

# Image Analysis - Part2

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## Image Segmentation and Clustering

### Overview

The program takes a batch of images as input and runs a given set of operations on them (Edge Detection, Histogram Clustering, Histogram Thresholding, Dilation, and Erosion). All of these operations output a new image for each image in the batch. A TOML file in the source code's root directory is used to configure the operations, such as the strength of noise and the weights for filters. All critical functions are implemented from scratch except for array operations that use the third-party mathematics library **NumPy**.

Below operations are performed

- Edge Detection - Canny Edge Detection
- Image Dialtion
- Image Erosion
- Image Segmentation
  - Clustering using k-means
  - Histogram Thresholding
  - Watershed Segmentation Algorithm

### Usage

```
git clone https://github.com/jonaylor89/image-segmentation.git
cd image-segmentation
pip3 install --user pipenv
pipenv install
pipenv run python segmentation.py
```

### Implementation

The programming language of choice for this project was python. The high-level reason for deciding to write everything in python was that it gives the ability to rapidly develop and integrate each operation and python's **numpy** library, which allows for idiomatic and fast array operations. Python's fairly obvious downside is its speed. To mitigate the problem of speed for the image operations, **numba**, a third-party python library used for mathematics, is being used. **numba** has a python function decorator for just-in-time compiling functions to machine code before executing. Using this decorator on functions that use heavy math and looping (i.g. convolving) provides major speed increases with speeds similar to using lower-level compiled languages like C/C++ or Rust. Compilation time effects the first image in the batch, but every image after that uses the precompiled machine code. Image batches, rather than being operated on synchronously, are thrown into a process pool where a preconfigured number of worker processes pulls images off a queue and runs them through the operation pipelines.

### Full Third-Party Dependency List

```
# Pipefile
[packages]
pillow = "*"      # reading and writing images
numpy = "*"       # fast array operations
*click = "*"      # command line interface utility
numba = "*"       # just-in-time compiler for operations
toml = "*"        # reading configuration file
tqdm = "*"        # progress bar
```

*These can be found in the Pipfile at the root of the source code*

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## Results

### Input

The input is specified in `toml` defaulted to `dataset\input` folder. There are seven default images from each class of image in input folder

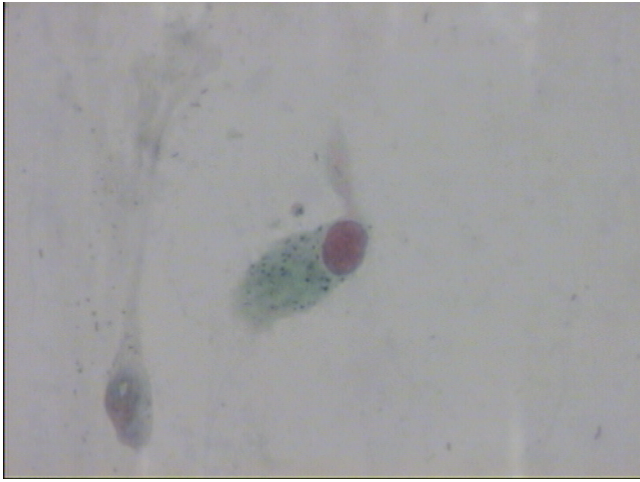
- cyl01
- inter01
- let01
- mod01
- para01
- super01
- svar01

### output

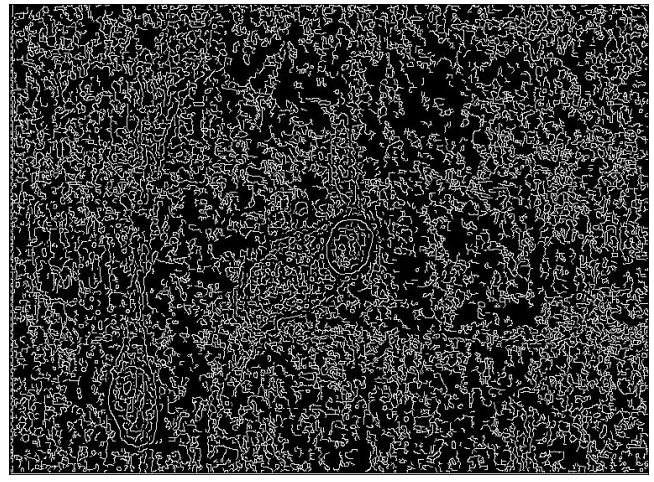
The results are in the output directory specified in the `toml` configuration file. In the output directory, there are the grayscale images from each operation with the file name. `{operation}_{original image}.BMP` (e.g. `edges_svar53.BMP`).

## EDGE DETECTION

```
def cannyEdgeDetection(img_array: np.array) -> np.array
```



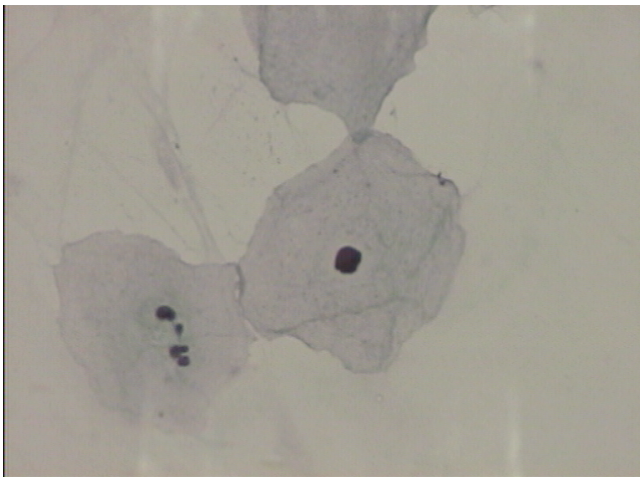
Original



Edge Detection

## DILATION

```
def dilate(img_array: np.array, win: int = 1) -> np.array
```



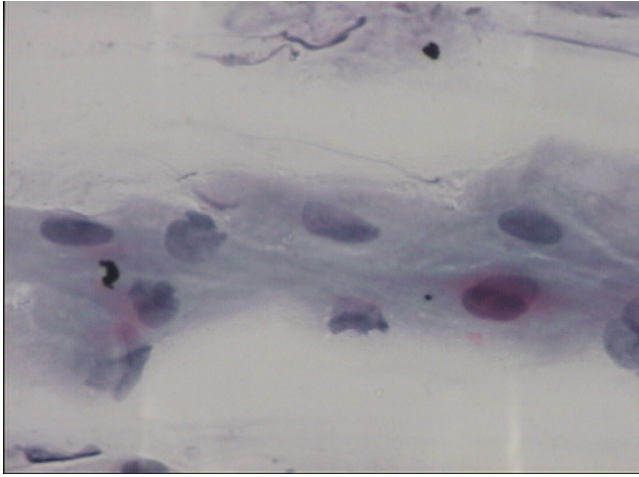
Original



Dialtion

## Erosion

```
def erode(img_array: np.array, win: int = 1) -> np.array
```



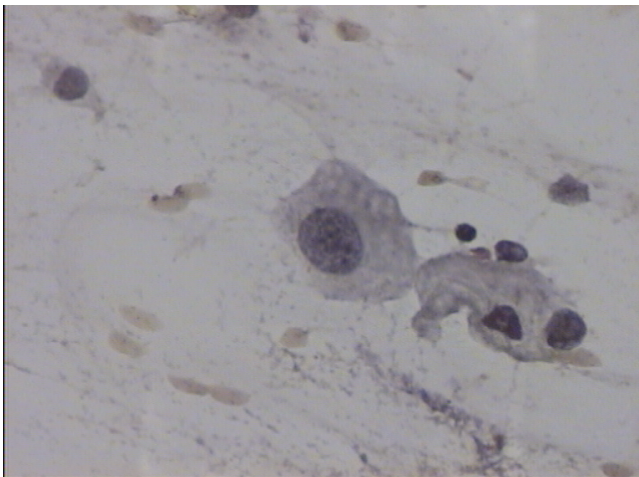
Original



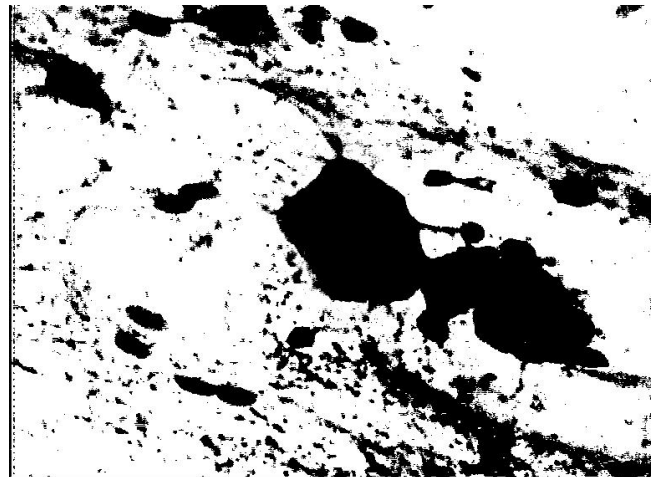
Erosion

### Segmentation - Histogram Thresholding

```
def histogramThresholding(img_array: np.array) -> np.array
```



Original

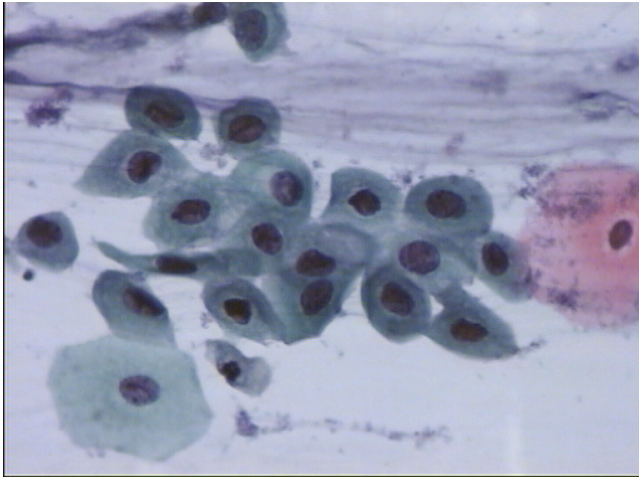


Segmentation - Histogram Thresholding

### Segmentation - Clustering (k-means)

```
def histogramClustering(img_array: np.array) -> np.array
```





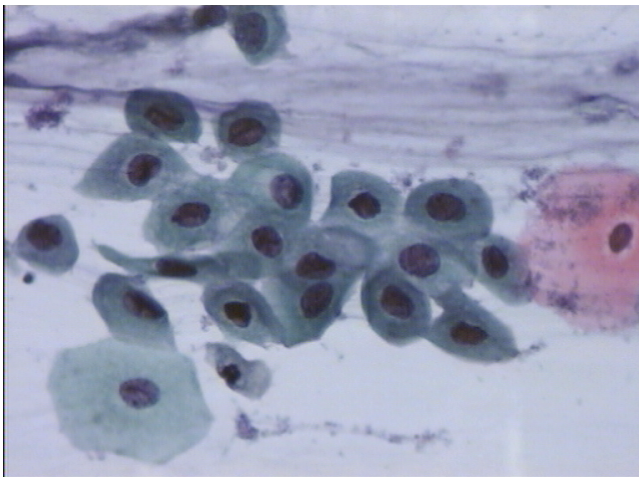
Original



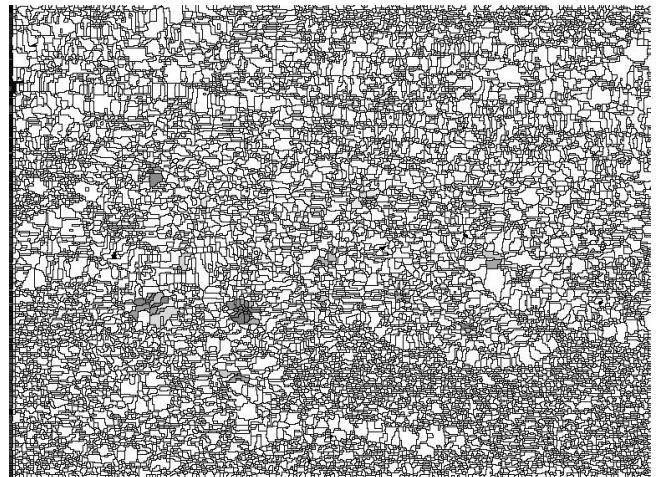
Clustering (k-means)

### Additional Algorithm - Watershed Segmentation

```
class Watershed(object)
```



Original



Watershed Segmentation