





# Testing a deep learning algorithm for automatic detection of prenatal ultrasound for under-resourced communities

Marika Toscano<sup>1</sup>, MD; Junior Arroyo<sup>2</sup>, BS; Ana C, Saavedra<sup>2</sup>, BS; Thomas J, Marini<sup>3</sup>, MD; Timothy M, Baran<sup>3</sup>, PhD; Kathryn Drennan<sup>1</sup>, MD; Ann M, Dozier<sup>4</sup>, PhD; Lorena Tamayo<sup>5</sup>, BS; Tina Zhao<sup>6</sup>, MD; Benjamin Castaneda<sup>2</sup>, PhD.

Considerate in triggers and its Techniques on Expension Publisher International Parts and New York Considerate Principles (Service Street Association (Service Association of Service Street Association (Service Ass

### Introduction

In rural and under-resourced communities, the scarcity of Obstetric ultrasound (OB US) imaging results in a considerable gap in the perinatal healthcare.

## Objective

To test a new automated diagnostic framework performed without an experienced sonographer or interpreting provider for assessment of fetal biometry measurements, fetal presentation, and placental position.

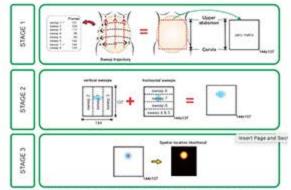
### Methods

- A standardized volume sweep imaging (VSI) protocol based solely on external body landmarks was used to obtain OB USN without an experienced sonographer
- A deep learning algorithm (U-net) was trained to automatically segment the fetal head and placental location from VSI OB US to evaluate fetal biometry, fetal presentation, and placental position without a radiologist

A deep learning algorithm accurately predicts fetal presentation, placental location, and fetal biometry from ultrasound images obtained by individuals without prior ultrasound training.

This offers a promising means for expanding access to vital Obstetric ultrasound imaging in rural and under-resourced

# Fig 1. Scheme for the generation of the spatial location likelihoods

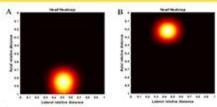


Stage 1: Volume sweep imaging (VSI) protocol (depicted by arrows) is performed on gravid abdomen by ultrasound-naïve operator with 8 hours of training and guided solely by external body landmarks.

Stage 2: Deep learning algorithm applied, with frames containing the target region (in this case fetal head) colored

Stage 3: A Gaussian filter is applied to produce the spatial location likelihood. Based on this map, our algorithm produces a diagnosis: non-cephalic fetus

# Fig 2. Heatmaps of cephalic (A) and non-cephalic (B) presentation.



Color signal corresponds to location of fetal head