

Learn2Reg:

Opportunities and Challenges for Data Fusion

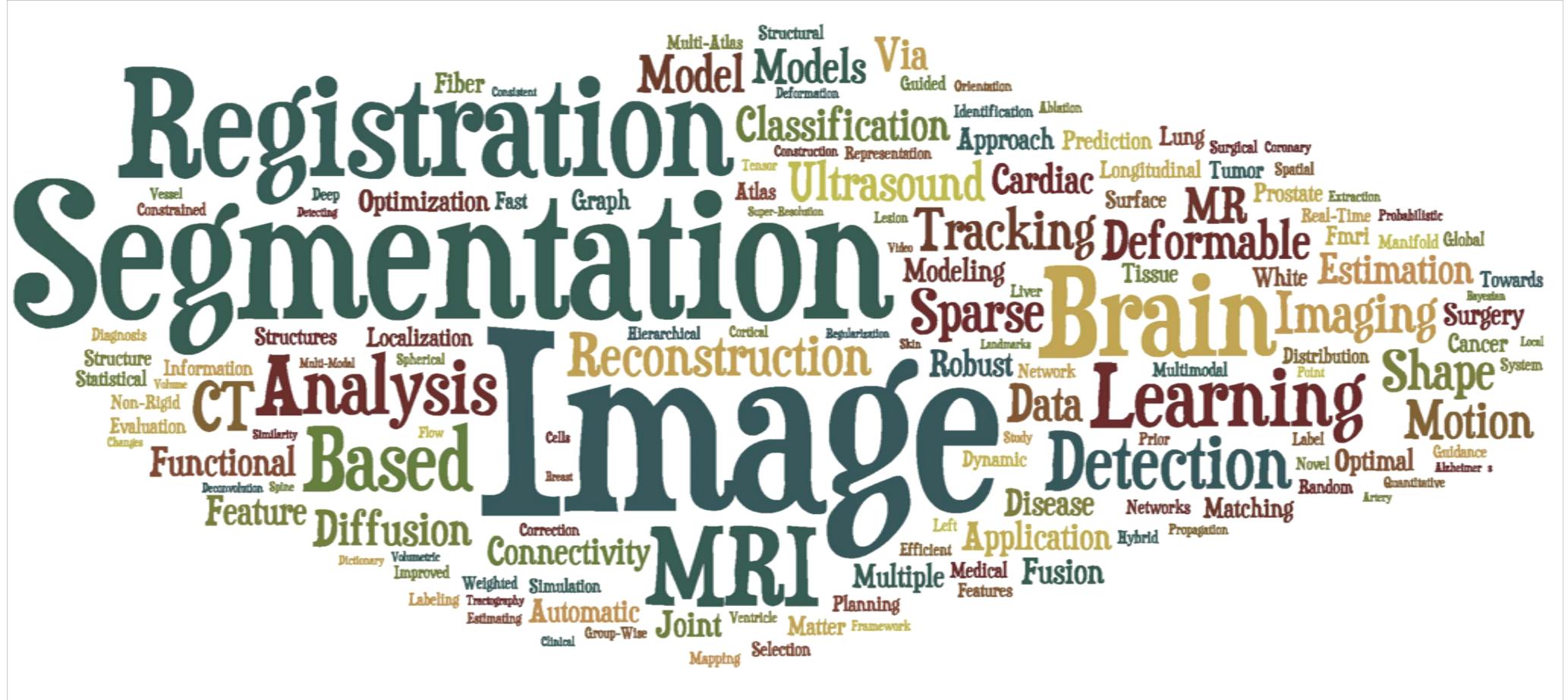
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Image registration – a brief MICCAI history (2013)

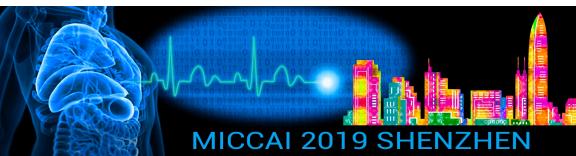


MICCAI 2019 SHENZHEN

MICCAI 2013, Nagoya, Japan
– courtesy of Kensaku Mori

Image registration – a brief MICCAI history (2018)

A word cloud visualization centered around the theme of medical image analysis and deep learning. The main words are 'learning', 'deep', 'using', 'segmentation', 'convolutional', 'network', 'registration', 'based', 'image', 'imaging', 'brain', 'analysis', 'ct', 'mri', 'mr', 'ultrasound', 'medical', 'breast', 'prostate', 'cancer', 'lung', 'respiratory', 'tracking', 'multi-scale', 'fully', 'recurrent', 'adversarial', 'estimation', 'functional', 'disease', 'automated', 'mapping', 'high', 'multiple', 'information', 'early', 'learning-based', 'prediction', 'tumor', 'structures', 'unsupervised', 'medical', 'ultrasound', 'spatial', 'semantic', 'jet', 'local', 'via', 'automated', 'functional', 'disease', 'quantification', 'adaptive', 'iterative', 'information', 'early'.



MICCAI 2018, Granada, Spain

KING'S
College
LONDON

Image registration – a brief MICCAI history (2018)

- Medical Image Analysis - MICCAI 2018 special issue had 2/9 published papers worked on deep learning for registration:

Unsupervised learning of probabilistic diffeomorphic registration for images and surfaces

Adrian V. Dalca, Guha Balakrishnan, John Guttag, Mert R. Sabuncu

October 2019

Pages 226-236

Adversarial learning for mono- or multi-modal registration

Jingfan Fan, Xiaohuan Cao, Qian Wang, Pew-Thian Yap, Dinggang Shen

December 2019

Page 101545



Image registration – a brief MICCAI history (2019)

- **MICCAI 2019** – registration papers:
 - 24 papers with the word “registration” in title
 - 17 papers with the word “motion” in title
 - 3 papers with the word “deformation” in title
 - (*did not check for tracking, image guidance, fusion....*)
 - That makes ~44 / 538 accepted papers (~8%)
- Satellite events – 3 registration tutorials but no workshops:
 - **Learn2Reg**
 - AirLab: Autograd Image Registration Laboratory
 - Image Registration during Minimally Invasive Catheter-based Interventions

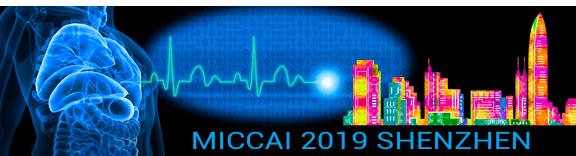


Image registration – a brief MICCAI history (2019)

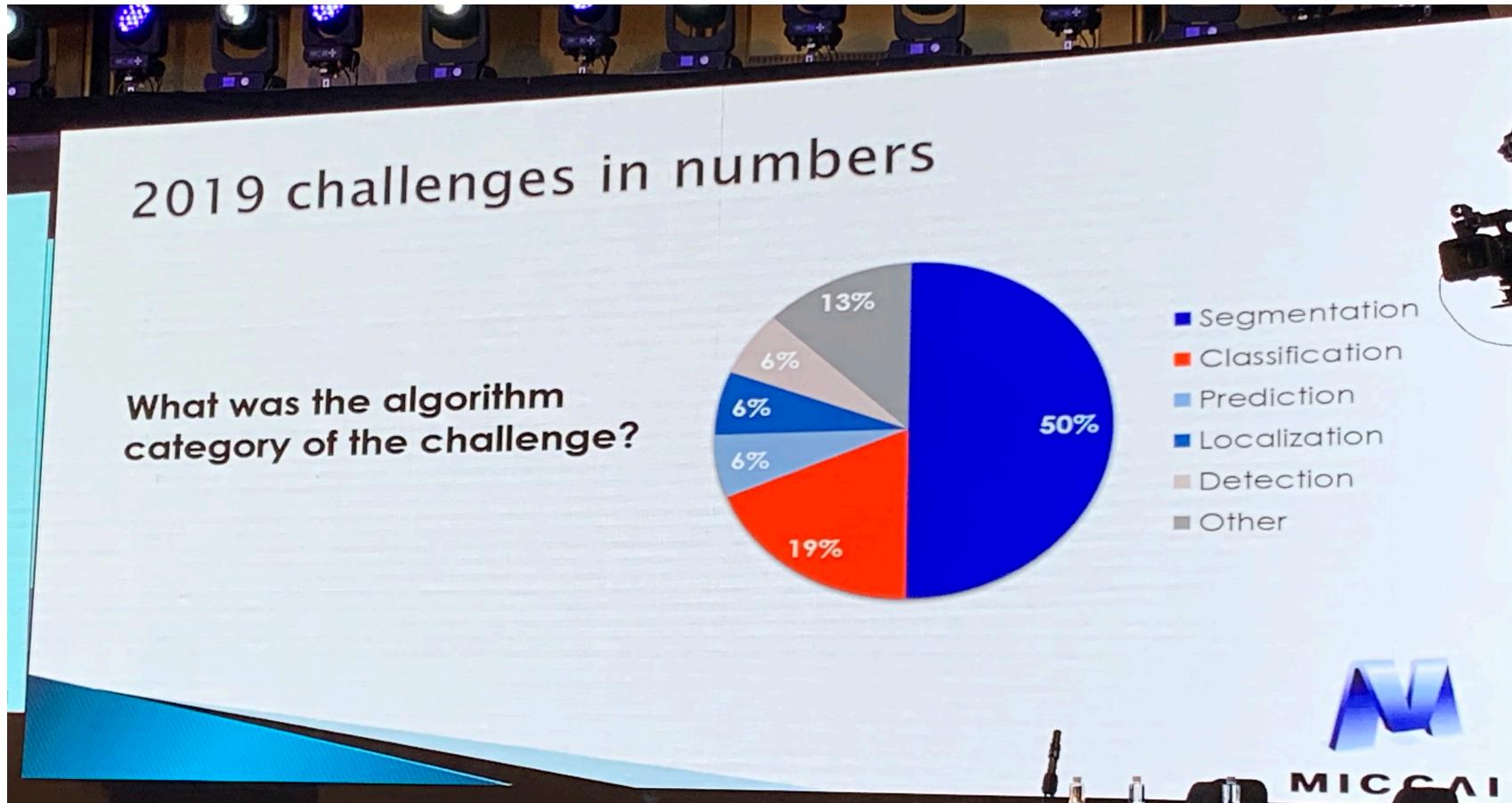


Image registration – a brief WBIR history

- 8th International Workshop on Biomedical Registration 2018:
 - Invited lecture: “*Is image registration a solved problem?*”



WBIR 2018 in Leiden, NL
Marius Staring and Stefan Klein

- *Apparently not!*



WBIR 2020 in Portorož, Slovenia
Žiga Špiclin, Jamie McClelland, Jan Kybic, and Orcun Goksel



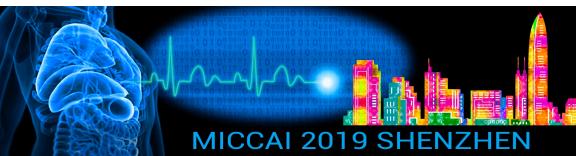
SHENZHEN

Winter of registration?



Opportunities and Challenges

- Deep Learning has made huge advances predominantly in image/disease classification, detection, prediction and segmentation
 - Due to availability of large, annotated/labelled data sets
 - Generally now considered to outperform previous, more traditional methods
 - Still challenges related to domain changes, presence of abnormalities etc



Opportunities and Challenges

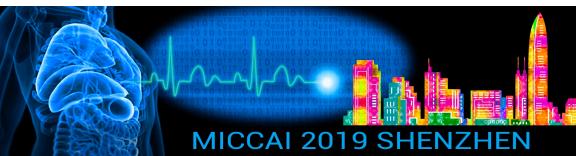
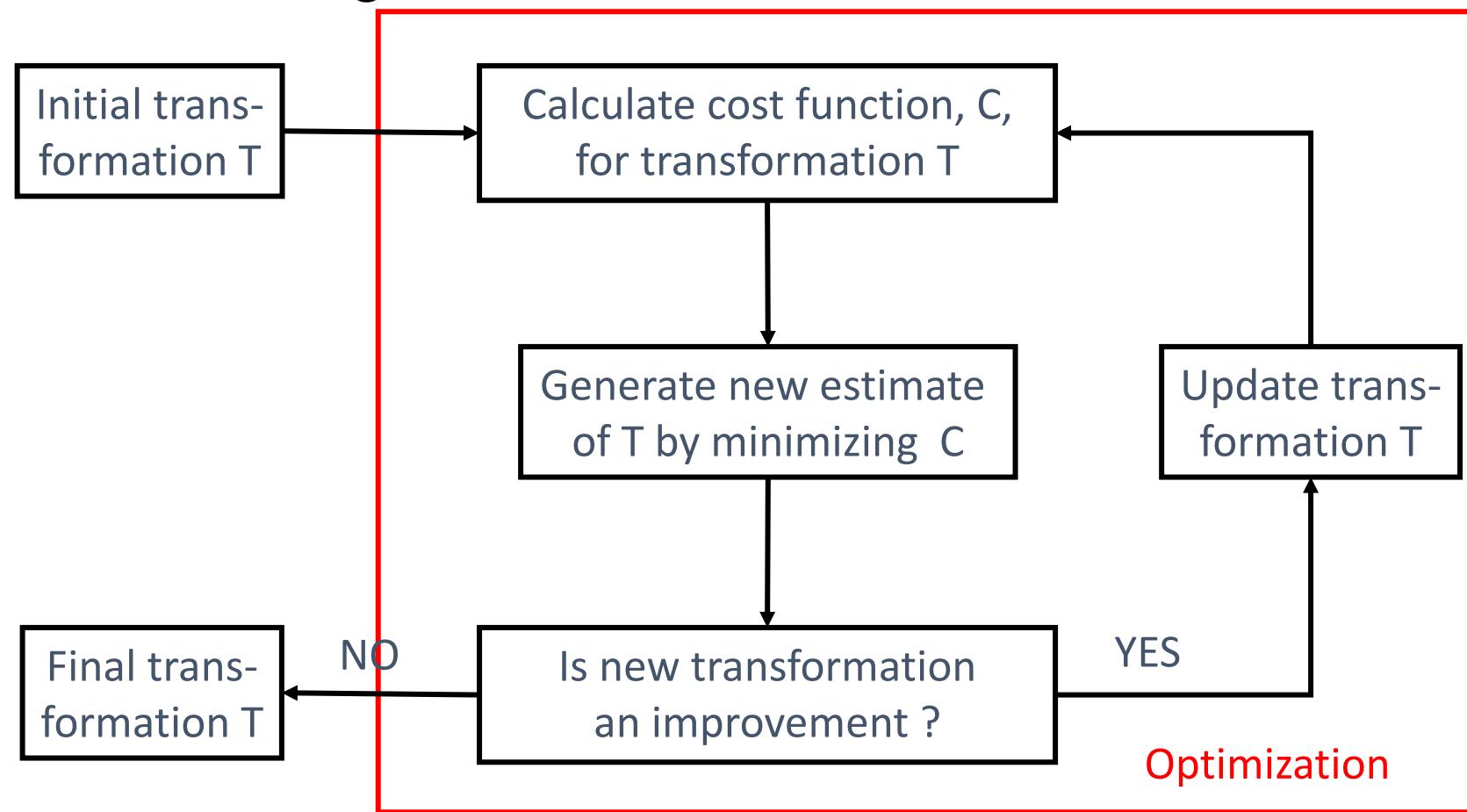
- Image registration still somewhat “limping behind”
 - Due to lack of ground truth annotations, true deformation field etc
 - In deep learning methods, registration often used for data augmentation, preprocessing etc.
- **New deep learning approaches for registration are emerging***



*Introductions to unsupervised, self-supervised, semi- and weakly supervised methods have been given this morning!

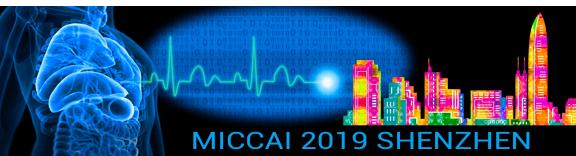
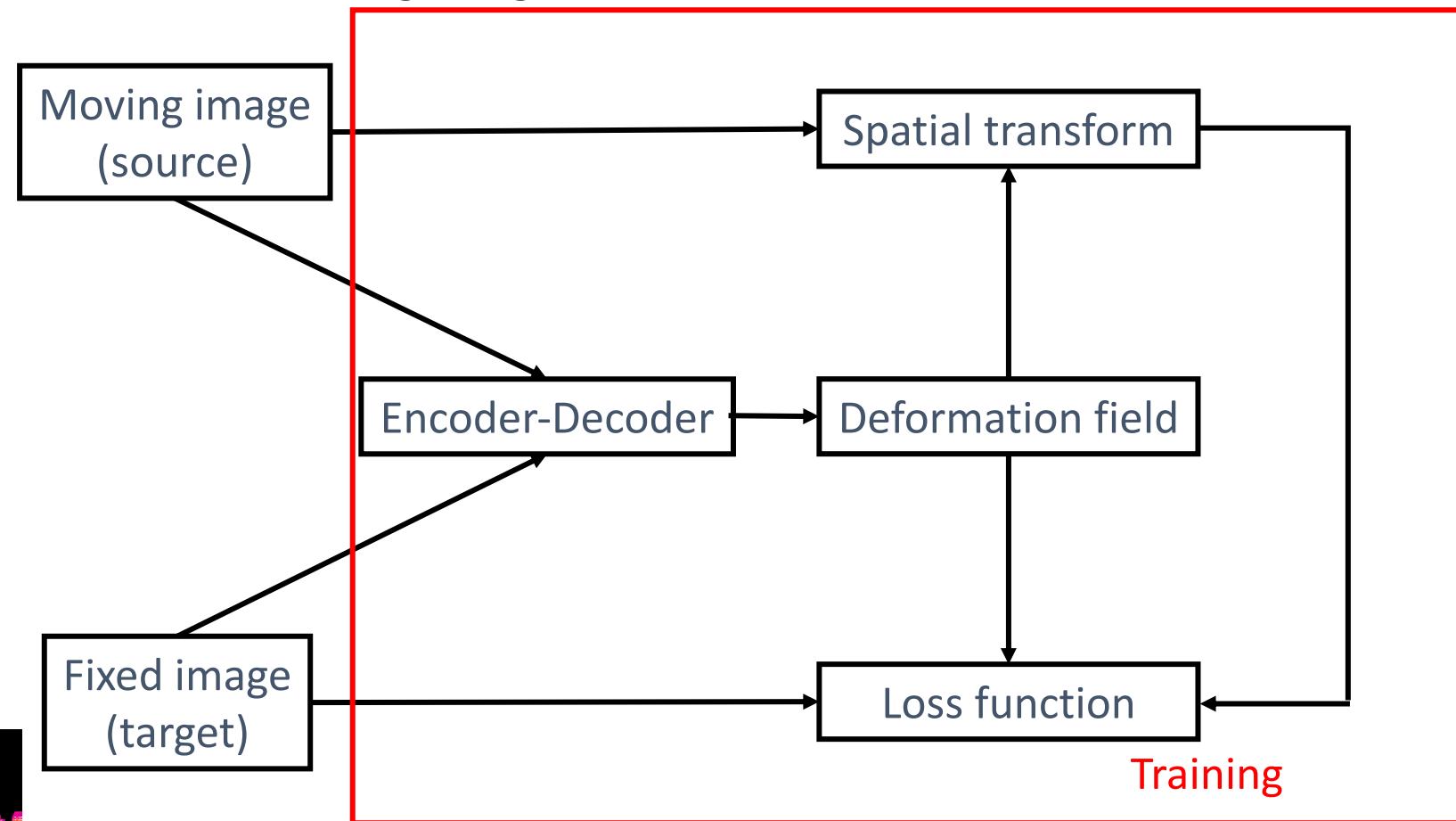
Opportunities and Challenges

- Recall conventional registration:



Opportunities and Challenges

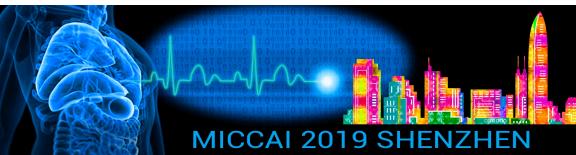
- Model for deep learning registration:



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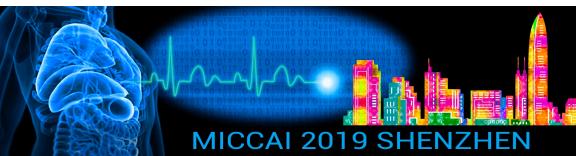
Opportunities and Challenges

- Conventional registration:
 - Requires domain knowledge
 - Single/multi-modal image registration, image protocol used, which organ...
 - Requires robust parameter tuning and setting
 - Weighted cost function, number of resolution levels, ...
 - Requires prior information on expected types of deformation:
 - Rigid, affine, deformable, composite, sliding, volume preserving, ...



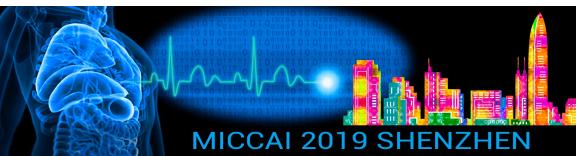
Opportunities and Challenges

- Deep learning registration:
 - Conventional cost function, transformation, optimisation methods have counterparts in loss function, transformer network, training
 - Training on many image pairs replaces optimisation on single image pair
 - Network architecture design and associated hyperparameters still remain important
 - Super-fast at inference time – but requires training data and training time!
 - And some conventional image registration methods have also near real-time performance and do not require training data – in many cases they now work “out of the box”



Opportunities and Challenges

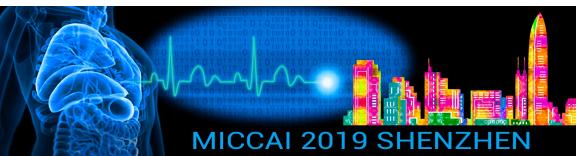
- Loss function has direct relation to conventional registration cost function!
 - Ie here one can play with almost all existing image similarity and regularization methods **as long as they are differentiable** (as required for backpropagation)
- But this handcrafted loss function also has drawbacks
 - We fall into the same trap as before, as we do not directly learn the image similarity measure (nor the transformation model)
 - Instead, we indirectly tune the weightings in regard to the loss function, by using many representative data pairs
 - We also still face the issue of weighted terms in the loss function



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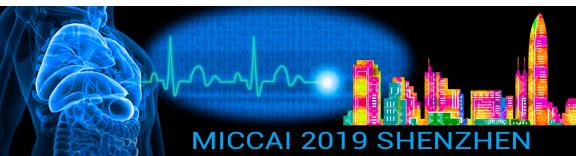
Opportunities and Challenges

- Same **challenges** as for conventional image registration!
- Lack of sufficient validation criteria:
 - Dice coefficient, target registration error, intensity based measures: *these are all surrogate measures*
 - A lot of focus on inter-subject brain registration (due to availability of labelled data)
 - How good is good enough?
- Clinical value is unclear:
 - Perhaps more clear in image guidance and tracking applications
 - But does image registration – and subsequent image fusion – help in clinical decision making?



Opportunities and Challenges

- Danger of standing still! We are currently re-creating of what we had before:
 - Rigid and deformable registration
 - Multi-modal registration
 - Diffeomorphic registration
 - *With often superior results on standard datasets* (LBP40 etc)
- Yet researchers are shying away from some of the hard, clinically more relevant problems:
 - Anatomical and disease variability, normal vs abnormal growth patterns, complex physiological motion, new emerging imaging protocols, motion during image acquisition...



Opportunities and Challenges

- Treating deep learning as a black box in registration (or anywhere else) would be a mistake:
 - Need to incorporate prior knowledge:
 - anatomy, physiology, imaging physics
 - Need to discover and study new knowledge obtained but hidden:
 - motion models, weightings and losses
 - Need to find generic solutions that can transfer between domains
 - E.g. low-dose vs high-dose CT, different MR protocols, different radiotracers...
 - *Within reason!*
 - Need to cater for patient-specific vs population-generic problems*
 - Multi-modal, dynamic or longitudinal registration is patient specific
 - Atlases and cross-sectional registration need to work for a population



*We could make that same argument for disease-specific or organ-specific registration

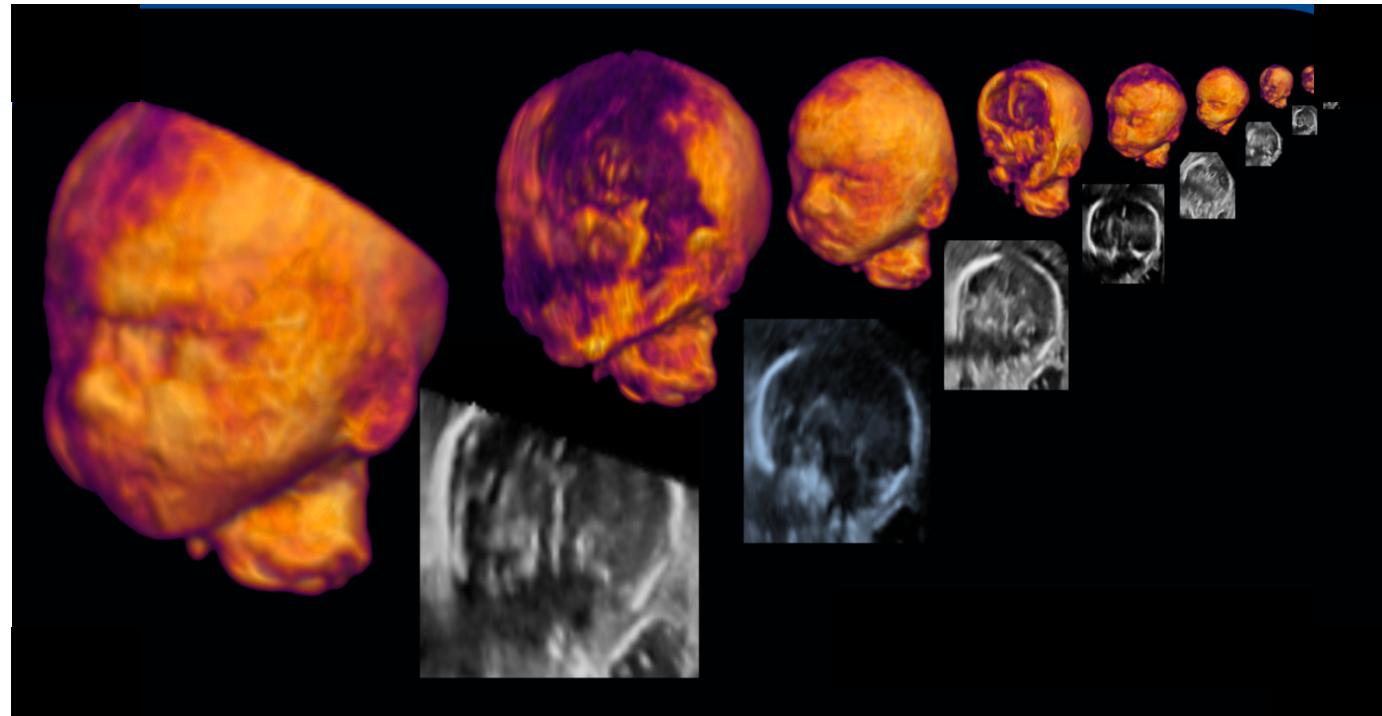
Some opportunities – and challenges – illustrated:

- Case study 1: Fetal head compounding
 - We are following a moving target ...
 - ... with a manually held ultrasound probe ...
 - ... making this highly operator dependent ...
 - ... and don't even mention the artefacts (shadows due to attenuation)!



Some opportunities – and challenges – illustrated:

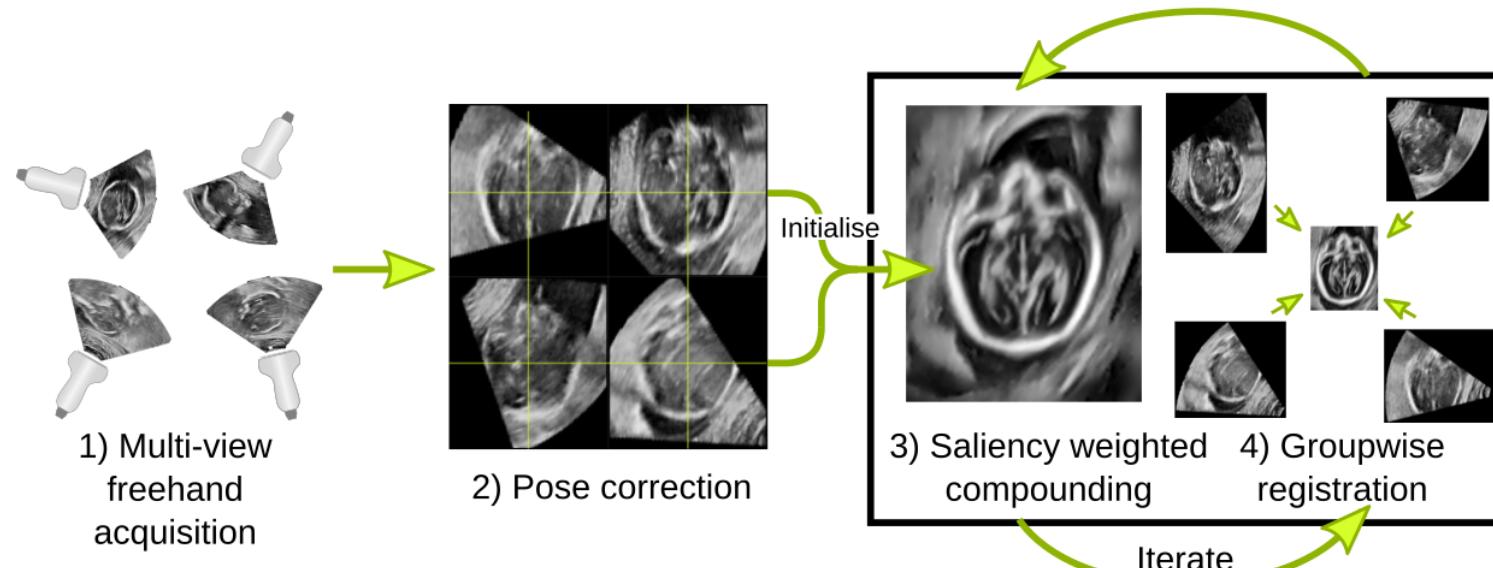
- Case study 1: Fetal head compounding
 - Register the moving head (we can stay rigid)



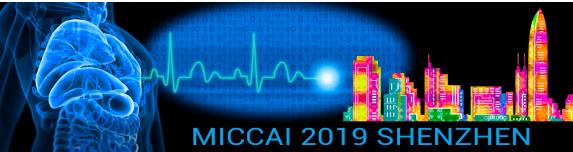
Robert Wright et al. "Complete Fetal Head Compounding from Multi-View 3D Ultrasound", MICCAI 2019

Some opportunities – and challenges – illustrated:

- Case study 1: Fetal head compounding
 - Iterative spatial transformer network gives a precise alignment given any initial orientation.
 - Robust to fast fetal / probe movement where previous methods fail.



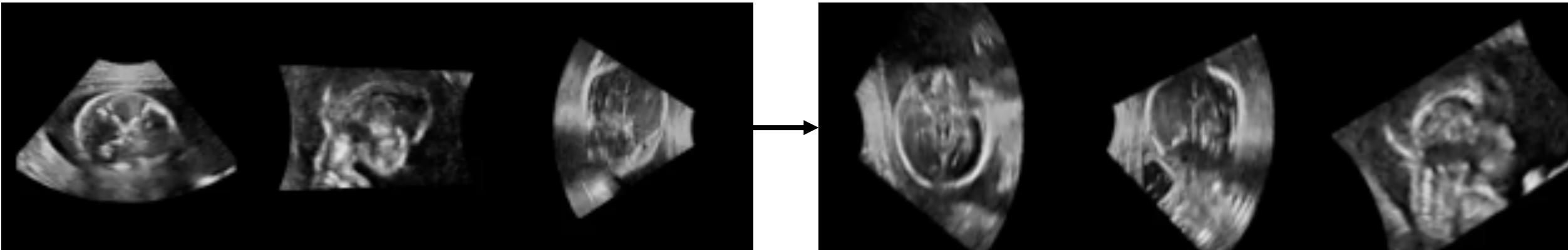
Robert Wright et al. "Complete Fetal Head Compounding from Multi-View 3D Ultrasound", MICCAI 2019



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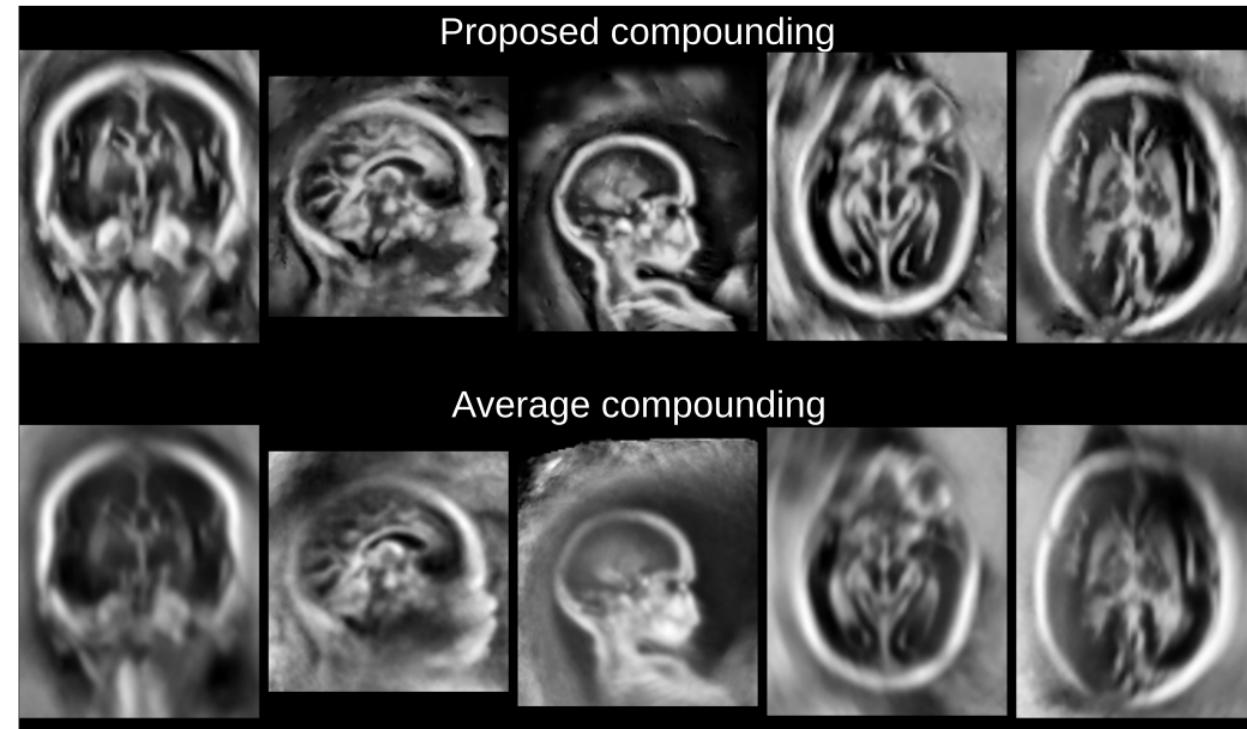
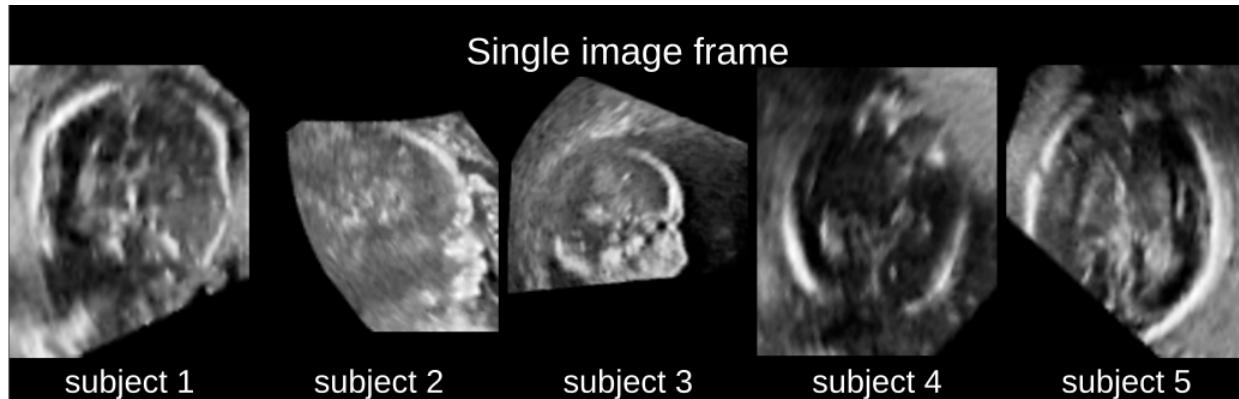
Some opportunities – and challenges – illustrated:

- Case study 1: Fetal head compounding
 - Pose correction of multi-view images:



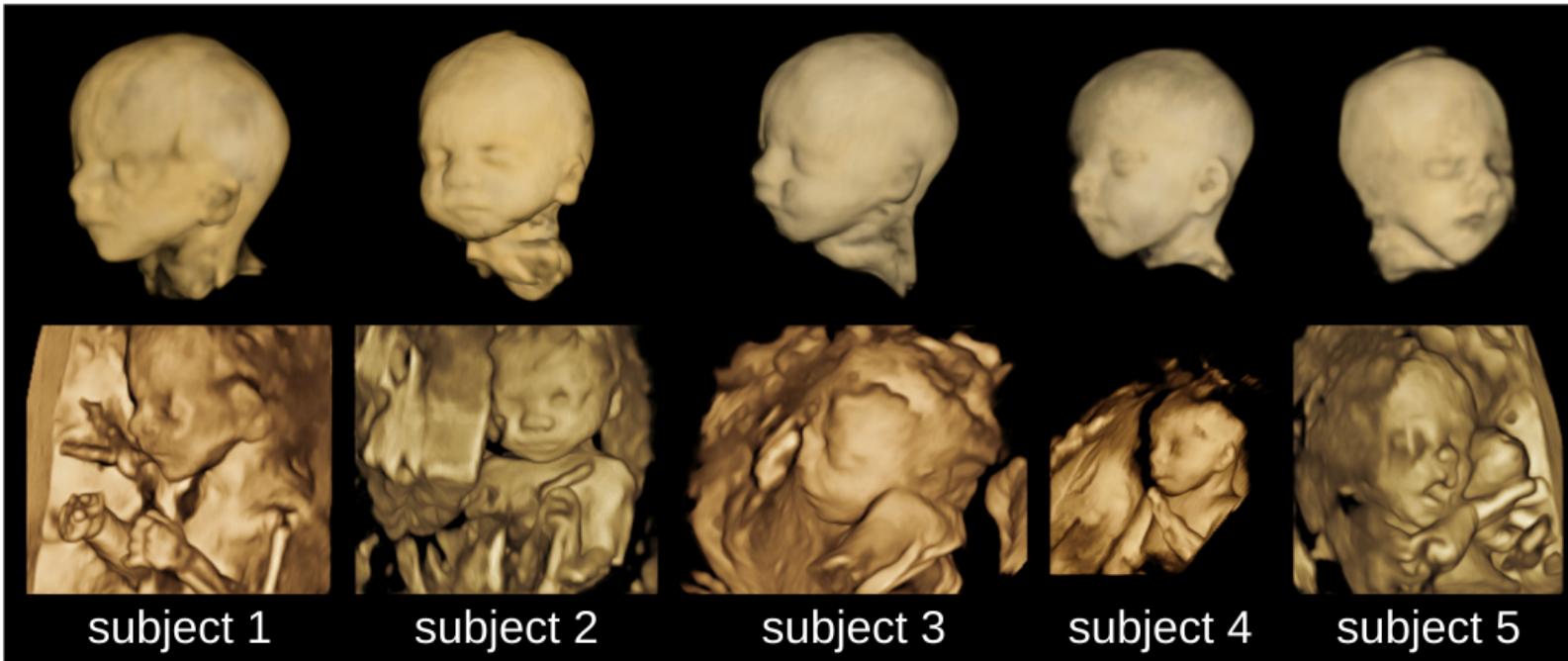
Some opportunities – and challenges – illustrated:

- Case study 1: Fetal head compounding
 - Iterative compounding with saliency weighting:



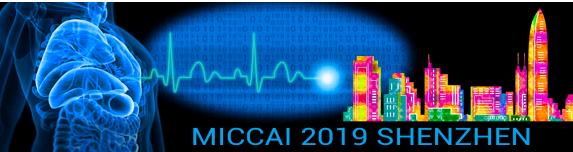
Some opportunities – and challenges – illustrated:

- Case study 1: Fetal head compounding
 - Rendering result for visualization:



Some opportunities – and challenges – illustrated:

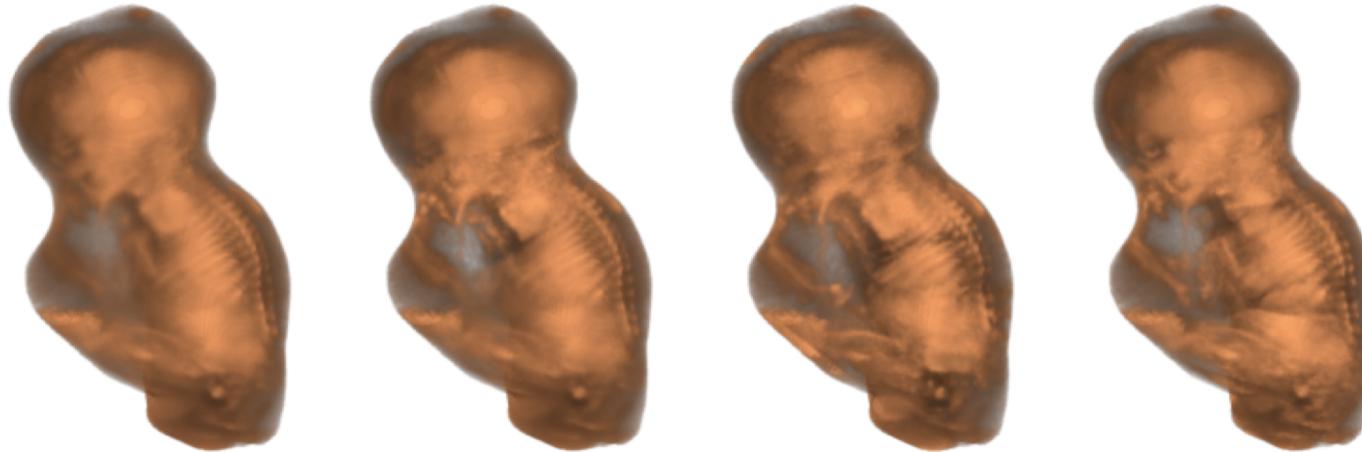
- Case study 2: Fetal head compounding
 - Remaining challenges:
 - Ultrasound frame rate – ultimately we want to do this real-time
 - Ultrasound distortion – currently solved with conventional, low-parameter B-spline registration
 - Pose correction still requires initial registration to an atlas – which has been formed by conventional registration
 - Small deformations of fetus – swallowing fluid, opening/closing lips, ...
 - Large deformations of fetus – moving limbs near face, sucking thumb, ...



Robert Wright et al. "Complete Fetal Head Compounding from Multi-View 3D Ultrasound", MICCAI 2019

Some opportunities – and challenges – illustrated:

- Case study 2: Whole fetal body compounding
 - We are still following a moving target with ultrasound ...
 - ... which is way too big to image in one go ...
 - ... and now also moving limbs and doing summersaults!



Thank you!



<- Today at 16:30
MLMIR El Chino 4



Contact me (julia.schnabel@kcl.ac.uk) on:

- PhD places available in Smart Medical Imaging
- PostDoc positions available in AI for early detection of disease and prognosis of outcome

(Some even require novel registration!)

