



Investigating Effect of Noise on the Error Rate of Recognizing Faces/Houses under the Limited Perceptual Load

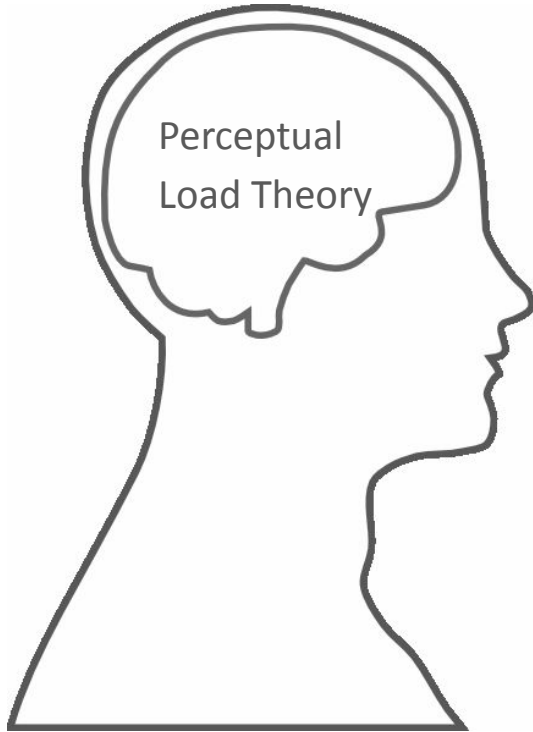
Pod: Snarling Coyotes

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Background Information



- **Perceptual load** theory suggests that the efficiency of **selective attention** is dependent on both perceptual and cognitive loads.
- Perception is often analysed as a process in which causal events from the environment act on a subject to produce states in the brain.
- **Face recognition** is one of the methods to study perceptual load.

Effect of Noise in Face stimuli on Perceptual Load

Averaged broadband responses to face stimuli from the noisy task.

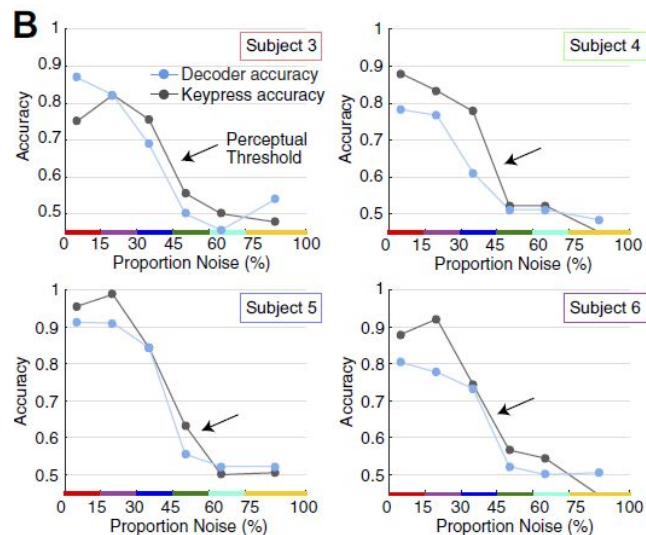


Fig. 3

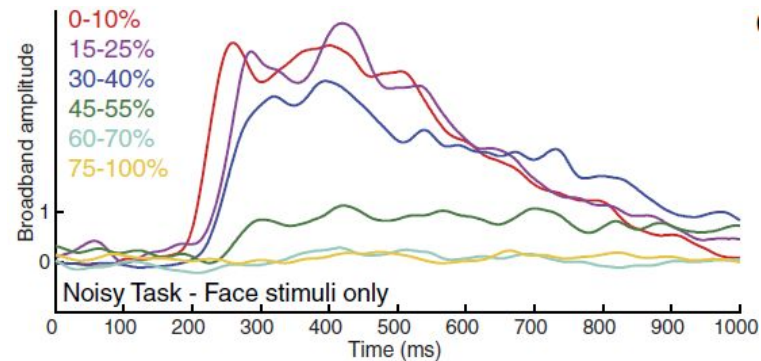


Fig. 2

Decoder and subject keypress accuracy in 4 subjects, as a function of noise.

Data and Experimental Protocol

ECoG data + Noise

Subdural grids
Strips of platinum {electrodes}

Miller's Dataset

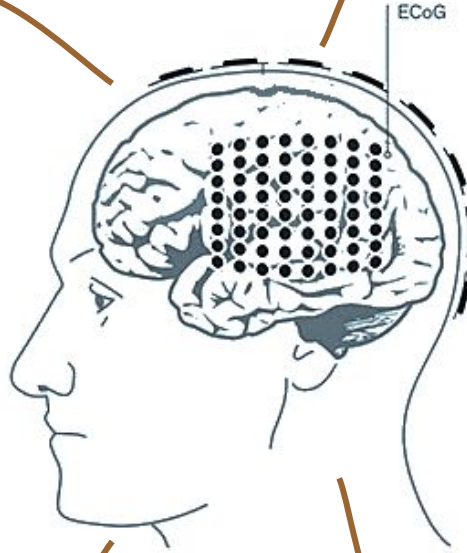
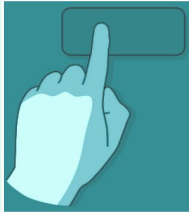
Images < Faces, Houses >

Seven subjects

< *Epileptic Patients* >

4 males
3 females

>> only 5 pressed



Stimulus = [1: 600]

Sampling rate = 1000Hz

Notch-filtered @ 60, 120, 180, 240
and 250 Hz

Continuous voltage data

< channel, time >

Stimulus category

{ 1 = House
2 = Face }

Noise % = [0: 100]

Onset time

Offset time

Methods Pipeline

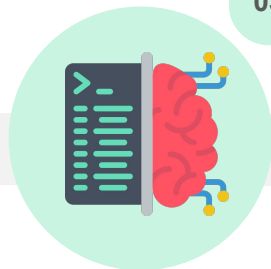
Raw Voltage
Data

01



Machine Learning
Models

03



Channel
Mapping

05



Broadband
power spectra

02



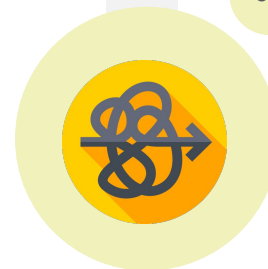
Feature
Selection

04



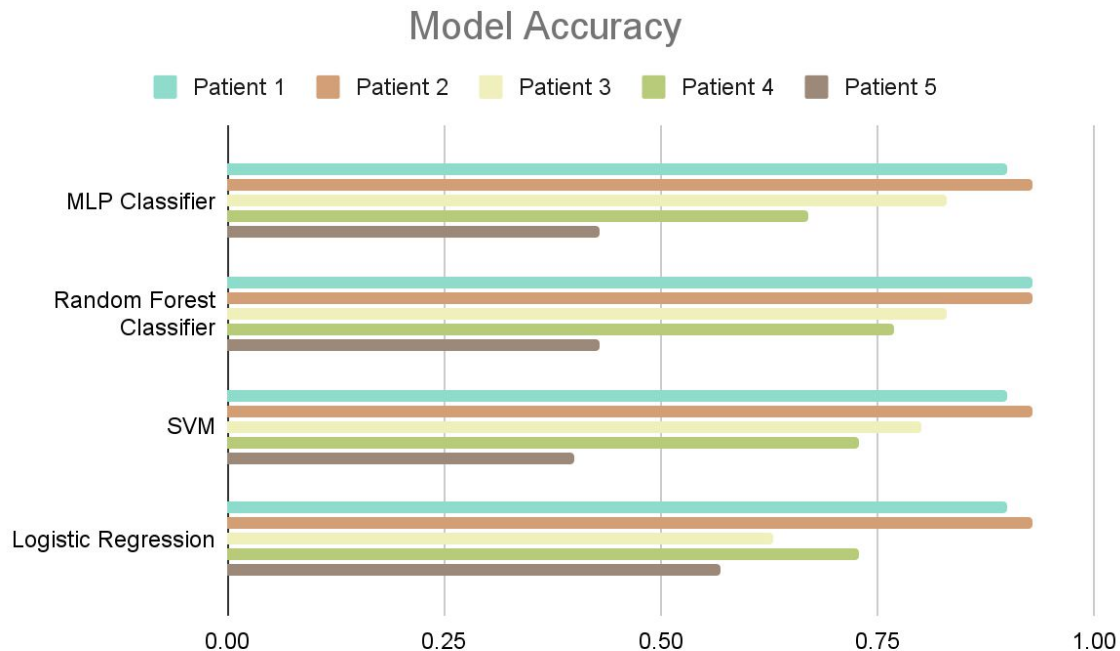
Noise

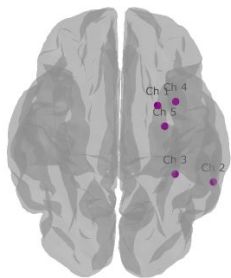
06



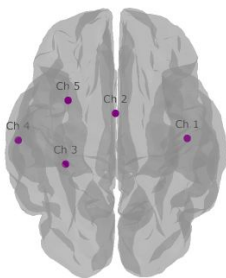
Model Accuracies that has been trained and tested with patients' responses (after tuning hyperparameters)

Machine learning based classifiers
were used to predict the response of
each patient for different stimuli.

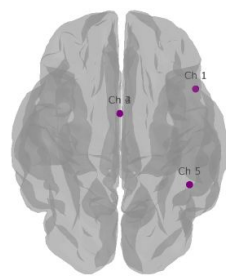




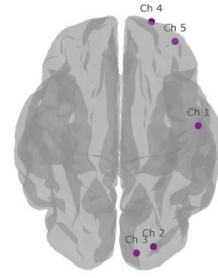
Patient 1



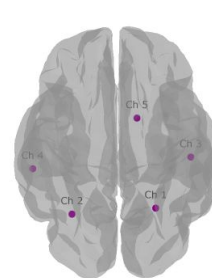
Patient 2



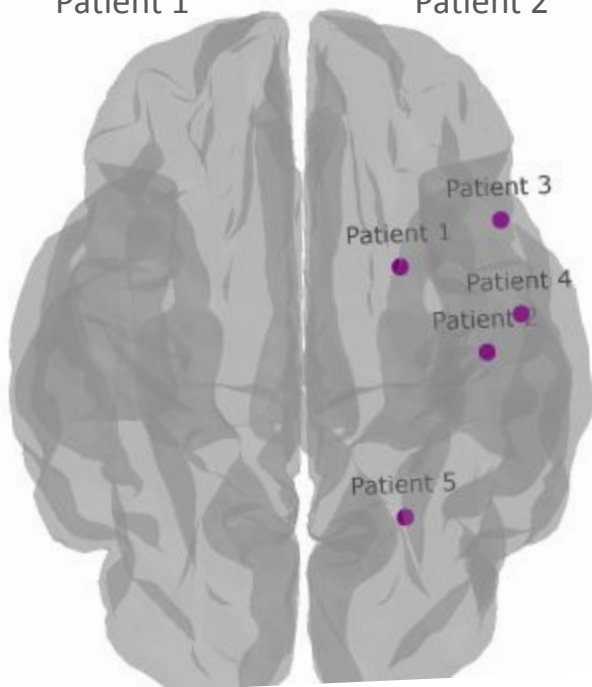
Patient 3



Patient 4

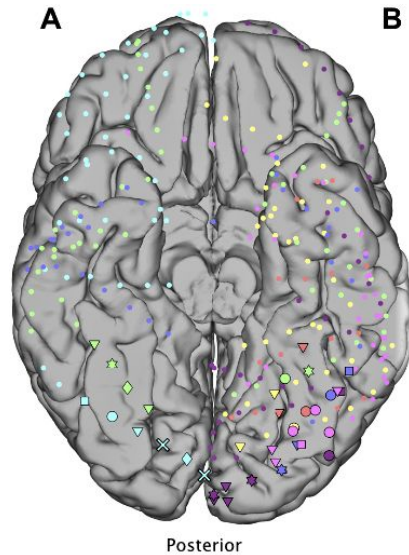


Patient 5

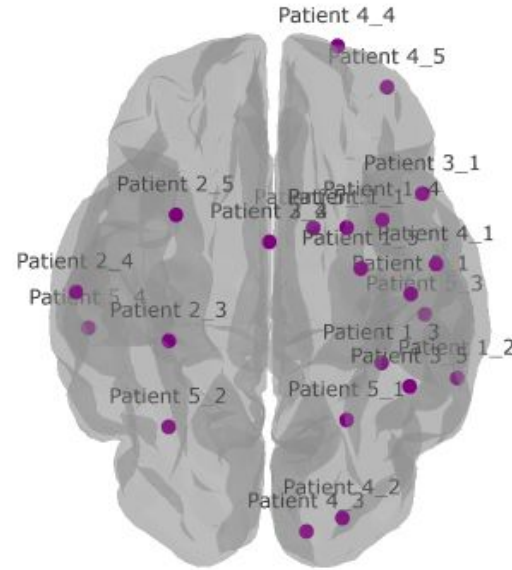


The Most Important Channel for Each Patient According to Logistic Regression Model

The Top Five Most Important Channels for Each Patient According to Logistic Regression Model



All Patient in one 3D brain graph
(according to Miller's model)

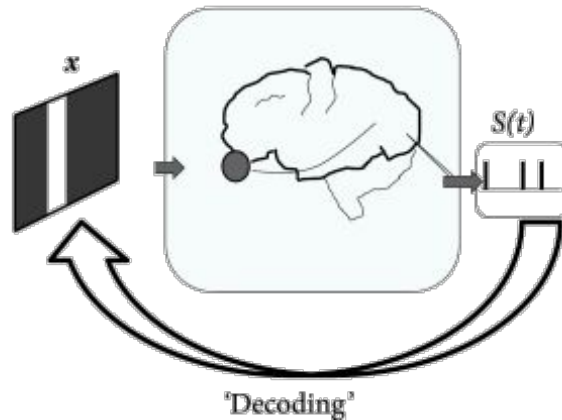


All Patient in one 3D brain graph
(according to our model)

Concentrated in
a region that
matches with
the paper's
prediction

Conclusion

Since our model predictions for actual labels match with that of patient predictions, we can say that our model is a good representative of neural decoding in patients. So we can go ahead with analysing the effect of noise on patients' decisions by using the models.



Future Work

- We could train the classifiers on data from experiment 1 and use them to predict accuracies on our test set from experiment 2. The difference in accuracy, if any, could then be attributed to the presence of noise in experiment 2.
- We can also check how the accuracy changes when we remove non-important / negatively contributing channels from our input.
- We can extract out different frequency bands from the data and see if one kind of oscillations are more relevant than the others.

References

1. Murphy, G., Groeger, J.A. & Greene, C.M. (2016) Twenty years of load theory—Where are we now, and where should we go next?. *Psychon Bull Rev* 23, 1316–1340
2. Gillett, G. (1989). Perception and Neuroscience. *The British Journal for the Philosophy of Science*, 40(1), 83-103.
3. Lopatina Olga L., Komleva Yulia K., Gorina Yana V., Higashida Haruhiro, Salmina Alla B. (2018). Neurobiological Aspects of Face Recognition: The Role of Oxytocin. *Frontiers in Behavioral Neuroscience*. 12, 195.
4. Miller, K. J., Hermes, D., Pestilli, F., Wig, G. S., & Ojemann, J. G. (2017). Face percept formation in human ventral temporal cortex. *Journal of neurophysiology*, 118(5), 2614-2627.