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# Automated Visual Evaluation (AVE) For Cervical Cancer Screening and Its Challenges

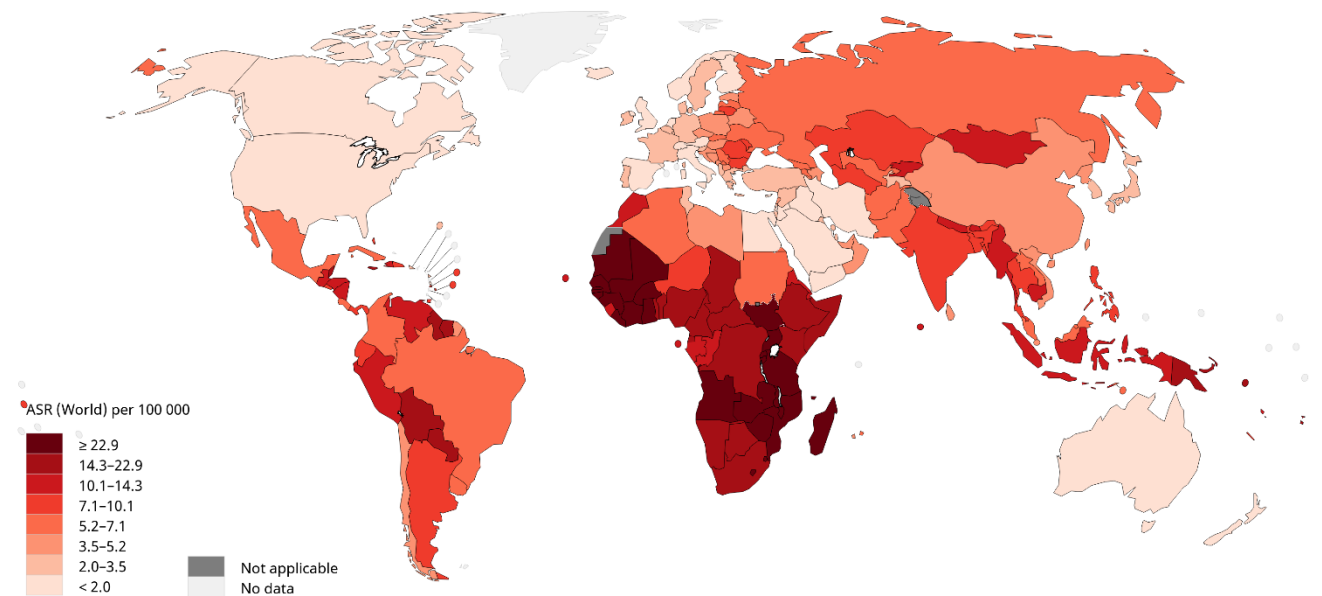
Zhiyun Xue



# Cervical Cancer

- The 4th highest cause of cancer mortality in women
  - 570 000 new cases and 310 000 deaths in 2018
- Especially prevalent in lower resource regions
  - 90% deaths and 85% occurring in low and middle-income countries

Estimated age-standardized mortality rates (World) in 2018, cervix uteri, all ages



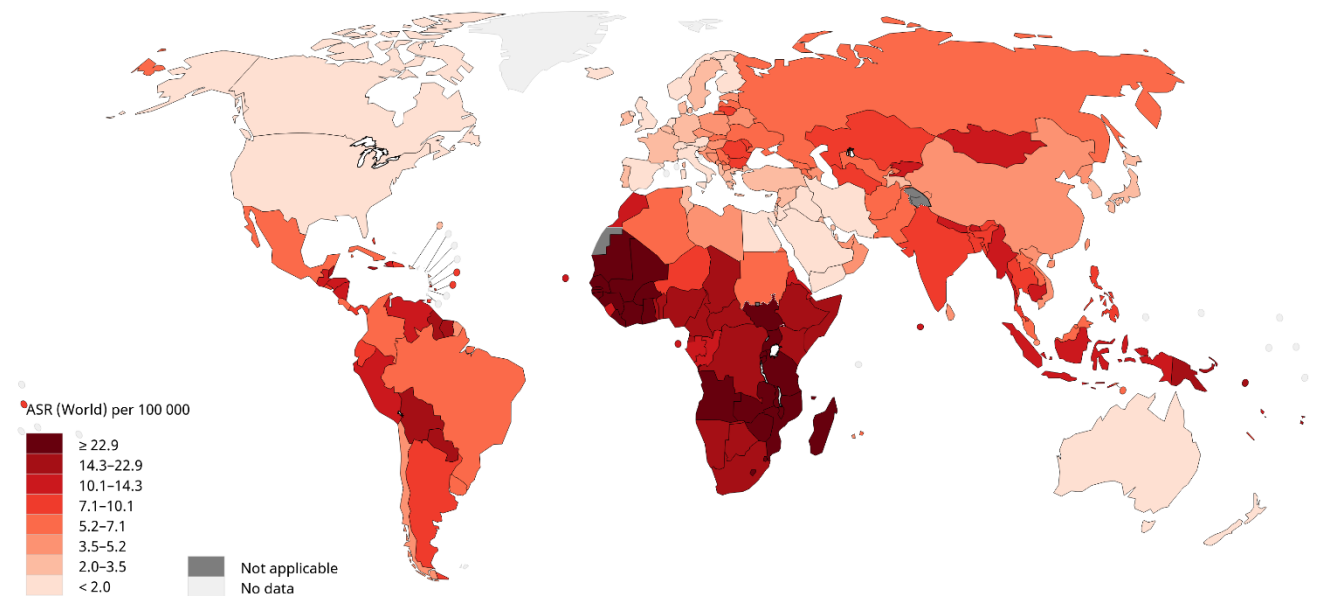
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Data source: GLOBOCAN 2018  
Graph production: IARC  
(<http://gco.iarc.fr/today>)  
World Health Organization

# Cervical Cancer

- Caused by persistent infection of certain types of Human Papillomavirus (HPV)
  - Two HPV types (16 and 18) cause 70% of cervical cancers and pre-cancerous cervical lesions
  - HPV infection often goes away on its own
  - HPV infection can be preventable with vaccination

Estimated age-standardized mortality rates (World) in 2018, cervix uteri, all ages

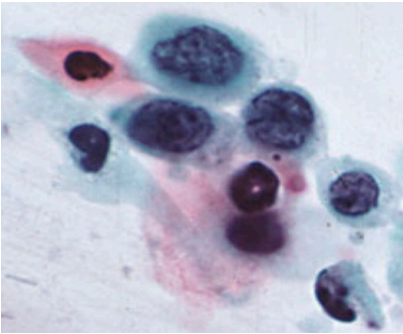


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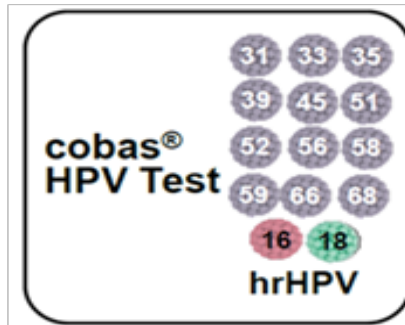
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# Cervical Cancer Screening

## Cytology



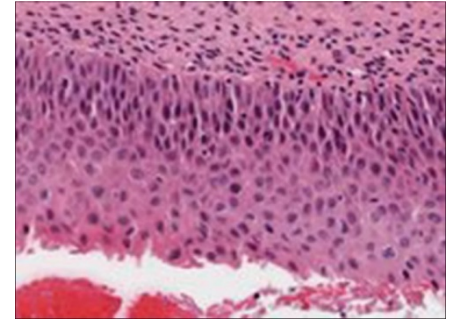
## Molecular



## Visual



## Histology



Data Collection

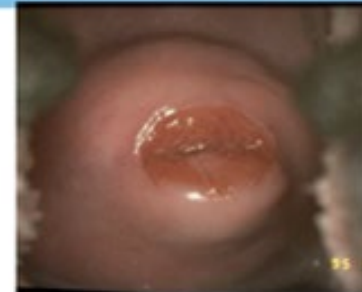
Data Hosting

Analysis

# Cervical Cancer Screening – Visual Inspection with Acetic Acid (VIA)

## What Is VIA?

- Use bright white light to visualize cervix with unaided eye
- Clean cervix with dilute 3-5% acetic acid solution (white vinegar)
- Wait at least one minute
- Abnormal tissue temporarily appears white (acetowhite)
- Get *immediate* results
- Promotes linkage of screening with treatment



NEGATIVE



POSITIVE

# Cervical Cancer Screening – Automated Visual Evaluation (AVE)

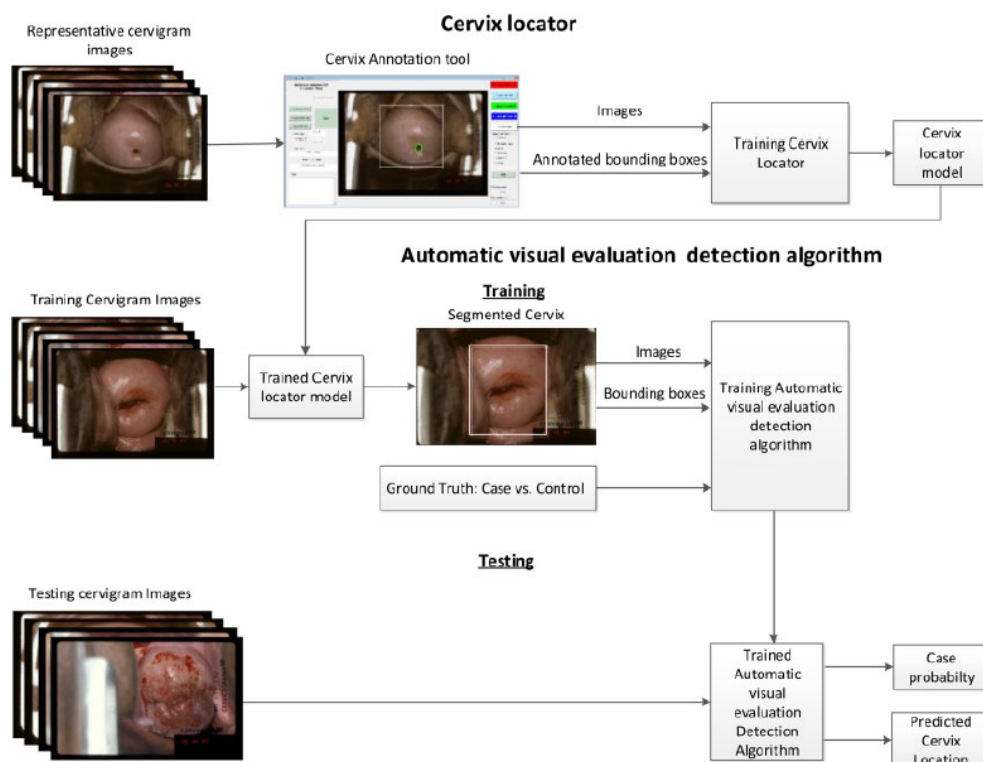
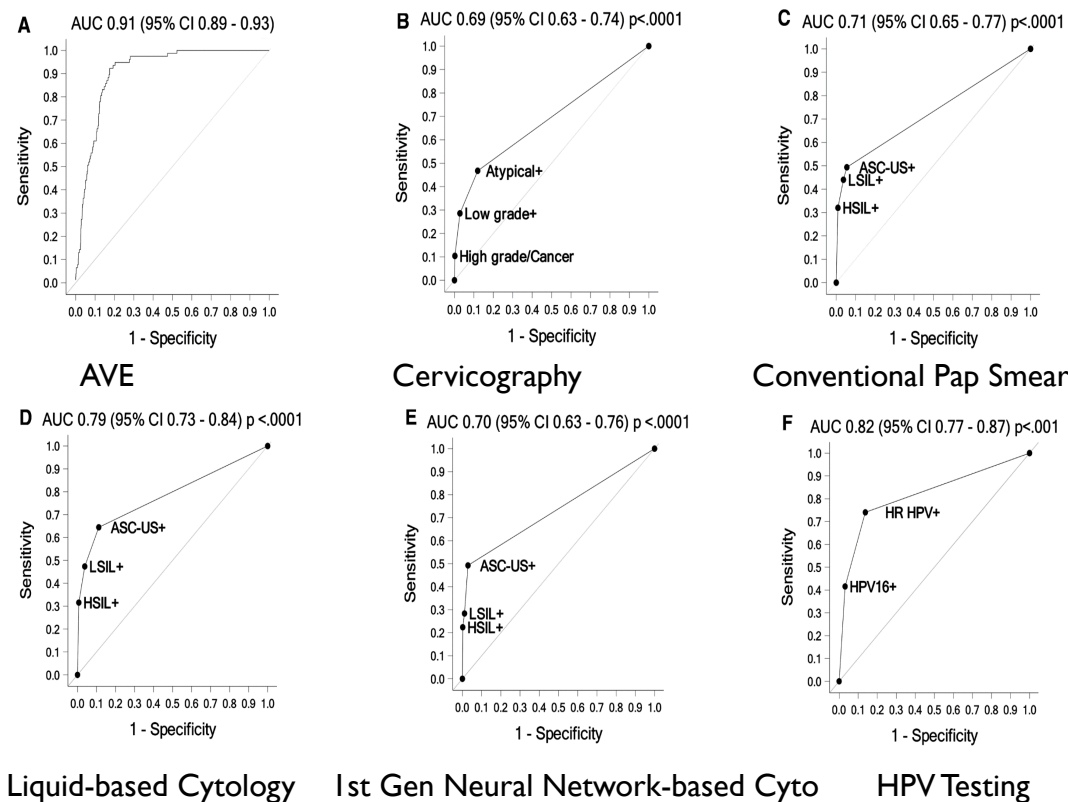


Figure 2. The system architecture of the automated visual evaluation algorithm. Two models are trained: a cervix locator (top), and the automated visual evaluation detection algorithm (bottom). The final validation algorithm incorporated both cervix locator and automated visual evaluation.

- Population-based natural history study of cervical neoplasia - Guanacaste
  - 7 year (1993 to 2001) longitudinal study conducted by NCI
  - 9400+ women
  - 60,000 cervical images (cervigrams) collected
  - Cytology, HPV, and histology results

# Cervical Cancer Screening – Automated Visual Evaluation (AVE)



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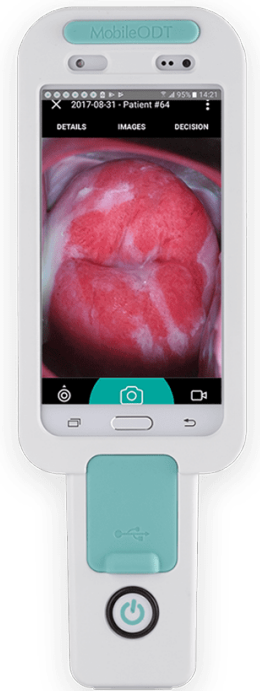
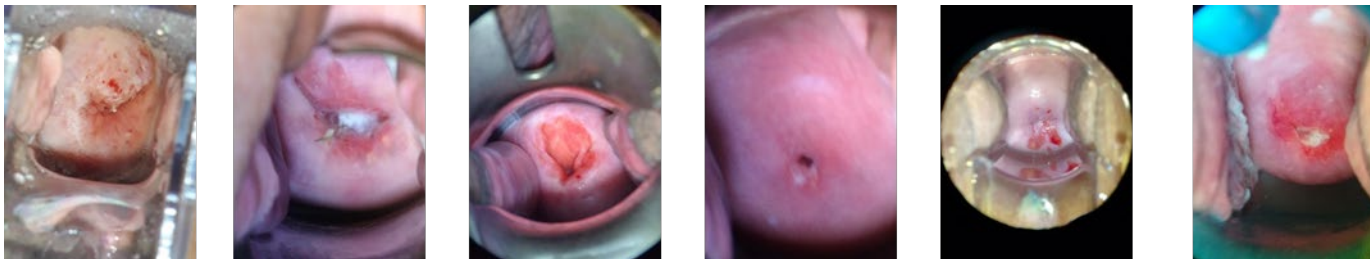


**Hu L, Bell D, Antani S, Xue Z, ..., Schiffman M. An Observational Study of Deep Learning and Automated Evaluation of Cervical Images for Cancer Screening, JNCI: Journal of the National Cancer Institute, Jan. 2019**



# Cervical Cancer Screening – AVE using Smartphones

- MobileODT image dataset
  - Collected from clinical sites around the world using MobileODT's EVA system
  - Larger variance in image quality and many were for documentation
  - Each woman had only one visit, but multiple images



**Xue Z, Novetsky AP, ..., Schiffman M, Antani S.** A demonstration of automated visual evaluation of cervical images taken with a smartphone camera, IJC: International Journal of Cancer, April 2020

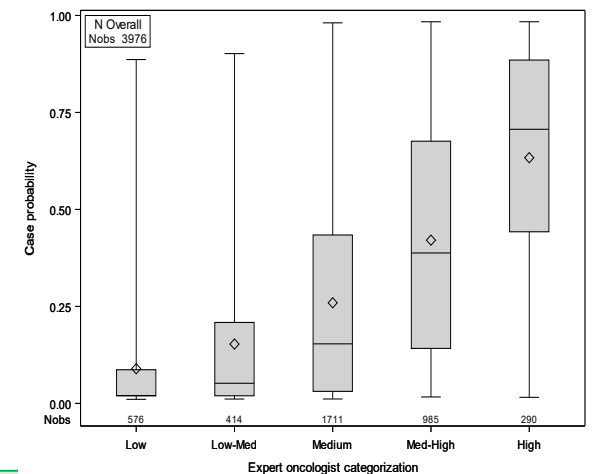
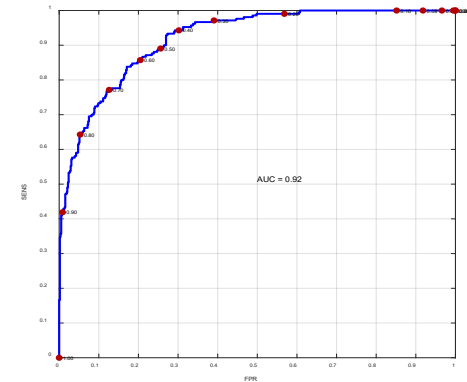


# Cervical Cancer Screening – AVE using Smartphones

- A “good set” of MobileODT dataset was evaluated by a panel of gynecologic oncologists at Rutgers University
  - Risk of having “precancer”
  - Image categories: High, Low, Medium (Uncertain and Post-Cryotherapy out)
  - Women categories (based on the categories of all her images): Low risk, Low-medium risk, Medium risk, Medium-high risk, and High risk

# Cervical Cancer Screening – AVE using Smartphones

- Used Low risk and High risk women to train, validate and test AVE
  - Multiple random splits
  - Used two deep learning object detection networks
  - The AUC values were above 0.9 for all the ROC curves
- Also included all Intermediate (Low-med, Med, Med-high) risk women to test AVE models
- The AVE severity score were strongly associated with gynecologic oncologist impression



**Xue Z, Novetsky AP, ..., Schiffman M, Antani S.** A demonstration of automated visual evaluation of cervical images taken with a smartphone camera, IJC: International Journal of Cancer, April 2020

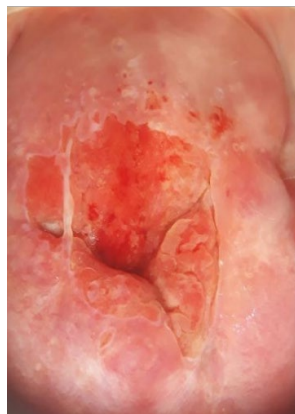
## Challenges for AVE

- A small amount of labeled (histopathology ground truth) dataset
- A highly imbalanced dataset (much fewer cases and a lot more controls)
- Robustness of AVE for image variabilities across multiple devices
- Image quality control and its influence on the AVE algorithm
- Selection of AVE cutpoint and adjustment of AVE scores

## Challenges for AVE – Variability Across Devices



Samsung S8 cellphone



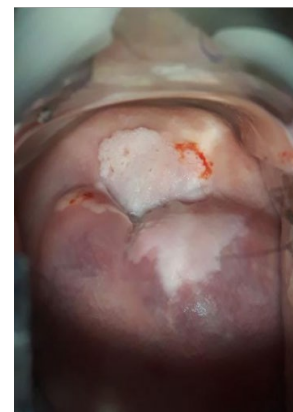
MobileODT EVA



Colposcope with Nikon camera



Samsung S8 cellphone



MobileODT EVA



Colposcope with Nikon camera



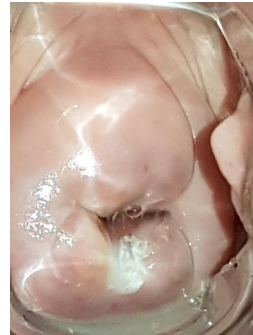
Motorola G3



Samsung J530

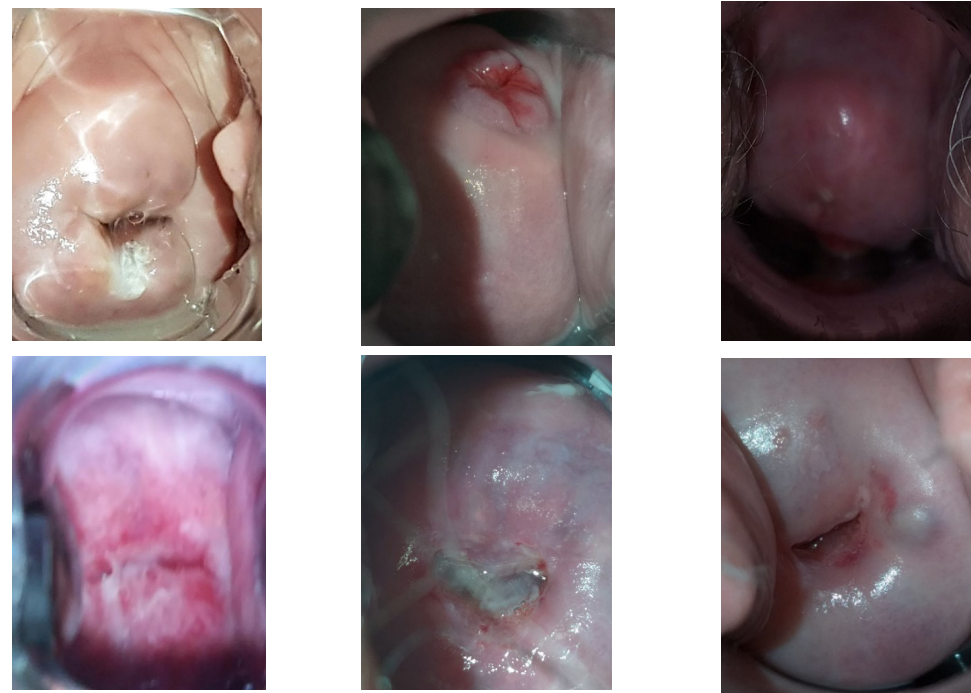
# Challenges for AVE – Image Quality

- Glare on the image
- Shadow
- Blur (shake, motion, digital magnification)
- Focus
- Occlusion and cervix visibility



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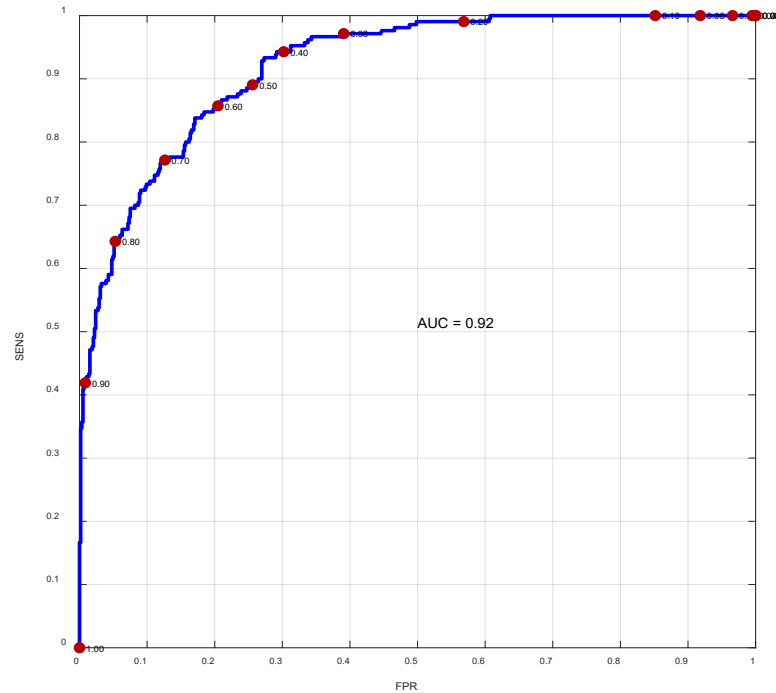


Guo P, Singh S, Xue Z, Long LR, Antani SK. **Deep Learning for Assessing Image Focus for Automated Cervical Cancer Screening**. IEEE EMBS International Conference on Biomedical & Health Informatics (BHI), May 2019

Ganesan P, Xue Z, Singh S, Long LR, Ghoraani B, Antani SK. **Performance Evaluation of a Generative Adversarial Network for Deblurring Mobile-phone Cervical Images**. IEEE Engineering in Medicine and Biology Conference (EMBC), July 2019

Guo P, Xue Z, Mtema Z, ..., Schiffman M, Antani SK. **Ensemble Deep Learning for Cervix Image Selection toward Improving Reliability in Automated Cervical Precancer Screening**, MDPI Diagnostics, July 2020

# Challenges for AVE – AVE Score Cutpoint Selection and Adjustment





## Conclusions

- AVE offers a great opportunity to address an important women's health problem
- Insight into clinical practices, health data acquisition, and gaps are key to identifying creative opportunities for intelligent automation
- Addressing challenges such as dataset size, variety, and quality are key to obtaining reliable results

