



# fMRI analysis



GitHub <a href="https://github.com/dcdace/fMRI\_training">https://github.com/dcdace/fMRI\_training</a>

# Dace Apšvalka [Datza]



Winter, 2023





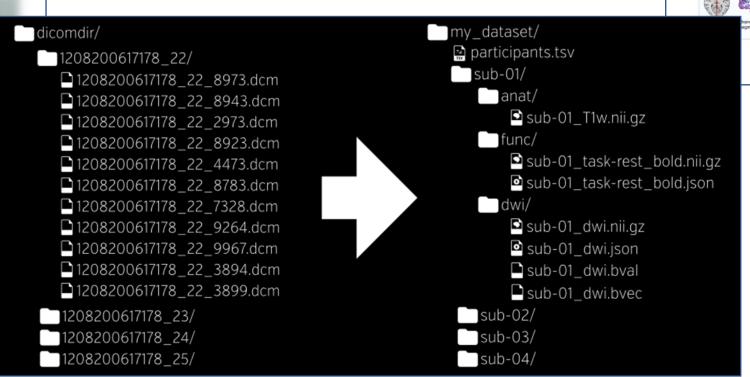
Pre-process

#### Collect the data

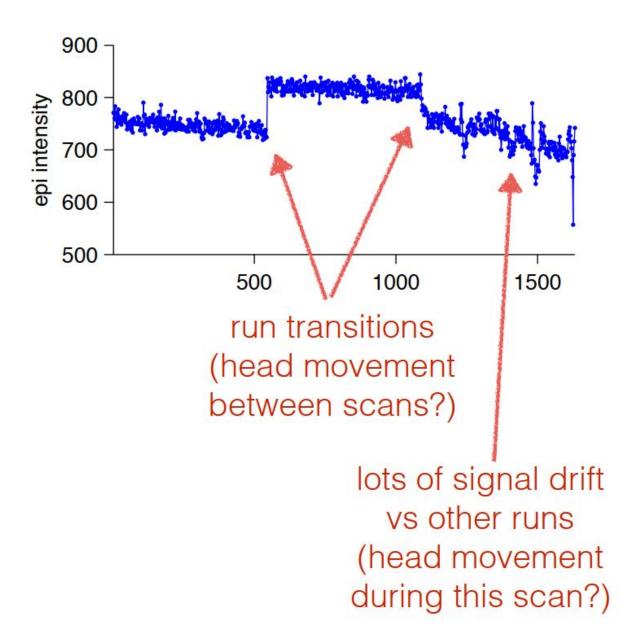


#### Why can't we analyse right away?

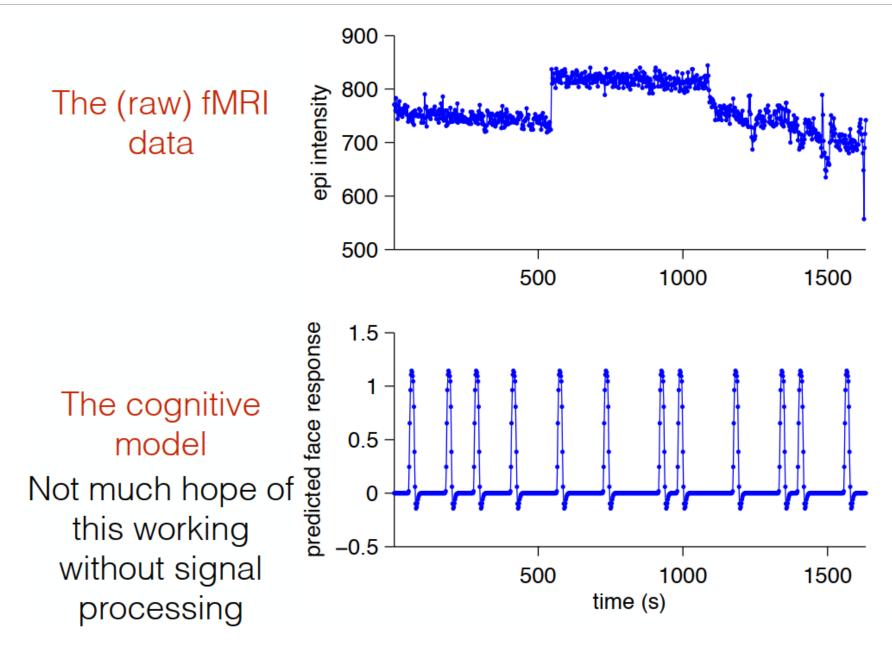
Pre-process & Analyse

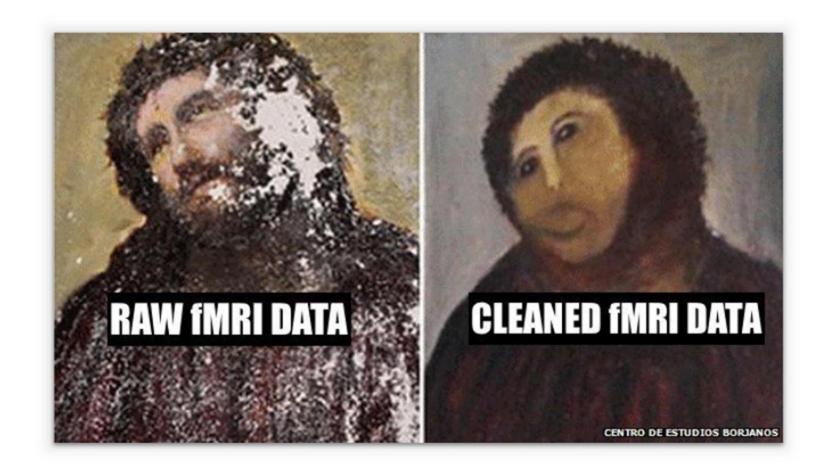


# Raw data in example voxel



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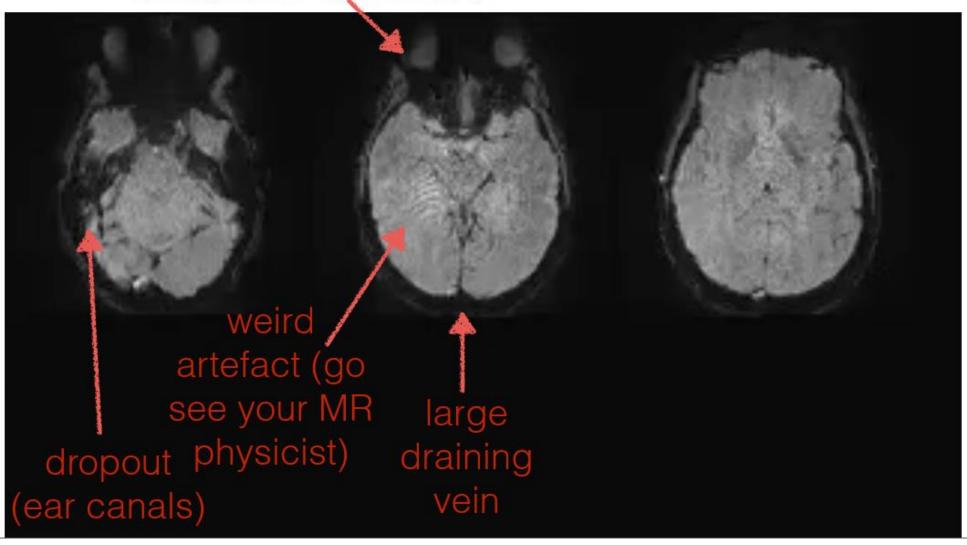


- Artefacts
- Head motion
- Slice timing
- Brain differences
- Noise
- ... and many others

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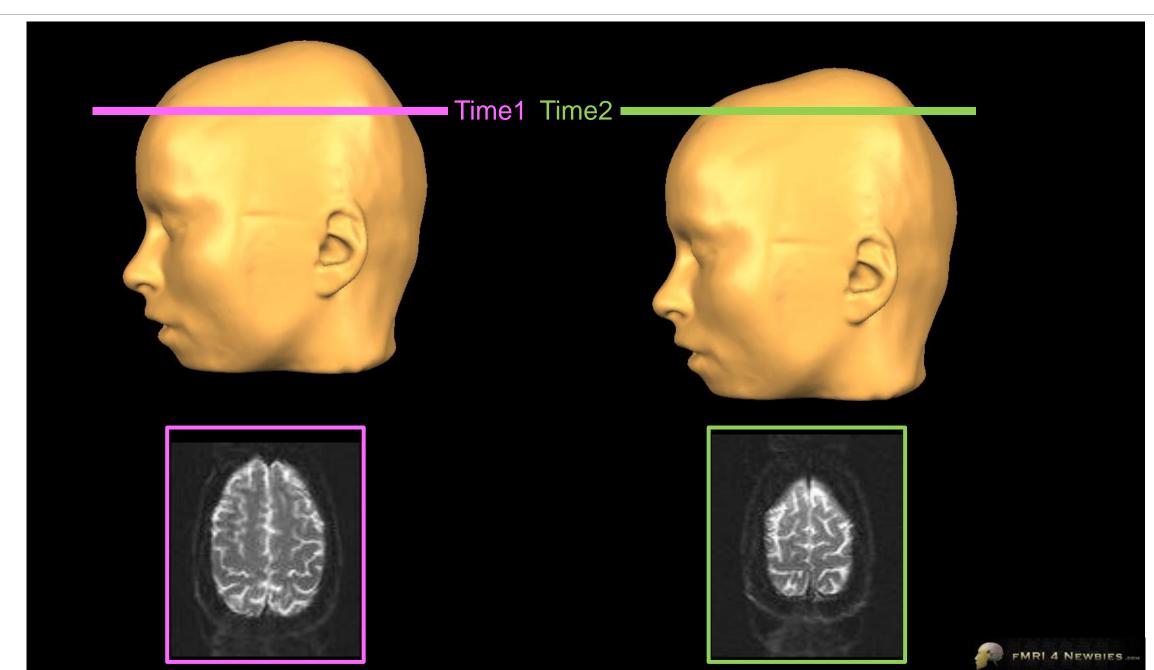
### Artefacts

eye blink / movement effect (not just in eye balls!)



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# Head motion



### Head motion correction

- Prevention is the best remedy
- Tell your subjects how to be good subjects
- Make sure the subject is comfortable going in
- Discourage any movements that would displace the head between scans



Bite bar



Thermoplastic mask

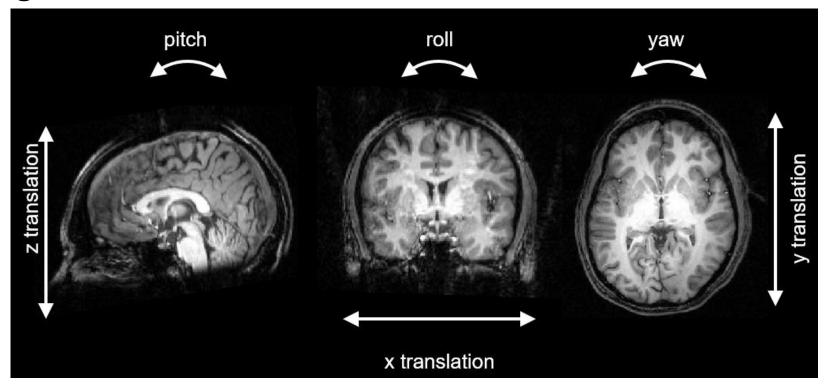


Vacuum pack

Often a bunch of foam padding works as well as anything

#### Head motion correction

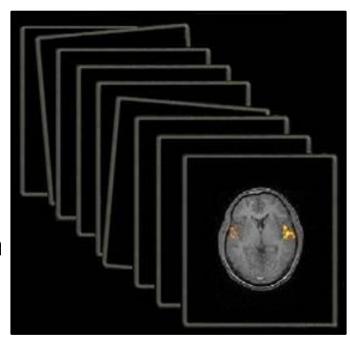
Spatial realignment



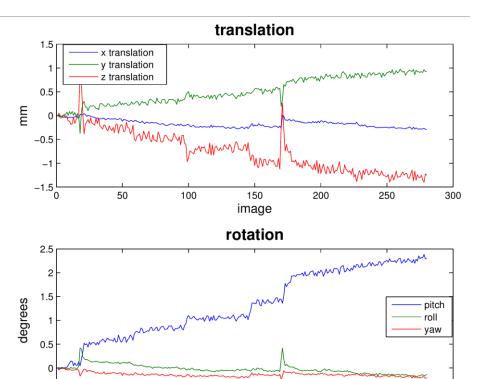
- Most algorithms assume a rigid body (= the brain doesn't deform with movement)
- Align each volume of the brain to a target volume using 6 parameters: 3 translations and 3 rotations
- Target volume: e.g., the mean image of the first run

### Head motion correction

- Spatial realignment
  - Not expected to work perfectly, but it helps
  - Be suspicious of data with a lot of head movement!



Translation (mm)		Rotation (rad)		
У	Z	Pitch	Roll	Yaw
0.3348	1.8892	0.0182	0.0065	0.001
0.418	1.5832	0.0165	0.0063	0.0061
0.4925	0.8524	0.0234	0.0123	0.0269
1.5353	1.8285	0.0327	0.0222	0.0939
0.3014	0.7574	0.0043	0.0197	0.0094
0.9508	1.0428	0.026	0.0171	0.0164
	y 0.3348 0.418 0.4925 1.5353 0.3014	y z  0.3348 1.8892 0.418 1.5832 0.4925 0.8524 1.5353 1.8285 0.3014 0.7574	y z Pitch  0.3348 1.8892 0.0182  0.418 1.5832 0.0165  0.4925 0.8524 0.0234  1.5353 1.8285 0.0327  0.3014 0.7574 0.0043	y     z     Pitch     Roll       0.3348     1.8892     0.0182     0.0065       0.418     1.5832     0.0165     0.0063       0.4925     0.8524     0.0234     0.0123       1.5353     1.8285     0.0327     0.0222       0.3014     0.7574     0.0043     0.0197



150

image

200

250

300

50

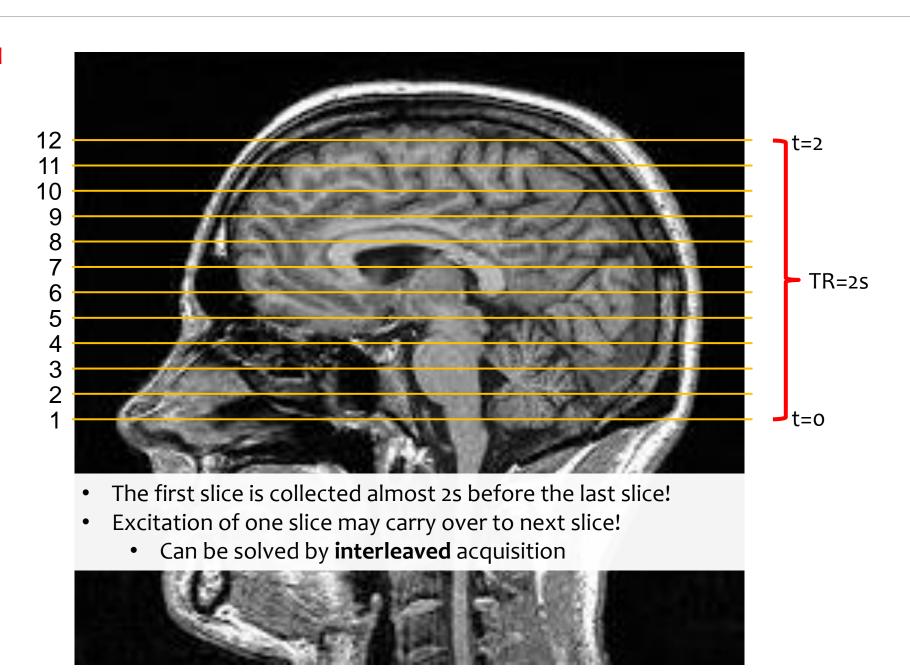
100

- Artefacts
- Head motion
- Slice timing
- Brain differences
- Noise
- ... and many others

# Slice timing

The fMRI data are acquired slice-by-slice – not whole brain all at once!

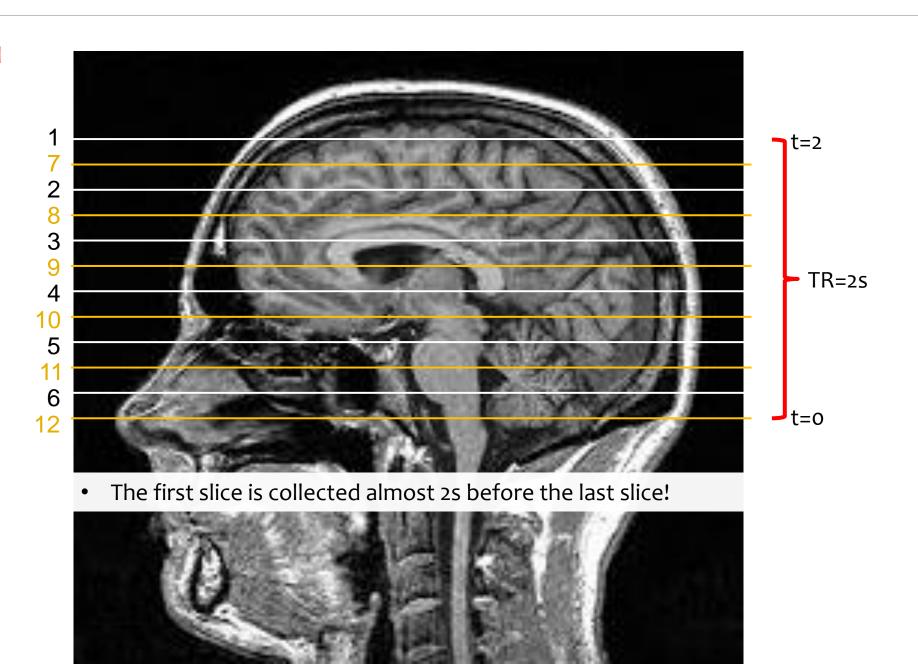
Order: **Ascending** 



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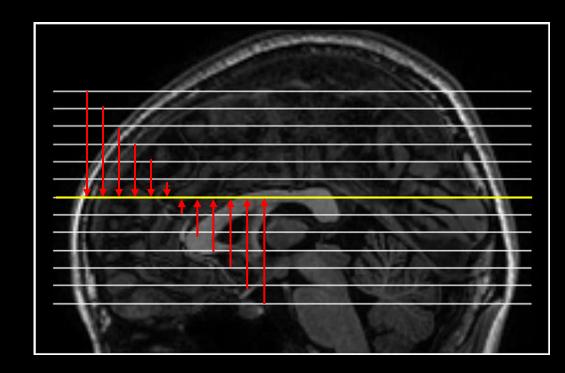
**Descending-Interleaved** 

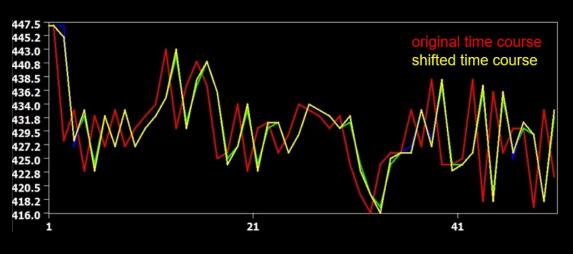


#### Slice time correction

### Temporal interpolation

- Adjusts the timings of the slices so that it is as if they had been collected simultaneously
- Uses information from nearby time points to estimate the amplitude of the MR signal at the onset of each TR
- Key parameters
  - Reference slice
    - The only slice that is not interpolated
    - Typical choice is to use the middle slice
  - Slice order
  - TR
- No need to slice time correct if TR is very short (e.g., multi-band: multiple slices at the same time)







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# Brain differences – between subjects

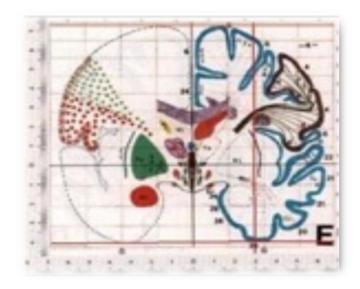
- Brain size can differ (up to 30%)
- Brain shape can differ

- Solution: normalise to a standard template
  - stretching, squeezing and warping each brain to the "standard brain"

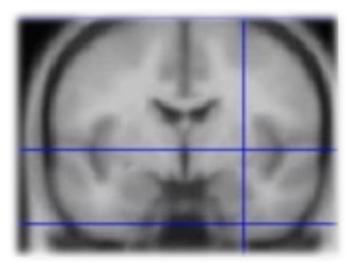


# Standard brain template

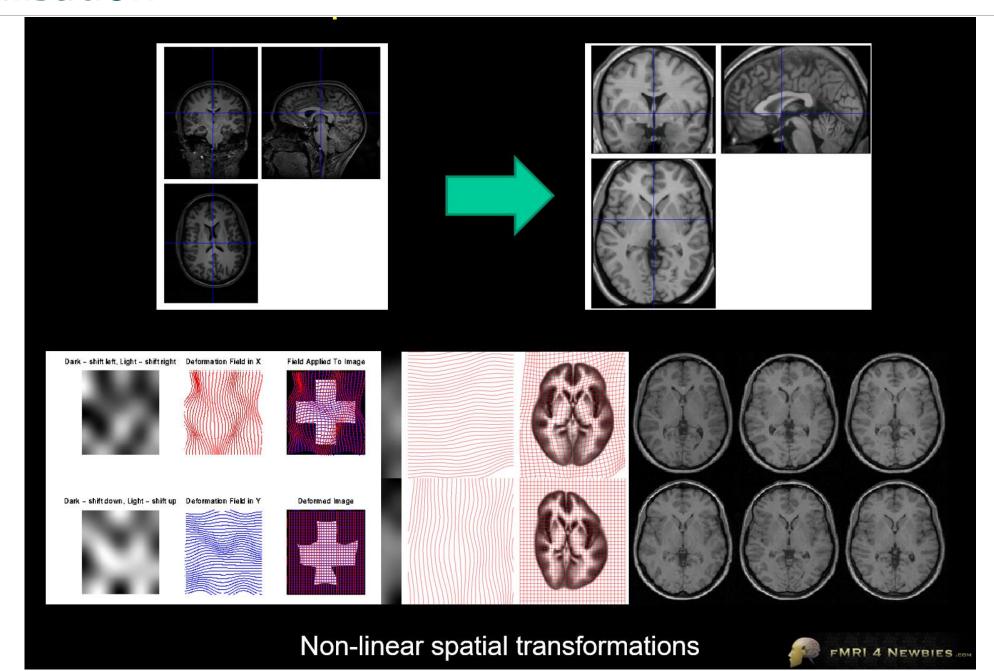
- Talairach space
  - From atlas of Talairach and Tournoux (1988)
  - Based on single subject (60y, Female, Cadaver)



- MNI Montreal Neurological Institute
  - Average scan of 152 'normal' right-handed subjects
  - Approximated to Talairach space
  - Slightly larger



# Normalisation



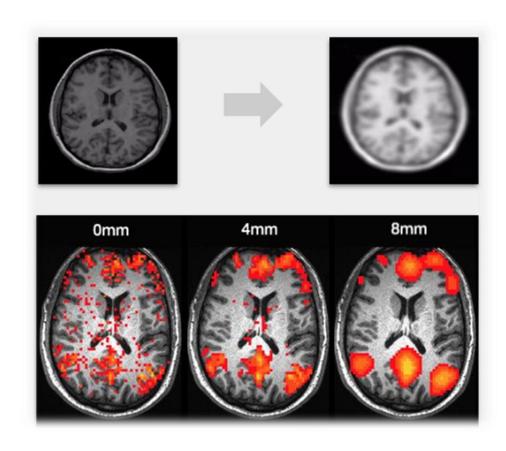
# Spatial smoothing

 Helps make residual errors in previous spatial pre-processing stages less problematic (e.g., imperfect normalisation, motion correction)

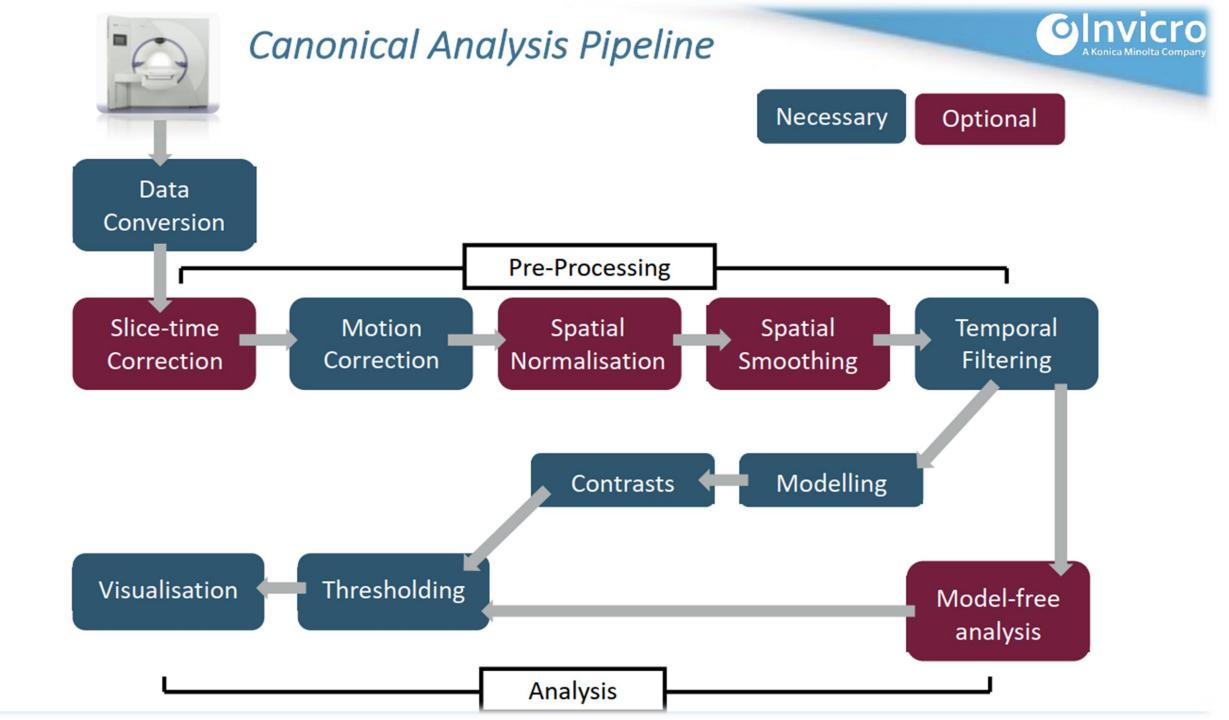
Can increase signal-to-noise ratio

 Necessary for validity of multiple corrections within random field theory

But it reduces the image resolution



- Artefacts
- Head motion
- Slice timing
- Brain differences
- Noise can be modelled in GLM
- ... and many others



# Pre-processing tools

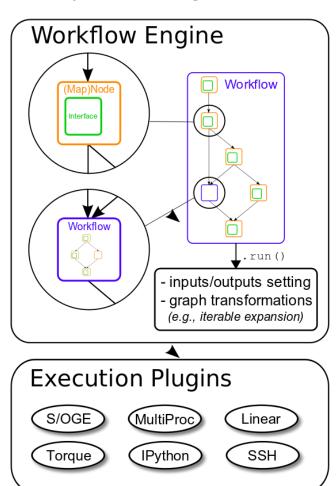


# Pre-processing tools



#### Interfaces Uniform Python API SPM **FSL** FreeSurfer Interface Interface Interface FreeSurfer SPM **FSL** (Command-line (Command-line (Matlab functions) programs) programs)

Idiosyncratic, Heterogeneous APIs



# Pre-processing tools

- fMRIPrep <a href="https://fmriprep.org/en/stable/">https://fmriprep.org/en/stable/</a>
  - Fully automated fMRI data pre-processing tool
  - The workflow is based on Nipype and encompasses a large set of tools from well-known neuroimaging packages, including FSL, ANTs, FreeSurfer, AFNI, and Nilearn. This pipeline was designed to provide the best software implementation for each state of pre-processing, and will be updated as newer and better neuroimaging software becomes available.
  - **Robustness** The pipeline adapts the pre-processing steps depending on the input dataset and should provide results as good as possible independently of scanner make, scanning parameters or presence of additional correction scans (such as fieldmaps).
  - Ease of use Thanks to dependence on the BIDS standard, manual parameter input is reduced to a minimum, allowing the pipeline to run in an automatic fashion.
  - "Glass box" philosophy Automation should not mean that one should not visually inspect the results or understand the methods. Thus, fMRIPrep provides visual reports for each subject, detailing the accuracy of the most important processing steps. This, combined with the documentation, can help researchers to understand the process and decide which subjects should be kept for the group level analysis.

# **fMRIPrep**



T1-weighted

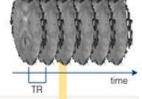
One or more (e.g. in longitudinal studies) T1w images





**BOLD** run

One run of one task (or resting-state) time-series of blood-oxygen level (BOLD) measurements





**Fuse & Conform** 

All T1w images are aligned and averaged to form a 3D reference image NIfTI headers are checked for validity



#### **INU Correction**

The T1w reference is run through the N4 algorithm to correct for intensity nonuniformity (INU)



#### Skull-stripping

Atlas-based brain extraction is performed on the reference T1w image



Functional preprocessing

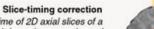
#### Generate reference & brain mask

Time-points showing non-steady state artifacts (excess of T1 contrast) are aligned and averaged to generate a reference image in native space



#### Estimation of head-motion

Parameters representing bulk head motion (due to involuntary drift, swallowing, etc.) of each timepoint with respect to the reference are estimated



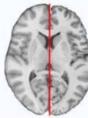
(Optional) When the acquisition time of 2D axial slices of a given timepoint is available, temporal dynamics are estimated and all slices resampled to the mid-timepoint of that TR



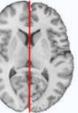


Brain atlas

Default: MNI152 nonlinear asymmetric v2009c

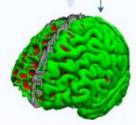


Spatial normalization Non-linear, spatial



Brain tissue segmentation

The brain-extracted image is alignment to the brain classified into CSF, GM and WM

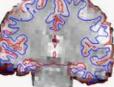


Surface reconstruction

Surfaces of the cortical sheet are reconstructed from the anatomical information (T1w reference, T2w)



Registers activity in BOLD voxels to anatomical location



#### Susceptibility distortion estimation

(Optional) Find a deformation field that compensates for the distortion, when adequate acquisitions are present



#### Sample on surface

Sample the BOLD signal on the surfaces reconstructed from the anatomical data



#### Sample in template

Resample the BOLD signal in atlas-space, concatenating all pertinent transformations

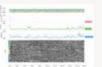


resampling of the BOLD signal in its original grid, applying corrections





Calculate and store nuisance regressors such as noise components, motion parameters, global signals, etc.







Let's see the 03\_nb\_Quality-Control-and-Preprocessing.ipynb notebook

Pre-process