

An image dataset related to automated macrophage detection in immunostained lymphoma tissue samples --Manuscript Draft--

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Abstract:	<p>Background: We present an image dataset related to automated segmentation and counting of macrophages in diffuse large B-cell lymphoma (DLBCL) tissue sections. For the classification of DLBCL subtypes as well as for providing a prognosis of the clinical outcome, the analysis of the tumor microenvironment and, particularly, of the different types and functions of tumor-associated macrophages, is indispensable. Until now, however, most information about macrophages is obtained either in a completely indirect way by gene expression profiling or by manual counts in immunohistochemically (IHC) fluorescence stained tissue samples while automated recognition of single IHC stained macrophages remains a difficult task. In an accompanying publication, a reliable approach to this problem has been established, and a large set of related images has been generated and analyzed.</p> <p>Results: Provided image data comprise a) fluorescence microscopy images of 44 multiple immunohistostained DLBCL tumor subregions, captured at four channels corresponding to CD14, CD163, Pax5 and DAPI; b) cartoon-filtered versions of these images, generated by Rudin-Osher-Fatemi (ROF) denoising; c) an automatically generated mask of the evaluation subregion, based on information from the DAPI channel, and d) automatically generated segmentation masks for macrophages, B-cells and the total of cell nuclei, using information from CD14, CD163, Pax5 and DAPI channels, respectively.</p> <p>Conclusions: A large set of IHC stained DLBCL specimens is provided together with segmentation masks for different cell populations generated by a reference method for automated image analysis, thus featuring considerable reuse potential.</p>	
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DATA NOTE

An image dataset related to automated macrophage detection in immunostained lymphoma tissue samples

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Abstract

Background: We present an image dataset related to automated segmentation and counting of macrophages in diffuse large B–cell lymphoma (DLBCL) tissue sections. For the classification of DLBCL subtypes as well as for providing a prognosis of the clinical outcome, the analysis of the tumor microenvironment and, particularly, of the different types and functions of tumor-associated macrophages, is indispensable. Until now, however, most information about macrophages is obtained either in a completely indirect way by gene expression profiling or by manual counts in immunohistochemically (IHC) fluorescence stained tissue samples while automated recognition of single IHC stained macrophages remains a difficult task. In an accompanying publication, a reliable approach to this problem has been established, and a large set of related images has been generated and analyzed.

Results: Provided image data comprise a) fluorescence microscopy images of 44 multiple immunohistostained DLBCL tumor subregions, captured at four channels corresponding to CD14, CD163, Pax5 and DAPI; b) cartoon-filtered versions of these images, generated by Rudin–Osher–Fatemi (ROF) denoising; c) an automatically generated mask of the evaluation subregion, based on information from the DAPI channel, and d) automatically generated segmentation masks for macrophages, B–cells and the total of cell nuclei, using information from CD14, CD163, Pax5 and DAPI channels, respectively.

Conclusions: A large set of IHC stained DLBCL specimens is provided together with segmentation masks for different cell populations generated by a reference method for automated image analysis, thus featuring considerable reuse potential.

Key words: lymphoma, DLBCL, macrophage, multiple immunohistochemical staining, automated cell counting, ROF filtering, floating threshold, rule-based detection, image dataset

Data Description

Context

We present an image dataset generated as a part of an accompanying publication, which is concerned with method development and comparison for automated segmentation and counting of macrophages in diffuse large B-cell lymphoma (DLBCL) tissue sections [1]. DLBCL is an aggressive cancer disease which is characterized by a large heterogeneity of pathological, clinical and biological features [2]. Therefore, a crucial step for the classification of DLBCL subtypes as well as for providing a prognosis of the clinical outcome is the analysis of the tumor microenvironment in terms of counts, local distributions and functions of the different cell populations and, particularly, of the tumor-associated macrophages occurring there [3].

Until now, most information about macrophages is obtained either by gene expression profiling [4] or by manual counts in immunohistochemically (IHC) stained tissue microarrays resp. high-power fields, thus either gathering information in a completely indirect way or accepting extreme subsampling rates [5]. A reliable approach for fully automated segmentation, identification and counting of IHC stained macrophages within whole tissue slides has been addressed in [1].

Our dataset contains monochrome fluorescence microscopy images of 44 DLBCL tissue samples wherein different macrophage populations (using antibodies against CD14 and CD163) and B-cells (using antibody against Pax5) as well as the total of cell nuclei (using DAPI) have been stained and imaged at different wavelengths. Further, we supply processed images, comprising cartoons (generated by Rudin-Osher-Fatemi filtering) as well as results of the automated macrophage segmentation. For this publication, we completed these data by automated segmentation of B-cells and the total of cell nuclei.

Methods

a) Preparation and staining of DLBCL tissue. From the files of the Lymph Node Registry Kiel, 44 DLBCL biopsy specimens have been selected. For every specimen, from formalin-fixed paraffin-embedded tissue a slice of 2 μm thickness has been obtained. In order to detect specific macrophages and its relation to B-cells, a triple IHC staining has been done, using primary antibodies against CD14 (Clone EPR3653; Cell Marque, Rocklin, CA, USA; 1:10), CD163 (Clone 10D6; Novocastra, Leica Biosystems, Wetzlar, Germany; 1:100) and Pax5 (polyclonal; Santa Cruz Biotechnology, Heidelberg, Germany; 1:100) labelled with donkey anti rabbit Alexa 488, donkey anti mouse Alexa 555 and donkey anti goat Alexa 647 (all from Invitrogen, Thermo Fisher Scientific, Waltham, MA, USA; 1:100) as secondary antibodies. Subsequently, the slices have been incubated with DAPI (Invitrogen, Thermo Fisher Scientific, Waltham, MA, USA; 1:5000) and cover-slipped with mounting medium. Use of tissue was in accordance with the guidelines of the internal review board of the Medical Faculty of the Christian-Albrechts-University Kiel, Germany (No. 447/10).

b) Selection of tumor subregions and image acquisition. Within every tissue sample, the tumor area was defined and marked by a pathologist based on inspection of conventional Haematoxylin-Eosin (HE) staining in a neighboring reference slice. Subsequently, within the IHC stained slice, a rectangular subregion of the tumor area has been selected, taking care for acceptable tissue and staining quality. Maximum size of tumor subregions is 10 mm^2 .

Images of tumor subregions within the IHC stained slides have been captured by Hamamatsu Nanozoomer 2.0 RS slide

scanner (Hamamatsu Photonics, Ammersee, Germany) with $20\times$ magnification at four wavelengths, resulting in single images for the CD14, CD163, Pax5 and DAPI channels, respectively. Pixel size is $0.45\times 0.45\ \mu\text{m}$ in all images. Single-channel raw images have been converted into uncompressed .tif format and sliced into tiles of 1000×1000 px format (at right and lower border, the sizes may be smaller), using the software package ImageJ with the extension ndpertools [6]. The resulting monochrome images have been further converted from RGB into greyscale mode using the modulus of the RGB vector and finally saved in losslessly compressed .png format. We refer to them as to images of type "original". Let us remark that image acquisition and tiling have been performed in such a way that no spatial misalignment between the scans at the different wavelengths occurred.

c) Image processing. For every tile, the segmentation method from [1] has been applied to the CD14, CD163, Pax5 and DAPI channel images, resulting in ROF-filtered cartoons of the images (saved as type "cartoon"), a mask for the evaluation subregion within the tile, indicating the presence of tissue at all, as inferred from DAPI channel information (saved as type "eval-mask"), and segmentations of macrophages within the CD14 and CD163 channels (saved as type "segment"). Due to the large inhomogeneity of IHC staining, even across a single target macrophage, we provide two further masks containing the convex hulls of the segmented features instead of the features themselves (saved as type "convhull"). The segmentation masks for double-stained macrophages are saved as type "multiple". For a general description of the ROF filter based segmentation method, we refer to [1]. Here, we describe in more detail the generation of segmentations for the Pax5 and DAPI channels, which are new in this paper.

Let us recall the notation from [1] where the indices i and j count the current intensity threshold and the features to be inspected at this stage, $s(F_j)$, $c(F_j)$ and $r(F_j)$ denote the size of a feature F_j itself, the size of its convex hull and the ratio of the principal axes' lengths of the smallest ellipse covering the feature, respectively. s_{min} , s_{max} , c_{max} and r_{max} denote the minimal and maximal feature size (in px), the maximal area excess of the convex hull (in percent) and the maximal ratio of axes, respectively.

In order to obtain a segmentation of the DAPI channel, the ROF-filtered cartoon has been subjected to a local Narendra-Fitch contrast enhancement [7]

$$p(k, l)_{enhanced} = m(k, l) + \frac{c}{\sigma(k, l)} \cdot \left(p(k, l)_{original} - m(k, l) \right) \quad (1)$$

where $c > 0$ is a weight parameter and $m(k, l)$, $\sigma(k, l)$ denote the mean and standard deviation of the intensities within a subregion centered at the pixel $p(k, l)_{original}$, respectively. We used $c = 0.75$ and a square subregion of 11×11 px size. Then, in a first run, Steps 3 – 10 of the ROF filter based segmentation have been applied, using the bounds $s_{min} = 60$ and $s_{max} = 119$ for the feature size but modifying geometrical rule No. 3) for feature classification from [1] as follows: If $s_{min} \leq s(F) \leq s_{max}$ then test whether the feature satisfies both of the criteria 3b) $r(F_j) \leq r_{max}$ (the feature is not too elongated) and 3d) $c(F_j)/s(F_j) \leq 1 + c_{max}/100$ (the deviation from circular shape is bounded from above). If yes, save the feature F_j into the output mask, interpreting it as a cell nucleus, and mask it in $I^{(3)}(i)$. If not then neglect the feature and mask it in $I^{(3)}(i)$ as well. Here, we used the parameter values $r_{max} = 2.5$ and $c_{max} = 150$. In a second run, Steps 3 – 10 of the ROF filter based segmentation have been repeated with the parameter settings $s_{min} = 120$ and $s_{max} = 180$, using again the described modification of rule No. 3) but saving only those features into the output mask which are completely disjoint to the output of the first run. Finally, the

results of both runs have been combined into a single mask (saved as type "segment"). Within a further result mask of type "convhull", the convex hulls of the detected features have been stored.

For the segmentation of the B-cells, the ROF-filtered cartoon of the Pax5 channel has been subjected to a moderate Narendra-Fitch contrast enhancement as well, using the parameter $c = 0.1$ and a square subregion of 15×15 px size. To the result, Steps 3 – 10 of the ROF filter based segmentation have been applied, using the bounds $s_{min} = 80$ and $s_{max} = 159$ as well as the described modification of rule No. 3) with parameters $r_{max} = 2.5$ and $c_{max} = 150$ but saving into the output mask (of type "segment") only features which intersection with the convex hull of some cell nucleus, as obtained in the segmentation of the DAPI channel, is nonempty. Thus, numerous artifacts appearing in the Pax5 staining will be excluded. Again, the convex hulls of the detected features have been stored within a further mask of type "convhull".

multiple, evalmask]_mode_[gs, bw].png. Size of losslessly compressed .png image files has been minimized by application of OptiPNG routine [8]. Moreover, a logfile named specimen_xx_tile_yy_zz_logfile.txt is provided, containing detailed information about procedures, parameters and results of automated segmentation.

Reuse potential

Although there is a vast number of publications concerned with the composition of tumor microenvironment in various types of lymphoma disease, image datasets of IHC stained cancer tissue are rarely publicly accessible if at all, cf. the discussion in [9]. Most data generated for the purpose of such analyses are not findable or not even accessible. For example, the Genomic Data Commons Data Portal of the National Cancer Institute [10, 11] currently lists only 48 cases of mature B-cell

Table 1. Image files available within a given second-level folder.

Channel	Description	Type	Mode
CD14	CD14 staining, original single-channel image	original	gs
	ROF filtered cartoon derived from original	cartoon	gs
	mask highlighting the segmented macrophages	segment	bw
	mask highlighting the convex hulls of the segmented macrophages	convhull	bw
	mask highlighting the segmented macrophages bearing CD163 staining as well	multiple	bw
CD163	CD163 staining, original single-channel image	original	gs
	ROF filtered cartoon derived from original	cartoon	gs
	mask highlighting the segmented macrophages	segment	bw
	mask highlighting the convex hulls of the segmented macrophages	convhull	bw
	mask highlighting the segmented macrophages bearing CD14 staining as well	multiple	bw
Pax5	Pax5 staining, original single-channel image	original	gs
	ROF filtered cartoon derived from original	cartoon	gs
	mask highlighting the segmented B-cells	segment	bw
	mask highlighting the convex hulls of the segmented B-cells	convhull	bw
DAPI	DAPI staining, original single-channel image	original	gs
	ROF filtered cartoon derived from original	cartoon	gs
	mask representing the evaluation subregion	evalmask	bw
	mask highlighting the segmented cell nuclei	segment	bw
	mask highlighting the convex hulls of the segmented cell nuclei	convhull	bw

Dataset structure

Image data are organized by tissue specimens (top-level folders) and tiles (second-level folders), the latter ones ordered by position. Top-level folders are named specimen_01, ... , specimen_44; second-level folders are named e.g. specimen_01_tile_01_01, ... , specimen_01_tile_09_08. Within each second-level folder, 19 image files in greyscale ("gs") or black-and-white ("bw") mode are stored in losslessly compressed .png format. Table 1 and Figure 1 summarize the different images available at a given tile. The filenames are built as specimen_xx_tile_yy_zz_channel_[CD14, CD163, Pax5, DAPI]_type_[original, cartoon, segment, convhull,

lymphoma with an image of a HE-stained slide available, while IHC stainings are missing at all. In this situation, the image dataset presented in this note constitutes a document of interest in itself.

We will outline the most important options for further use of the data. First, it allows for a detailed morphometrical investigation of the imaged macrophages and B-cells with respect to distribution of geometrical parameters as size, diameter, perimeter, etc., as well as to overall shape patterns. Second, the data may be used for validation, calibration and comparison of cell segmentation methods (manual, automated) and related software packages, making available a large reference dataset together with the output of a reference method as described in [1]. In particular, the original images as well as the segmentations presented here could be used for the generation

of a sufficiently large training set for automated macrophage detection by machine learning methods. Third, the data may be used for study of co-localization and clustering of macrophages and B-cells within lymphoma tissue and cancer microenvironment, employing appropriate methods of point-pattern statistics [12, 13]. Note that the segmentation masks provided here are well-suited for a further processing of the obtained features (e.g. extraction of barycenters, replacement of the features by circles or squares of equal size). Finally, the dataset enables a closer study of the double-stained macrophage subpopulation.

To illustrate the described reuse potential, we include a set of composite figures, each combining information from several separate images. Figure 2.A shows an original image at CD14 channel (greyscale, contrast enhanced by factor 3.5, inverted) with superimposition of the mask of the evaluation sub-region, as obtained from the DAPI channel (light blue), and the segmentation of the CD14-stained macrophages (olive green). Figure 2.B shows the same tile as imaged at the Pax5 channel (greyscale, inverted) with superimposition of the cell nuclei segmentation from DAPI channel (blue, convex hulls) and the segmentation of the CD163-stained macrophages (dark yellow). In Figure 2.C, for the same tile, both macrophage segmentations (olive green resp. dark yellow, convex hulls) are combined in order to reveal double-stained parts (light yellow). In Figure 2.D, we superimposed to Figure 2.C the segmentation of B-cells from the Pax5 channel (magenta resp. grey, convex hulls). Observe that some B-cells are positioned inside of macrophages, indicating that they are engulfed by the macrophages for phagocytosis. It is obvious that co-localization and clustering patterns as empirically noticeable here must be investigated on a sound base of statistical methodology.

Availability of supporting data

All image data are publicly accessible at the Leipzig Health Atlas (LHA) repository [14] and can be reached from the address [15]. Each top-level folder can be downloaded as .zip file and bears a separate identifier, e.g. <https://health-atlas.de/lha/7WT490WEG2-7> within the repository, see Table 2. Two folders with total size larger than 1 GB (Nos. 04 and 44) have been splitted into a pair of files.

Declarations

List of abbreviations

CD14: a monocyte receptor protein; CD163: a macrophage receptor protein; DAPI: 4',6-diamidino-2-phenylindole; DLBCL: diffuse large B-cell lymphoma; HE: Haematoxylin-Eosin; IHC: immunohistochemical(ly); LHA: Leipzig Health Atlas; Pax5: a B-cell lineage specific activator protein; ROF: Rudin-Osher-Fatemi.

Ethical Approval

Tissue usage is covered by a statement of the internal review board of the Medical Faculty of the Christian-Albrechts-University Kiel, Germany (No. 447/10).

Consent for publication

Not applicable.

Table 2. Datasets available at the Leipzig Health Atlas.

Name	Size (MB)	Identifier
specimen_01.zip	167	7WT490WEG2-7
specimen_02.zip	146	7WT4A21HRJ-0
specimen_03.zip	125	7WT4AVVFEX-0
specimen_04a.zip	644	7WT5EP6YKD-6
specimen_04b.zip	731	7WT5J3GEK2-8
specimen_05.zip	674	7WT5P9AE2F-3
specimen_06.zip	173	7WT5QAKG27-4
specimen_07.zip	422	7WT5TVU42M-3
specimen_08.zip	409	7WT5W48QCX-6
specimen_09.zip	291	7WT5XWDOGT-7
specimen_10.zip	380	7WT614A02V-9
specimen_11.zip	729	7WT64Y97CY-7
specimen_12.zip	372	7WT67E4C8A-0
specimen_13.zip	154	7WT68MRXMT-1
specimen_14.zip	290	7WT69YN7KV-3
specimen_15.zip	128	7WT6C66UQ0-8
specimen_16.zip	150	7WT6GH4HQ0-7
specimen_17.zip	377	7WT6JKPKWV-5
specimen_18.zip	168	7WTH7R24UK-4
specimen_19.zip	109	7WTHAMJ6K4-2
specimen_20.zip	432	7WTJQXW8X-6
specimen_21.zip	445	7WTJP21QET-7
specimen_22.zip	481	7WTJWJ3GY7-5
specimen_23.zip	343	7WTK8KRE71-9
specimen_24.zip	476	7WTKDDOPPG-7
specimen_25.zip	775	7WTKH27N4C-8
specimen_26.zip	658	7WTKVQ8W1-5
specimen_27.zip	141	7WTKNFV28U-8
specimen_28.zip	232	7WTKPNXD2T-6
specimen_29.zip	567	7WTKT4NFNW-5
specimen_30.zip	345	7WTXJ4XWAD-0
specimen_31.zip	224	7WTXMW3UT9-1
specimen_32.zip	303	7WTXNVQ3X6-8
specimen_33.zip	341	7WTXQACWNG-8
specimen_34.zip	488	7WTXWFTC37-6
specimen_35.zip	294	7WTXYM32XF-3
specimen_36.zip	232	7WTY1FY9M5-0
specimen_37.zip	580	7WTY3EF3AA-2
specimen_38.zip	539	7WTY57UPU6-8
specimen_39.zip	907	7WTY81JGEU-0
specimen_40.zip	396	7WTY9TGTX1-3
specimen_41.zip	436	7WTYCTQU5A-8
specimen_42.zip	118	7WTYEECKGR-9
specimen_43.zip	619	7WTYGX9DNJ-2
specimen_44a.zip	524	7WTYPGFOCK-2
specimen_44b.zip	466	7WTYTTQ9KUT-5

Competing Interests

The authors declare that they have no competing interests.

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Author's Contributions

MW performed the image processing and wrote the manuscript. SR performed the IHC staining and image generation. RH curated the large-size image datasets and managed the storage within the Leipzig Health Atlas repository. WK identified the cohort. UDB contributed to the Context, Methods and Reuse potential sections. All authors read and approved the final manuscript.

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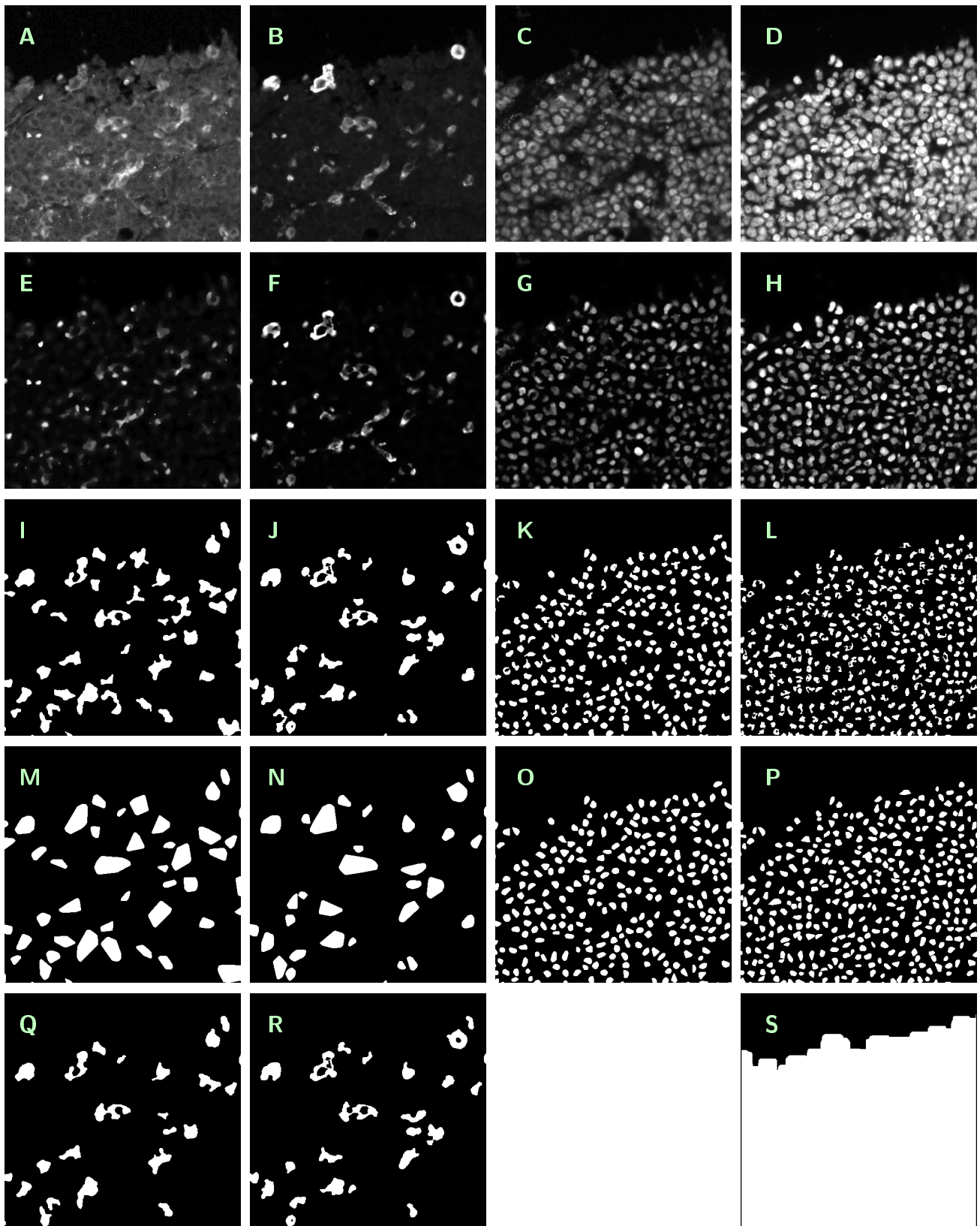


Figure 1. Summary of images available at a given tile (cutouts of 500×500 px size). Contrast enhanced in A by factor 3.5, in E by factor 7 and in F, G and H by factor 2.

Originals (A–D). A — specimen_02_tile_01_06_channel_CD14_type_original_mode_gs.png, B — specimen_02_tile_01_06_channel_CD163_type_original_mode_gs.png, C — specimen_02_tile_01_06_channel_Pax5_type_original_mode_gs.png, D — specimen_02_tile_01_06_channel_DAPI_type_original_mode_gs.png,

Cartoons (E–H). E — specimen_02_tile_01_06_channel_CD14_type_cartoon_mode_gs.png, F — specimen_02_tile_01_06_channel_CD163_type_cartoon_mode_gs.png, G — specimen_02_tile_01_06_channel_Pax5_type_cartoon_mode_gs.png, H — specimen_02_tile_01_06_channel_DAPI_type_cartoon_mode_gs.png,

Segmentations (I–L). I — specimen_02_tile_01_06_channel_CD14_type_segment_mode_bw.png, J — specimen_02_tile_01_06_channel_CD163_type_segment_mode_bw.png, K — specimen_02_tile_01_06_channel_Pax5_type_segment_mode_bw.png, L — specimen_02_tile_01_06_channel_DAPI_type_segment_mode_bw.png,

Convex hulls (M–P). M — specimen_02_tile_01_06_channel_CD14_type_convhull_mode_bw.png, N — specimen_02_tile_01_06_channel_CD163_type_convhull_mode_bw.png, O — specimen_02_tile_01_06_channel_Pax5_type_convhull_mode_bw.png, P — specimen_02_tile_01_06_channel_DAPI_type_convhull_mode_bw.png,

Various (Q–S). Q — specimen_02_tile_01_06_channel_CD14_type_multiple_mode_bw.png, R — specimen_02_tile_01_06_channel_CD163_type_multiple_mode_bw.png, S — specimen_02_tile_01_06_channel_DAPI_type_evalmask_mode_bw.png.

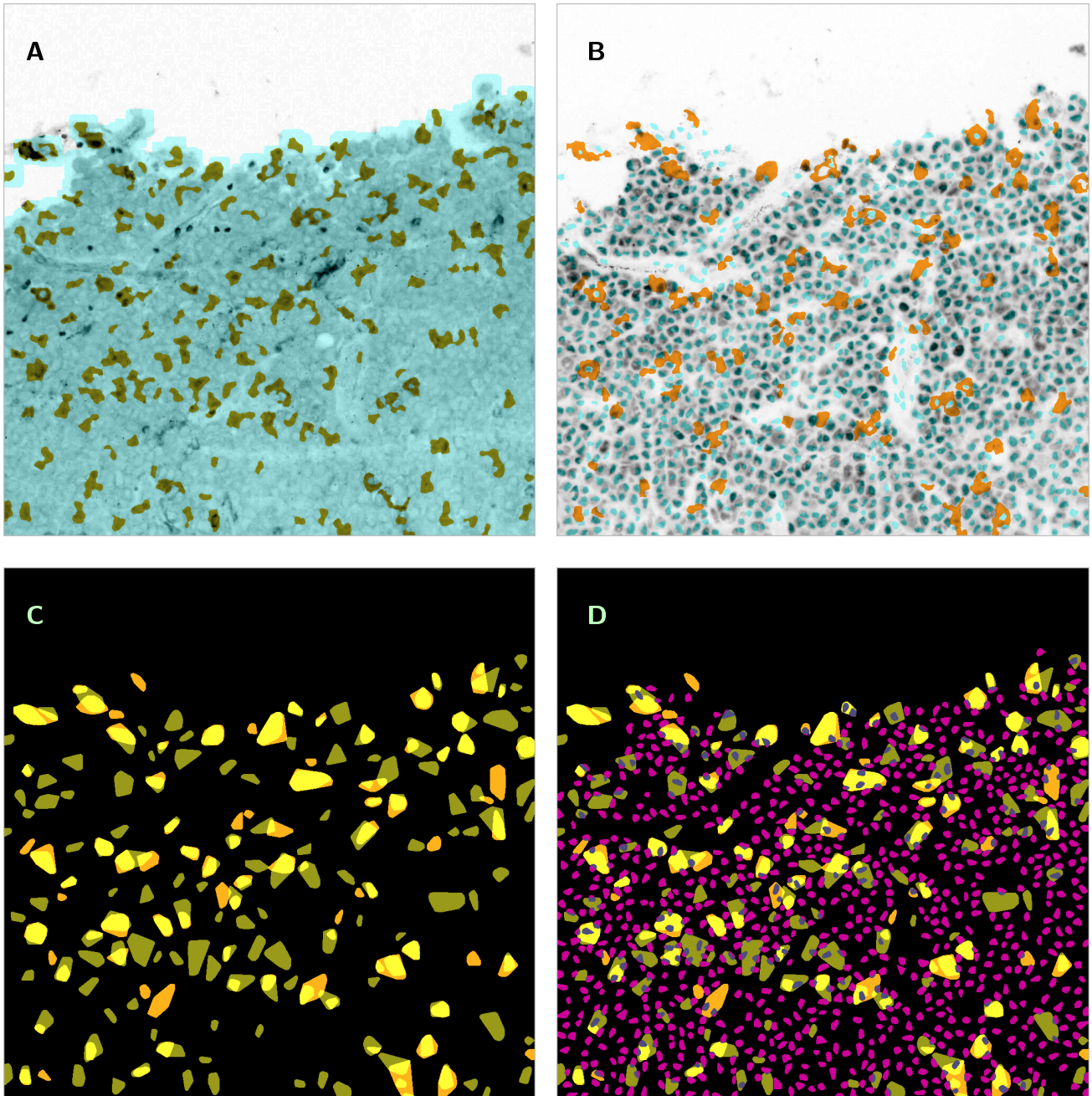
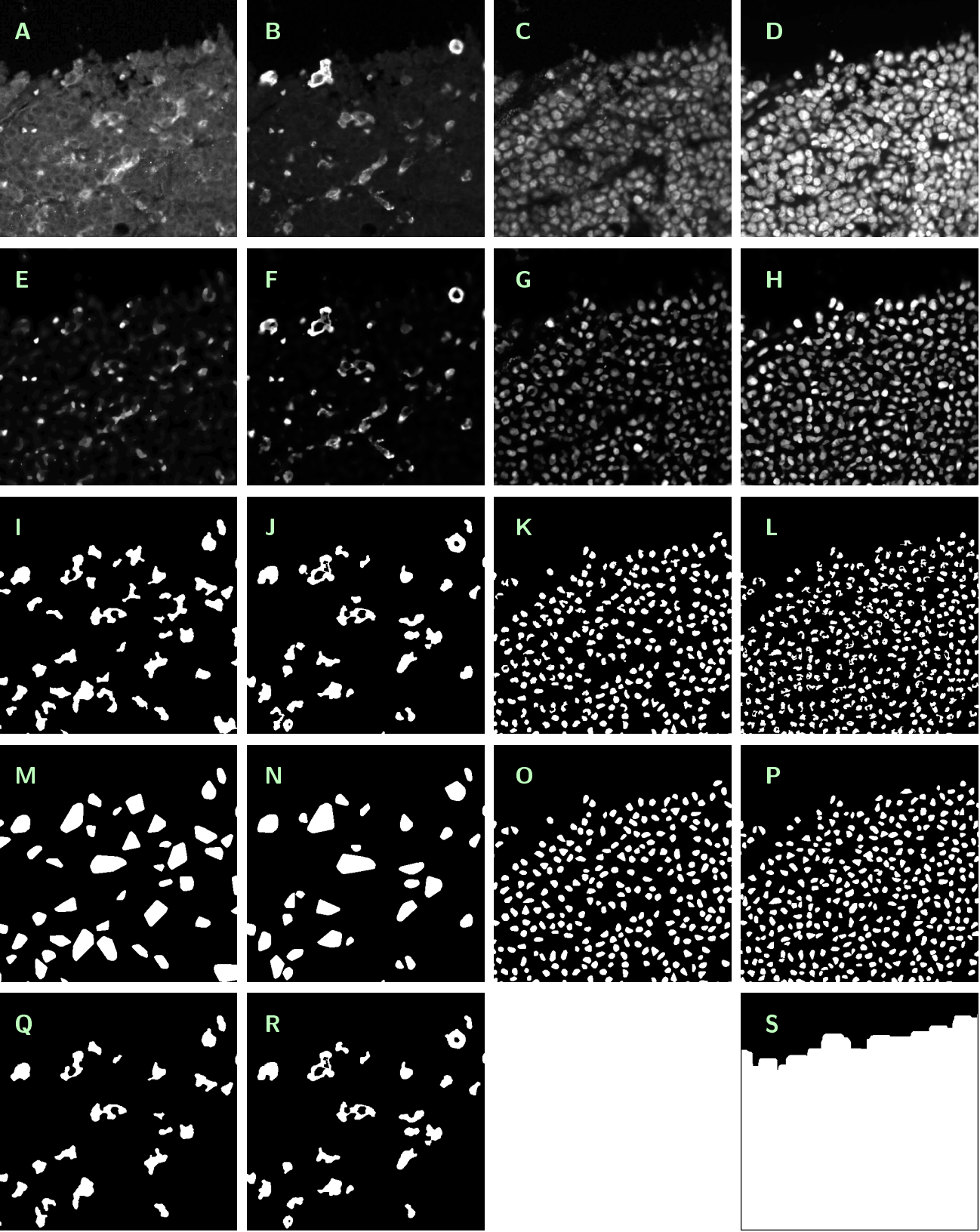
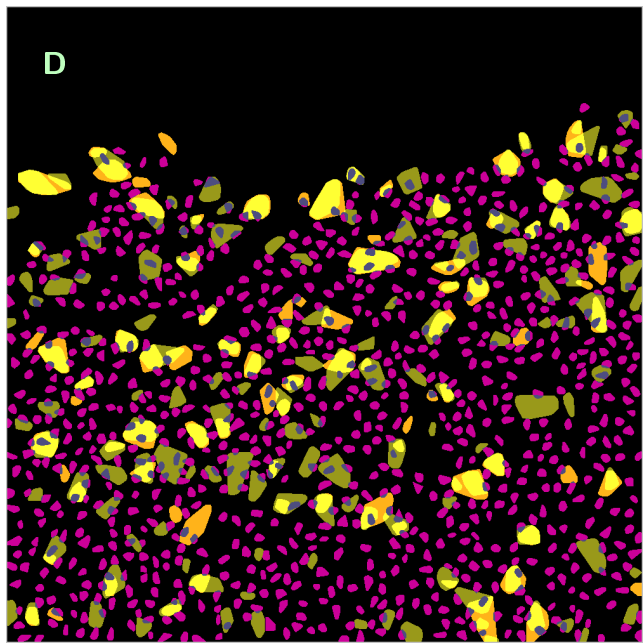
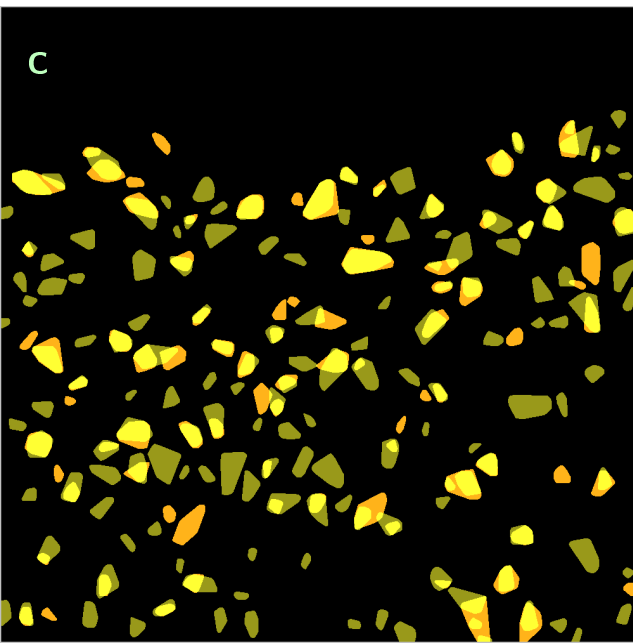
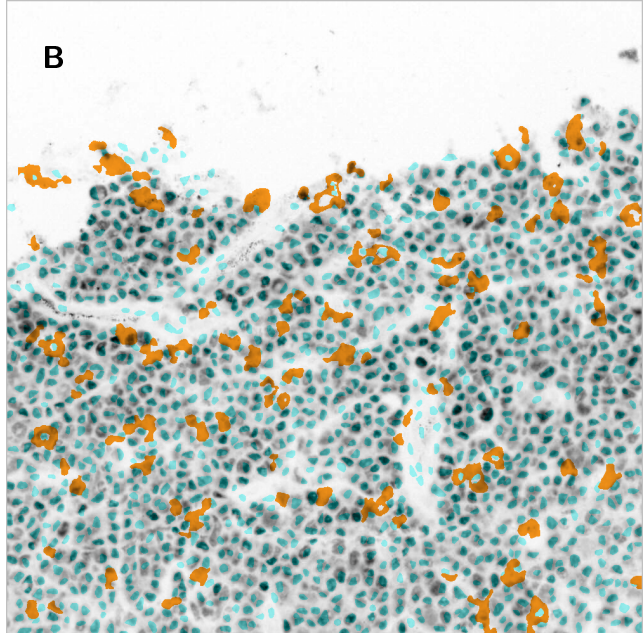
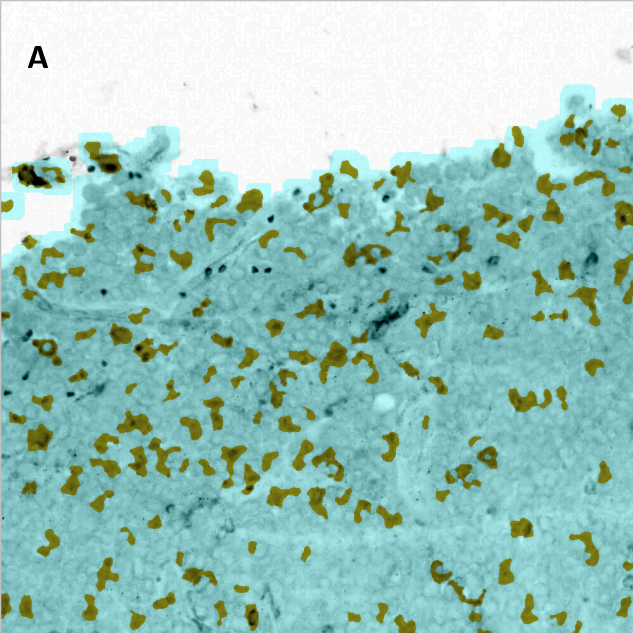


Figure 2. Examples of combined information from several images (see detailed description above). Image size is 1000×1000 px ($450 \times 450 \mu\text{m}^2$). Composed images are based on specimen_02_tile_01_06_ ...

A — channel_CD14_type_original_mode_gs.png / channel_DAPI_type_evalmask_mode_bw.png / channel_CD14_type_segment_mode_bw.png,
 B — channel_Pax5_type_original_mode_gs.png / channel_DAPI_type_convhull_mode_bw.png / channel_CD163_type_segment_mode_bw.png,
 C — channel_CD14_type_convhull_mode_bw.png / channel_CD163_type_convhull_mode_bw.png,
 D — channel_CD14_type_convhull_mode_bw.png / channel_CD163_type_convhull_mode_bw.png / channel_Pax5_type_convhull_mode_bw.png.





Leipzig, September 04, 2019

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*GigaScience
Editor-in-Chief
Laurie Goodman, PhD*

Dear Dr. Goodman,

hereby we submit our manuscript

“An image dataset related to automated macrophage detection in immunostained lymphoma tissue samples”

for publication as a Data Note in GigaScience. In our view, the note fits well into the aims and scope of the journal, making a well-curated image dataset with considerable reuse potential publicly findable, accessible und usable. Data presented here were generated as a part of an accompanying publication, which has recently appeared. They are hosted in the Leipzig Health Atlas repository and published there under CC0 license.

*With best regards,
on behalf of the authors*

Marcus Wagner.