



A spatio-temporal atlas of the growing brain for fMRI studies

M. Murgasova, V. Doria, L. Srinivasan, P.
Aljabar, D. Edwards, D. Rueckert

Department of Computing
Imaging Sciences Department

Imperial College London
United Kingdom



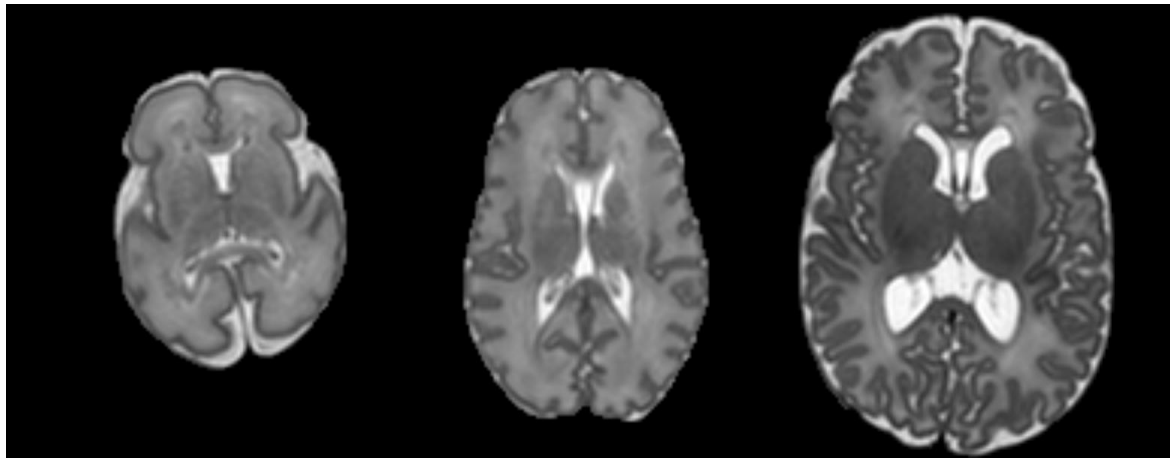
Outline

- Motivation
- Average models of brain and brain growth
- B-spline transformations for modeling average brain growth
- Anatomical template of the growing brain
- Visualising development of brain networks from resting state fMRI
- Conclusion



Motivation

- The aim: to understand brain development during the 3rd trimester of pregnancy
- Specifically: development of activation maps such as default mode and motor network
- Challenge: data scattered across different subjects and ages



29 weeks GA

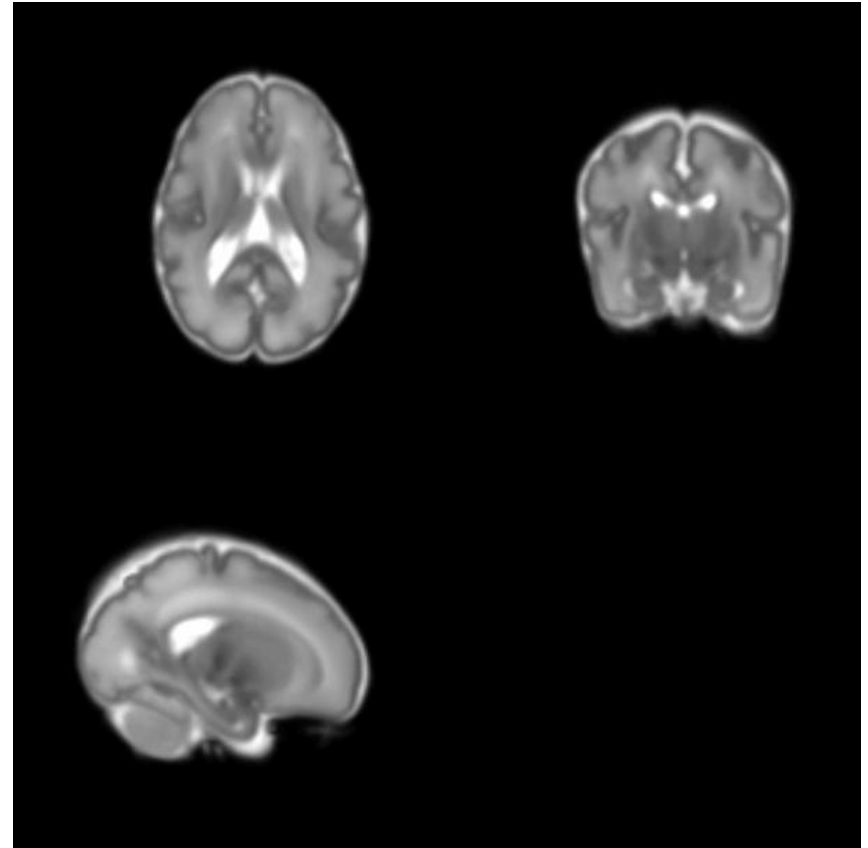
35 weeks GA

42 weeks GA



4D model of average growth

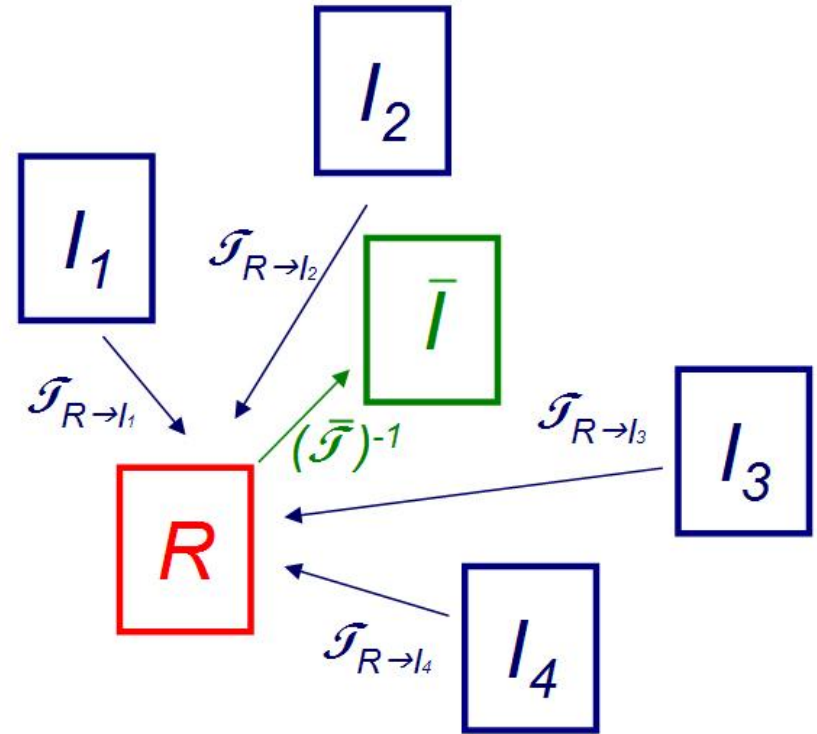
- The aim:
 - to construct average 4D spatio-temporal model of growing brain (3D for brain anatomy + 1D for age)
- Solution:
 1. Construct the average anatomy of growing brain
 2. Create average model from additional information attached to the anatomy



Non-rigid template
B-spline registration 2.5 mm
(103 images 28-44 weeks GA)

Average models of brain (3D)

- To visualise average brain anatomy
- Training images are registered to a reference subject
- Inverse of the average transformation is used to find the average brain space
- Training images are transformed to the average space and their intensities averaged



$$\bar{\mathcal{T}} = \frac{1}{n} \sum_{k=1}^n \mathcal{T}_{R \rightarrow I_k}$$



Average models of brain growth (4D)

- To calculate age-dependent average anatomical templates, average transformation is replaced by age-dependent weighted average:

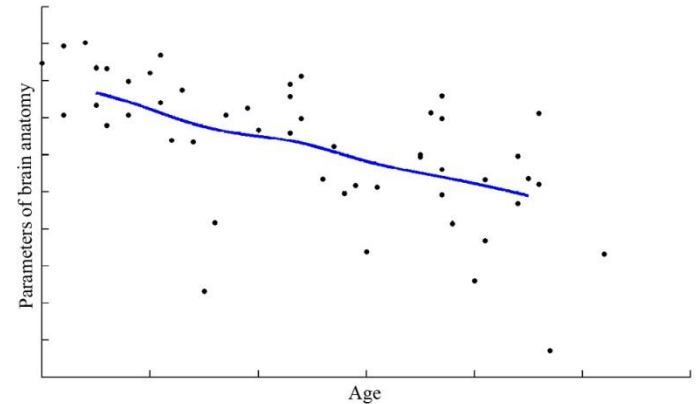
$$\bar{\mathcal{T}}(t) = \frac{\sum_{k=1}^n w(t_k, t) \mathcal{T}_{R \rightarrow I_k}}{\sum_{k=1}^n w(t_k, t)}$$

$$w(t_k, t) = \frac{1}{\sigma \sqrt{2\pi}} \exp \frac{-(t_k - t)^2}{2\sigma^2}$$

Age of the subject (green arrow pointing to t_k)
 Age for the template (red arrow pointing to t)

Template for the age t (blue arrow pointing to $\bar{\mathcal{T}}(t)$)

$$\bar{I}(t) = \frac{\sum_{k=1}^n w(t_k, t) I_k \circ \mathcal{T}_{R \rightarrow I_k} \circ \bar{\mathcal{T}}^{-1}(t)}{\sum_{i=1}^n w(t_k, t)}$$





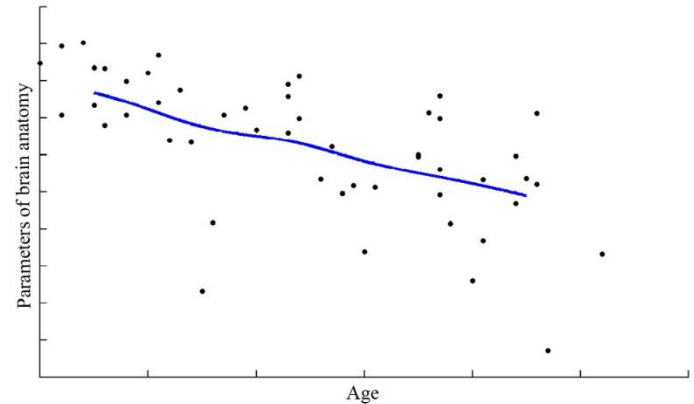
Average models of brain growth (4D)

- To calculate age-dependent average anatomical templates, average transformation is replaced by age-dependent weighted average:

$$\bar{T}(t) = \frac{\sum_{k=1}^n w(t_k, t) \mathcal{T}_{R \rightarrow I_k}}{\sum_{k=1}^n w(t_k, t)}$$

$$w(t_k, t) = \frac{1}{\sigma \sqrt{2\pi}} \exp \frac{-(t_k - t)^2}{2\sigma^2}$$

Age of the subject (green arrow pointing to t)
 Age for the template (red arrow pointing to t_k)



Template for the age t (blue arrow pointing to $\bar{I}(t)$)

Registration? (red circle around $\mathcal{T}_{R \rightarrow I_k}$)

$$\bar{I}(t) = \frac{\sum_{k=1}^n w(t_k, t) I_k \circ \mathcal{T}_{R \rightarrow I_k} \circ \bar{T}^{-1}(t)}{\sum_{i=1}^n w(t_k, t)}$$



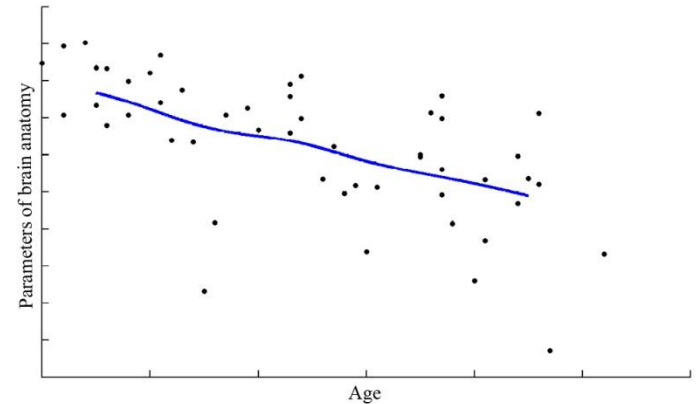
Average models of brain growth (4D)

- To calculate age-dependent average anatomical templates, average transformation is replaced by age-dependent weighted average:

$$\bar{T}(t) = \frac{\sum_{k=1}^n w(t_k, t) \mathcal{T}_{R \rightarrow I_k}}{\sum_{k=1}^n w(t_k, t)}$$

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Template for the age t (blue arrow pointing to $\bar{I}(t)$)

$$\bar{I}(t) = \frac{\sum_{k=1}^n w(t_k, t) I_k \circ \mathcal{T}_{R \rightarrow I_k} \circ \bar{T}^{-1}(t)}{\sum_{i=1}^n w(t_k, t)}$$

Registration? (red circle around $\mathcal{T}_{R \rightarrow I_k}$)

Averaging transformations? (orange circle around $\bar{T}^{-1}(t)$)



Registration

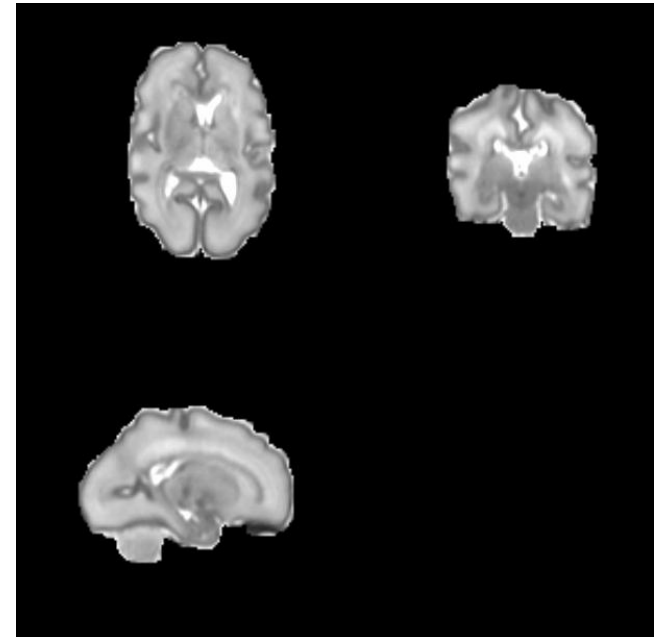
- We used B-spline FFD registration (Rueckert *et al.* 1999)
- Transformation model:

$$T(v) = A(v) + B(v)$$

Global affine component: 12 parameters (translation, rotation, scaling and skewing)

Local component – B-spline

- Similarity measure:
 - Normalised mutual information (NMI)
 - Challenge: NMI too weak – boundaries of the brain not always aligned correctly
 - Solution: removing extra-cranial CSF



Longitudinal registration: subject at 30, 34 and 40 weeks GA



Model for the transformation

- Global shape change is not important when assessing the development of activation network => images are affinely aligned in pre-processing step
- We model only local shape change using B-spline transformation:

$$\mathcal{T}(v) = \sum_i B_i N_i(v)$$

Control points

Basis functions

- All transformations are defined on the same lattice in the space of the reference image



Averaging B-spline transformations

- We assume small deformation setting
- Transformations defined on the same lattice
=> the averaging of B-spline transformations can be approximated by averaging the control points:

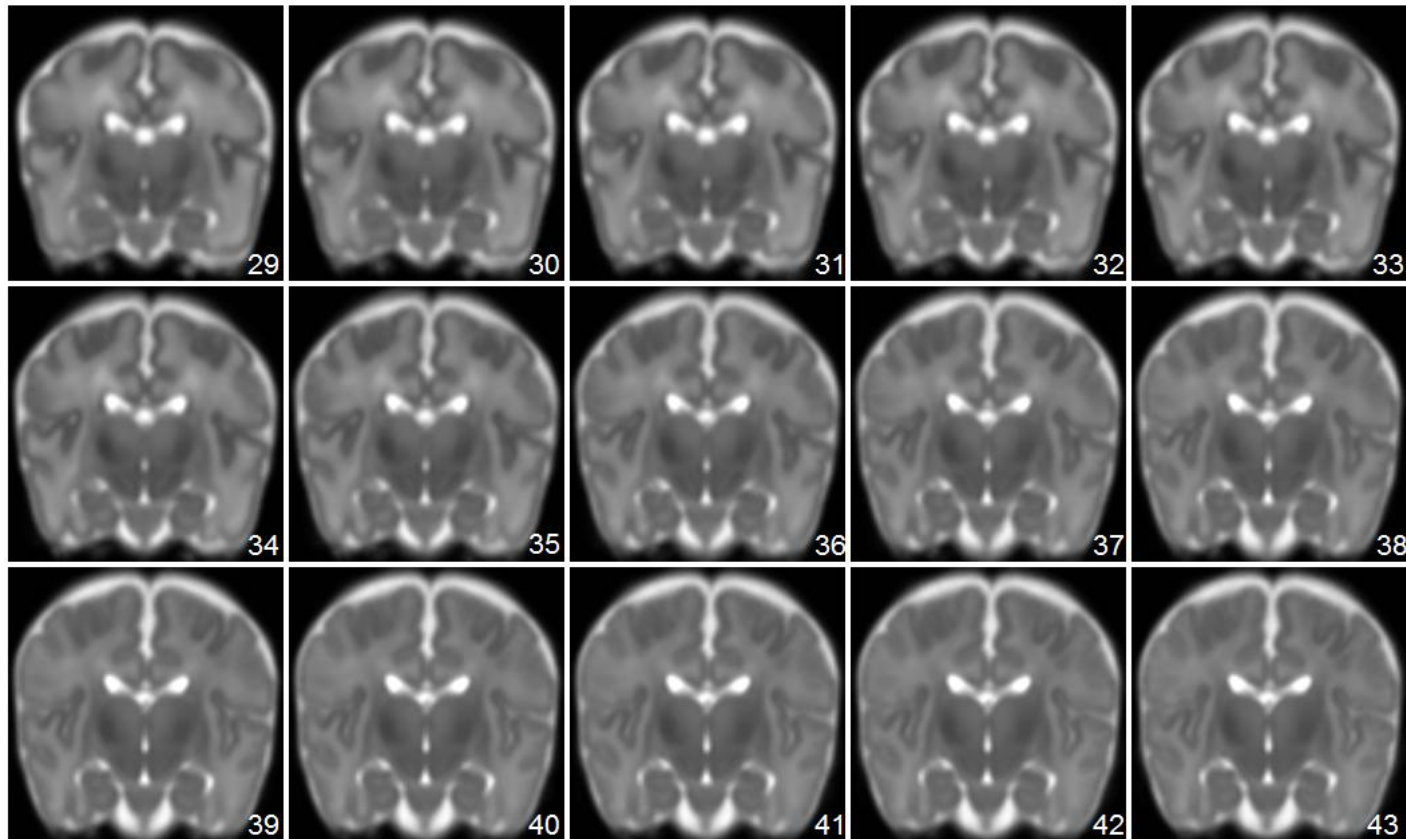
$$\bar{T}(v) = \sum_i \bar{B}_i N_i(v), \quad \bar{B}_i = \frac{1}{n} \sum_{k=1}^n B_{i,k}$$

- Inverse transformation is not a B-spline, but can be numerically approximated as a B-spline



Anatomical template of growing brain

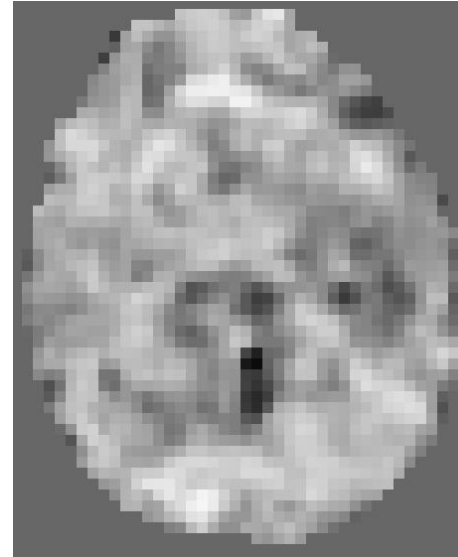
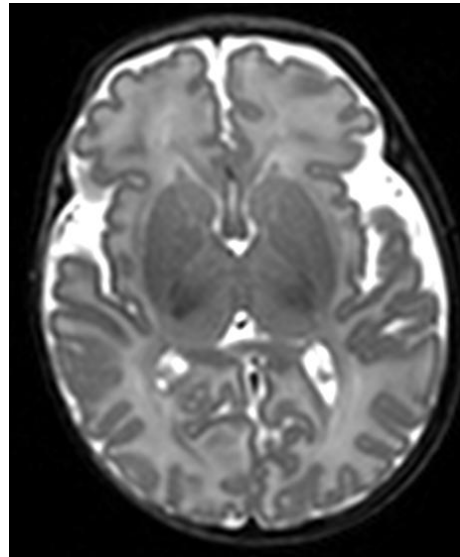
- Created using 103 T2-weighted MR images of premature babies with ages 28-44 weeks GA
- B-spline control point spacing 5mm
- Standard deviation 2 weeks for calculating weights





fMRI data

- 39 subjects (29-44 weeks GA) consisting of anatomical template and aligned resting-state fMRI data
- Correlation maps were estimated from fMRI data by placing the seed in left motor cortex/medial prefrontal cortex and calculating correlation of time-series with the rest of the brain using FSL





Creating average activation maps

- The 4D anatomical template has been created
- Only 39 fMRI datasets => we chose to use average template from 103 images as a basis for the model
- The anatomical images with the attached correlation maps can now be registered to the average anatomical template to create average 4D correlation maps



Creating average activation maps

- 39 anatomical images were registered to the average template of corresponding age
- Corresponding correlation maps were then transferred not only to the corresponding template but also to neighbouring frames => smooth model
- Average time-dependent correlation maps were obtained by weighted averaging at each time-point

Average correlation map at time-point t

Correlation map

to the corresponding template frame

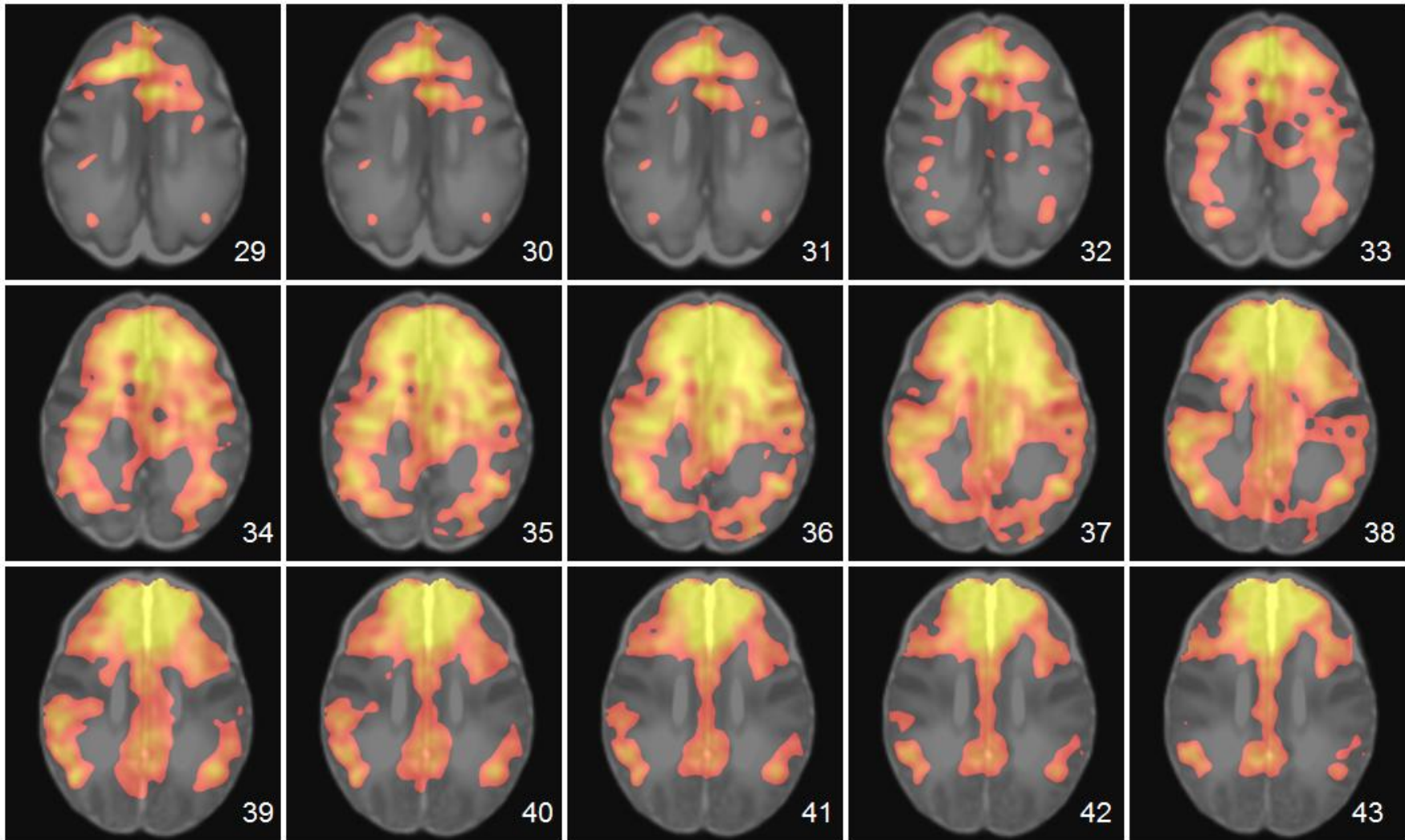
to the reference space

to the template frame at time-point t

$$\bar{C}(t) = \frac{\sum_{k=1}^m w(t_k, t) C_k \circ \mathcal{T}_{\bar{I}(t_k) \rightarrow F_k} \circ \bar{\mathcal{T}}(t_k) \circ \bar{\mathcal{T}}(t)^{-1}}{\sum_{i=k}^m w(t_k, t)}$$

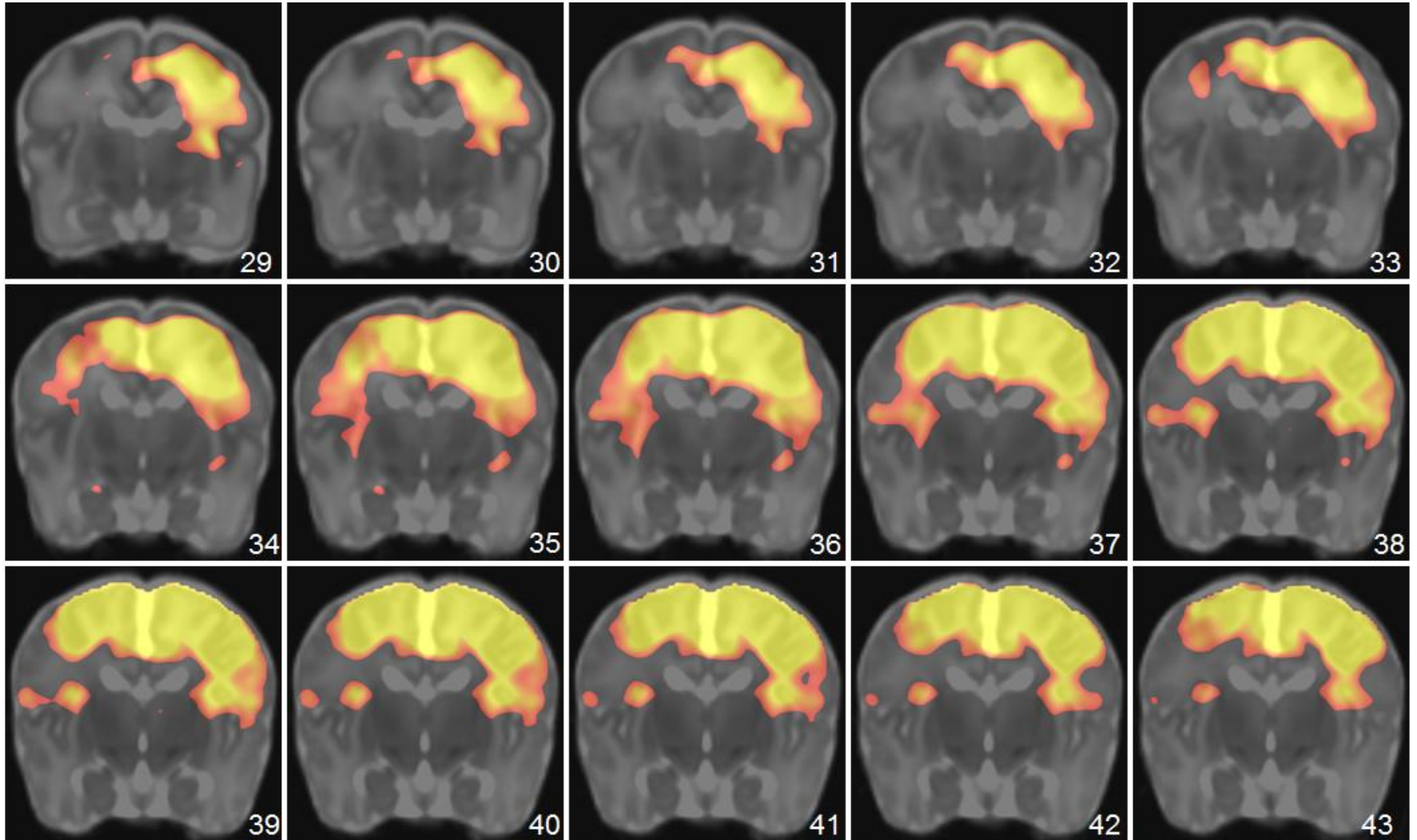


Model of developing default network





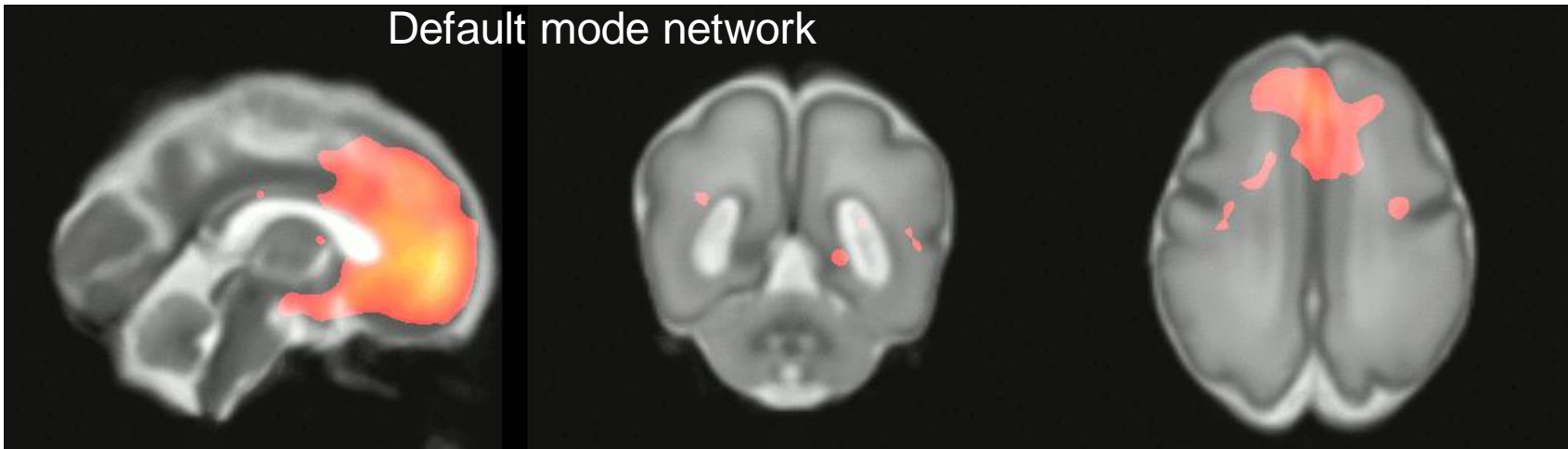
Model of developing motor network



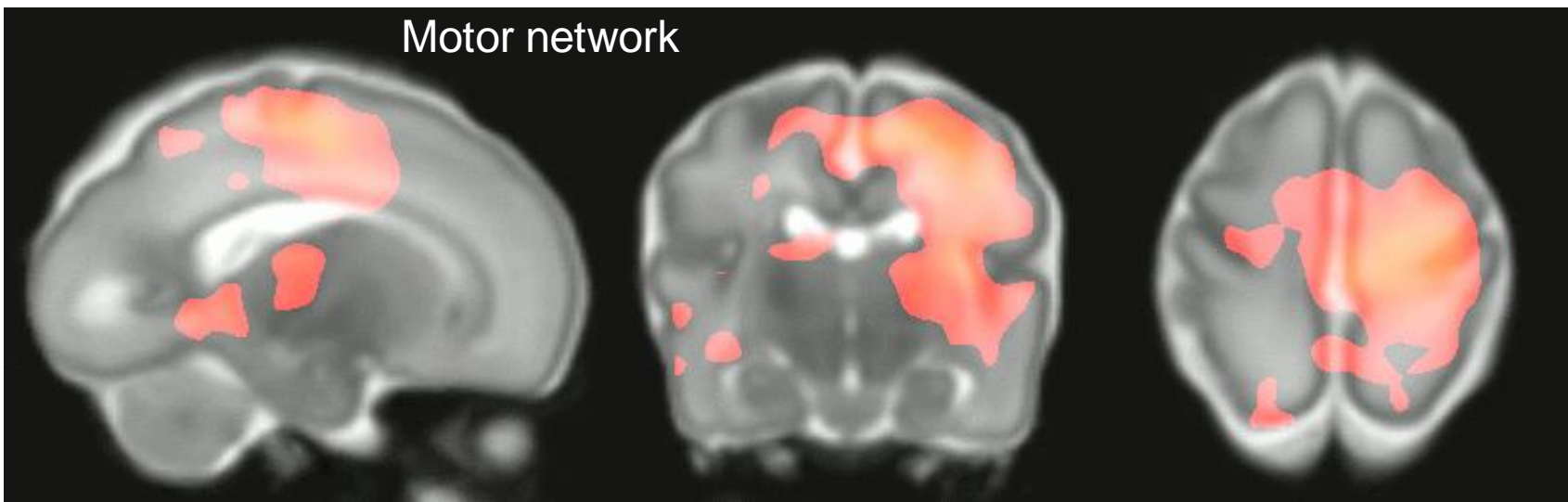


Models of developing brain networks

Default mode network



Motor network





Conclusion

- The presented model revealed the development of brain networks during 3rd trimester of pregnancy
- The results have shown for the first time the presence of default mode network before birth
- The detailed paper by Valentina Doria et al. describing this important medical finding and more statistical analysis is under review
- Future work: Including more data in the model and validation



Thank you for your attention

4D probabilistic atlas

- Similar methodology with affine registration is being used to build 4D brain tissue probability maps (still under construction)

