Image-based cell phenotyping with deep learning

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Images can be quantified for all kinds of phenotypes

Muscle structure



David Thomas

Patient biopsy tissue



Margaret Shipp/Scott Rodig

Image Mass Spec



Michael Angelo

3D

Muscle structure

Control human iPS

Isogenic Duchenne-like iPS



Olivier Pourquie

Allen Institute for Cell Science



Screen for specific phenotypes using images

Treatment for AMKL (leukemia)



DNA stain with outlines identifying the nuclei

Clinical trials underway for Alisertib in adults with AMKL. Wen Q, et al. (2012). Cell 150(3):575-89





What is image-based profiling?



Caicedo J.C., Singh S., Carpenter A. "Applications of Image-Based Profiling of Perturbations". Current Opinion in Biotechnology - 2016.

ML for image-based profiling - Overview

- 1. Cell segmentation
- 2. Single-cell representation learning

1. Cell segmentation

Diversity of cell segmentation problems





Cell segmentation as face detectors



Dataset and challenge



Data Science Bowl — Organization

Create annotated dataset







Training models for segmentation



Diversity of models for image segmentation



U-Nets

Feature Pyramid Nets

Mask RCNN

A single model improves accuracy and reduces time





Caicedo, et al. 2019 Nature Methods

Existing tools for generic cell segmentation



Stringer et al. 2020, Nat Meth

U-Nets



Hollandi et al. 2020, Cell Systems

Mask RCNN



Greenwald et al. 2021, BioRxiv

Feature Pyramid Nets

Open challenges

- Collecting larger publicly available datasets
- Learning from few examples with active learning

2. Single cell representation learning

Representation learning for image-based profiling

1. Raw images



2. Segmented images



3. Single-cell feature matrices



4. Population profiles of treatments



5. Downstream statistical analysis



Classical approach to measuring cell morphology

Engineer measurements Define and compute useful properties



Perturbation experiments



Weakly supervised learning of single-cell feature embeddings



Main goal: Treatment-level profiling



Evaluating Image-Based Profiling



13 data points

78 pairwise connections

Learned representations improve profiling performance

Number of folds of enrichment for top connections to have the same MOA/pathway vs. rest of the connections



% top connections

Deep learning

Classical features





Determining variant impact







EGFR Wild Type









EGFR Mutant



Determining variant impact



Correcting for batch effects



Correcting for batch effects — Gradient Reversal Layer



Minimizing batch effect confounders

0 CTRL • 1 DS1 . 0 DGK 1 DS2 10 10 0 FIG4 1 DS3 0_GDAP1 1_DS4 0 GUK1 1 DS5 8 8 0 KIF1A 1 DS6 0 MFN2 1 DS7 6 6 0 RRM1 1_DS8 0 VCP 1 DS9 4 4 2 2 0 0 -2 -2 -4 0.0 2.5 5.0 7.5 10.0 12.5 15.0 0.0 2.5 5.0 7.5 10.0 12.5 15.0 ResNet18GRL - validation 0 CTRL •

0 DGK 10 0 FIG4 0 GDAP1 0 GUK1 8 0 KIF1A 0 MFN2 6 0 RRM1 0 VCP 4 2 0 -2 -4 0.0 2.5 5.0 7.5 10.0 12.5 15.0



ResNet18GRL - training

Open challenges

- Domain adaptation and batch effect correction
- Explainable models and interpretation capabilities

Image-based Profiling

Extracting information from biomedical images

1. Raw images



1. Raw diamonds

2. Single cells



- 2. Pieces of diamond

3. Feature extraction



3. Diamond polishing

4. Aggregated profiles





4. Jewelry

Thank you!