

Spatial Biology Congress |2024-12-07

## Spatial atlas of inflammatory skin disease and cutaneous T cell lymphoma

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## Introduction

LΡ

**Benign** Malignant Benign Inflammatory skin disease Cutaneous T Cell Lymphoma (CTCL) PS AD Eczema **Ps**oriasis MF SS DAR **M**ycosis **S**ézary **F**ungoides *Syndrom* Lichen **Dar**ier **P**lanus Disease

- Benign inflammatory skin diseases and CTCL are both deregulation of T cells in the skin
- To improve cure of the malignant diseases, multiple challenges remain:
  - **Enhance diagnostic** ٠
  - Predict response in malignant disease

Roediger et al., T cells in the skin: Lymphoma and inflammatory skin disease (2022)

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## Multiplexed protein imaging of skin diseases



- 1) What characterizes malignant skin disease ?
  - 2) Can we predict response to treatment ?

### But ... Challenges in analyzing highly multiplex fluorescence images

- Manageable
  - artifacts

Folds 



- Varying intensity range
- CD14 300-700

Stains

Low Signal/Noise

Whole punch spills 



**Spillover** 

**More challenging** artifacts







### But ... Challenges in analyzing highly multiplex fluorescence images

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- Varying intensity
   range
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## Image vs average intensity

Spillover





Averaging intensity **hides the pattern**.

3000

0

"Classical" tools, based on the average intensity suffer from this fact.

#### Deep-learning based cell phenotyping reveals heterogeneous cell composition

B cell

CD8T

Treg

#### CellSighter<sup>1</sup>

Annotated 17 celltypes in 11 images (>8000 cells)



#### Prediction vs Ground Truth (Validation)

Accuracy = 0.862; F1 = 0.690



<sup>1</sup>CellSighter: a neural network to classify cells in highly multiplexed images, Amitay et al., Nat. Comm. (2022)

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#### Deep-learning based cell phenotyping reveals heterogeneous cell composition



#### Deep-learning based cell phenotyping reveals heterogeneous cell composition



## "Hacking" CellSighter for image-based positive marker detection



<sup>1</sup>CellSighter: a neural network to classify cells in highly multiplexed images, Amitay et al., Nat. Comm. (2022)

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## Leveraging deep-learning for positive marker detection



ROI-19 ; CD3+ mask - 10µm



## Back to the biology

### Cell composition reveals strong heterogeneousity within diseases



- Strong heterogeneity of cell composition
- Some malignant patients have similar composition to healthy / benign

## Increased T cell densities and CD3 loss in malignant skin disease



#### Strong reduction of the hallmark CD3 marker in T helper in malignant disease

Sano et al., Immunological study on CD3 defective cutaneous T cell lymphoma cells from a patient with Sézary syndrome. Clin Exp Immunol. (1998) Liu et al., Single-cell transcriptomics links malignant T cells to the tumor immune landscape in cutaneous T cell lymphoma. Nat Commun 13, 1158 (2022)

## Spatial configuration : the «immune pathological infiltrate»



- Building from cell-in-contact network we find communities of immune cells using
- louvain clustering (keeping those > 30 cells).

## Spatial configuration : the «immune pathological infiltrate»



- The IPI is increased in malignant compared to most benign skin NCD.
- They are highly lymphocytic compared to the rest and are significantly more vascularized.

## Diving further : finding «immunological synapses»



## Diving further : finding «immunological synapses»



## Diving further : finding «immunological synapses»



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## T cytotoxic fraction improves response to treatment



CAILS score evolution of the biopsied lesion in CTCL

- Classifying patients according to evolution of CAILS score (Responder vs Non Responder)
- Responsive patients have in average 2-3 times greater proportions of cytotoxic
- T cells at time of biopsy (validated on survival of additionnal 34 patients)



- We build a spatial atlas of immune skin diseases with precisely phenotyped cells and functional markers
- T helper cells in CTCL shift their program to evade immune surveillance by cytotoxic T cells
- The best predictor for lesion evolution is the fraction of cytotoxicic T cells at time of biopsy

## See the story online





## https://bit.ly/skin-atlas

#### Spatial atlas of noncommunicable skin diseases �

Skin is our body's largest organ, serving as the first line of defense against infection. Within the skin are various immune cells that protect us from gathogens. Yet sometime, skin diseases arise from deregulations in T cells which are also strongly influenced by all other immune cells resent in the skin.

Multiplex imaging technology provides unprecedented access to the spatial organization and protein expression of single cells within the skin. This powerful tool allows us to better characterize different immune diseases, particularly distinguishing between benign conditions like. Psoriasis and Eczemä, and malignant conditions such as Cutaneous T-cell Lymphoma (CTL) which can be line-threatening.

To achieve this, we collected several skin biopsies from patients with various skin diseases and employed multiplex imaging to map their protein landscapes.

Get ready for the tour

Image adapted from "T cells and the skin: from protective immunity to inflammatory skin disorders", Ho and Kupper, Nat. Rev. 2019, DOI

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