





QEEG Clinical Report BrainLens V0.4

Report Description

Personal & Clinical Data

Name	Omidmohammadi34mr	Date of Recording	2025-09-22		
Date of Birth - Age	1990-04-24 - 35.5	Gender	Male		
Handedness(R/L)	Right	Source of Referral	Navid Karami		
Initial Diagnosis	Anxiety-Drug Abuse-MDD				
Current Medication		-			

Navid Karami

Summary Report





Arousal Level

TMS Responsibility

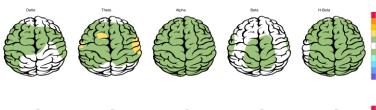




rTMS Response Prediction

Probability

Z-score Information





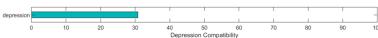
Absolute Power

Relative Power

Cognitive Performance



Compatibility with Depression

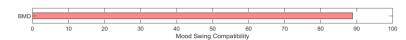




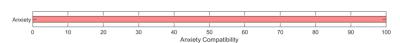
Non-responder

Posterior APF-EC= 11.00

Compatibility with Mood Swing



Compatibility with Anxiety

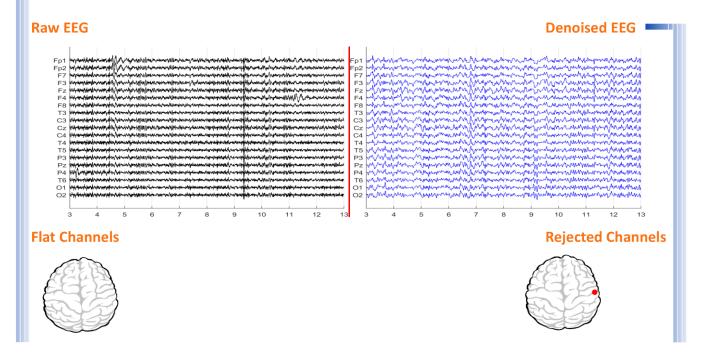


To investigate QEEG-based predicting medication response, please refer to the Report.





Denoising Information (EC)



Number of Eye and Muscle Elements				Low Artifact Percentage	
Eye	1	Muscle	0	0	
Total Artifact Percentage				High Artifact Percentage	
EEG Quali	ity	good		Total Recording Time Remaining 235.78 sec	





Pathological assessment for mood disorders and adult ADHD

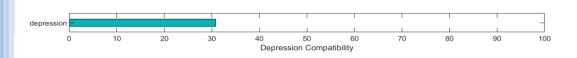
Compare to Mood Disorders Database

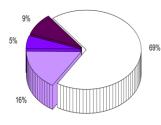


Compare to Adult ADHD Database

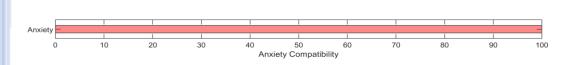


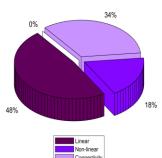
EEG Compatibility with Depression Diagnosis



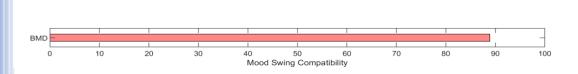


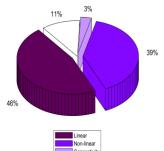
EEG Compatibility with Anxiety Diagnosis





EEG Compatibility with Mood Swing Diagnosis *



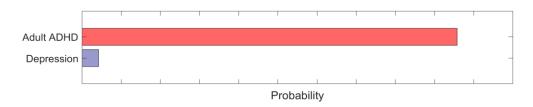


^{*} This index can only be investigated if there are symptoms of mood swings (R/O BMD or R/O mood swings).





Depression and Adult ADHD Diagnosis Probabiliy



Cognitive Functions Asessment



Arousal Level Detection

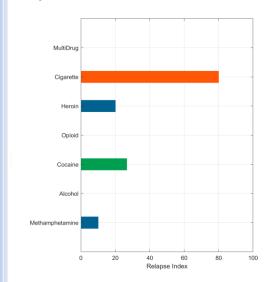




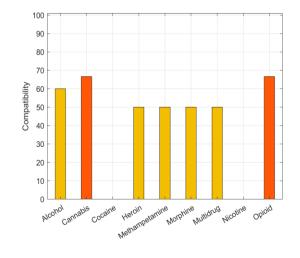


Pathological Assessment for Substance Abuse

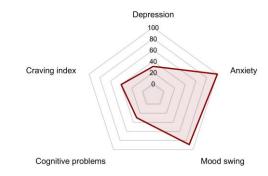
Relapse Index



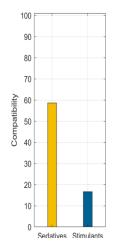
Subsance Abuse Compatibiliy



Comorbid Symptoms



The relapse graph shows the relapse index based on a combination of EEG neuromarkers. If the type of substance your patient uses is included in this chart, you can read its relapse rate. The condition for using this chart is that the patient consumes each substance specified in the chart. If your patient does not consume each of the substances specified in the chart, the index shown is not valid.



The compatibility graph shows the compatibility of the patient's EEG neuromarkers and the alternations that the specific substance causes in the EEG. In other words, this chart indicates that your patient has how percentage of validated neuromarkers due to the use of specific substances.

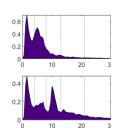
Using this chart, you can figure out how substances have affected EEG and if multiple drugs were used, which one has the most dominant effect. If your patient does not consume each of the substances specified in the chart, the index shown is not valid.

Functional Problems Source Detection

Eyes Closed

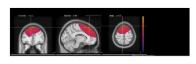


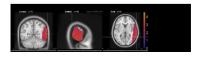












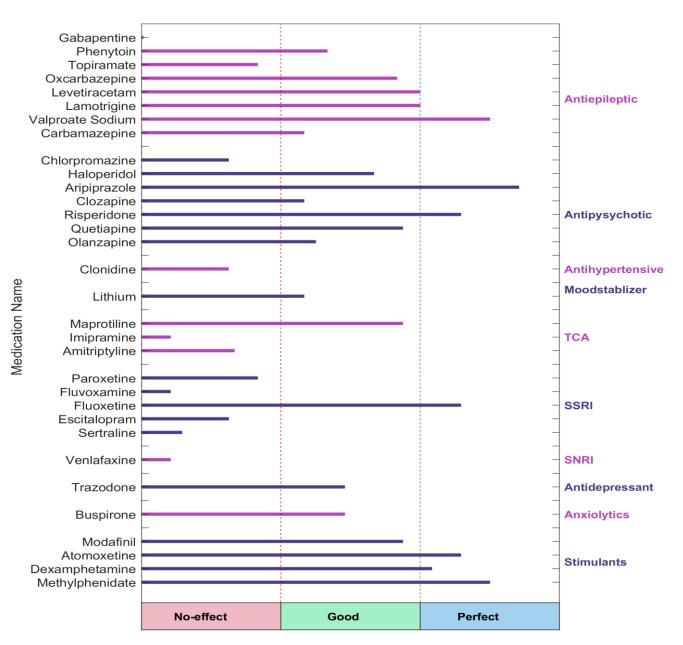
Brodmann area 8 Superior Frontal Gyrus

Middle Temporal Gyrus





QEEG based predicting medication response



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.

two charts, calculate 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.



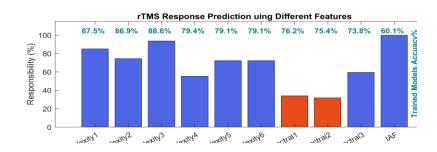


rTMS Response Prediction

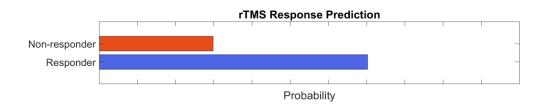
Network Performance

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

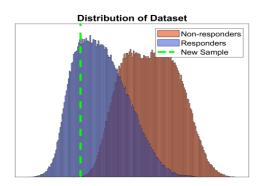
Features Information



Responsibility



Data Distribution



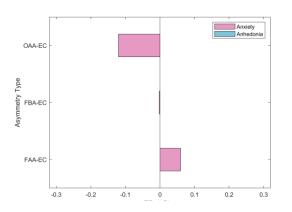
About Predicting rTMS Response

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.

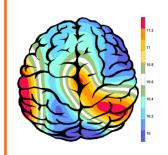




Alpha Asymmetry(AA)



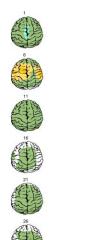
APF(EC)

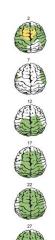


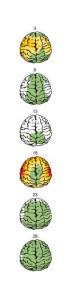
Frontal APF= 10.25

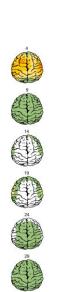
Posterior APF= 11.00

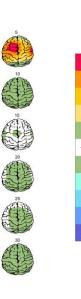
🚃 Absolute Power-Eye Closed (EC) 🌮





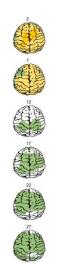


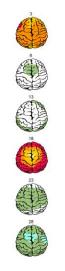


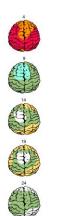


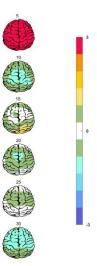
Relative Power-Eye Closed (EC) 🌮







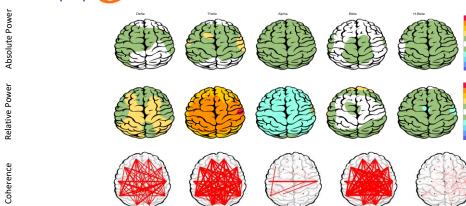




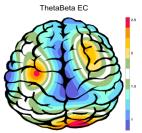


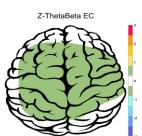


Z Score Summary Information (EC)



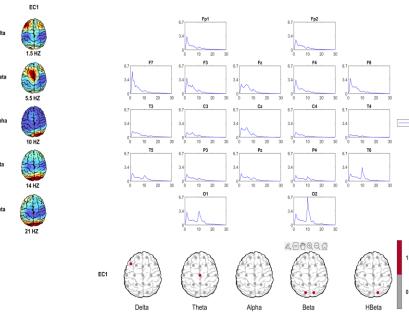
E.C.T/B Ratio (Raw- Z Score)





EEG Spectra





Arousal Level

