Brain Imaging Data Structure (BIDS) tutorial

StratNeuro Retreat 2024 - Exclusive day for PhD students

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KØBENHAVNS UNIVERSITET



What do you hope to learn from the tutorial?

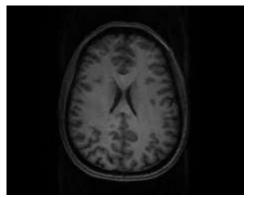
- How to structure MRI data in a good way/ How to improve data structuring / Learn a method to organize data that I can apply in future PostDoc positions (even if different field from mine now)
- I expect to get started with BIDS since I do not have previous experience and I might need it in the future / To learn how to work with BIDS
- Apply BIDS for PET with focus on specificity of PET metadata
 Learn more on MRI and MEG BIDS
- "I'm interested in multimodal method, and one of the modality I wish to use in the future is brain imaging. I have interacted with EEG data before but not PET or other imaging data.
- So I wish to learn the basic of dealing with brain imaging and the analyse techniques."

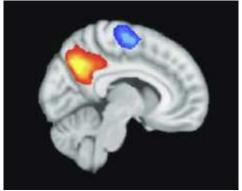


Do you have any specific questions or topics?

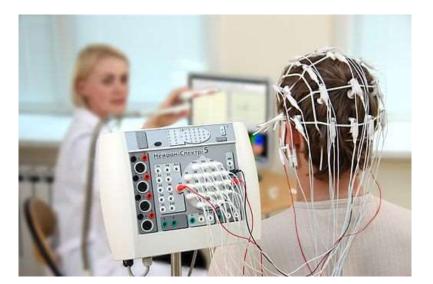
- What is the optimal way to structure repeat scans for the same individual(s)?
 - BIDS specification uses sessions (like baseline and followup) or runs (test-retest like re-scans)
 - See definitions here: https://bids-specification.readthedocs.io/en/stable/common-principles.html
- How BIDS can be useful in machine learning applications if applicable
 - Let's take a little detour...

Current neuroimaging in a nutshell

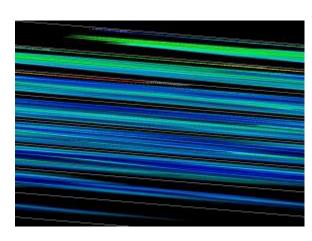


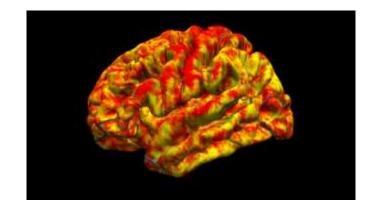


Magnetic resonance imaging (MRI) – both structural and functional (fMRI)



Electroencephalography (EEG)

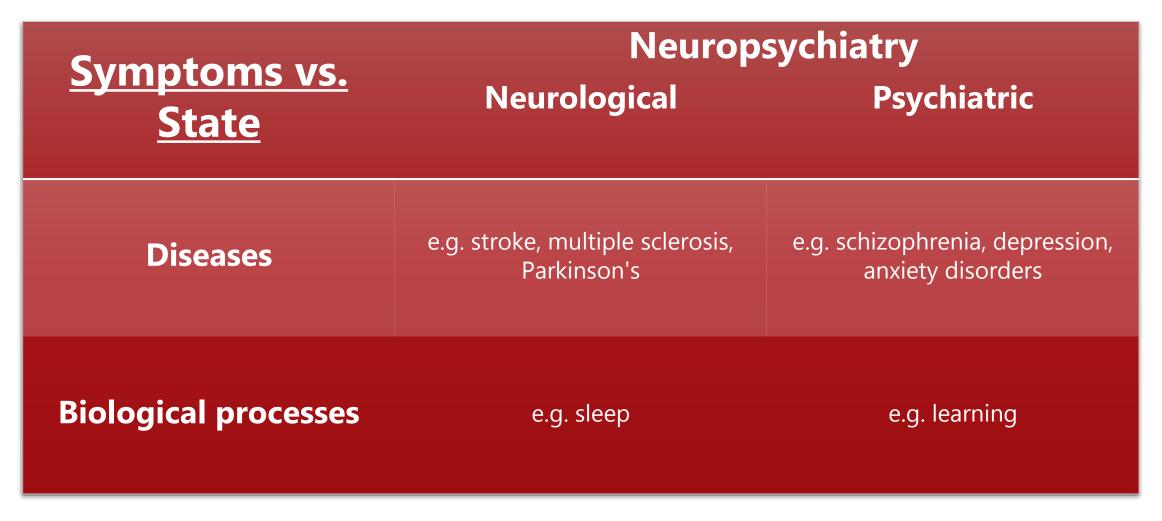




Positron emission tomography (PE

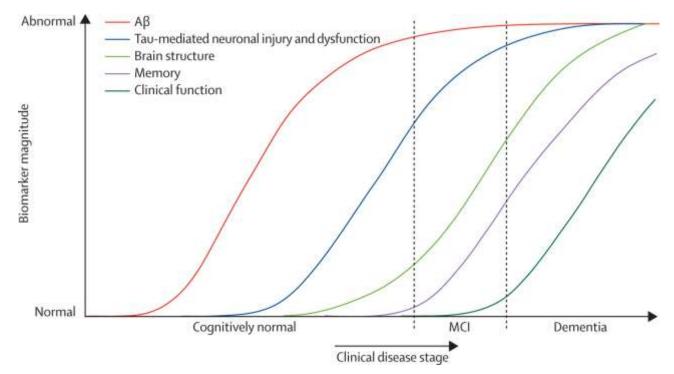
- + many other MRI based modalities (diffusion, arterial spin labelling, spectroscopy)
- + Magnetoencephalography (MEG)
- + CT (limited)
- +...
- +demographics
- +neurospsychological questionaires

Areas of interest in neuroimaging



Why neuroimaging?

AD



Jack et al. *The Lancet Neurology* 12.2 (2013): 207-216.

MDD

SSRI:

- 40–60% of patients will respond
- 30-45% will achieve remission

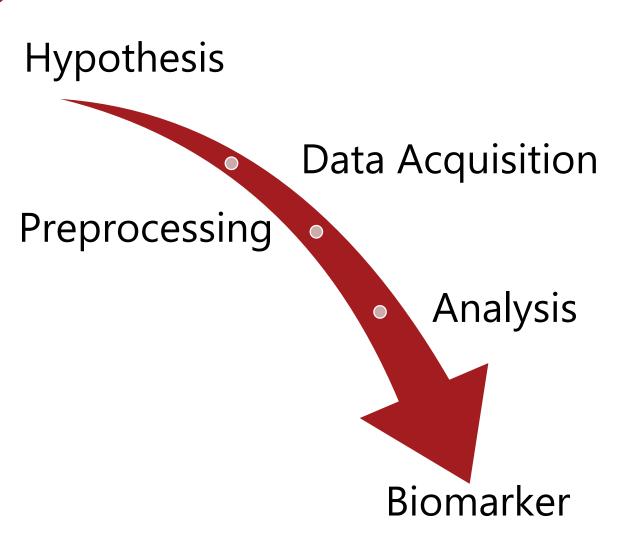


Khin et al. J Clin Psychiatry 72 (2011) Carvalho et al. J Clin Pharm Ther 32 (2007): 415–428. Thase et al. Br J Psychiatry 178 (2001): 234–241.



A typical neuroimaging study

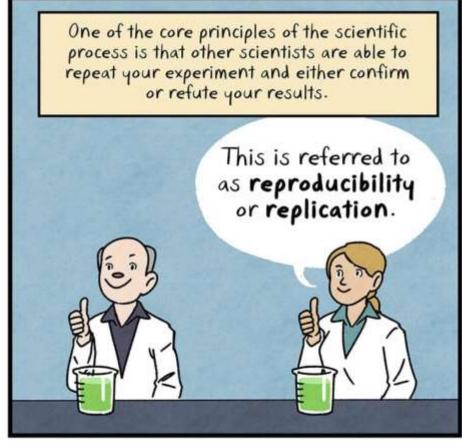






Replication crisis

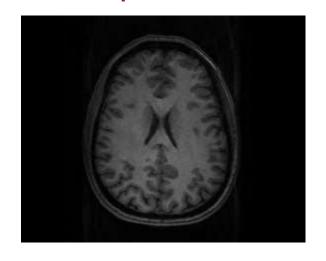


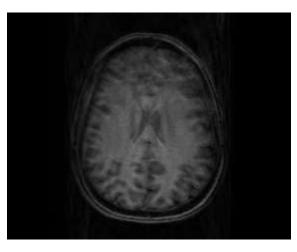






Acquisition

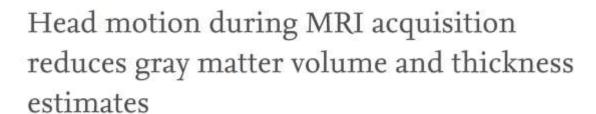






Neurolmage Volume 107, 15 February 2015, Pages 107-115





Martin Reuter ^{a, b, c, d} № M. Dylan Tisdall ^{b, d, 1}, Abid Qureshi ^{a, d}, Randy L. Buckner ^{b, d}, André J.W. van der Kouwe ^{b, d}, Bruce Fischl ^{b, c, d}







Preprocessing I

ORIGINAL RESEARCH article

Front. Neuroinform., 24 April 2015 | https://doi.org/10.3389/fninf.2015.00012

Reproducibility of neuroimaging analyses across operating systems

RESEARCH ARTICLE



The Effects of FreeSurfer Version, Workstation Type, and Macintosh Operating System Version on Anatomical Volume and Cortical Thickness Measurements

Ed H. B. M. Gronenschild , Petra Habets, Heidi I. L. Jacobs, Ron Mengelers, Nico Rozendaal, Jim van Os, Machteld Marcelis

Published: June 1, 2012 • https://doi.org/10.1371/journal.pone.0038234

Preprocessing II



Contents lists available at Science Direct

NeuroImage





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Review Article

NEWS AND VIEWS · 20 MAY 2020

Cerebra measur on acqu PET ce

Neuroimaging results altered by varying analysis pipelines

Martin Nør Masanori le Ramin V Pa Mark Slifst Peter S Tal and Gitte I

Seventy laboratories that analysed the same neuroimaging data each produced different results. This finding highlights the potential consequences of a lack of standardized pipelines for processing complex data.

Martin Lindquist

Article

Variabi neuroir







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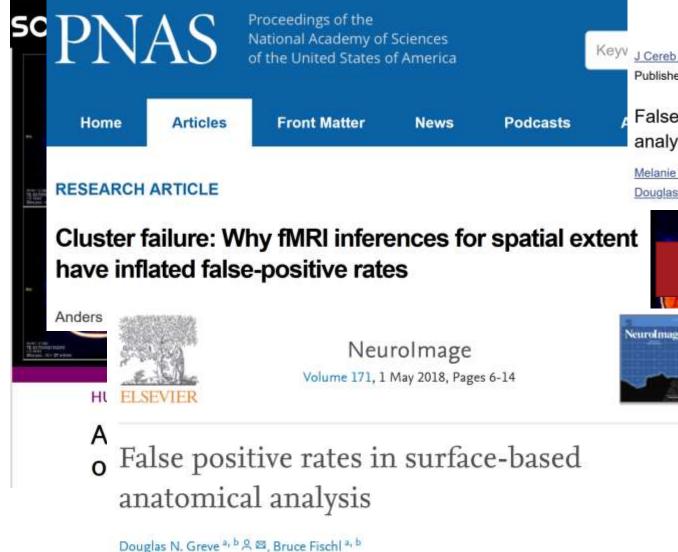
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NARPS - https://www.narps.info

Different preprocessing strategies lead to different conclusions: A [11C]DASB-PET reproducibility study

Martin Nørgaard, 1,2 Melanie Ganz, 1,3 Claus Svarer, 1 Vibe G Frokjaer, 1 Douglas N Greve, 4 Stephen C Strother, 5 and Gitte M Knudsen 1.2 [WIP]

Statistical Analysis



JCBFM Journal of Cerebral Blood Flow & Metabolism

J Cereb Blood Flow Metab. 2021 Jul; 41(7): 1647-1657.

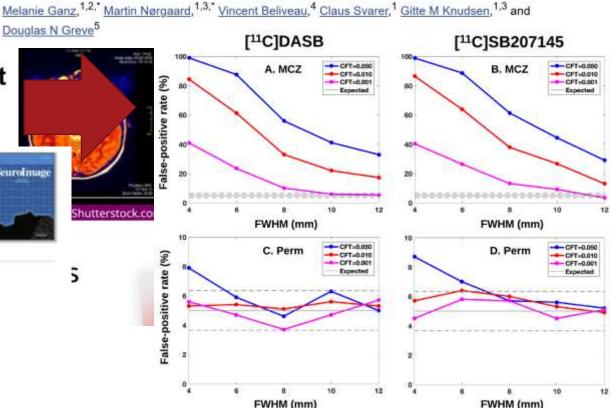
Published online 2020 Nov 26. doi: 10.1177/0271678X20974961

PMCID: PMC8221774

PMID: 33241770

False positive rates in positron emission tomography (PET) voxelwise

analyses



NeuroImage: Clinical



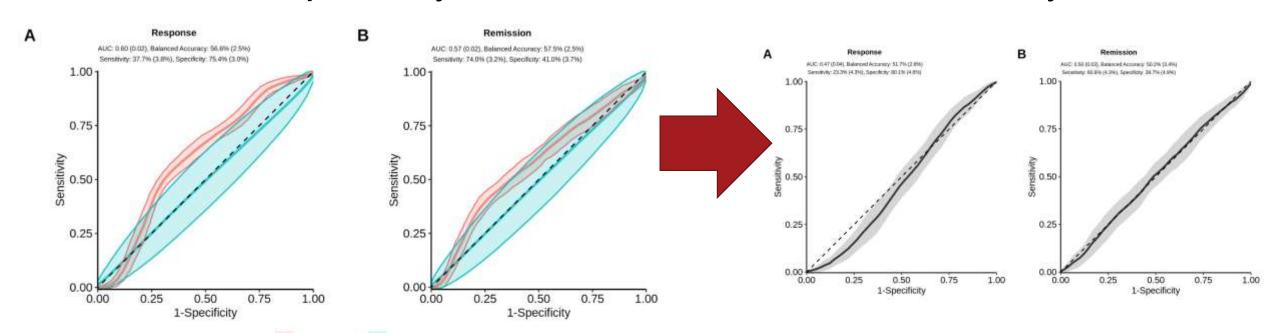
Deployment of biomarker

Generalizability of treatment outcome prediction in major depressive disorder using structural MRI: A NeuroPharm study

Vincent Beliveau ° b R. S., Ella Hedeboe °, Patrick M., Fisher °, Vibeke H., Dam °,
Martin B., Jargensen ° c d, Vibe G., Frokjaer ° c d, Gitte M., Knudsen ° c d, Melanie Ganz ° s

Neuropharm study

EMBARC study



In sample

Permuted

Not-Permuted

Out of sample

Solutions?



Reproducible Research. Image adapted from The Turing Way handbook DOI: 10.5281/zenodo.3332807 https://the-turing-way.netlify.app/welcome.html

Solutions I

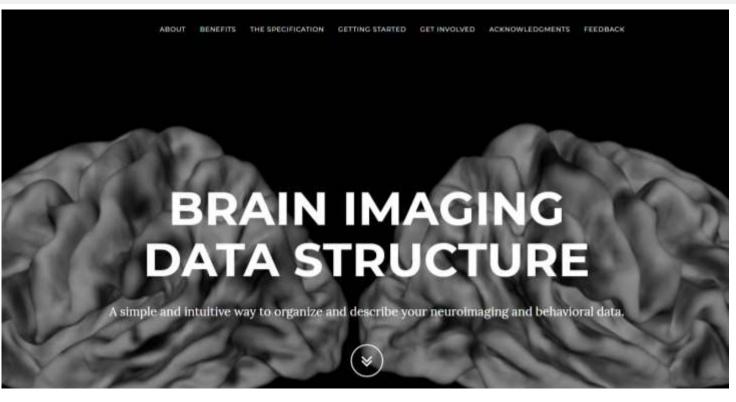
Review Article

Guidelines for the content and format of PET brain data in publications and archives: A consensus paper

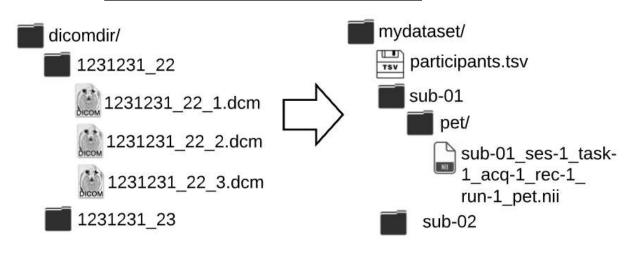
Gitte M Knudsen¹, Melanie Ganz¹, Stefan Appelhoff², Ronald Boellaard³, Guy Bormans⁴, Richard E Carson⁵, Ciprian Catana⁶, Doris Doudet⁷, Antony D Gee⁸ Douglas N Greve⁶, Roger N Gunn⁹, Christer Halldin¹ Peter Herscovitch¹¹, Henry Huang⁵, Sune H Keller¹², Adriaan A Lammertsma3, Rupert Lanzenberger13, Jeih-San Liow¹⁴, Talakad G Lohith¹⁵, Mark Lubberink¹⁶, Chul H Lyoo¹⁷, J John Mann¹⁸, Granville J Matheson¹⁰, Thomas E Nichols¹⁹ , Martin Nørgaard¹ , Todd Ogden²⁰, Ramin Parsey²¹, Victor W Pike¹⁴, Julie Price⁶, Gaia Rizzo⁹, Pedro Rosa-Neto^{22,23}, Martin Schain²⁰, Peter JH Scott²⁴, Graham Searle⁹, Mark Slifstein²¹, Tetsuya Suhara²⁵, Peter S Talbot²⁶, Adam Thomas²⁷, Mattia Veronese²⁸, Dean F Wong²⁹, Maqsood Yaqub³, Francesca Zanderigo³⁰, Sami Zoghbi¹⁴ and Robert B Innis¹⁴



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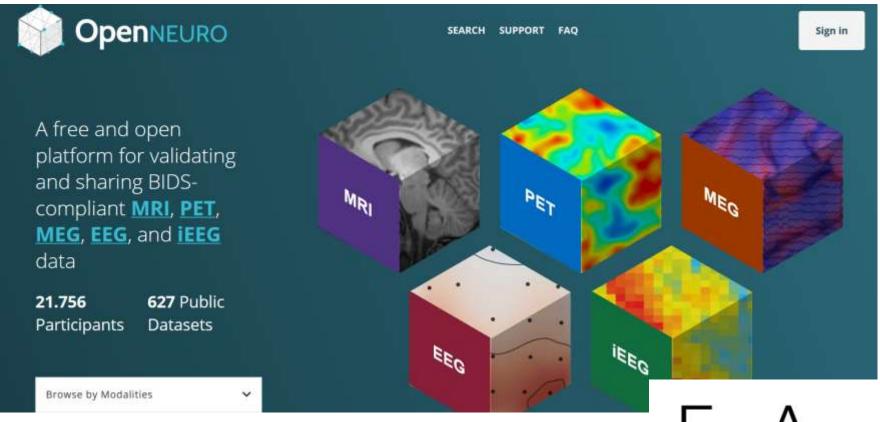


From raw data to BIDS format





Solutions II















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Solutions III

End-user applications

Middleware

Software nfrastructure



The first NiPrep for functional MRI data.

Projects maintained by the NiPreps community



dMRIPrep Diffusion MRI "Prep"

The AliPrep for diffusion MRI scans, typically used in analysis of tractography and white-matter



MRIQC

Quality Control (QC) of MRI (structural and functional)

MRIGC produces visual reports for the efficient screening for quality of MRI data, and estimates quality metrics to learn machines flag subper data.



CrowdMRI

A database of crowdsourced neuroimaging features

sMRIPrep

Structural MRI "Prep"

The MPrep for structural (T1-weighted, and T2w) MRI scans, typically used in morphometry analysis and acquired for spatial reference of IdMRI.

An Internet Service endpoint to collect statistics about data quality, extensible to any crowdsourcing



NiWorkflows Neuroimaging

processing commodities

SDCFlows Susceptibility distortion correction workflows

Library of tools for estimating and correcting for susceptibility-derived distortions typically affecting EPI images (d/fMRI).



MRIQCnets

Machine learning for QC and protocol assessment

Deep Learning models to support higher-level application, such as the deep-MRIOC, the Euler Number Predictor, or the FaceDetector



TemplateFlow

ccess, interface, and ow across templates

An archive and a client tool to allow pro-grammatic access to neuroimaging templates and attases, by humans and machines.





Library supporting the spatial transformation data formats, and an easy-to-use interface for their application and manipulation.



Library supporting the neuroimaging data formats (e.g., NiTTi, GIFTI, and CIFTI2).



NiPype Neuroimaging workflows and Interfaces in Python

The workflow engine supporting the execution graph and run time management imaging tasks, communication, data flow).

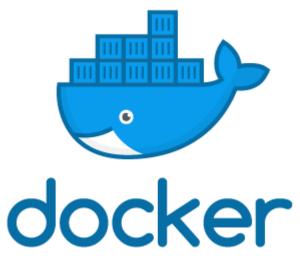


BIDS + PyBIDS Brain Imaging Data

Structure

The data specification prescribing the formal structure for the neuroimaging data inputs and outputs.







Additional comments

 "I would like to have demo PET data and metadata needed for BIDS conversion.



- Moreover, both fMRI (resting state and task) and structural MRI(dwi, t1, t2) Arterial Spin labeling and MEG. For me it would very important to have these demo data to convert them as a separate dataset but also to have a tool convert all these modalities into a single multimodal dataset."
- "I'm currently working on the deep learning methods with eye movement data. It would be very interesting to see the potential multimodal methods between eye tracking data and brain image."
 - BEP020 https://bids.neuroimaging.io/bep020



Additional comments

 "Since it is for phd students probably approaching new postdoc positions soon, I think it would be great to make the method we learn in the tutorial as transferrable as possible to other environments, working in slightly different fields, so that it is a skill we can bring with us to the next labs."







Questions?

BIDS resources:

Website https://bids.neuroimaging.io

Starter Kit https://bids-standard.github.io/bids-starter-kit/

YouTube channel https://www.youtube.com/channel/UCxZUcYfd nvIVWAbzRB1tlw

Data sharing resources:

Openneuro https://openneuro.org/

OpenNeuroPET https://openneuropet.github.io/

PublicNeuro https://publicneuro.eu/

Serotonin atlas https://xtra.nru.dk/FS5ht-atlas/

Benzodiazepine atlas https://xtra.nru.dk/BZR-atlas/

SV2A atlas ...coming online soon!





Thank you!

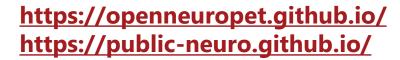




































Rigshospitalet