## **Supplementary Information**

## Higher-order interactions shape collective dynamics differently in hypergraphs and simplicial complexes

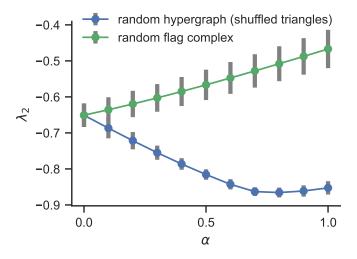
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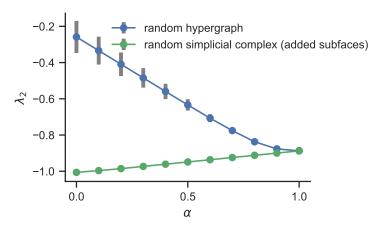
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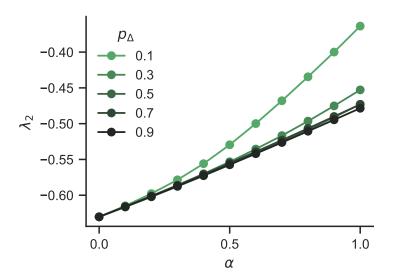


Supplementary Figure S1: Synchronization is enhanced by higher-order interactions in random hypergraphs but is impeded in simplicial complexes. Here, we build the simplicial complexes as random flag complexes by generating ER graphs with wiring probability p and filling all closed triangles with 2-simplices. For each realisation, we build a corresponding random hypergraph by keeping the same 1-hyperedges but move each 2-hyperedge to a random location. This ensures that the random hypergraph has the same number of 1-hyperedges and 2-hyperedges as the original simplicial complex. The maximum transverse Lyapunov exponent  $\lambda_2$  is plotted against  $\alpha$  for random hypergraphs (blue) and simplicial complexes (green). As  $\alpha$  is increased, the coupling goes from first-order-only ( $\alpha=0$ ) to second-order-only ( $\alpha=1$ ). Each point represents the average over 5 independent hypergraphs or simplicial complexes with n=100 nodes. The error bars represent standard deviations. We set p=0.4 for the simplicial complexes.

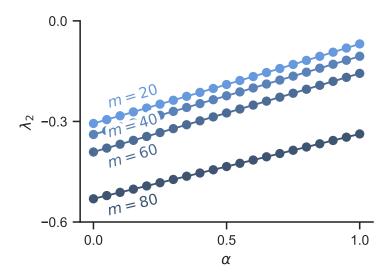
<sup>\*</sup>Y.Z. and M.L. contributed equally to this work.



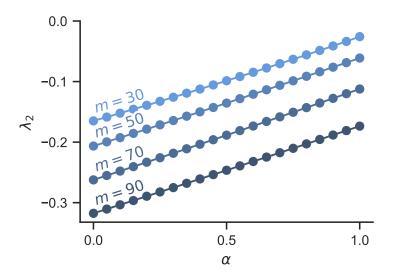
Supplementary Figure S2: Synchronization is enhanced by higher-order interactions in random hypergraphs but is impeded in simplicial complexes. Here, we build the random hypergraphs by generating each possible 1-hyperedge and 2-hyperedge with probabilities p and  $p_{\triangle}$ , respectively. For each realisation, we build a corresponding random simplicial complex by adding 1-hyperedges to satisfy the inclusion condition. The maximum transverse Lyapunov exponent  $\lambda_2$  is plotted against  $\alpha$  for random hypergraphs (blue) and simplicial complexes (green). As  $\alpha$  is increased, the coupling goes from first-order-only ( $\alpha=0$ ) to second-order-only ( $\alpha=1$ ). Each point represents the average over 5 independent hypergraphs or simplicial complexes with n=100 nodes. The error bars represent standard deviations. We set  $p=p_{\triangle}=0.1$  for the random hypergraphs.



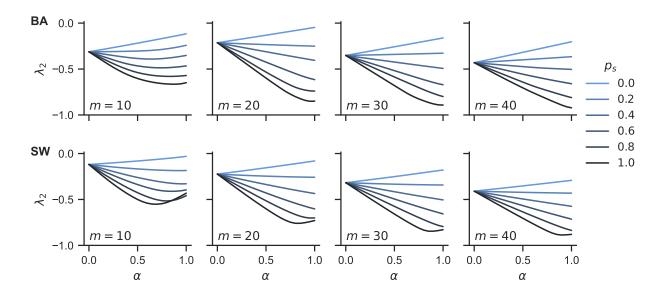
Supplementary Figure S3: Synchronization is impeded by higher-order interactions in simplicial complexes regardless of the probability of closed triangles being filled. Here, we build the simplicial complexes as random flag complexes by generating ER graphs with wiring probability p and filling closed triangles with probability  $p_{\triangle}$ , for a range of  $p_{\triangle}$  values. The maximum transverse Lyapunov exponent  $\lambda_2$  is plotted against  $\alpha$ . As  $\alpha$  is increased, the coupling goes from first-order-only ( $\alpha=0$ ) to second-order-only ( $\alpha=1$ ). Each point represents the average over 50 independent simplicial complexes with n=100 nodes. We set p=0.4 for the generation of the original ER graphs.



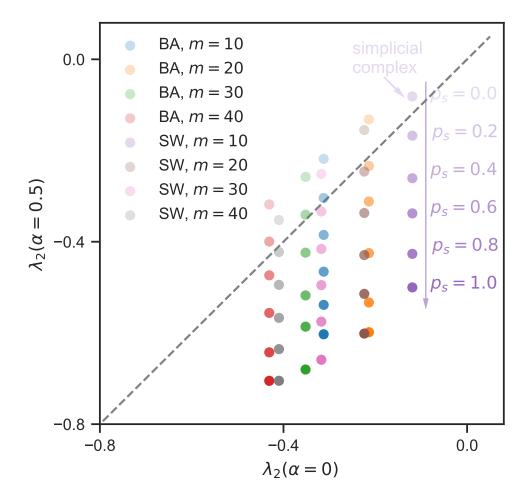
Supplementary Figure S4: Nonpairwise interactions impede synchronization in simplicial complexes constructed from scale-free networks. Here, each curve corresponds to a simplicial complex constructed from a scale-free network with n=300 nodes and mean degree 2m. A scale-free network of n nodes is grown by starting from an m-clique and attaching each new node to m existing nodes. The probability of attachment is proportional to the current degrees of the existing nodes.



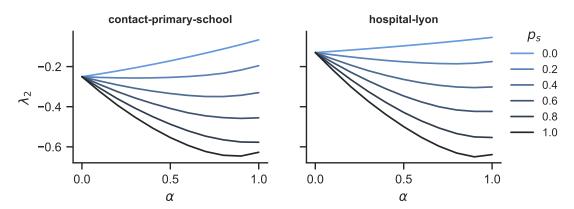
Supplementary Figure S5: Nonpairwise interactions impede synchronization in simplicial complexes constructed from small-world networks. Here, each curve corresponds to a simplicial complex constructed from a small-world network with n=300 nodes and mean degree m. A small-world network of n nodes is constructed by starting from a ring network with each node joined with its m nearest neighbors and randomly rewiring each link with probability  $p_r=0.15$ .



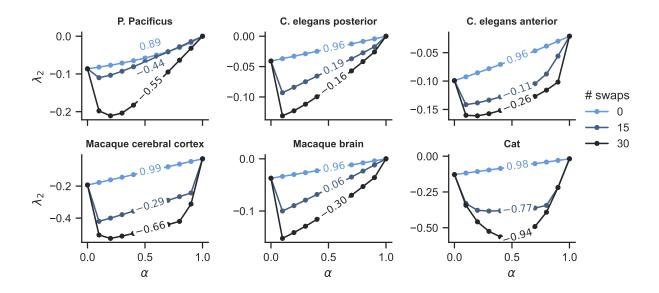
Supplementary Figure S6: Synchronization stability of hypergraphs constructed from synthetic networks. Analog of main text Fig. 5, but for scale-free (BA) networks and small-world (SW) networks. The hypergraphs are constructed in the same way as in main text Fig. 5. For each value of  $p_s$ , we plot synchronization stability  $\lambda_2$  (averaged over 100 independent realizations) as a function of the control parameter  $\alpha$ . We fix the network size to n=100 for both classes of networks. The BA networks have a mean degree of 2m, whereas the SW networks have a mean degree of m. Regardless of the mean degree, the role of higher-order interactions always transitions from impeding synchronization to promoting synchronization as  $p_s$  is increased.



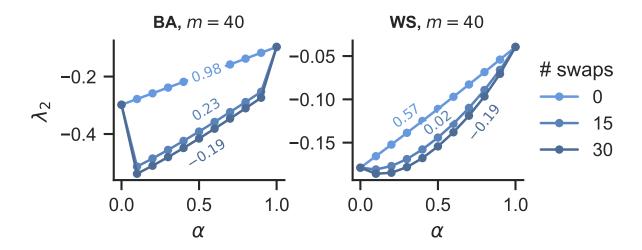
Supplementary Figure S7: Nonpairwise interactions enhance synchronization in hypergraphs constructed from scale-free (BA) networks and small-world (SW) networks, except when the hypergraph structure is close to being a simplicial complex. This plot uses the same data as in Supplementary Fig. S6 and is the analogy of main text Fig. 6 for synthetic networks.



Supplementary Figure S8: Analog of main text Fig. 5 based on real-world hypergraphs. The hypergraphs come from social contacts in a Lyon hospital and a primary school. They are known to be close to being simplicial complexes [1]. From these datasets, we keep hyperedges up to order two and consider the largest connected component. For each value of  $p_s$ , we plot synchronization stability  $\lambda_2$  (averaged over 100 independent realizations) as a function of the control parameter  $\alpha$ . The role of higher-order interactions quickly transitions from impeding synchronization to promoting synchronization as  $p_s$  is increased.



Supplementary Figure S9: Cross-order degree correlation affects synchronization stability in systems with mixed pairwise and nonpairwise interactions ( $0 < \alpha < 1$ ). This plot uses the same networks as in main text Fig. 5 and extends main text Fig. 7. Curves are colour-coded by the number of pairs of nodes that were selected to swap their 2-simplices memberships: 0, 15, or 30. The resulting degree correlation is indicated on each curve.



Supplementary Figure S10: Cross-order degree correlation affects synchronization stability in systems with mixed pairwise and nonpairwise interactions ( $0 < \alpha < 1$ ). The hypergraphs are constructed from scale-free and small-world networks, with the same parameters as in Supplementary Figs. S4 and S5: n = 300, m = 40, and  $p_r = 0.15$ . Curves are colour-coded by the number of pairs of nodes that were selected to swap their 2-simplices memberships: 0, 15, or 30. The resulting cross-order degree correlation is indicated on each curve.

## References

[1] Lotito, Q. F., Musciotto, F., Montresor, A. & Battiston, F. Higher-order motif analysis in hypergraphs. *Commun. Phys.* **5**, 79 (2022).