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Subject Section

Gating Mass Cytometry Data by Deep Learning-**Supplementary Materials**

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Abstract

Motivation: Mass cytometry or CyTOF is an emerging technology for high-dimensional multiparameter single cell analysis that overcomes many limitations of fluorescence-based flow cytometry. New methods for analyzing CyTOF data attempt to improve automation, scalability, performance, and interpretation of data generated in large studies. Assigning individual cells into discrete groups of cell types (gating) involves time-consuming sequential manual steps, untenable for larger studies.

Results: We introduce DeepCyTOF, a standardization approach for gating, based on deep learning techniques. DeepCyTOF requires labeled cells from only a single sample. It is based on domain adaptation principles and is a generalization of previous work that allows us to calibrate between a target distribution and a source distribution in an unsupervised manner. We show that DeepCyTOF is highly concordant (98%) with cell classification obtained by individual manual gating of each sample when applied to a collection of 16 biological replicates of primary immune blood cells, even when measured across several instruments. Further, DeepCyTOF achieves very high accuracy on the semi-automated gating challenge of the FlowCAP-I competition as well as two CyTOF datasets generated from primary immune blood cells: (i) 14 subjects with a history of infection with West Nile virus (WNV), (ii) 34 healthy subjects of different ages. We conclude that deep learning in general, and DeepCyTOF specifically, offers a powerful computational approach for semi-automated gating of CyTOF and flow cytometry data.

Availability: our codes and data are publicly available at https://github.com/KlugerLab/ deepcytof.git.

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Supplementary information: Supplementary data are available at Bioinformatics online.

1 Data Pre-processing

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Given a blood samples A, we first perform an elementary logarithmic transformation $A_{i,j} \leftarrow \log(1 + A_{i,j})$

In addition, we standardize each column of A to have zero mean and and unit variance.

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Fig. S1: Third hidden layer representation of a blood sample (the unlabeled cells are omitted), obtained from DeepCyTOF without a calibration step. Each color corresponds to a cell type. Different cell types are concentrated in different regions of the code space.





Fig. S3: Confusion matrix for visualizing the performance of DeepCyTOF when applied to sample 15 without (top) and with (bottom) a calibration step (the unlabeled cells are omitted). The rows represent the actual cell type label and the columns represent the predicted cell type label. The F-measure associated with the top panel (0.8857) is significantly lower than the F-measure associated with the bottom panel (0.9609).



Fig. S2: MMD between the source (sample 2) and each of the of other 15 samples in the multi-center dataset before (white) and after (black) calibration. The MMD values were computed based on random batches of size 1000 from each sample.

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Fig. S4: t-SNE plot of CD8+ T cell from sample 15 before (top) and after (bottom) with DeepCyTOF.

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Fig. S5: MMD between the reference sample and each of the remaining 15 samples in the multi-institute cytometry dataset. White: without denoising. Black: with denoising.

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