

Neural Correlates of Human Performance

Computational Neuroscience



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Outline

- Introduction
- Independent components of the EEG
- Tonic and phasic EEG correlates of task performance
- Independent neuromodulators of components
- The effects of arousing feedback on subject behaviors and the EEG
- Cognitive-state monitoring and management

EEG Correlates of Neurocognitive Performance

- Lapses of attention or drowsiness can lead to catastrophic incidents for ship, air, truck, rail, or plant operators, air traffic controllers, security officers, and workers in many other occupations.
- The US National Highway Traffic Safety Administration (NHTSA) reported that ~25% of police-reported accidents were related to driver inattention.
- National Sleep Foundation (NSF) reported that 60% of adult drivers had driven a vehicle while feeling drowsy and 37% had actually fallen asleep.

EEG Correlates of Human Performance during Sustained Attention Tasks

| Study | Task(s); Measure(s) | Electrode Sites or Brain Regions | δ | θ | α | β |
|---|--|---|------|-----------------------|--------|---|
| Badia et al. (1994) | Sleep onset | F3, C3, O1 | | + | +/- | |
| Baulk et al. (2001) | Simulated driving task in an immobile car, secondary auditory detection task; lane crossing incidents, RT, Karolinska Sleepiness Scale (KSS) | C3-A1 | | + | + | |
| Beatty et al. (1974) | Radar monitoring task; target detection time | O1-P3 | | + | | |
| Belyavin and Wright (1987) | Visual vigilance and letter discrimination tasks; RT, error/missing rate | P3-O1, P4-Oz | + | + | + | - |
| Campagne et al. (2004) | Simulated driving on mobile platforms; running-off- road incidents, speed variations | F3, C3, P3, O1 (C3, P3 shown) | | + | + | |
| Cante | Iches have demons | $t_{F_{p}1,F_{p}2,F_{3},F_{4},P_{3},P_{4},O_{1},O_{2}}$ | orre | eląt | es. | _ |
| Gillbor al. fiber Ctua | and lap time per cycle Son and the time per cycle | nce during su | stai | ine | d* | |
| Harrison and Horne (1986) Hasar Gild How Hon Libon | S.D. of lane position, KSS, RT Multiple slop latence test AVSLT Order | Of cone secon | d t | 0 ⁺ | * * | |
| Horne and Baulk (2004) | Simulated driving task in an immobile car; KSS, lane | (C3-A1) | | + | + | |
| Huang et al. (2001) | Auditory and visual vigilance tasks; correct rate | C3, C4 | | + | + | |
| Huang et al. (2008) | Compensatory tracking task; tracking error, reaction time | 70 EEG channels; occipital independent components | + | + | + | |
| Huang et al. (2009) | Event-related lane departure during simulated driving (static): reaction time | 256 EEG channels; occipital and parietal independent components | + | + | + | |
| Jung et al. (1997) | Auditory oddball task; error rate | Cz, Pz/Oz | | + | - | * |
| Kecklund and Åkerstedt (1993) | Real truck driving; KSS, self-rated performance capacity | Cz-Oz | | + | + | |
| Lal and Craig (2002, 2005) | Simulated driving in a static car frame; facial features (from video) of the driver | 19 EEG channels | + | + | | |
| Lowden et al. (2009) | Simulated driving on a moving base; speed, lateral position, steering wheel angle, KSS | Fz-A1, Cz-A2, Oz-Pz | | | + | + |
| Makeig and Inlow (1993) | Auditory oddball task; local error rate | 13 EEG channels | + | + | _ | |
| Makeig and Jung (1995, 1996) | Auditory oddball task, visual target detection; local error rate | Cz, Pz/Oz | + | + | - | * |
| Makeig et al. (2000) | Compensatory tracking task; tracking error | F3, C4, P4, O1 (C4 shown) | + | + | | |
| Ogilvie and Wilkinson (1984) | Auditory response task; reaction time | Cz, Pz | | | | |
| Ogilvie et al. (1991) | Auditory response task; reaction time | 14 EEG channels (C3, C4 shown) | + | + | - | - |
| Ota et al. (1996) | Auditory response task; reaction time | 18 EEG channels (F1, F2, O1, O2 shown) | | + | +/- | |
| Otmani et al. (2005) | Simulated driving on a mobile base; S.D. of lateral position steering wheel angle KSS | F3, C3, P3, O1 | | + | + | |

Objectives of this Study

- To investigate tonic spectral changes during continuous sustained-attention tasks in a realistic environment.
- To explore basic induced by land-deviation events, subject responses, or arousing feedback presented to drowsy participants.

• To build a brain-machine interface that can continuously monitor brain dynamics and cognitive states of participants active Efferni B ord har Atom in real operational environments.

Driving is a Complicated Task

- Visual Motion Processing (Sensory)
- Visual Spatial Attention
- Sensory-motor coordination
- Vigilance/Alertness
- Error Detection/Correction; Response Selection
- Goal Planning/ Expectation/ Uncertainty
- Navigation (Cognitive Map)
- Divided Attention (Dual- / Multi-Task)
- Vestibular/Kinetic Processing

A VR-based Dynamic Driving Simulator



Paradigm: Single Trials Embedded in Continuous Driving



Cruising Speed: 100 km/hr Linear deviation (D=c T) Inter-Deviation-Interval: 5 ~ 10 sec Deviation: 50% leftward, 50% rightward deviation

From Huang et al., 2005, 2007.

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- Data Acquisition: 256channel (Biosemi system)
- Sampling rate: 256 Hz
- 4 infrared cameras
- Driving simulator (OPEN GL/Performer)
- Scene: An endless straight road at night
- 10 subjects, two 1-hour sessions per subject
- Response: Arrow keys
- Behavioral log: 256 Hz













Time (sec)

Left



Signal Mixing





From Jung et al., 2000.

Data Analysis using ICA



Component Stability: Cross-subject clustering analysis of ICA components



After Clustering



Deviation-induced Brain Dynamics



Huang et al, in preparation

Deviation-induced Brain Dynamics









Spectral Perturbations as a Function of RT



| | _ | Trends | | | | | | | |
|---|------------------------------------|------------|------------|--------|--------|------------|----|------------|--------------|
| | | Motionless | | | Motion | | | | |
| | Cluster / Figure | δ | θ | α | β | δ | θ | α | β |
| Ô | Occipital (bilateral) / Fig. 3 | ** | /* | \sim | \sim | /* | /* | \sim | \sim |
| | Occipital (medial) / Fig. S6 | 1 | 1 | \sim | \sim | * * | 7* | ∕∖* | \sim |
| | Occipital (tangential) / Fig. S8 | ** | * * | ** | \sim | 1 | 1 | \nearrow | \nearrow |
| | Medial posterior parietal / Fig. 4 | 7* | /* | _ | _ | 7* | 1 | _ | - |
| | Left somatomotor / Fig. 5 | 1 | 1 | _ | _ | 1 | 1 | — | _ |
| Ì | Right somatomotor / Fig. S10 | 1 | ** | _ | _ | 1 | 1 | 7 | \mathbf{A} |
| | Central medial / Fig. 6 | 7* | ** | /* | 1 | * * | 7* | 1 | 1 |
| Ó | Frontal medial / Fig. 7 | 1 | 1 | _ | _ | 1 | ** | _ | - |

 $\mspace{1.5mm}$ monotonic increase, $\mspace{1.5mm}$ monotonic decrease, $\mspace{1.5mm}$ biphasic trend, — no difference, * power changes significantly different (p < 0.01) from mean reference power of each respective frequency band at 3-s RT.

Temporal Dynamics of Component Spectra and Subject Performance





There are modulators that mediated spectral activations of the cortical areas by intra-cortical feedback loops, or controlled by thalamo-cortical feedback loops.

Independent Mochatort Analysis



brain areas

From Onton & Makeig, 2006.

Independent Modulator Decomposition



Single-Subject IM Decomposition







Neuromodulators Medicate Spectral Activations of ICs

Modulator's Activity vs Driving Performance



Modulator's Activity vs Driving Performance





Driving error moving average (DEMA)

Real-time Cognitive-State Monitoring

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Correlation between Power Spectra and Driving Performance





1. Training



References:

- 1. Jung, et al., IEEE TBME, 44:60-9, 1997.
- 2. Lin, et al., EURASIP J Applied Signal Processing, 19:3165-74, 2005.
- 3. Lin, et al., IEEE TCAS I, 52(12):2726-38, 2005.
- 4. Lin, et al., IEEE TCAS I, 53(11): 2469-76, 2006.
- 5. Lin, et al., Proc. of the IEEE, 96(7):1167-83, 2008.



Sample Results









Arousing Feedback Rectifies Lapse in Performance

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From Lin et al, NeuroImage, 2010.



From Lin et al, *NeuroImage*, 2010.





From Lin et al, NeuroImage, 2010.



EEG Dynamics following Feedback





From Jung et al, IEEE EMBC 2010.



| | Tria | Overall | | | |
|---------------|-----------|-------------|----------|--|--|
| Trials | offootivo | inoffootivo | Accuracy | | |
| classified as | enective | menecuve | (%) | | |
| effective | 320 | 223 | | | |
| ineffective | 23 | 120 | 64.0% | | |

From Jung et al, IEEE EMBC 2010.



Missing Link



Clinical Neurophysiology Volume 118, Issue 9, September 2007





Missing Link

Laboratory EEG





EEG-based Cognitive-State Monitoring



Estimating Cognitive states of the drivers



Sample Results



Lin at. al., Proc. of the IEEE, July 2008.



Summary



- This study has reported both tonic and phasic spectral dynamics of independent components in response to lanedeviations during a continuous lane-keeping driving task.
- Independent Modulator Analysis (IMA) separated spectra of statistically independent sources into the weighted sum of maximally temporally independent modulators (IMs).
- Across subjects, we found common modulators which mediated activations of the several cortical areas associated with subjects' task performance.
- The modulator activations were highly correlated with the driving performance in the realistic lane-keeping driving task.







- Arousing auditory feedback delivered to the cognitively challenged subjects immediately agitated subject's responses to the events.
- The improved behavioral performance was accompanied by concurrent spectral suppression in the theta- and alphabands of a lateral occipital component.
- We also showed that continuous, accurate, noninvasive and near real-time estimation of subject's cognitive level is feasible in a realistic operational environment.
- It is feasible to integrate novel dry sensors, advanced signalprocessing algorithms and miniature supporting hardware into a mobile & wireless cognitive-state monitoring and management system.