The MNE package for M/EEG data processing

MGH/HST Athinoula A. Martinos Center for Biomedical Imaging

A. Gramfort¹, M. Luessi², E. Larson³, D.A. Engemann⁴, D. Strohmeier⁵, C. Brodbeck⁶, M. Hämäläinen²

¹Telecom ParisTech, CNRS LTCI - CEA/Neurospin ²MGH, Harvard Med. School, USA ³U. of Washington, USA ⁴CNS (INM-3), Juelich Research Center, Germany ⁵Ilmenau U., Germany ⁶New York U., USA



Features of the MNE Software Family

Preprocessing

MASSACHUSETTS
GENERAL HOSPITAL

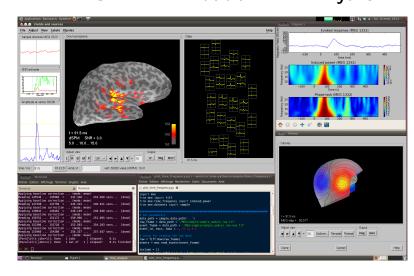
HST

- Review raw data, filter, correct ECG / EOG with SSPs, ICA Forward & inverse modeling
- FreeSurfer structural data: Automatic forward modeling
- ■MNE dSPM sLORETA (TF-)M×NE LCMV
- Statistics (sensor and source spaces)
- Time-Frequency (Phase-Locking, Induced Power)
- Parametric and non-parametric stats, with clustering Connectivity (sensor and source spaces)
- Functional and effective connectivity measures

Project Vision and Goals

- ■State of the art, many examples, documented and tested
- Open development, collaboration between several centers
- Sharing best practices, promoting reproducible research

MNE-C - MNE-Matlab - MNE-Python



http://martinos.org/mne

A. Gramfort, M. Luessi, E. Larson, D. Engemann, D. Strohmeier, C. Brodbeck, L. Parkkonen, M. Hämäläinen MNE software for processing MEG and EEG data, Submitted.

MNE-Python

- Python: general-purpose, high-level language
- Free: can run on a cluster without license problems
- Permissive BSD license: allows use in commercial products
- Many third-party packages easily integrated, e.g., ML
- Open, 29 contributors so far: ≈ 8 person years of effort

Lines of Code May 2013 50k Code: 32,223 Comments: 17,380 Blanks: 9,794 2013

Learn more

- Mailing list: mne_analysis@nmr.mgh.harvard.edu
- ■http://martinos.org/mne/ (general doc)
- http://martinos.org/mne/python_tutorial.html
- http://martinos.org/mne/auto_examples/ (> 70 demos)
- ■http://mne-tools.github.com/mne-python-intro-slides

Induced power (MEG 0933) Frequency (Hz) 50 100 150 200 250 300 350 time (ms)

From raw to dSPM in < 30 lines of code

http://github.com/mne-tools

```
import mne
# load data
fname = 'raw.fif'
raw = mne.fiff.Raw(fname)
raw.info['bads'] = ['MEG 2443', 'EEG 053'] # mark bad channels
# band-pass filter data in beta band, and save it
raw.filter(13.0, 30.0, filter_length=4096, n_jobs='cuda')
raw.save(fname[:-4] + '_beta.fif')
picks = mne.fiff.pick_types(raw.info, meg=True, eeg=True, eog=True, exclude='bads')
events = mne.find_events(raw)
epochs = mne.Epochs(raw, events, event_id=1, tmin=-0.2, tmax=0.5, proj=True, picks=picks, baseline=(None, 0), preload=True,
                          reject=dict(grad=4000e-13, mag=4e-12, eog=150e-6))
\# compute evoked response and noise covariance, and plot evoked
        = epochs.average()
cov = mne.compute_covariance(epochs, tmax=0)
evoked.plot()
   compute inverse operator
fwd_fname = 'sample_audvis-meg-eeg-oct-6-fwd.fif'
fwd = mne.read_forward_solution(fwd_fname, surf_ori=True)
inv = mne.minimum\_norm.make\_inverse\_operator(raw.info , fwd , cov , loose=0.2)
# compute inverse solution
stc = mne.minimum_norm.apply_inverse(evoked, inv, lambda2=1 / 3.0 ** 2, method='dSPM')
# morph it to average brain for group study
stc_avg = mne.morph_data('sample', 'fsaverage', stc, 5, smooth=5)
stc_avg.plot()
```

