





QEEG Clinical Report BrainLens V0.4

Report Description

Personal & Clinical Data

Name	Izadyar Moghadasi	Date of Recording	2025-05-28	
Date of Birth - Age	2008-08-27 - 16.75	Gender	Male	
Handedness(R/L)	Right	Source of Referral	Dr Mahboubeh Karimi	
Initial Diagnosis	ADD-ADHD-Anxiety-Attention and Concentration Problem-Attention Deficit			
Current Medication		-		

Dr Mahboubeh Karimi

Summary Report



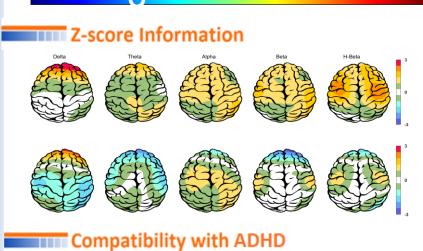




Probability







Relative Power Absolute Power

0 10 20 30 40 50 60 70 80 90 ADHD Compatibility

Arousal Level



APF

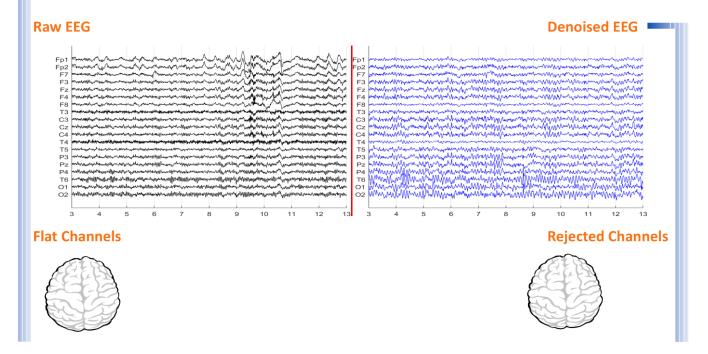
Posterior APF-EC= 09.75

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





Denoising Information (EC)



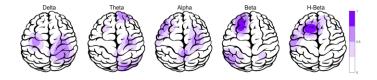
Number of Eye and Muscle Elements				Low Artifact Percentage	
Eye	2	Muscle	2	0	
Total Artifact Percentage				High Artifact Percentage	
				0	
EEG Quali	ity	good		Total Recording Time Remaining 248.10 sec	



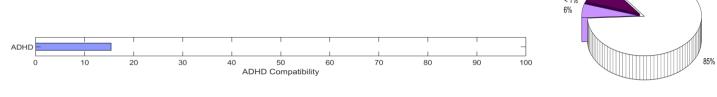


Pathological assessment for ADHD

Compare to ADHD Database

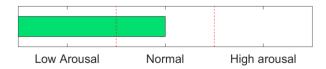


EEG Compatibility with ADHD Diagnosis





Arousal Level Detection



ADHD Clustering *

1.

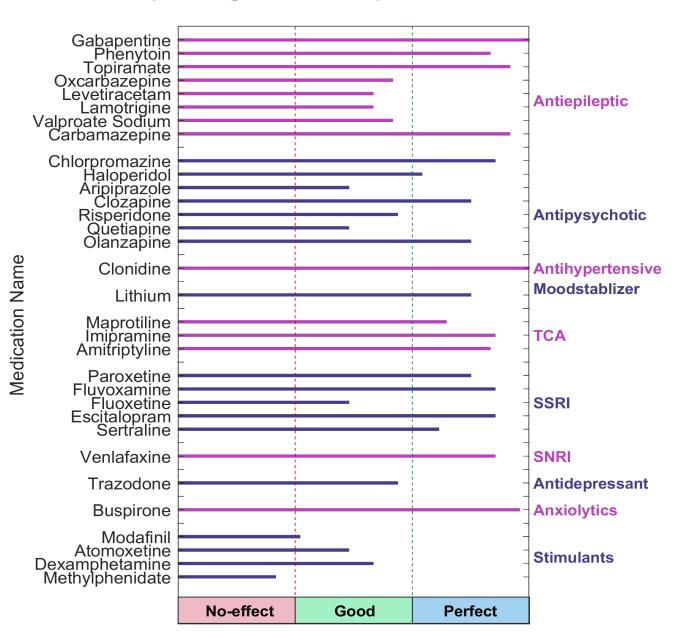
* If there is Paroxymal epileptic discharge in EEG data, this case needs sufficient sleep and should avoid high carbohydrate intake.

You can consider anticonvulant medications.





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.



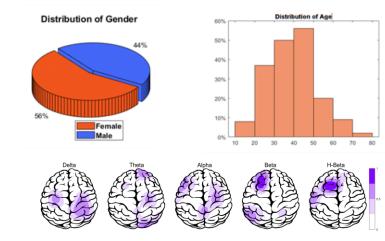


rTMS Response Prediction

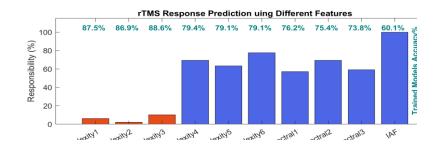
Network Performance

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

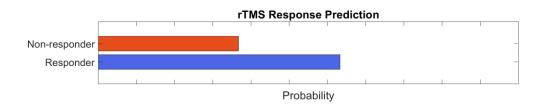
Participants Information



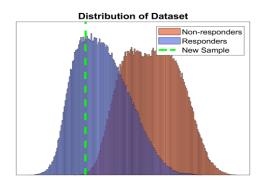
Features Information



Responsibility



Data Distribution



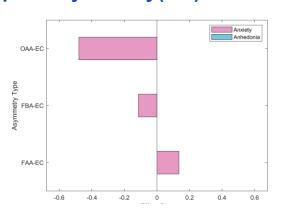
About Predicting rTMS Response

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.

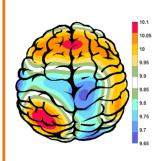




Alpha Asymmetry(AA)



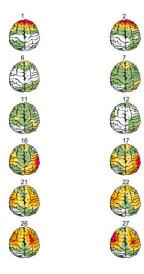
APF(EC)

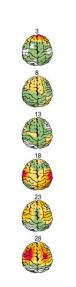


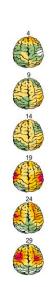
Frontal APF= 09.83

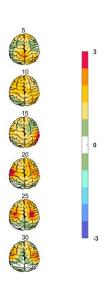
Posterior APF= 09.75

🚃 Absolute Power-Eye Closed (EC) 🠠



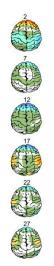






Relative Power-Eye Closed (EC) 🌮







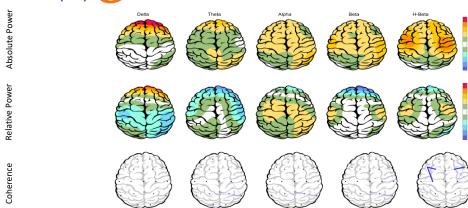




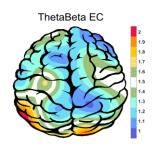


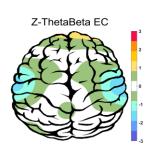


Z Score Summary Information (EC)

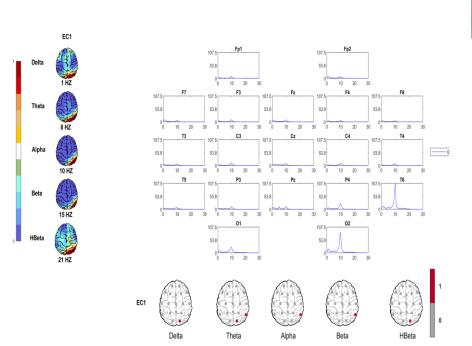


E.C.T/B Ratio (Raw- Z Score)





EEG Spectra



Arousal Level

