

1 **Supplementary Material 8**

2 **Global Network Metrics**

3 **Methods**

4 *Assortativity*

5 Assortativity (R) quantifies the tendency of a network's nodes to connect to other
6 nodes that are similar in some specified attribute, such as degree, which in this context
7 refers to the number of connections a node has. High assortativity indicates that nodes
8 tend to link to others with a similar degree, while low assortativity (or disassortativity)
9 suggests the opposite. The formula for calculating assortativity, focusing on degree
10 assortativity, is:

$$R = \frac{\sum_{ij} jk(e_{jk} - q_j q_k)}{\sigma_q^2}$$

11 Here, e_{jk} is the proportion of edges that connect nodes of degrees j and k , q_j and
12 q_k are the expected proportions of these degrees under randomness, and σ_q^2 is the
13 variance of the degree distribution.

14 *Hierarchy*

15 The hierarchy coefficient is used to identify the presence of a hierarchical
16 organization in a network that measures the extent to which a network is organized in
17 a hierarchical manner, with some nodes acting as hubs that are more central than
18 others. This can be quantified by examining the distribution of paths or the levels of
19 organization within the network, where a high level of hierarchy implies a structured
20 tier of node interactions.

$$\text{bzscore} = \frac{b - \text{mean}(\text{brand})}{\text{std}(\text{brand})}$$

21 Where b is hierarchy of network, $1 \times N$ array, N is the number of threshold sequences;
22 bzscore is the z -score of hierarchy of network, $1 \times N$ array, N is the number of
23 threshold sequences. brand is a $R \times 1$ array, R is the number of randomized network.
24 It is the hierarchy of randomized network.

25 *Synchronization*

26 Synchronization refers to the extent to which the activity of different nodes (or
27 regions in the context of brain networks) in a network becomes synchronized,
28 indicating coordinated activity. In brain networks analyzed through fMRI,
29 synchronization often involves measuring the statistical correlation or coherence of
30 the BOLD signal between regions over time. While synchronization itself is a
31 property observed through these correlations, rather than being calculated through a
32 simple mathematical formula, it can be quantified using measures such as phase
33 locking value (PLV) or Pearson correlation coefficients between pairs of nodes:

Synchronization between nodes i and j = Correlation ($BOLD_i$, $BOLD_j$)

34 Clustering coefficient (C_p)

35 The clustering coefficient (C_p) indicates the likelihood that the neighbors of a given
36 node i in a network are also neighbors with each other. The network's overall
37 clustering coefficient is the average of the clustering coefficients for all nodes within
38 the network. This metric is used to measure the degree of cliquishness or local
39 groupings within the network. The formula for calculating the clustering coefficient is
40 as follows:

$$C_p(G) = \frac{1}{N} \sum_{i=1}^N \frac{2a_i}{k_i(k_i - 1)}$$

41 Here, k_i is the degree of node i , a_i is the actual number of edges between the neighbors
42 of node i , and N is the total number of nodes in the network.

43 Shortest Path Length (L_p)

44 The shortest path length (L_p) represents the optimal path for information to travel
45 from one node to another within a network. The network's shortest path length (L_p) is
46 the average value of the shortest paths between all pairs of nodes in the network. This
47 metric is used to measure the efficiency of global path connectivity for information
48 transmission across the network. The formula to calculate the shortest path length is:

$$L_p(G) = \frac{1}{N(N-1)} \sum_{i \neq j \in N} L_{ij}$$

49 where L_{ij} is the length of the shortest path between nodes i and j in the network G , and
50 N is the total number of nodes in the network.

51 Small-worldness (Sigma)

52 Regular networks are characterized by high clustering coefficients and longer shortest
53 path lengths, in contrast to random networks, which have lower clustering coefficients
54 and shorter shortest path lengths. Small-world networks possess the high clustering
55 coefficients of regular networks as well as the short shortest path lengths of random
56 networks. The small-world property index (Sigma) represents the ratio of the
57 normalized clustering coefficient (C_p) to the normalized shortest path length (L_p).
58 The formula for calculating Sigma is:

$$\text{Sigma} = \frac{C_p / C_{p_{\text{rand}}}}{L_p / L_{p_{\text{rand}}}}$$

59 where $C_{p_{\text{rand}}}$ and $L_{p_{\text{rand}}}$ are the average clustering coefficient and average shortest
60 path length for 5000 random networks, respectively. When the ratio exceeds 1, the
61 network is considered to have small-world properties.

62 Global Efficiency (E_{global})

63 Global efficiency (E_{global}) reflects the overall capacity of a network for transmitting
64 information. The higher the global efficiency of a network, the faster the rate at which
65 information can be exchanged between nodes. The formula for calculating global
66 efficiency is:

$$E_{\text{global}}(G) = \frac{1}{N(N-1)} \sum_{i \neq j \in G} \frac{1}{L_{ij}}$$

67 Here, L_{ij} is the length of the shortest path between nodes i and j in the network G ,
68 and N is the total number of nodes in the network.

69 Local Efficiency (E_{local})

70 Local efficiency (E_{local}) reflects the capacity of a network for local information
71 transmission, and, in a sense, represents the network's resilience to random attacks.

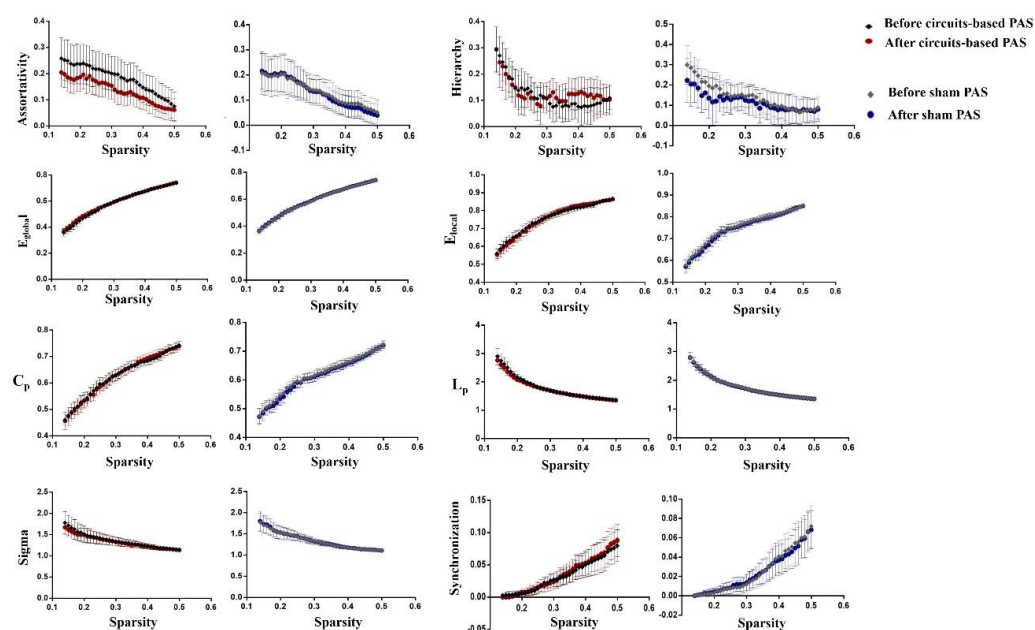
72 The formula for calculating local efficiency is:

$$E_{\text{local}}(i) = \frac{1}{N-1} \sum_{i \neq j \in G} \frac{1}{L_{ij}}$$

73 Here, L_{ij} is the shortest path length between nodes i and j within the subnetwork G
 74 formed by the neighbors of node i , and N is the total number of nodes in the network.

75 Results

76 No significant variations were detected in global network metrics under sparsities
 77 ranging from 0.14 to 0.5, when comparing the effects of circuits-based PAS and sham
 78 interventions in aMCI patients. Analysis covered various graph theory metrics,
 79 including assortativity, hierarchy, synchronization, clustering coefficient (C_p), shortest
 80 path length (L_p), small-world index (Sigma), global efficiency (E_{global}), and local
 81 efficiency (E_{local}) (all $p > 0.05$) (Figure S8-1).



82
 83 **Figure S8-1. Graphical representation of brain network attributes before and after circuits-based**
 84 **PAS and sham intervention in aMCI patients.** Global graph theory metrics, including assortativity,
 85 hierarchy, synchronization, clustering coefficient (C_p), shortest path length (L_p), small-worldness
 86 (Sigma), global and local efficiency (E_{global} , E_{local}), were assessed at varying sparsity thresholds from
 87 0.14 to 0.5. Data points show no significant difference between pre- and post-intervention states for
 88 both treatments (all $p > 0.05$). PAS: paired associative stimulation; aMCI: amnesic mild cognitive
 89 impairment.