

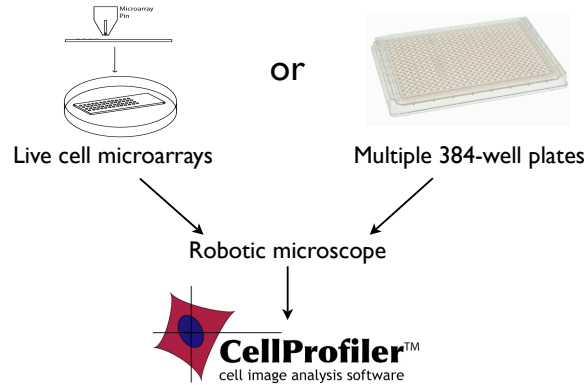
# Methods for High-Content, High-Throughput Image-Based Cell Screening



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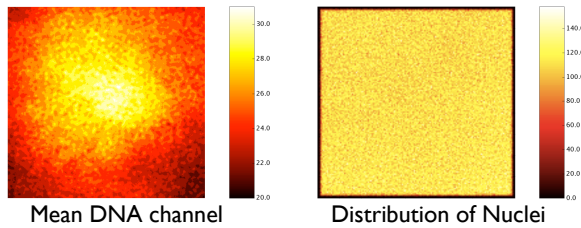
Our goal is to make meaningful biological predictions from large numbers of images of cells grown under different conditions (e.g., RNAi gene knockdowns).



## Processing steps:

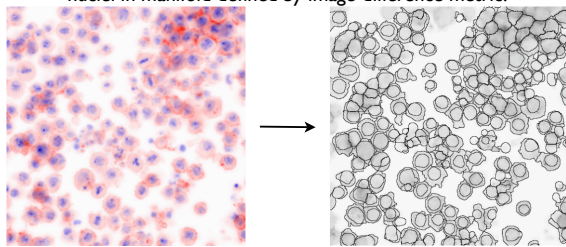
### a) Illumination correction

Estimate mean cell and stain distributions to find and correct illumination variation.



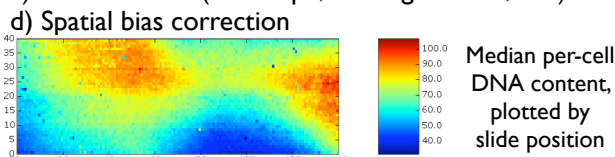
### b) Segmentation

Identify nuclei, then cell boundaries as Voronoi diagram of nuclei in manifold defined by image-difference metric.

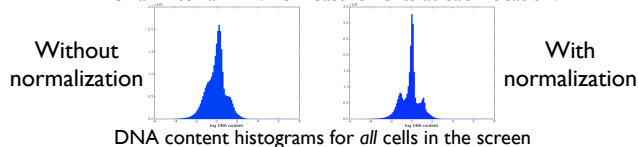


Jones et al, *Voronoi-Based Segmentation of Cells on Image Manifolds*, CVBIA, 2005

### c) Measurement (cell shape, staining texture, etc.)



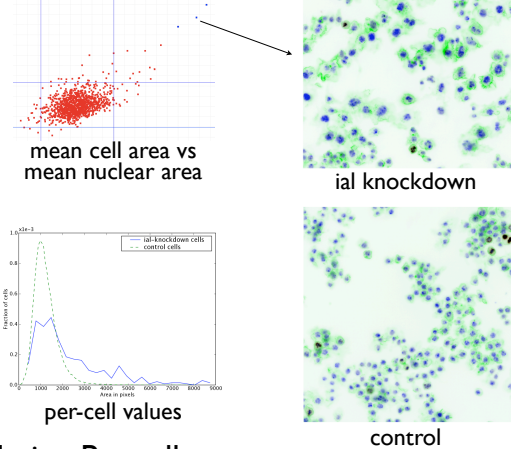
Median filter and divide measurements at each location:



DNA content histograms for all cells in the screen

## Analysis - Per-image

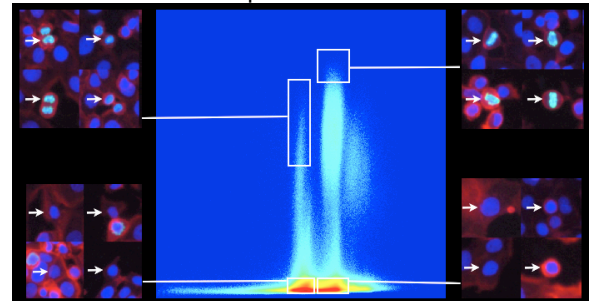
Dramatic changes in phenotype for a large number of cells can be detected using per-image measurements.



## Analysis - Per-cell

Some phenotypes can be detected per-cell using a small set of measurements. Per-cell analysis is necessary because such phenotypes are often present in only a small fraction of cells per-image.

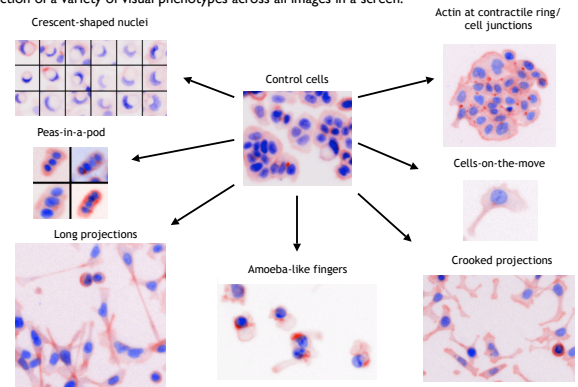
### Mitotic phase screen



DNA vs phospho-histone H3 (mitosis marker)

## Analysis - Automatic Classifier, per-cell

Other phenotypes require a complex combination of multiple features for identification. Machine learning, with training sets interactively created by biologists, allows automatic detection of a variety of visual phenotypes across all images in a screen.



Morphology screen in human cells

Per-cell approach gives better statistical power and more accurate predictions.

Biological verification and followup in progress.

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