# **AutoFB: Automating Fetal Biometry Estimation** from Standard Ultrasound Planes

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### **UCL**

# Introduction

Ultrasound (US) examination during pregnancy can

- Assess fetal size according to standardized biometry charts
- Biometry assesses fetal well-being and growth trajectory

Reproducible and accurate fetal biometry requires

- Identification of 3 standard 2D US planes by the sonographer
- Manually marking the key anatomical landmarks to clinically measure biometry







# **Fetal Biometry**

#### Three Standard Ultrasound Planes



Transventricular plane in the head



Transabdominal plane in the abdomen



Femur plane

HC – Head Circumference

**BPD** - BiParietal Diameter

OFD - Occipito-Frontal Diameter

AC - Abdominal CircumferenceTAD - Transverse Abdominal DiameterAPAD - Anterior-Posterior Abdominal Diameter

FL - Femur length



# **Motivation**

- Time-consuming task especially for less experienced sonographers
- Subject to intraoperator and interoperator variabilities

#### Automating fetal biometry

- Help in minimizing variabilities
- Serve as expert for trainee sonographers

#### Existing methods

• Rely on prior knowledge of which measurement to perform on a given image [1, 2, 3]



[Dromey, Prenat. Diagn. 2020]



Sobhaninia, et al. "Fetal ultrasound image segmentation for measuring biometric parameters using multi-task deep learning." EMBC 2019.
Khan, et al. "Automatic measurement of the fetal abdominal section on a portable ultrasound machine for use in low and middle income countries." IUS 2016.
Hermawati, et al. "Automatic femur length measurement for fetal ultrasound image using localizing region-based active contour method." Journal of Physics, 2019.

# **AutoFB - Contributions**

• A unified automated framework that estimates all relevant measurements for fetal biometry assessment from all three standard ultrasound planes









- Identifies and segments head, abdomen and femur
- Ablation study of UNet [1] and Deeplabv3+ [2] with different backbones





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[1] Ronneberger, et al. "UNet: Convolutional networks for biomedical image segmentation." MICCAI, 2015. [2] Chen, et al. "Encoder-decoder with atrous separable convolution for semantic image segmentation." ECCV, 2018.





- Fitting an ellipse on the head or abdomen masks
- Fitting a bounding box on the femur mask







- Template matching detects the ruler markers on the caliper visible on the US images
- Templates of ruler markers:



Ratio of millimetre (mm) to pixel gives the scaling factor







- Fitted ellipse circumference gives HC (AC) estimates
  - Major and minor axes gives BPD and OFD (TAD and APAD) estimates
- Bounding box diagonal gives FL estimates





# **Clinical Ultrasound Data Collection**

- Patients attending University College London Hospital for US examination enrolled and pseudo-anonymized after written consent and ethics approval
- Segmentation ground-truth obtained using the VIA tool [1]
- Clinically obtained measurements provided by the clinicians

# Subjects	# US Images	# Head	# Abdomen	# Femur
42	346	135	103	108

#### **Observations**

- Large intra-class variability
- Class imbalance









[1] Dutta and Zisserman, "The VIA Annotation Software for Images, Audio and Video." MM, 2019.

# **Results – Multiclass Segmentation**

- 4-fold cross-validation applied on Deeplabv3+ [1] and UNet [2] models
- Models trained with MobileNetv2 (MNv2) and ResNet50 backbones
- Either cross-entropy (CE) or weighted cross entropy (wCE) loss was used

Overall and per-class mean Intersection over Union (mIOU) for methods under comparison

Method	mIoU	mIoU-BG	mIoU-H	mIoU-A	mIoU-F
Deeplabv3+ (MNv2-CE)	$0.87 {\pm} 0.02$	$0.95 {\pm} 0.02$	$0.93 {\pm} 0.02$	$0.89 {\pm} 0.03$	$0.61 {\pm} 0.03$
Deeplabv3+ (MNv2-wCE)	$0.88{\pm}0.01$	$0.95{\pm}0.01$	$0.93{\pm}0.02$	$ 0.89{\pm}0.02$	$0.61{\pm}0.02$
UNet (MNv2-CE)	$0.82\pm0.05$	$0.93\pm0.03$	$0.89\pm0.05$	$0.85 \pm 0.05$	$0.56 \pm 0.03$
UNet (MNv2-wCE)	$0.86 \pm 0.01$	$0.94\pm0.01$	$0.91\pm0.02$	$0.86 \pm 0.02$	$0.58 \pm 0.01$
UNet (Resnet-CE)	$0.75\pm0.06$	$0.88\pm0.05$	$0.84\pm0.07$	$0.77 \pm 0.05$	$0.53 \pm 0.03$
UNet (Resnet-wCE)	$0.78\pm0.04$	$0.87 \pm 0.03$	$0.83 \pm 0.04$	$0.75 \pm 0.06$	$0.53 \pm 0.02$



- background

– Head

Abdomen <u>Femu</u>r

[1] Chen, et al. "Encoder-decoder with atrous separable convolution for semantic image segmentation." ECCV, 2018. [2] Ronneberger, et al. "UNet: Convolutional networks for biomedical image segmentation." MICCAI, 2015.



# **Results – Fetal Biometry Estimation**

Comparison between the best performing models and absolute error between the clinically measured and predicted fetal biometry



For the best performing segmentation network (Deeplabv3+):

Error is lowest with a median of

- 0.80*mm* for BPD
- 1.30*mm* for OFD
- 2.67*mm* for HC

Error is lowest with a median of

- 2.39*mm* for TAD
- 3.82*mm* for APAD
- 3.77*mm* for AC

FL Error is comparable

- median of 2.1*mm*
- but with fewer outliers



# **Results – Qualitative Comparison**

• Inaccurate segmentation resulted in fetal biometry estimation failure







# Conclusion

- AutoFB framework with Deeplabv3+ outperformed other models despite large intra-class variability
- Obtained errors are lower than the ±15% error permissible during routine clinical assessment

### **Future work**

- Online detection of standard planes and biometry estimation in ultrasound videos
- Experts and novices comparison with the AutoFB can provide evidence supporting clinical translation







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# Thank you

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