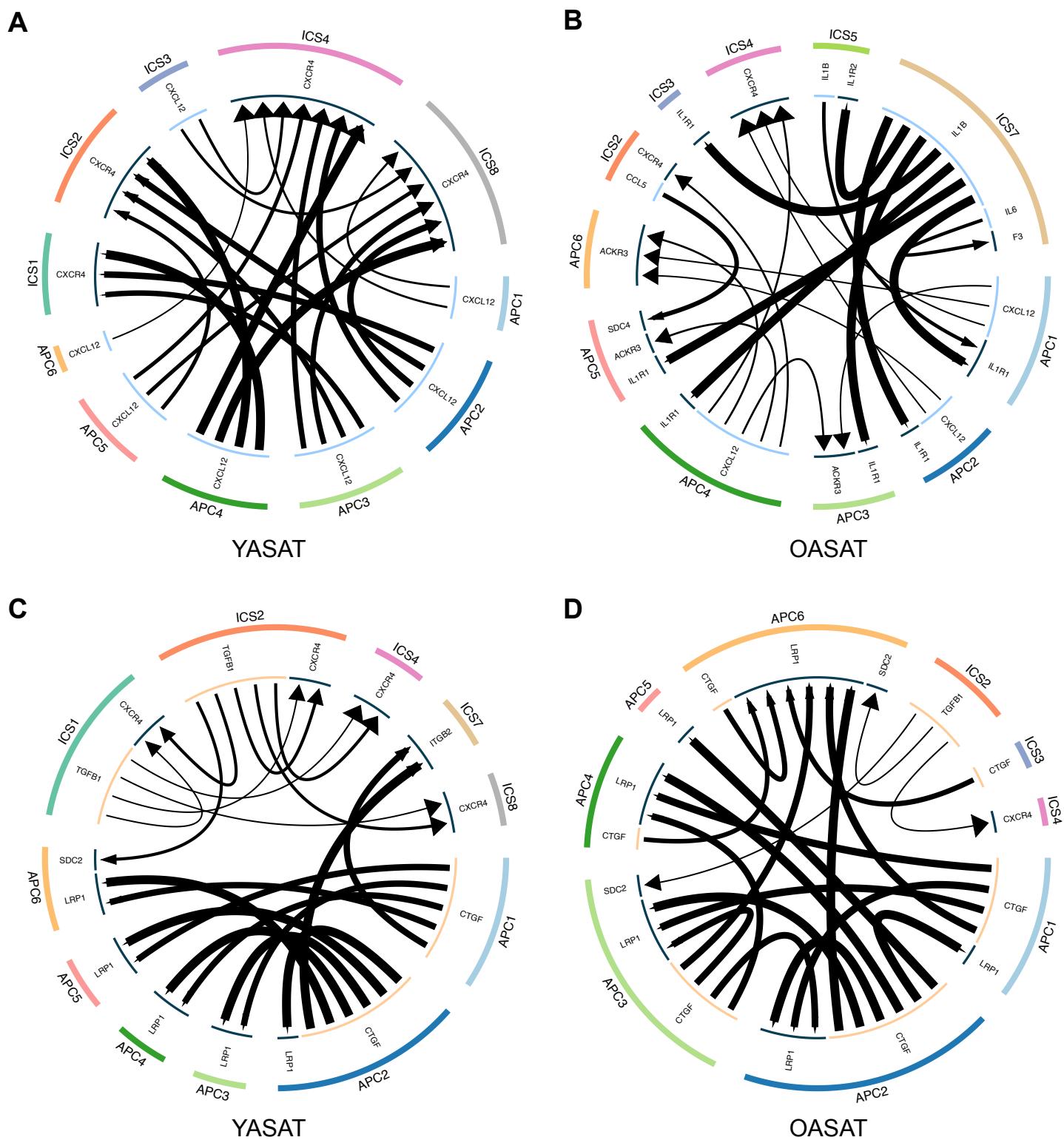


Figure 6 Multi-lineage interactions in young and old ASAT. (A-B) Circus plots showing top 20 chemokines mediated ligand-receptor interaction for all APC and immune cell subpopulations. (C-D) Circus plots showing top 20 growth factors mediated ligand-receptor interaction for all APC and immune cell subpopulations. YASAT: young abdominal subcutaneous adipose tissue; OASAT: old abdominal subcutaneous adipose tissue.

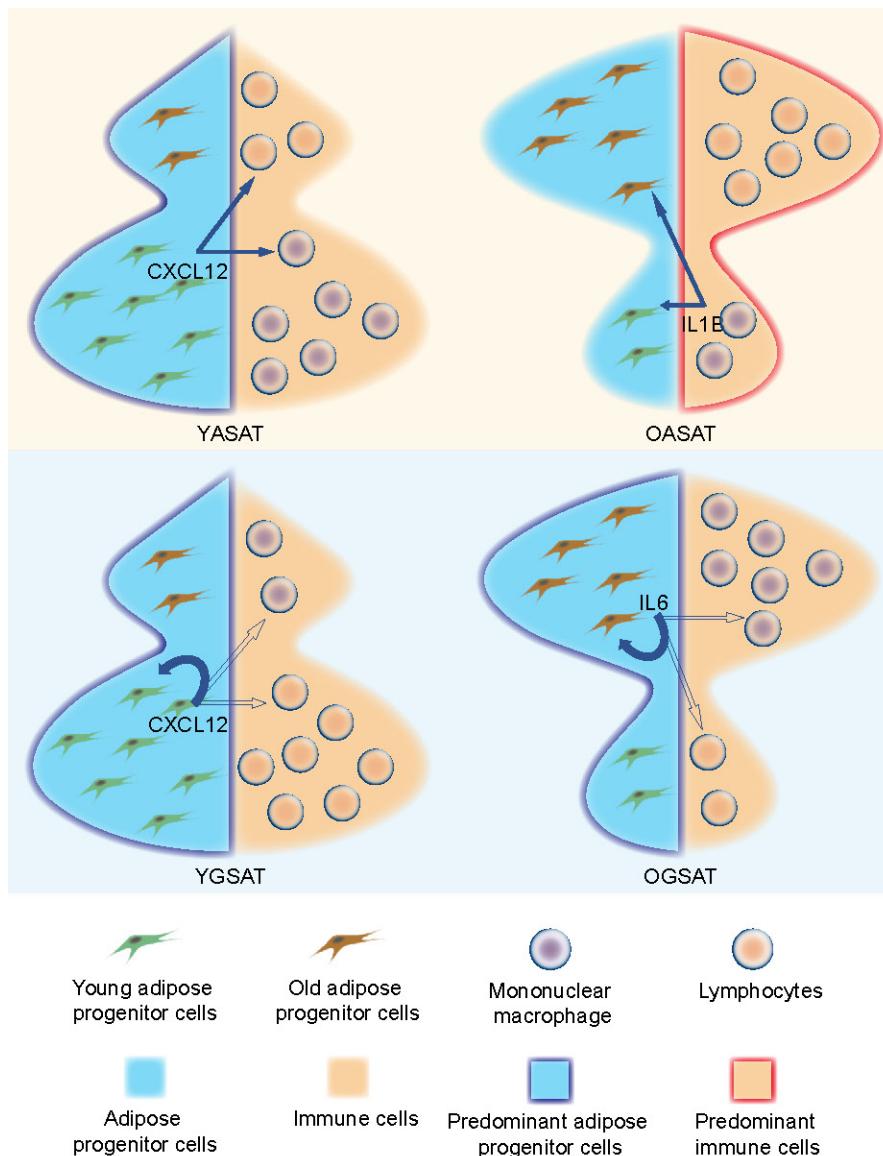


Figure.7 The schematic diagram of cell populations' crosstalk during the aging process of ASAT and GSAT. In YASAT, adipose progenitor cells dominate the tissue by secreting CXCL12 which acts on immune cells. In OASAT, immune cells dominate the tissue by secreting IL1 which acts on adipose progenitor cells. In YGSAT, adipose progenitor cells dominate the tissue by secreting CXCL12 which acts on both immune cells and itself. In OGSAT, adipose progenitor cells dominate the tissue by secreting IL6 which acts on both immune cells and itself.