CUGH & NCI Cervical Cancer Webinar Series 2: Latest scientific advances, tools, & approaches to address cervical cancer control

August 5, 2020

11:00am-12:00pm EDT



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Co-Moderator



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Co-Moderator



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CUGH & NCI Cervical Cancer Webinar Series 2: Latest scientific advances, tools, & approaches to address cervical cancer control _{August 5, 2020}

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Deciding How to Control Cervical Cancer in Lower-Resource Settings: Health Decision Modeling and the Natural History of HPV Infection

Nicole Gastineau Campos, PhD



Guiding Cervical Cancer Prevention Policy in the Era of COVID-19

- Health care systems will face scarce human and economic resources.
- More than ever, cervical cancer prevention will need to be
 - Simple
 - Safe
 - Effective
 - Affordable
 - Cost-effective (high value for money)
- Health decision models will be used to identify costeffective screening strategies in lower-resource settings.

Evaluating Complex Prevention Strategies

HPV Vaccination	Cervical Screening	Treatment of Precancer
Age at vaccination	Age(s) at screening	Eligibility for treatment / type of treatment
Valency of vaccine	Screening test	Post-treatment surveillance
Number of vaccine doses	Referral threshold for screening test result	Delivery mechanism (mobile clinics; brick and mortar clinic)
Delivery mechanism (e.g., school-based; campaign)	Triage test or co-test	
	Treatment threshold for triage test or co-test result	
	Routine screening interval / number of lifetime screens	
	Delivery mechanism (e.g., provider- vs self-collection of sample; number of visits for testing, results, and treatment; high vs. low throughput)	

Why Use Health Decision Models?

- Many complex strategies to be compared
- Long interval between HPV infection and cancer is not directly observable in clinical studies
- Models are the only tools that project lifetime costs and consequences of strategies

Parts of a Screening Program in Lower-Resource Settings

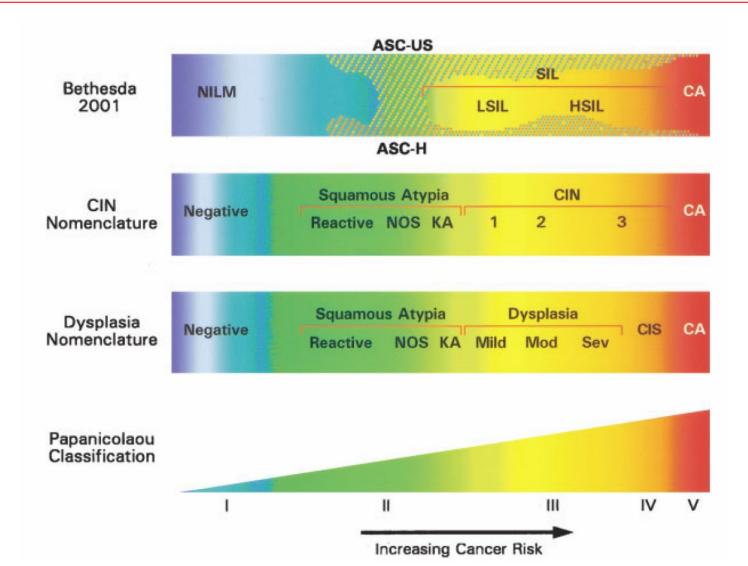
Screening: HPV test (self-collected sample)

> Triage: HPV genotyping; AVE

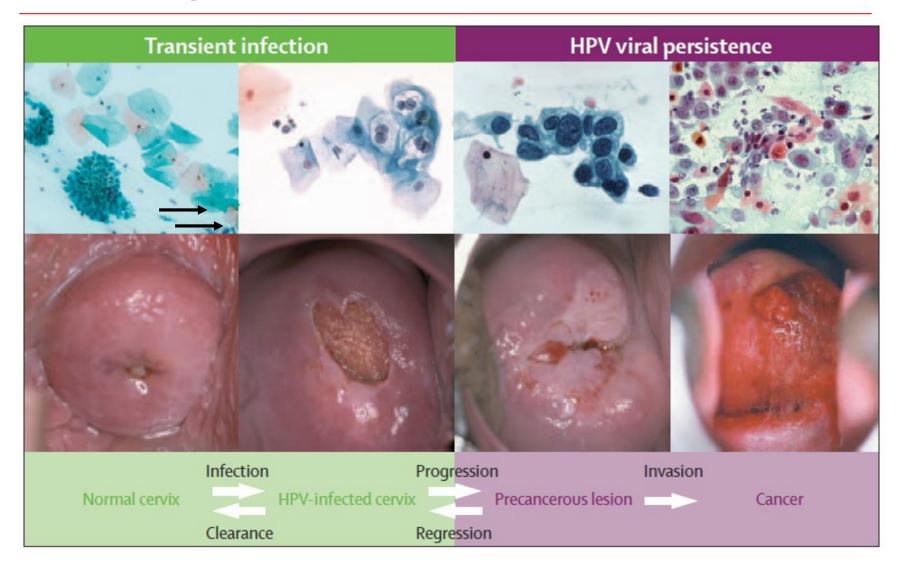
Treatment of precancer (eligible): Thermoablation

> Treatment of precancer (else): LEEP

Past and Current Models Use Clinical Definitions for Model Health States

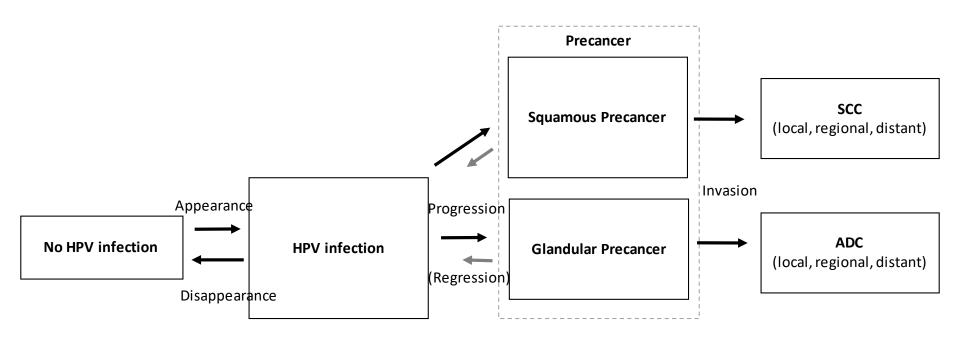


Multi-stage Causal Pathway of Cervical Carcinogenesis



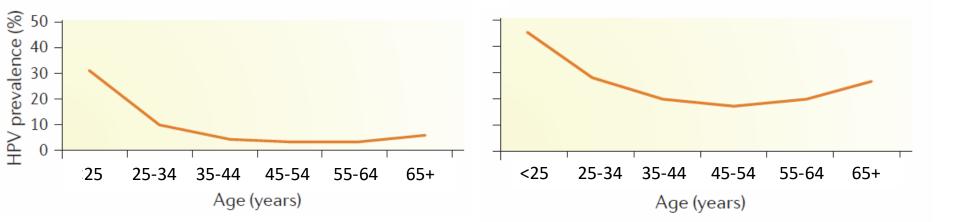
Source: Schiffman et al. Lancet 2007

New Health Decision Model Schematic: Universal Natural History

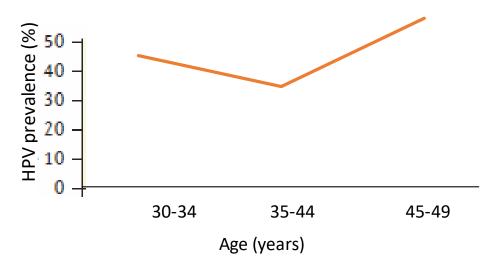


Global HPV Prevalence Patterns Vary by Region

Lower prevalence at optimal screening age (25-49 years) Higher prevalence at optimal screening age (25-49 years)



Women living with HIV (Nigeria)



Sources: Schiffman et al. Nature Reviews 2016; Denny et al. JNC/2010.

At Least Three HPV Natural History Patterns

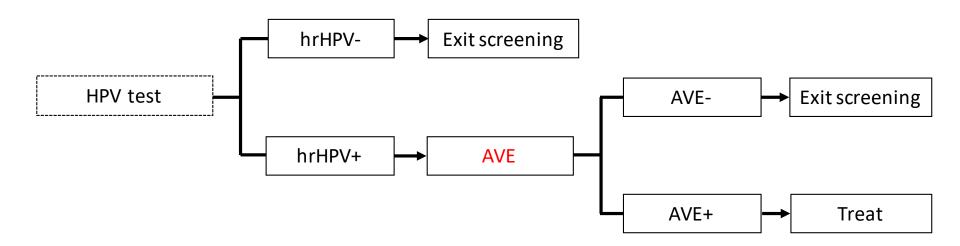
Pattern	Observed settings
Lower HPV Prevalence	North America; Oceania; Europe; Central/South America; Asia
Higher HPV Prevalence (non- HIV)	Sub-Saharan Africa
HIV	Women living with HIV

Data Availability for Transition Risks

Pattern	HPV Acquisition (by age, HPV type)	HPV Clearance (by HPV type, time since infection)	HPV Progression to Precancer (by HPV type, time since infection)	Invasion (by HPV type, duration of precancer)
Lower HPV Prevalence	Available	Available	Available	?
Higher HPV Prevalence (non- HIV)	Laboratory and data analysis in progress	Laboratory and data analysis in progress	Limited	?
HIV	Data analysis in progress	Data analysis in progress	Limited	?

Health decision models must account for population differences in transition risks in order to provide valid policy conclusions.

AVE Triage of hrHPV-Positive Women



Conclusions

- Health decision models are the only tool that can project cost-effectiveness over the lifetime for complex prevention strategies.
- To provide valid cost-effectiveness results, models must be based on
 - The multi-stage causal pathway of cervical carcinogenesis (universal)
 - Transition risks (vary by population HPV prevalence pattern)
- Development of a new modeling framework is underway.
- Transition risks for Higher HPV prevalence settings and WLHIV are urgently needed to inform valid health decision analyses.

New Tools and Approaches: Accelerating Cervical Cancer Control

Mark Schiffman, MD, MPH



August 5, 2020

Disclosure

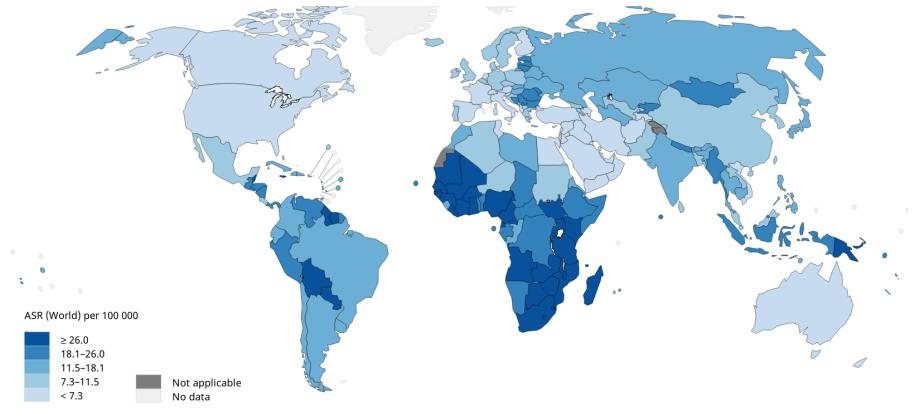
Our NCI research group has received cervical screening supplies and assay results at no cost for our independent evaluations of test performance. I have no commercial interest in any technology, and our research and this presentation are free from commercial influence.

The views I express are personal and do not necessarily represent NCI or any collaborator.

Acknowledging Many Collaborating Researchers

- I wish to acknowledge the large collaborative AVE research group.
- Special thanks to Silvia de Sanjose, Nicolas Wentzensen, Doug Lowy, NCI team, National Library of Medicine group, Global Research Labs group.

MOTIVATION: Cervical cancer is an avoidable disease with gross inequities (Globocan 2018) Progress in prevention lagging far behind our scientific knowledge



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



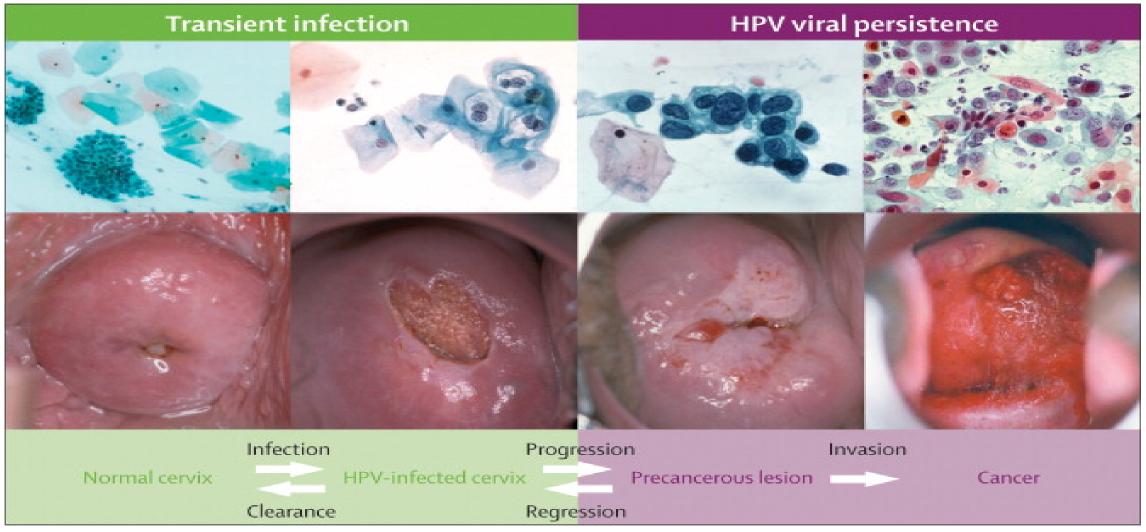
COVID-19 will tend to accentuate disparities

- Pandemic limits resources available for cervical cancer prevention
- Threat of spreading CoV-2 a serious consideration for prevention research and for screening programs
- In this environment, are currently planned programs still "better than nothing"?
- We need even better methods and strategies

The Time Lag Concern

- When we make decisions about how to control cervical cancer:
- To what degree can we anticipate improvements in prevention methods?
- Too much optimism is misleading
- What about "right around the corner" new technology?
 - There is no obvious answer but this talk is an example of high probability optimism

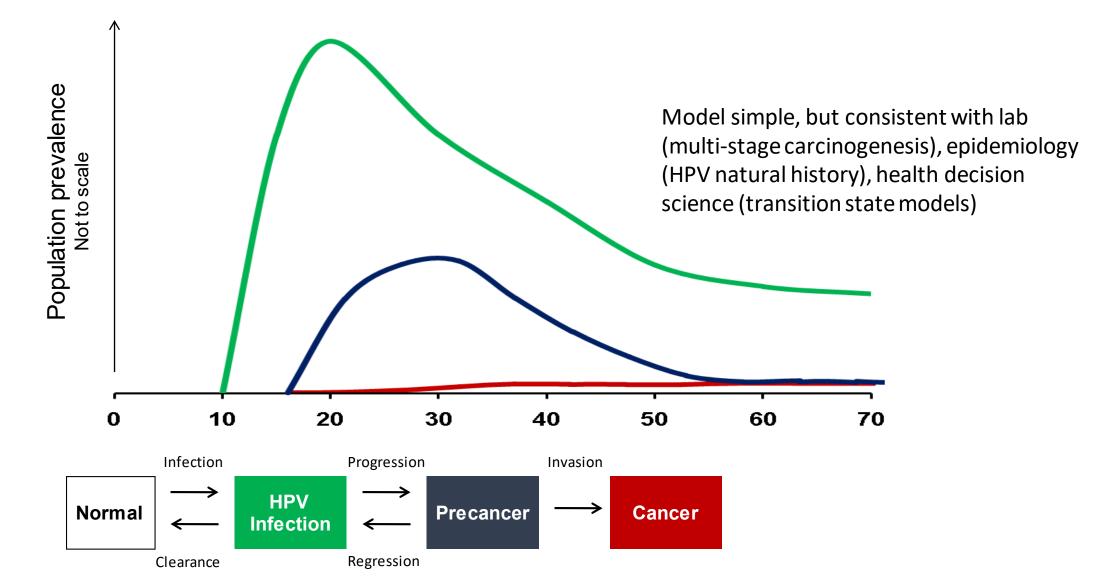
Simple View of Cervical Carcinogenesis Avoids Subjective Terms



Schiffman et al. Lancet 2007

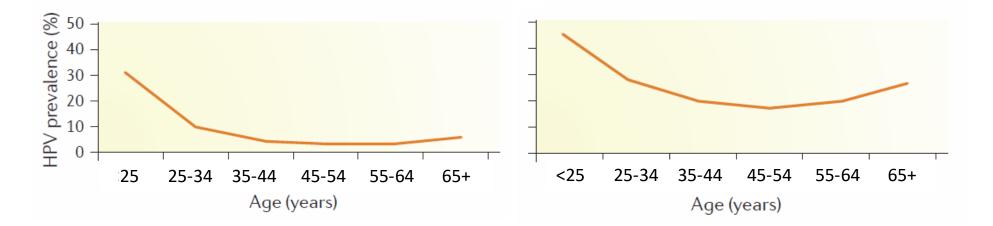
Causal Stages: Typical Age Curves

Note: Different in partly immunodeficient and WLWH populations

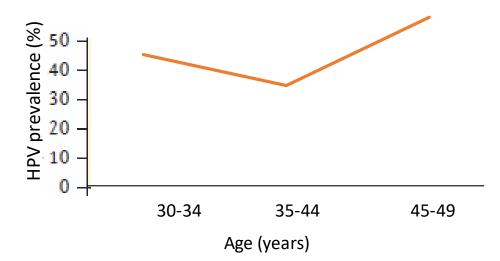


Global HPV Prevalence Patterns Vary by Region

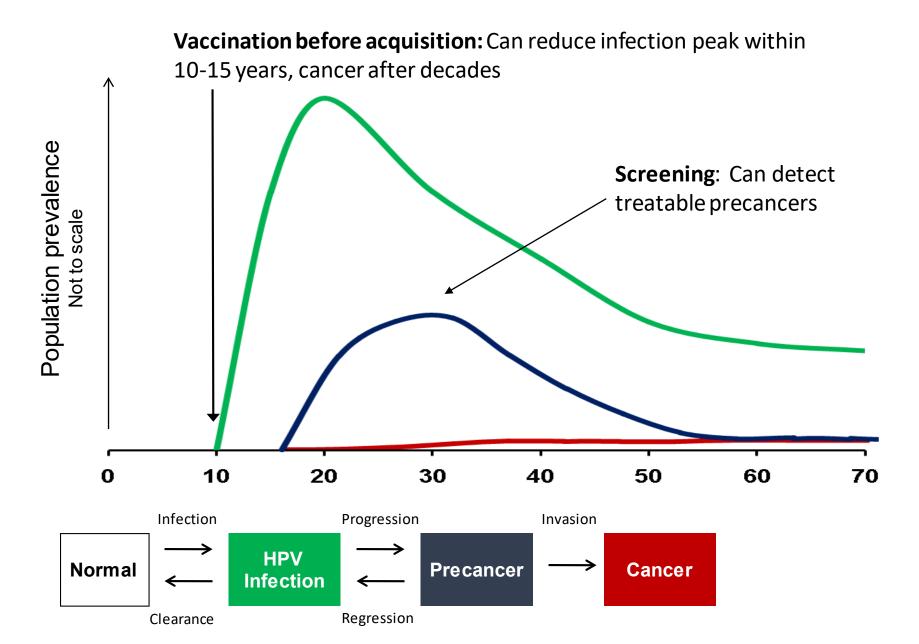
Lower prevalence at optimal screening age (25-49 years) Higher prevalence at optimal screening age (25-49 years)

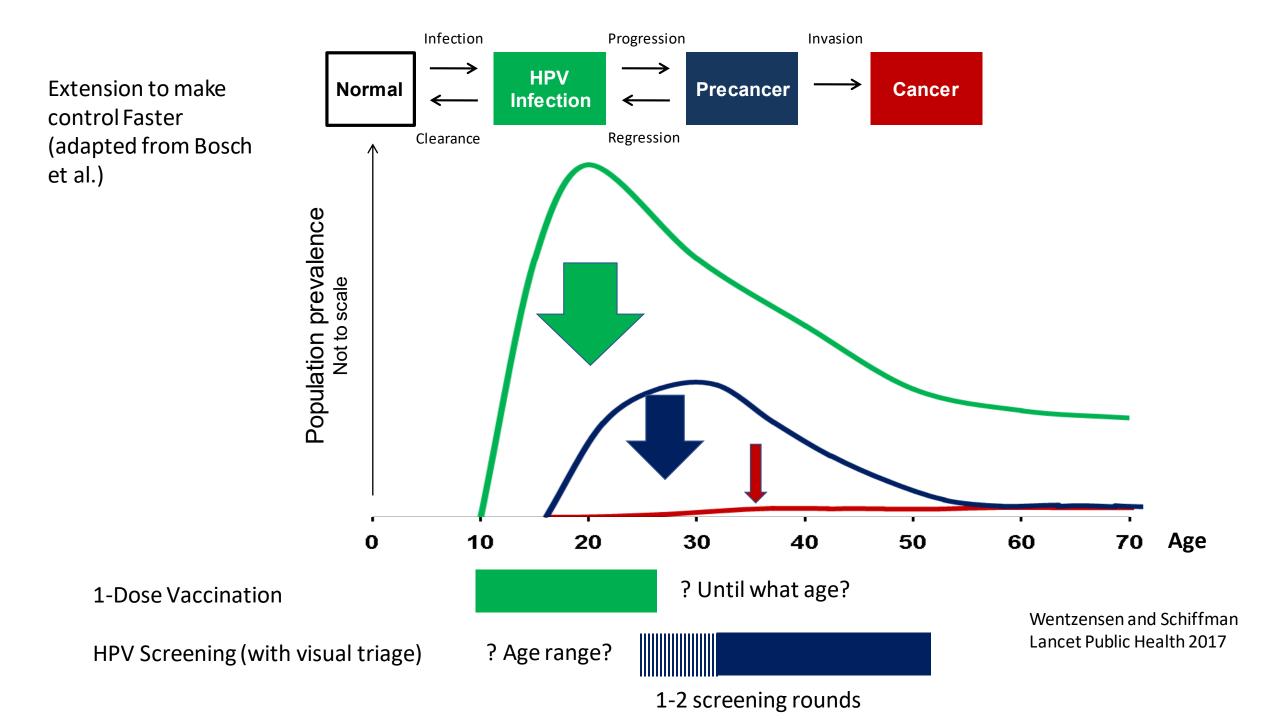


Women living with HIV (Nigeria)



Prevention Methods Tailored to Natural History

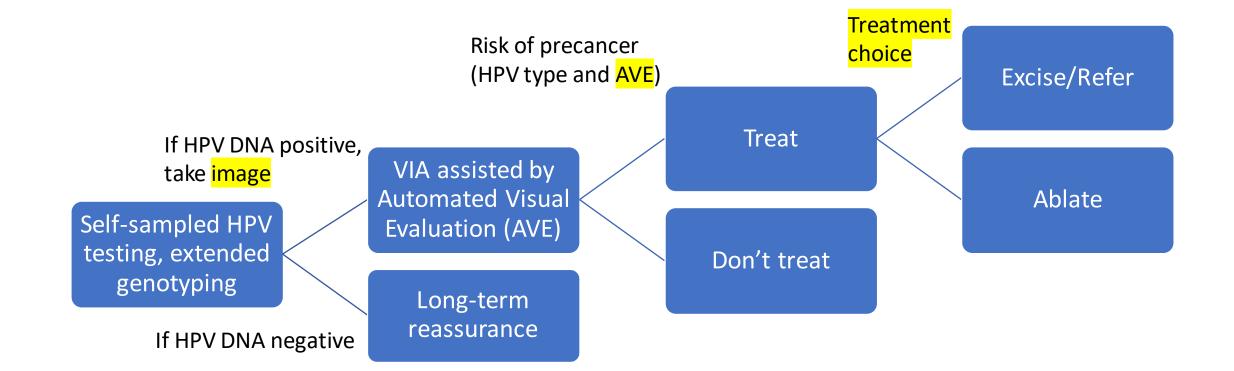




Personal estimation of a promising control strategy

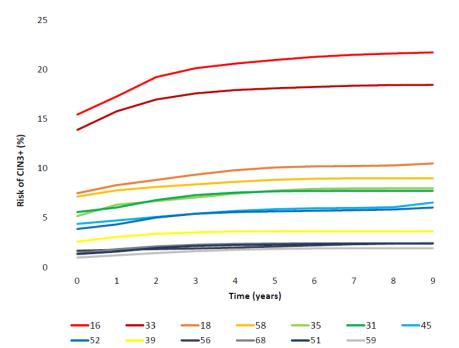
- (I predict that 1-dose HPV vaccine will protect long enough)
- Self-sampled HPV screening to provide reassurance to most women
 - Extended HPV typing using a technology like LAMP (isothermal loop mediated amplification)
 - Gives extended type group
- **Triage** using deep-learning evaluation of cervical images (assisted VIA)
 - Deep-learning algorithm #1 gives assurance of adequate image
 - AVE algorithm (#2) gives confidence score for whether HPV-positivity represents precancer combined with prediction from HPV typing
- Thermal ablation if feasible
 - Treatment choice algorithm (#3) gives deep-learning assistance on ablation vs. excision
- Excision restricted to those most at need

Summary of proposed strategy: HPV screening and visual triage assisted by <mark>3 deep learning algorithms</mark>



HPV type restriction might justify "extended" genotyping for prognosis

Type-specific cumulative risk of progression to CIN3+ of single HPV infections



Type group	% infections	7-yr CIN3+ risk
16	26	22
18, 45	5	>5, elevated cancer
31, 33, 35, 52, 58	39	>5
39, 51, 56, 59, 68	23	<5

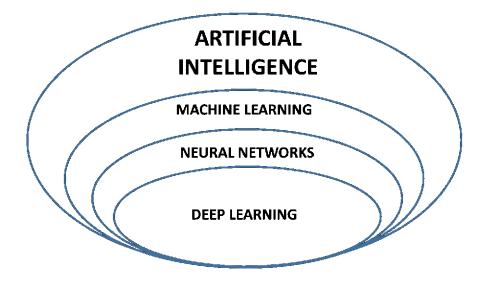
	Invasive cervical cancer		
	N tested	% pos	95% CI
HPV16	14595	54.4	53.6-55.2
HPV18	14387	15.9	15.3-16.5
HPV33	13827	4.3	4.0-4.6

Demarco, Hyun, et al. ECM, 2020. Schiffman, et al. Infect Agent Cancer, 2009.

Automated Visual Evaluation (AVE) for Triage of HPV-positive women

Key Concepts

- Artificial Intelligence the theory and development of computer systems able to perform tasks that normally require human intelligence, such as visual perception, speech recognition, decision-making, and translation between languages.
- Machine Learning the study of algorithms that improve automatically through experience
- (Artificial) Neural Network a machine learning method which learns by adjusting weights between interconnected network "layers" and "nodes"; inspired by observation of neuron networks in the human brain
- Deep Learning machine learning by use of Deep Neural Networks, that is, Neural Networks with "many" layers



Rodney Long and National Library of Medicine colleagues

Types of NN Learning

Supervised

- Two types classification and regression
- In either case, we provide training data points and desired outcomes to teach network
- For classification, we call the desired outcomes the class labels

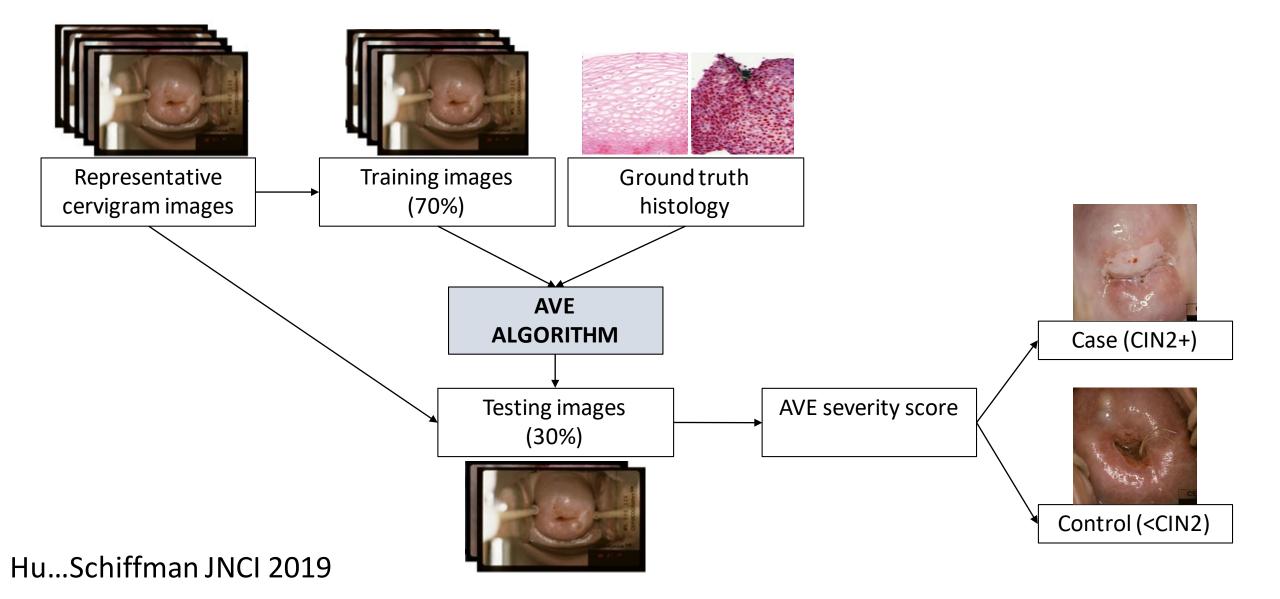
• Semi-supervised (classification)

- We provide:
 - A small number of training data points and class labels
 - A large amount of training data points w/o class labels
- Goal is (usually) to infer missing class labels by techniques such as clustering
- Then apply supervised learning to the completely-labelled training data

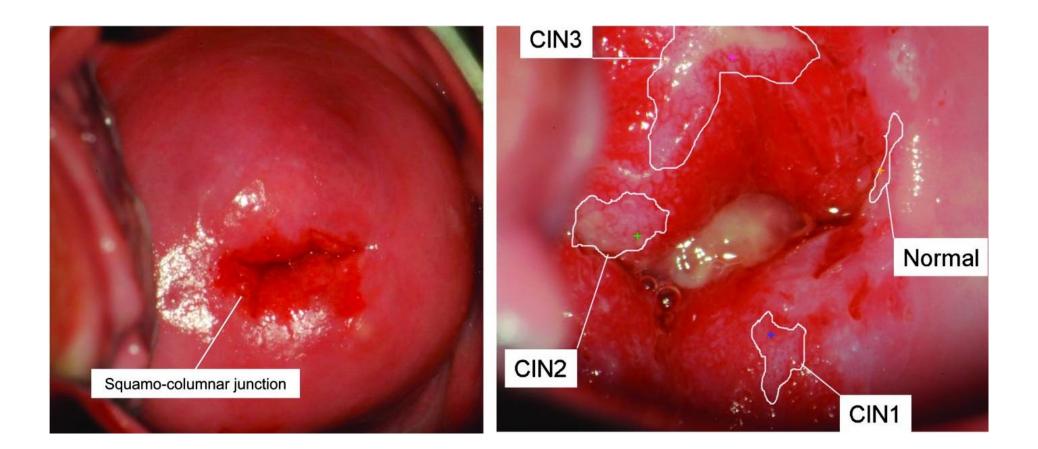
Unsupervised

- We provide training data points only
- Goal is to achieve a compact representation of the training data

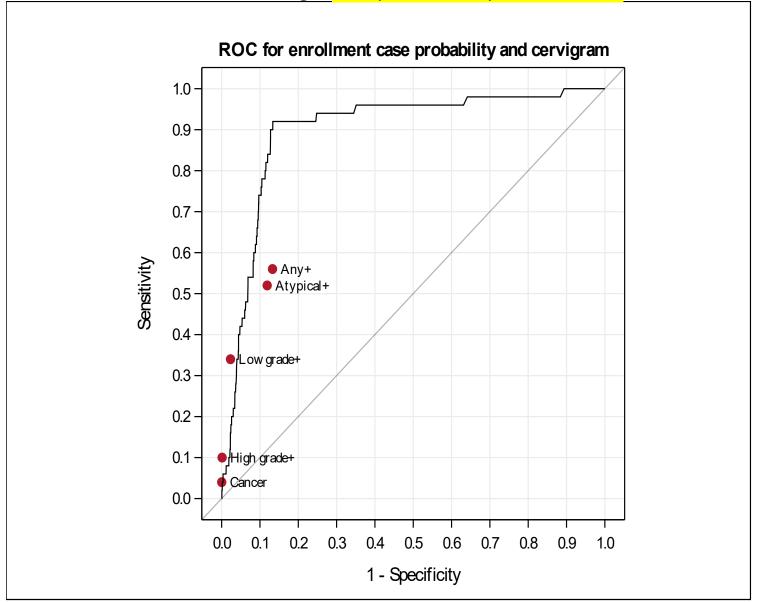
Deep Learning: Automated Visual Evaluation (AVE)



Visual Triage Not Easy for Human Observers (Can we match or exceed human experts?)



Guanacaste Cohort, Ages 30-49, General Screening AVE Algorithm Enrollment Image, Compared to Expert Reviewer



Contemporary digital image capture devices

- Proof-of-principle achieved using cervigrams
- AVE algorithm is sensitive to type of image, "fine tuning" between image types necessary and ease of "transfer learning" still not proven
- Large prospective collections of images paused due to COVID pandemic
- Confident that given enough images, we will succeed
 - We do have proof-of-principle that smartphone images can yield good AVE performance

Automated Visual Evaluation



Precancer

Ajenifuja and Desai et al., submitted

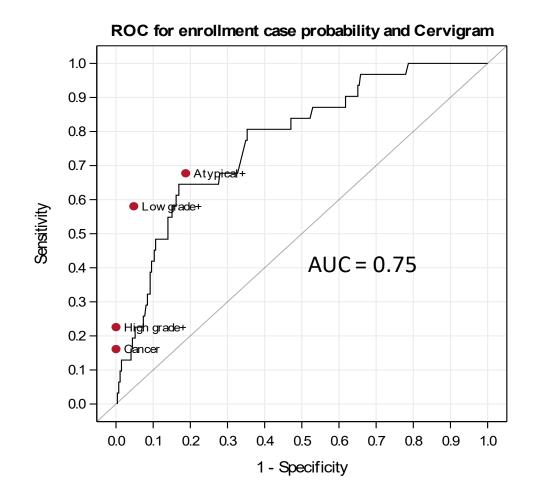
Normal

Adapting to different image collection devices (smartphones)

Current Limitation: AVE algorithm training needs more images than we have stored

- The more subtle the distinction we are trying to make, the more images we need
- Triage of HPV-positive women is even more difficult than general screening because the non-cases (HPV infections) are more like cases (precancer) than are HPV-negative controls
- Demonstration of why we need more images follows

Guanacaste Cohort, Ages 30-49, Triage AVE Algorithm Enrollment Image



Limitations of AVE discovered to date

- Preliminary evidence: AVE works mainly for classification of whether precancer is present "today", while HPV test/type predicts future (meaning, a negative triage test does not rule out future risk)
- Like all visual methods, deep learning works best when the cervical SCJ is fully visible (age restriction, we do not have good triage or treatment answers yet for older women)

Timeline for validation and dissemination

- The deep learning approach is valid and feasible
- The faster we accrue images for AVE, the faster we can compare to VIA alone and establish value (for general screening and especially for triage)
- Ethical constraints on launching large screening efforts
 - We have converted to a "stored image" strategy

Concluding Invitation

- We invite interested colleagues to join our AVE research community
- If you have collections of archived cervical images, consider collaboration
- Or, if you are interested in hearing more, we are starting a listserv for this new public health "specialty"
- Contact Silvia de Sanjose, Farideh Almani or me (preferably all of us).
 - Silvia <u>desanjose.silvia@gmail.com</u>
 - Mark <u>mark.w.schiffman@gmail.com</u>
 - Farideh <u>farideh.almani@nih.gov</u>

NCI-CGH/CUGH Webinar

Global Cervical Cancer: Beyond Technology....

Ophira Ginsburg, MSc MD Associate Professor, Department of Population Health Director High-Risk Cancer Genetics Program, Perlmutter Cancer Center NYU Grossman School of Medicine, NYU Langone Health



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I have no financial relationships to disclose.



First, do no harm.

 \succ Start and end with the women at risk.

Ensure you have considered all aspects, the needs and preferences of women, what matters to them.

Consider the whole picture, the cancer screening journey.

Do they have equitable access to affordable high-quality cancer health services? Including Rx for precancer and invasive cancer? Palliative care? Integrated *people-centred* health services means putting people and communities, not diseases, at the centre of health systems, and empowering people to take charge of their own health rather than being passive recipients of services.

Evidence shows that health systems oriented around the needs of people and communities are more effective, cost less, improve health literacy and patient engagement, and are better prepared to respond to health crises.

WHO 2016

https://www.who.int/servicedeliverysafety/areas/people-centred-care/ipchs-what/en/

Framework on integrated people-centered health services: an overview

Vision

	produced in a way that preferences, are cool comprehensive, safe, e	al access to quality health servi meets their life course needs a rdinated across the continuum effective, timely, efficient and a skilled and operate in a support	and respects social of care, and are cceptable; and all	
Strategy 1: Engaging and empowering people & communities	Strategy 2: Strengthening governance & accountability	Strategy 3: Reorienting the model of care	Strategy 4: Coordinating services within and across sectors	Strategy 5: Creating an enabling environment
Implementation principles Country-led Equity -focused Participatory Evidence-based Results-oriented Ethics-based Sustainable				

Systems strengthening

https://www.who.int/servicedeliverysafety/areas/people-centred-care/Overview IPCHS final.pdf?ua=1

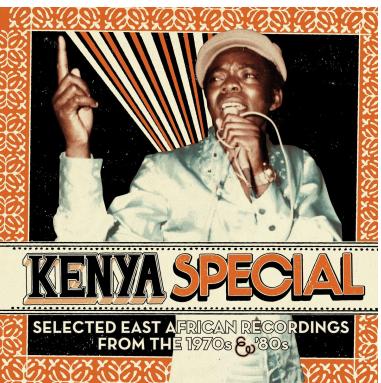


1970's....



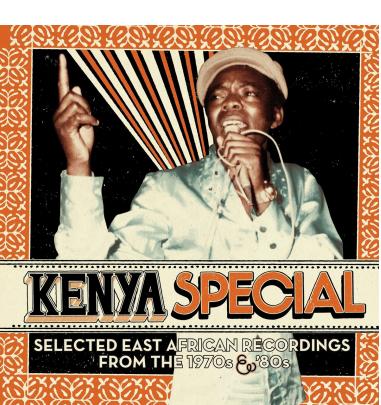






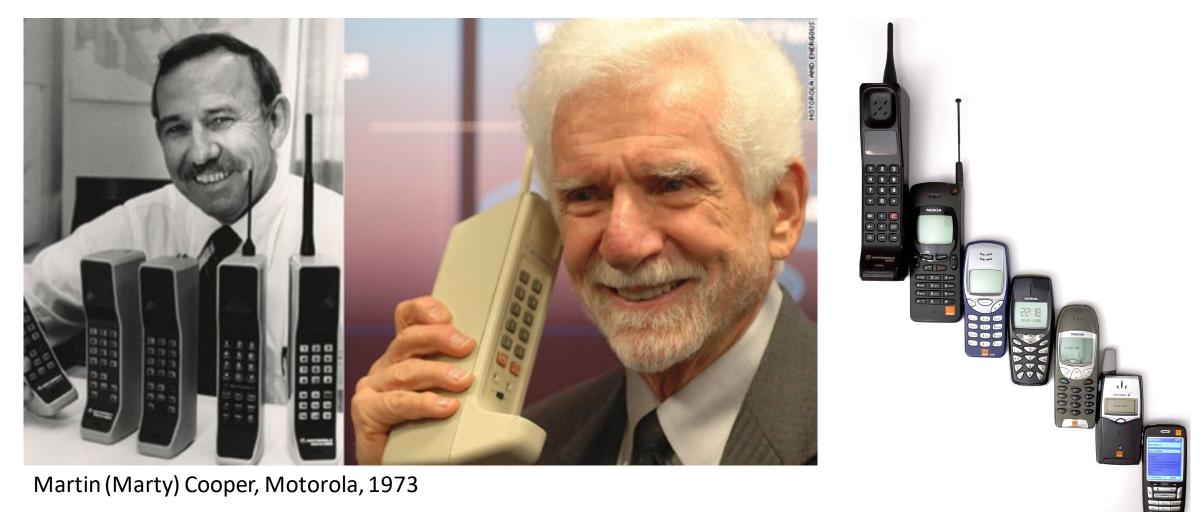








1973....





SEVENTY-FIRST WORLD HEALTH ASSEMBLY

Agenda item 12.4

WHA71.7 26 May 2018

Digital health

WHA 71/20 2018



"The spread of digital tech and global interconnectedness has significant potential to accelerate Member States' progress towards achieving universal health coverage, including ensuring access to quality health services."

from WHA 71/20



What is mHealth?

"Medical and public health practice supported by mobile devices, such as mobile phones, patient monitoring devices, personal digital assistants (PDAs), and other wireless devices."

mHealth: New horizons for health through mobile technologies, WHO, 2011

"The use of mobile wireless devices for public health"

WHA 71 DG report, 2018

Background (from WHA 71/20)

- ➢ 7 billion mobile phone subscriptions, 70% in LMIC
- In many LMICs, more people have access to a mobile phone than to clean water, a bank account, or electricity.
- Digital tech including mobile tech has the potential to revolutionize how populations interact with health services
- Mealth can improve quality and coverage of care, increase access to health information, services, and skills, and promote positive changes in health behaviors



International Telecommunications Union (ITU) 2015

BUT....

Governments have found it

challenging to assess, scale up, and

integrate mHealth "solutions".





Many pilot studies with no process for

scaling

Lack of interconnectedness between

apps, and of integration with existing

national eHealth & HIS infrastructures





WHO r on dig

17 April 2019 | News rel

WHO today released technology, accessib essential services.

WHO GUIDELINE

orld Health

RECOMMENDATIONS ON DIGITAL INTERVENTIONS FOR HEALTH SYSTEM STRENGTHENING



Ξ

Ξ

es can use digital health improve people's health and

Clients

- Targeted client communication
 - i.e. health promotion, education, countering myths, etc.
- 2-way SMS or info-line with nearest services





Health workers

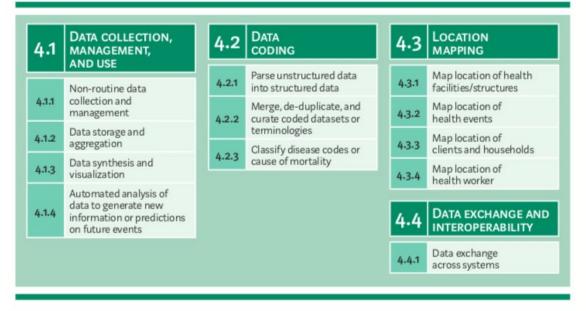
- Clinical decision support
- ➤ Training
- Quality assurance
- Referral coordination & tracking (navigation)

Health System Managers

- Supply chain management
- Facilities management
- HR management







Data Services

- Data collection/management/storage/ synthesis/visualization
- Location mapping: clients/patients/ facilities
- Data exchange & interoperability
- DHIS2, etc.
- Vital statistics, causes of death
- IARC: population-based cancer registries
- Cancer screening registry

WHO mERA Checklist

Item 1—Infrastructure: describe, in detail, the necessary infrastructure which was required to enable the operation of the mHealth programme

Item 2—Technology platform: describe, in sufficient detail to allow replication of the work, the software and hardware combinations used in the programme implementation

Item 3—Interoperability: describe how, if at all, the mHealth strategy connects to and interacts with national or regional Health Information Systems (HIS)/programme context

Item 4—Intervention delivery: elaborate the mode, frequency, and intensity of the mHealth intervention

Item 5—Intervention content: describe how the content was developed/identified and customised

Item 6—Usability testing: describe how the end-users of the system engaged in the development of the intervention

Item 7—User feedback: describe user feedback about the intervention or user satisfaction with the intervention

Item 8—Access of individual participants: mention barriers or facilitators to the adoption of the intervention among study participants

Agarwal S, LeFevre AE, Lee J, et al. BMJ. 2016;352:i1174

WHO mERA Checklist

Item 9—Cost assessment: present basic costs of the mHealth intervention

Item 10—Adoption inputs/programme entry: describe how people are informed about the programme or steps taken to support adoption

Item 11—Limitations for delivery at scale: present expected challenges for scaling up the intervention

Item 12—Contextual adaptability: describe appropriateness of the intervention to the context, and any possible adaptations

Item 13—Replicability: present adequate technical and content detail to support replicability

Item 14—Data security: describe security and confidentiality protocols

Item 15—Compliance with national guidelines or regulatory statutes

Item 16—Fidelity of the intervention

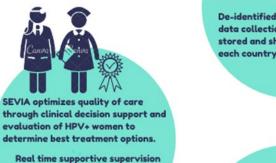
Agarwal S, LeFevre AE, Lee J, et al. BMJ. 2016;352:i1174

BE HE@LTHY BE MOBILE A handbook on how to implement

A handbook on how to implement mCervicalCancer

BE HE@LTHY BE MOBILE A handbook on how to implement

mCervicalCancer



Real time supportive supervision for provider training and quality assurance in cervical screening programs.



SEVIA facilitates care through supportive prompts for providers and patients: SMS/voice + calendar.

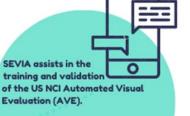
> Built-in and adaptable patient navigation tools to increase retention and reduce loss to follow-up.



De-identified image capture with data collection that is securely stored and shared according to each country's regulatory needs.

SEVIA

Smartphone-Enhanced Visual Assessment



De-identified images and data contribute to the US National Library of Medicine.

SEVIA's dashboard tracks 3 types of outcomes: process, programmatic and clinical.



Implementation Science: monitoring & evaluation in real-time.

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Proof of Concept Study in Tanzania

So Journal of Global Oncology An American Society of Clinical Oncology[®] Journal

Evaluation of a Smartphone-Based Training Strategy Among Health Care Workers Screening for Cervical Cancer in Northern Tanzania: The Kilimanjaro Method

Karen E. Yeates, Jessica Sleeth, [...], and Olola Oneko

Yeates KE, Sleeth J, Hopman W, Ginsburg O, Heus K, Andrews L, Giattas MR, Yuma S, Macheku G, Msuya A, Oneko O. Evaluation of a Smartphone-Based Training Strategy Among Health Care Workers Screening for Cervical Cancer in Northern Tanzania: The Kilimanjaro Method. J Glob Oncol. 2016 May 4;2(6):356-364. doi: 10.1200/JGO.2015.001768. PMID: 28717721; PMCID: PMC5493243.

original reports

Smartphone-Enhanced Training, QA, Monitoring, and Evaluation of a Platform for Secondary Prevention of Cervical Cancer: Opportunities and Challenges to Implementation in Tanzania

Karen Yeates, MD, MPH^{1,2,3}; Erica Erwin, MSc^{1,4}; Zac Mtema, PhD, MSc⁵; Frank Magoti⁵; Simoni Nkumbugwa, BSc⁵; Safina Yuma, MPH⁶; Wilma M. Hopman, MA⁷; Alyssa Ferguson, MPH⁸; Olola Oneko, MD⁹; Godwin Macheku, MMED¹⁰; Agnes Feksi Mtei, MD⁶; Carter Smith, BSc⁸; Linda Andrews, MPH⁸; Nicola West, BScN^{3,8}; Milena Dalton, MPH¹¹; Ashley Newcomb, MPH¹²; and Ophira Ginsburg, MD, MSc^{12,13}

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- Technology alone is never the solution to complex systems challenges.
- > Beware of hype. Look for and demand the evidence.
- Ask ourselves: is there a low-tech (or non-tech) intervention that might work as well? cost-effective? Affordable?

Acknowledgements

- Dr. Karen Yeates, Queen's University, Canada
- Dr. Zac Mtema, SkyConnect Ltd and Ifakara Health Institute, Tanzania
- Dr. Safina Yuma, CECAP Program, Ministry of Health, Tanzania
- Research Team: Erica Erwin, Frank Magoti, Simoni Nkumbugwa, Wilma M. Hopman, Alyssa Feguson, Olola Oneko, Godwin Macheku, Agnes Feksi Mtei, Carter Smith, Linda Andrews, Nicola West, Milena Dalton, Ashley Newcomb
- $\,\circ\,$ Grand Challenges Canada
- Pamoja Tunaweza Women's Centre, Kilimanjaro, Tanzania
- $\,\circ\,$ and most importantly the women who participated in all our studies....

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Nicole Gastineau Campos, PhD, Senior Research Scientist, Center for Health Decision Science Harvard T.H. Chan School of Public Health Boston, MA, U.S.A



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