





QEEG Clinical Report BrainLens V0.4

Report Description

Personal & Clinical Data

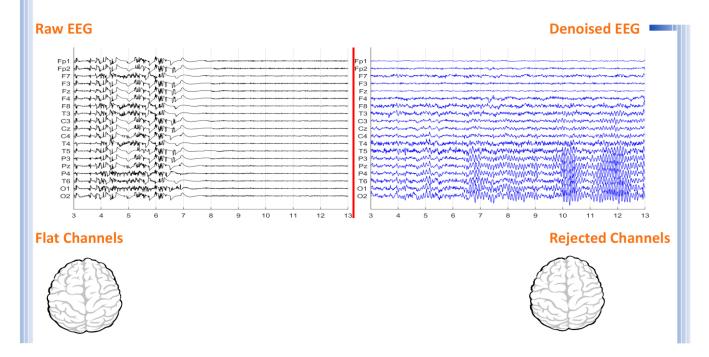
Name	Fateme Aliannezhadi	Date of Recording	01-Oct-2024						
Date of Birth - Age	21-Mar-1965 - 59.53	Gender	Female						
Handedness(R/L)	Right	Source of Referral	Dr Masjedi						
Initial Diagnosis	Migraine-Sleep Problems-Stress								
Current Medication	Medication Free								

Dr Masjedi





Denoising Information (EC)



Number of Eye and Muscle Elements			Low Artifact Percentage					
Eye	Eye 2 Muscle 2		2	0				
Total Artifact Percentage			High Artifact Percentage					
EEG Quali	ity	bad		Total Recording Time Remaining 411.34 sec				





Pathological assessment for mood disorders

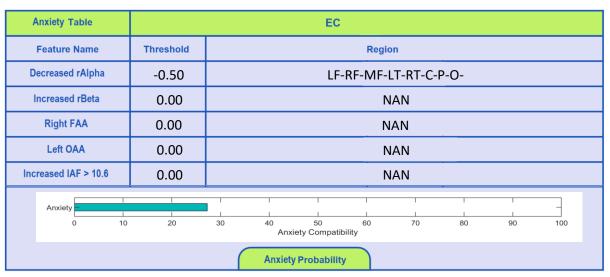
Compare to Mood Disorders Database



EEG Compatibility with Depression Diagnosis

Depression 1	Гable		EC								
Feature Na	me	Threshold	Threshold Region								
Increased Globa	ıl rAlpha	0.00	D NAN								
Increased globa	I rTheta	3.00		global							
Decreased r	Delta	-0.50	RF-RT-C-								
Increased rl	3eta	0.00				N/	AN .				
Left FAA		-0.21		Left FAA							
Right OA	A	0.08		Right OAA							
Decreased Cohere	nce (D, T)	-0.50		Decreased Coherence (D,T)							
Increased Cohere	nce (A, B)	0.00		NAN							
depression		10 20	30	40 Depre	50 ession Comp	60 atibility	70	80	90	100	
Depression Probability											

EEG Compatibility with Anxiety Diagnosis







EEG Compatibility with Mood Swings Diagnosis *

Mood Swings Ta	able		EC								
Feature Name	е	Threshold	Threshold Region								
Decreased rAlp	ha	-0.50	-0.50 LF-RF-MF-LT-RT-C-P-O-								
Increased (rDelta+r	Theta)	2.00	2.00 LF-RF-MF-LT-RT-C-P-O-								
Increased rBeta	a	0.00	00 NAN								
Decreased Alpha Col	herence	-0.50	0.50 Decreased Alpha Coherence								
Right FAA		0.00		NAN							
BMD0	10	20	30	40 Mood S	50 Swing Compa	60 atibility	70	80	90	100	
Mood Swings Probability											

* This index can only be investigated if there are symptoms of mood swings (R/O BMD or R/O mood swings).

Depression Severity



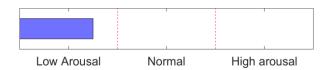
Anxiety Severity



Cognitive Functions



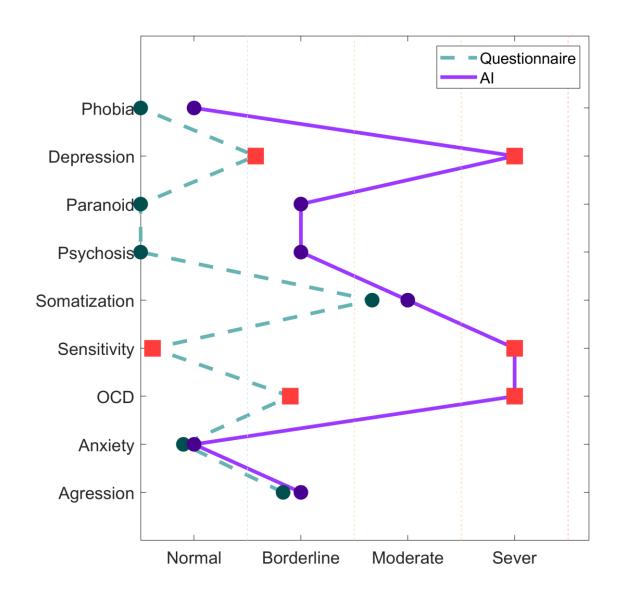
Arousal Level Detection







AI-Driven Psychometric Symptoms Assessing



Explanation

Note

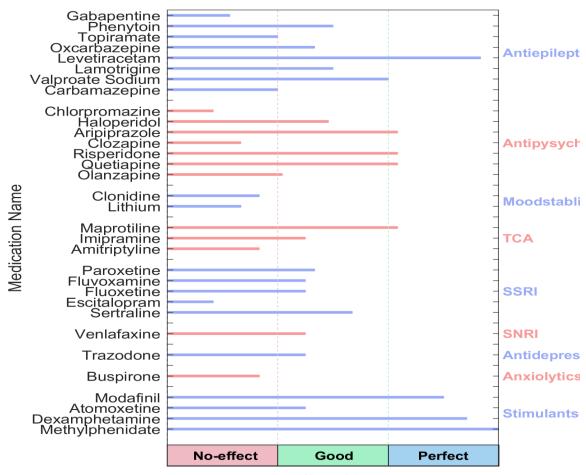
The above diagram illustrates the psychometric symptoms based on the SCL90 questionnaire of the subject (green line) and AI (purple line). Combination of non-linear EEG markers have been used to estimate these symptoms using AI. All the AI algorithms used in these analysis have an accuracy more than 97.60%, a sensitivity more than 97.54%, and a specificity more than 97.58%.

If a red square marker appears in the symptom, it means there is a remarkable difference between the subject's questionnaire score and AI estimate. In the other words, the subject's questionnaire score is in the normal to borderline area, but the AI estimate is in the moderate to extreme area or vice versa.





QEEG based predicting medication response



Effect Size

Explanation



Medication Recommendation

These two tables can be considered the most important finding that can be extracted from QEEG. To prepare this list, the NPCIndex Article Review Team has studied, categorized, and extracted algorithms from many authoritative published articles on predict medication response and Pharmaco EEG studies. These articles are published between 1970 and 2021. The findings extracted from this set include 85 different factors in the raw band domains, spectrum, power, coherence, and loreta that have not been segregated to avoid complexity, and their results are shown in these diagrams. One can review details in NPCIndex.com.

These two charts, calculate response probability to various medications, according only to QEEG indicators. Blue charts favor drug response and red charts favor drug resistance. The longer the bar, the more evidence there is in the articles. Only drugs listed in the articles are listed. These tables present the indicators reviewed in the QEEG studies and are not a substitute for physician selection.



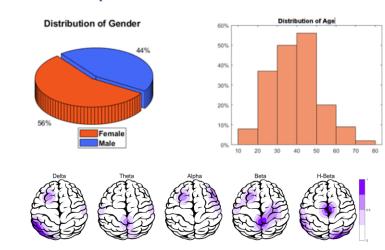


rTMS Response Prediction

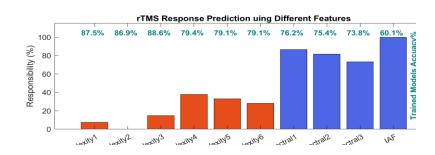
Network Performance

Accuracy: 92.1% Sensitivity: 89.13% Specificity: 97.47%

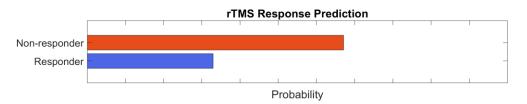
Participants Information



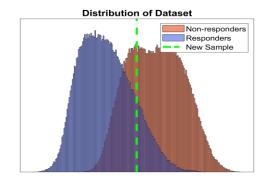
Features Information



Responsibility



Data Distribution



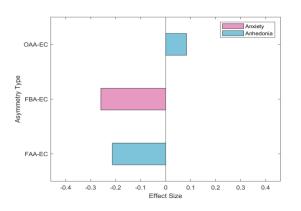
About Predicting rTMS Response

This index was obtained based on machine learning approaches and by examining the QEEG biomarkers of more than 470 cases treated with rTMS. The cases were diagnosed with depression (with and without comorbidity) and all were medication free. By examining more than 40 biomarkers capable of predicting response to rTMS treatment in previous studies and with data analysis, finally 10 biomarkers including bispectral and nonlinear features entered the machine learning process. The final chart can distinguish between RTMS responsive and resistant cases with 92.1% accuracy. This difference rate is much higher than the average response to treatment of 44%, in the selection of patients with clinical criteria, and is an important finding in the direction of personalized treatment for rTMS.

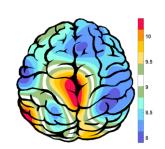




Alpha Asymmetry(AA)



APF(EC)



Frontal APF= 08.75

Occipital APF= 10.25

🚃 Absolute Power-Eye Closed (EC) 🌮































TBI Severity



Relative Power-Eye Closed (EC) 🐠







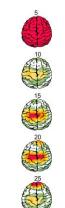






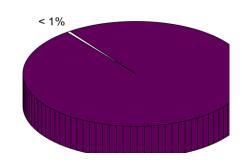






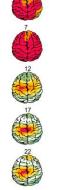
TBI Probability

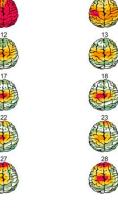
TBI Probability

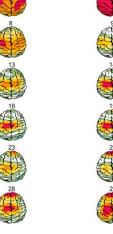








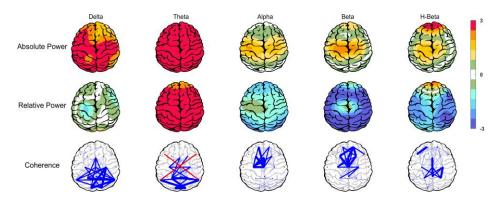




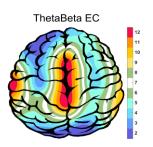


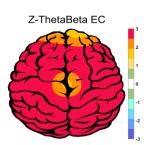


Z Score Summary Information (EC)

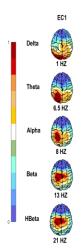


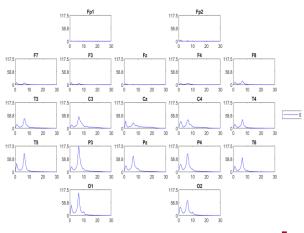
E.C.T/B Ratio (Raw- Z Score)





EEG Spectra















Arousal Level

