The MNE package for M/EEG data processing

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Features

Preprocessing

- Review raw data, filter, correct ECG / EOG with SSPs, ICA

Forward & inverse modeling

- Automatic BEM with FreeSurfer reconstruction
- MNE - dSPM - sLORETA - LCMV - (TF-)MxNE

Statistics (sensor & source space)

- Time-Frequency (Phase-Lock, Induced Power)
- Parametric and non-parametric stats, with clustering

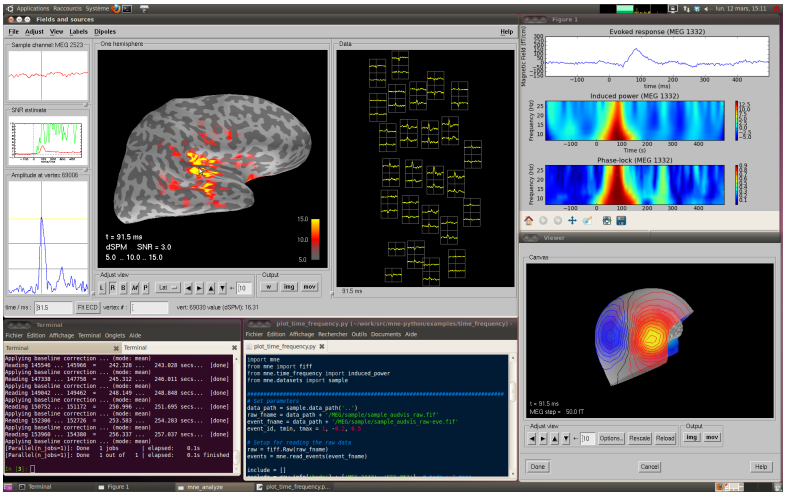
Connectivity (sensor & source space)

- Functional & effective connectivity measures

Project vision & Goals

- State of the art, many examples, documented & tested
- Open development, collaboration between many labs
- Sharing best practices, making reproducible research

C / Unix & Matlab & Python



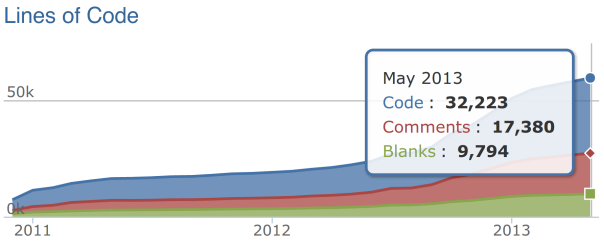
http://martinos.org/mne

http://github.com/mne-tools

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

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)

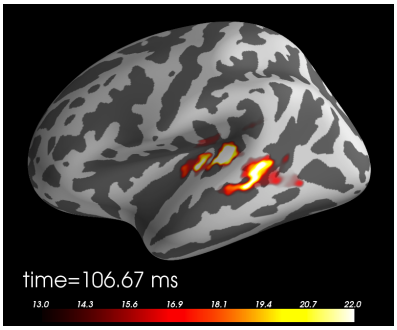


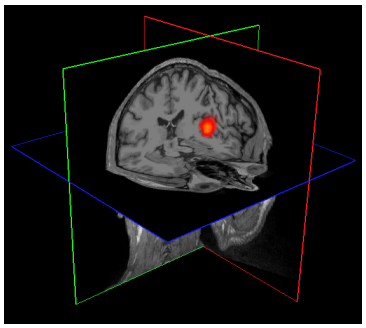
Learn more

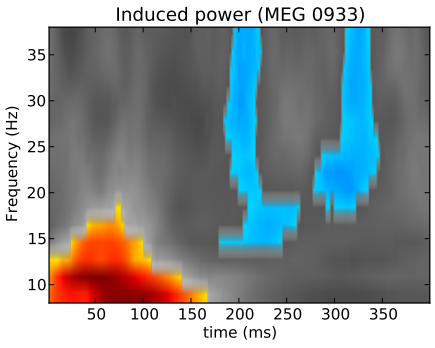
- Mailing list: mne_analysis@nmr.mgh.harvard.edu
- http://martinos.org/mne/ (general doc)
- http://martinos.org/mne/python_tutorial.html
- http://martinos.org/mne/auto_examples/ (> 40 demos)
- http://mne-tools.github.com/mne-python-intro-slides

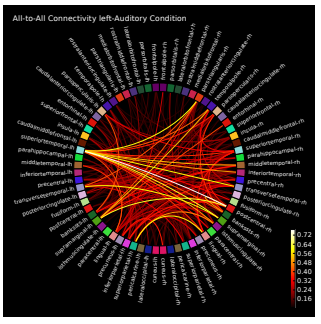
From raw to dSPM in < 30 lines of code

```
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')

# extract epochs
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
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()
```









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