





Unraveling Immunogenomic Diversity in Single-Cell Data

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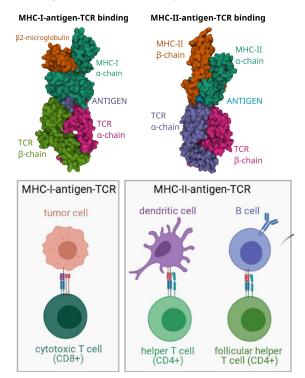
Demo material



https://github.com/ahmadalajami/scIGDWorkflowDemoBioC2024

Structural diversity of receptors

Receptors that coordinate antigen specificity in immune responses

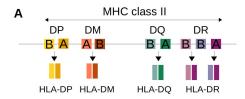


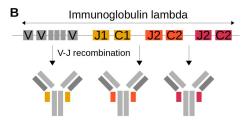
created with Biorender

Genetic mechanisms of structural diversity

Polygeny:

Multiple similar genes encode same function



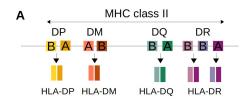


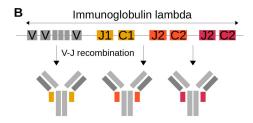
Gene group	Number of similar genes (incl. pseudogenes)
IGHG	5
HLA-DRB	8
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Genetic mechanisms of structural diversity

Polygeny:

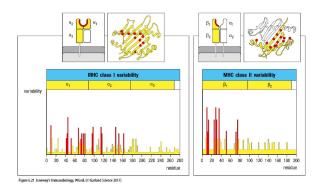
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Hyperpolymorphism:Multiple alleles for one gene

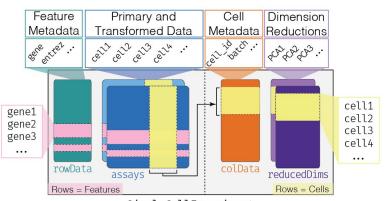


Gene group	Number of alleles (incl. pseudogenes)
IGHG	71
HLA-DRB	~4200

Source: IMGT and IPD-IMGT/HLA

Typical transcriptomic analysis workflow

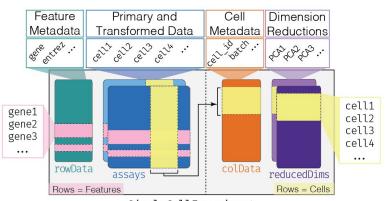
- 1. For each sequencing read:
 - a. identify cellular barcode
 - b. identify unique molecular identifier (UMI)
 - c. map read to reference genome
- 2. Count unique transcripts per gene and barcode using UMI
- 3. Generate feature count matrix (genes x cells)
- 4. Downstream analysis
 - a. dimension reduction
 - b. clustering
 - c. differential gene expression
 - d. ..



SingleCellExperiment

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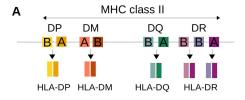


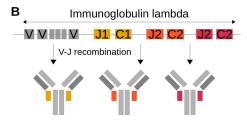
SingleCellExperiment

Allelic diversity and polygeny pose a bioinformatic challenge

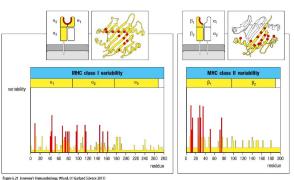
Polygeny:

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Hyperpolymorphism: Multiple alleles for one gene



Bioinformatic challenge

Cross-mapping: uniquely mapping to the wrong gene

Allelic diversity and polygeny pose a bioinformatic challenge

Polygeny:

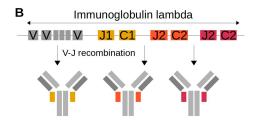
Multiple similar genes encode same function

MHC class II

DP DM DQ DR

BABBBA

HLA-DP HLA-DM HLA-DQ HLA-DR



MHC class I variability a1 a2 a3 MHC class I variability B1 A2 A2 MHC class II variability A1 A2 A3

Figure 6.21 Janeway's Immunobiology, 9th ed. (© Garland Science 2017)

Aim

To perform quantification of allele-specific expression for immunogenomic analysis

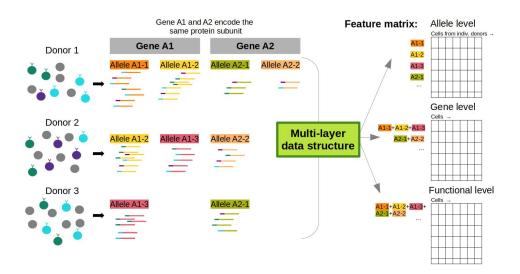
Hyperpolymorphism: Multiple alleles

for one gene

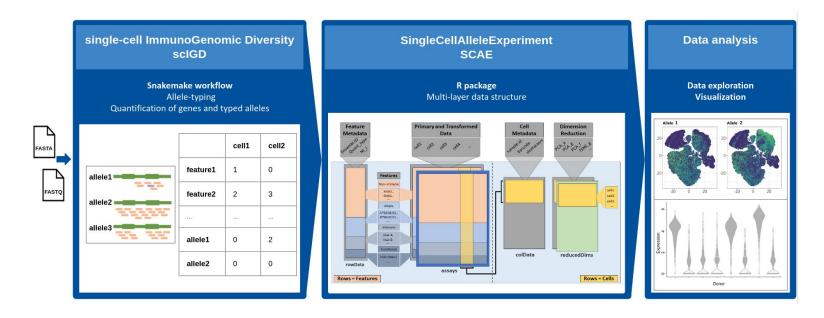
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Implementation of allele and functional information What we want

Building a multi-layer data structure for immune gene representation in single-cell data



Workflow How to get there



Example datasets present in scaeData – R/ExperimentHub data package

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Project partner

Federico Marini

scIGD



Group of Quantitative Immunology

Katharina Imkeller
Jonas Schuck

scaeData



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https://agimkeller.github.io

SingleCellAlleleExperiment











