

Efficient Graph-based Algorithm applied to Cell Detection

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Abstract

While object detection has impacted several industries, much remains unanswered in cancer cell detection using Pap smear microscopy. This project introduces a machine learning algorithm that detects and crops regions of interest (ROI) before scrutinizing cells for abnormal traits. Using digital images of cervical cells, we automatically detect ROI's following 3 main steps: (I) non-linear filters applied to RGB micrographs; (II) graph-based clustering using Felsenszwalb's[1]; (III) definition of super-pixels with Isodata binary classification. This process yielded the identification of the ROI with a precision average of 92% and a recall average of 95%. The resulting algorithm is at the core of the driver that controls the motorized stage of our future smart microscope that will enable scanning of full glass slides.

Cervical Cancer Declining Rate of Incidence & Mortality Death rate has been a second of the control of the cervic. It is usually a slow-growing cancer that may not have aymptoms but can be found with regular Pap tests. The FPV vaccine protects against HPV-16 and HPV-18 which cause about 70% of cervical cancers worldwide. The year survival rate is close to 75.0 for variety of 55.0 for variety of 55.0

Why Cell Detection?



Fig 2. Leica Microscope DMRX with Motorized stage. [3]

From the available
15,000 fields of a Pap
smear slide coverage
area, less than 100 fields
are manually inspected.
Our region proposal
algorithm aims to
automatically detect
ROIs with clinical value
from the whole slide
while speeding up the
work of cytopathologist.

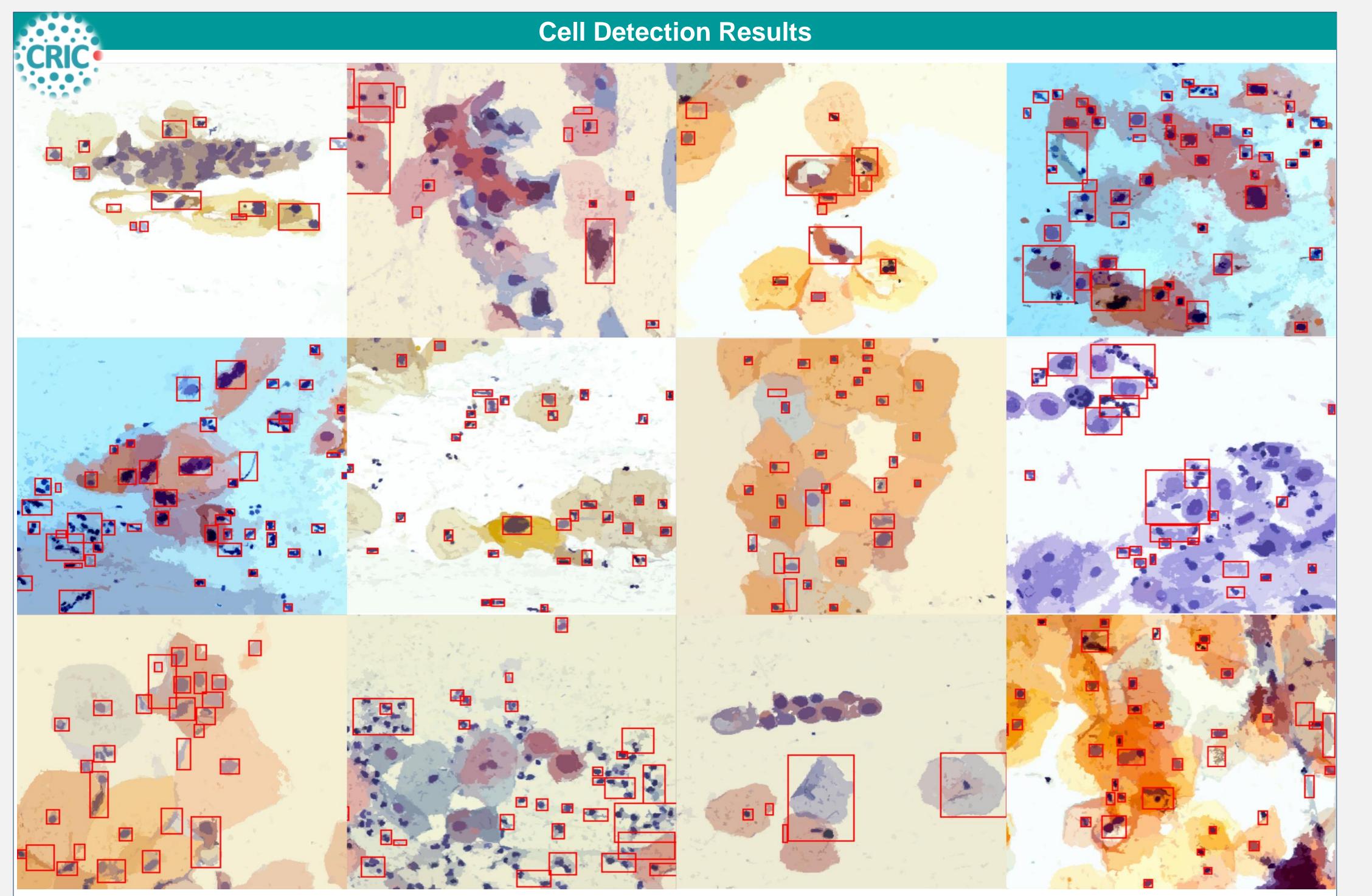
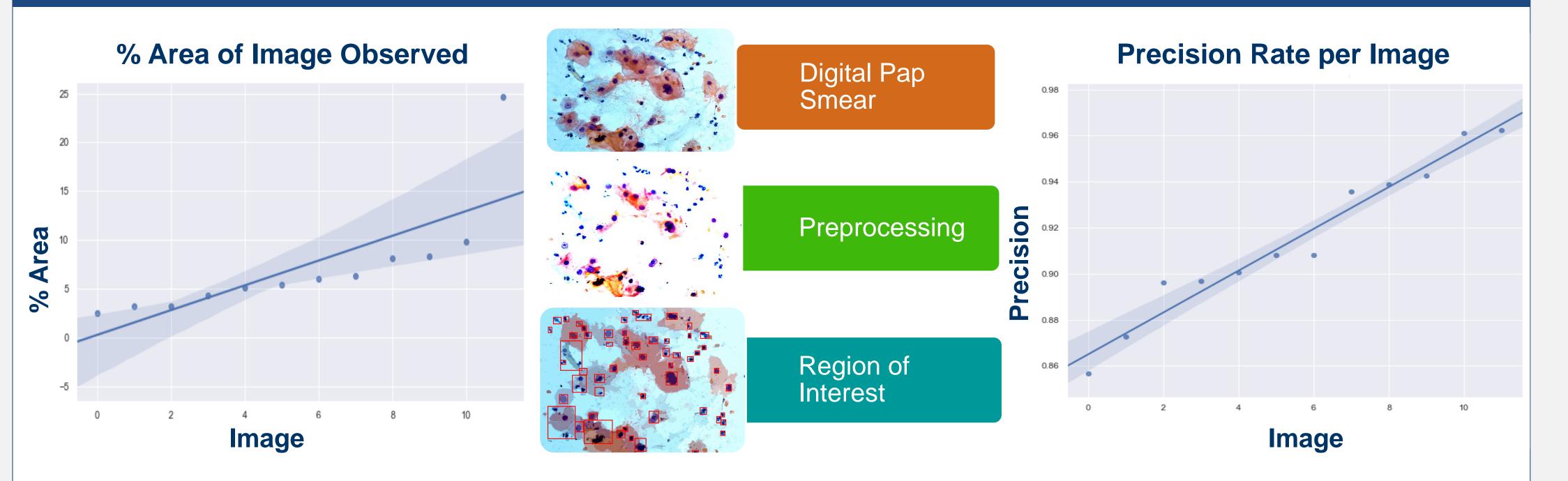


Fig 3. Region proposal from CRIC digital Pap smear images at 40X.

Methodology and Results



Figs. 4, 5, 6. Our method begins with the preprocessing of the CRIC images to later apply *Felzenszwalb's* unsupervised machine learning algorithm and obtain the ROIs where a cancer cell may be located. We evaluate our results using the Intersection over Union method between the calculated binary image mask of the ROI and the ground-truth. We also determined the %Area of a field after the location of the ROI's with clinical value.

Conclusions and Future Work

This work presented a proof-of-concept of the driver that will steer high-resolution image acquisition based on the detection of ROIs with clinical value.

Smart Microscope: No Cell left Behind

Fig 7. 40X-2500X LED Digital
Binocular Compound Microscope
with 3D Stage + USB Camera. [4].

Stage to be motorized and controlled to XY location of ROI

Automated Objective to higher resolutions for further inspection of abnormal cells

Fig 8. 8MTF - Motorized XY Scanning Stage. [5]

Computer Science:

Driven to detect individual cells and cell clusters with high clinical value.

Driven to control the stage based on spatial coordinates retrieved.

Control objectives to acquire high-resolution image on-demand



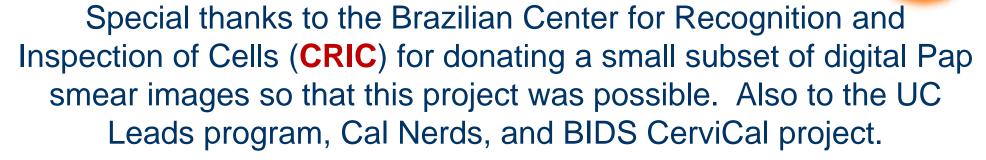
Engineering of Microscope:

Motorized stage for full glass slide.

Eyepiece Camera

Autofocus

Acknowledgements



Citations

[1] Efficient graph-based image segmentation, Felzenszwalb, P.F. and Huttenlocher, D.P. International Journal of Computer Vision, 2004.

[2] Figure 1: Infograph on Cervical Cancer: "Foundation Debuts Cervical Cancer Prevention Infographic." Infographic." Prevent Cancer Foundation Community Grants Comments.

[3] Figure 2. Leica Microscope DMRX with Motorized Stage • \$9,950.00." *PicClick*,
[4] Figure 7: 40X-2500X LED Digital Binocular Compound Microscope with 3D Stage + USB Camera." *AmScope*

Microscope
[5] Figure 8. Motorized Stage Adjustment Screws - Optical Positioners - Catalog - Opto-Mechanical Products – Standa.



