





# QEEG Clinical Report BrainLens V0.4

# Report Description

# Personal & Clinical Data

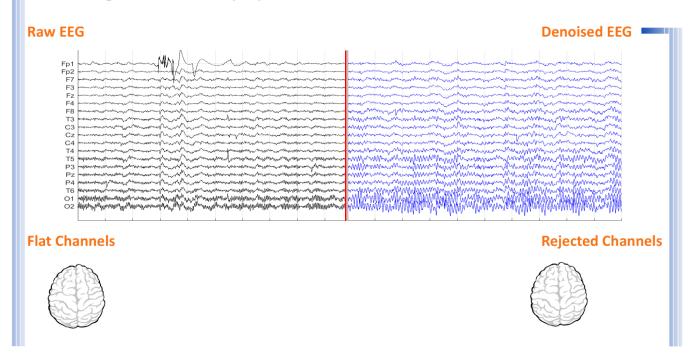
Name	Atefeh Shokofegi	Date of Recording	03-Aug-2024				
Date of Birth - Age	21-Mar-1988 - 36.37	Gender	Female				
Handedness(R/L)	Right	Source of Referral	Dr Masjedi				
Initial Diagnosis	Anxiety-Headache-Anxiety attack						
Current Medication	Medication Free						

Dr Masjedi





# Denoising Information (EC)



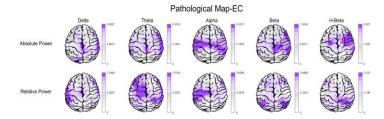
Number of Eye and Muscle Elements			Low Artifact Percentage					
Eye	2	Muscle	0	0				
Total Artifact Percentage				High Artifact Percentage				
0				()				
<b>EEG Quali</b>	ity	bad		Total Recording Time Remaining	285.60 sec			





# Pathological assessment for mood disorders

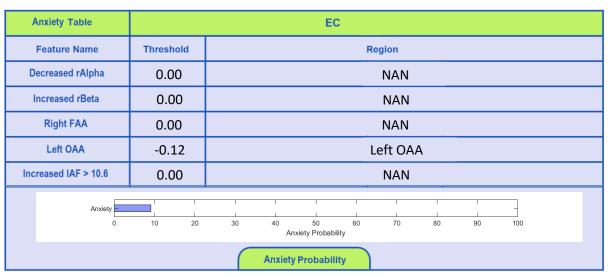
#### **Compare to Mood Disorders Database**



#### **EEG Compatibility with Depression Diagnosis**

Depression Table		EC							
Feature Name	Threshold	Threshold Region							
Increased Global rAlpha	0.00	NAN							
Increased global rTheta	0.50	0.50 global							
Decreased rDelta	0.00	NAN							
Increased rBeta	0.00	NAN							
Left FAA	-0.02	Left FAA							
Right OAA	0.00	NAN							
Decreased Coherence (D, T)	-0.50	Decreased Coherence (D,T)							
Increased Coherence (A, B)	1.00	Increased Coherence (A,B)							
depression 0	10 20	30 40 50 60 70 80 90 100  Depression Probability							
Depression Probability									

# **EEG Compatibility with Anxiety Diagnosis**







### **EEG Compatibility with Mood Swings Diagnosis \***

Mod	od Swings Table	EC										
Fe	eature Name	Threshold	Threshold Region									
Dec	creased rAlpha	0.00	0.00 NAN									
Increas	sed (rDelta+rTheta)	1.00		LF-RF-MF-LT-RT-C-P-								
Inc	creased rBeta	0.00		NAN								
Decreas	ed Alpha Coherence	0.00	NAN									
	Right FAA	0.00		NAN								
	BMD -	T T	ı	ı	-	-	ı	ı	1			
	0	10 20	30	40	50	60	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



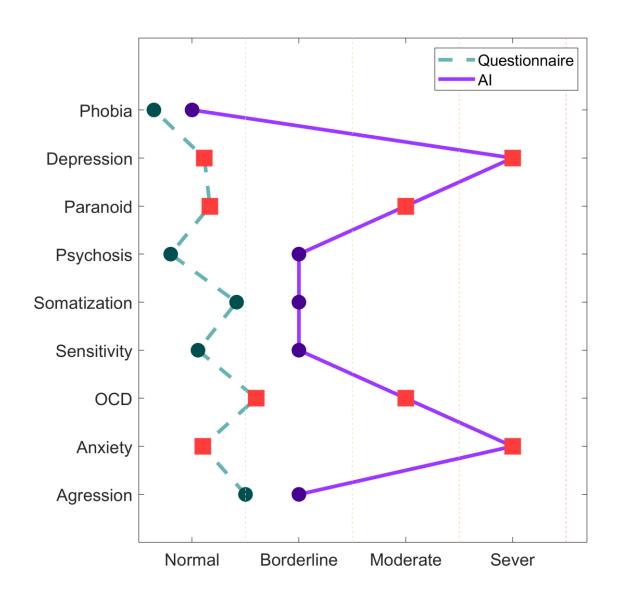
# Arousal Level Detection







#### **AI-Driven Psychometric Symptoms Assessing**



#### **Explanation**

Note

The above diagram illustrates the psychometric 

If a re

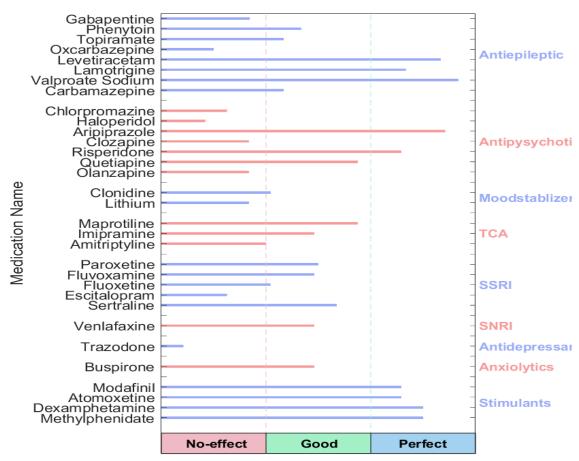
symptoms based on the SCL90 questionnaire of the subject (green line) and AI (purple line). Combination of non-linear EEG markers have been used to estimate these symptoms using AI. All the AI algorithms used in these analysis have an accuracy more than 97.60%, a sensitivity more than 97.54%, and a specificity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



# Explanation



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.



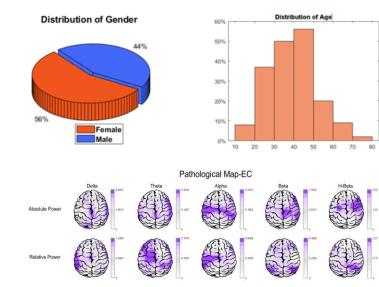


#### rTMS Response Prediction

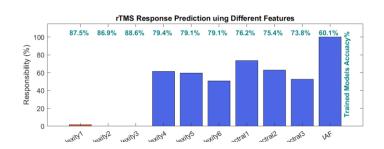
#### Network Performance

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

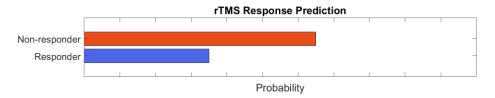
#### Participants Information



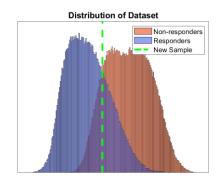
#### Features Information



#### Responsibility



#### Data Distribution



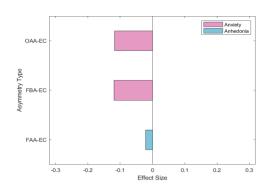
#### About Predicting rTMS Response

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.

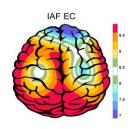




# Alpha Asymmetry(AA)



# APF(EC)



Frontal APF= 08.00

Occipital APF= 09.75

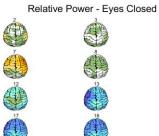
#### 🚃 Absolute Power-Eye Closed (EC) 🌮

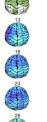


# TBI Severity









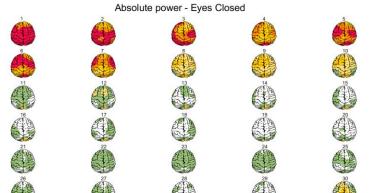


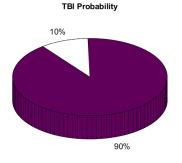


# Relative Power-Eye Closed (EC) 🌮



# TBI Probability



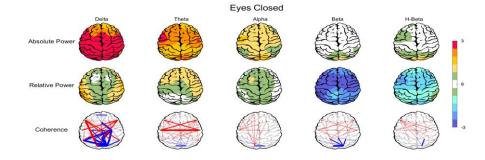






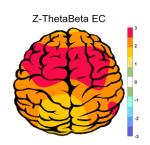
# Z Score Summary Information (EC)



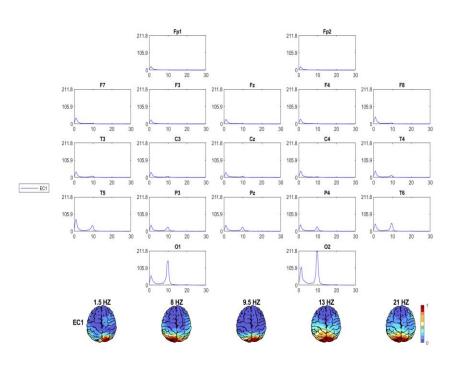


# E.C.T/B Ratio ( Raw- Z Score)

# ThetaBeta EC



# EEG Spectra



# Arousal Level

