Large-scale brain networks underlying domain-specific memory, intelligence, and academic performance



Poster M₇32



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Introduction

Academic performance is believed to rely, in part, on intelligence. This phenomena has been studied at length, but findings concerning the neural substrates of intelligence in intrinsic brain organization have been mixed, and little attention has been paid to how the relationships between intelligence and brain organization vary between sexes.

Here, we investigate the role of IQ in domain-specific learning in a historically male-dominated domain. We further probe how the intrinsic organization of large-scale brain networks reflects IQ and its role in learning, and how this relationship differs with respect to biological sex.

IQ is differentially related to brain network organization with respect to class type & sex Modeling class Resting-state

No significant relationships between resting-state brain network organization and IQ.

Methods

Participants

- 107 RH undergraduate students (48 female)
- No history of neurological/psychiatric diagnosis
- Enrolled in first calculus-based physics course (two course types: active learning instruction & traditional lecture)

Data collection

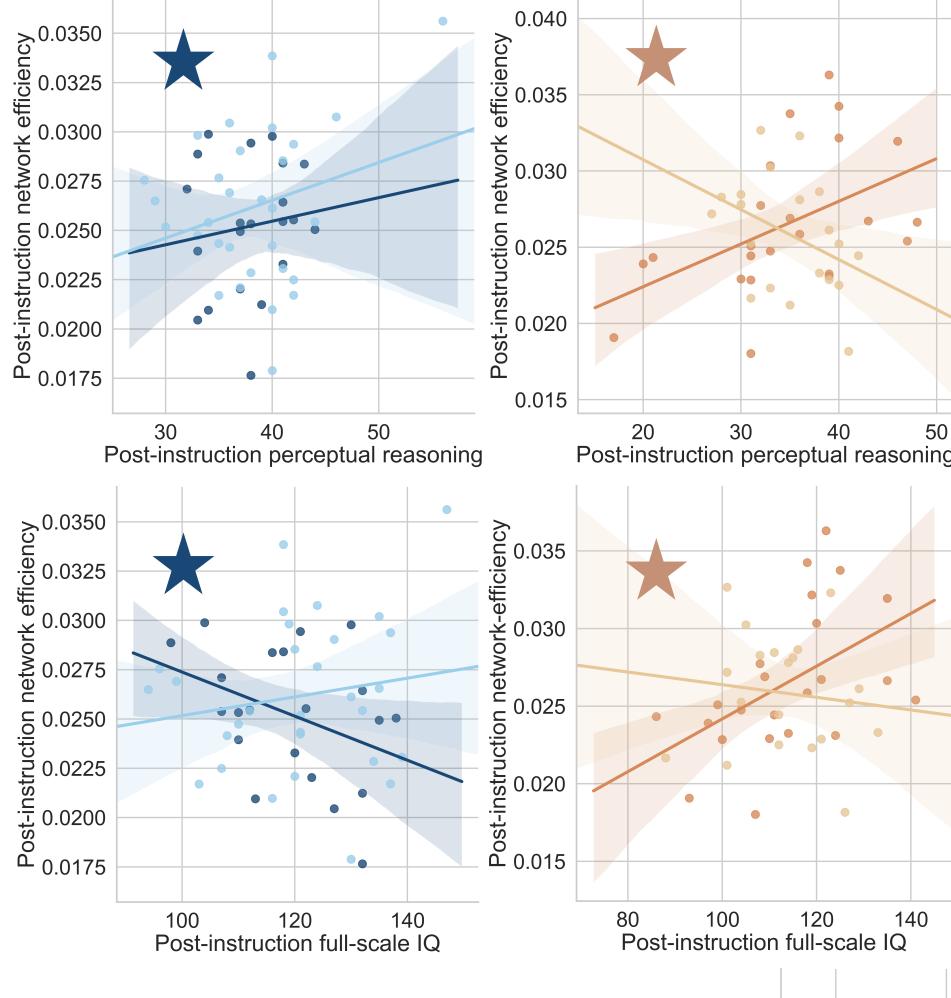
- All data were collected both pre- and postinstruction
- Behavioral data
 - o IQ (WAIS-IV)
 - Grade in the physics course
 - Accuracy during tasks in the scanner
- o MRI data
 - o T1-weighted structural images
 - Resting-state functional MRI (fMRI; 12 minutes)
 - Retrieval task (physics, general)
 - Physics reasoning task (Force Concept Inventory; FCI)
 - Transitive inference task

Data analysis

- Preprocessed to mitigate head motion, lowfrequency drift, aligned to MNI template
- BOLD signals were extracted, per region, using data-driven, functional brain parcellations (Fig. 1) and correlated, pairwise for connectivity matrices
- Efficiency, modularity, and path length calculated using bctpy for each subject, task, condition, session
- Paired t-tests for change in IQ scores pre- to post-instruction
- OLS regression: pre, post, change in IQ & class type, controlling for age, year in uni.
- OLS regression: pre, post, change in IQ & area under the curve of topological metrics, controlling for head size, average FD, year in uni., age
- Corrected for multiple comparisons under dependence (Li & Ji, 2005; Šidák, 1967)

Figure 1. Parcellations used for assembly of brain graphs

All topological measures were calculated twice, once for each parcellation, to verify that our results hold across network definitions. Both the Craddock (2012; above) and Finn/Shen (2015; below) are functionally defined, with 270 and 268 regions, respectively.



Force Concept Inventory

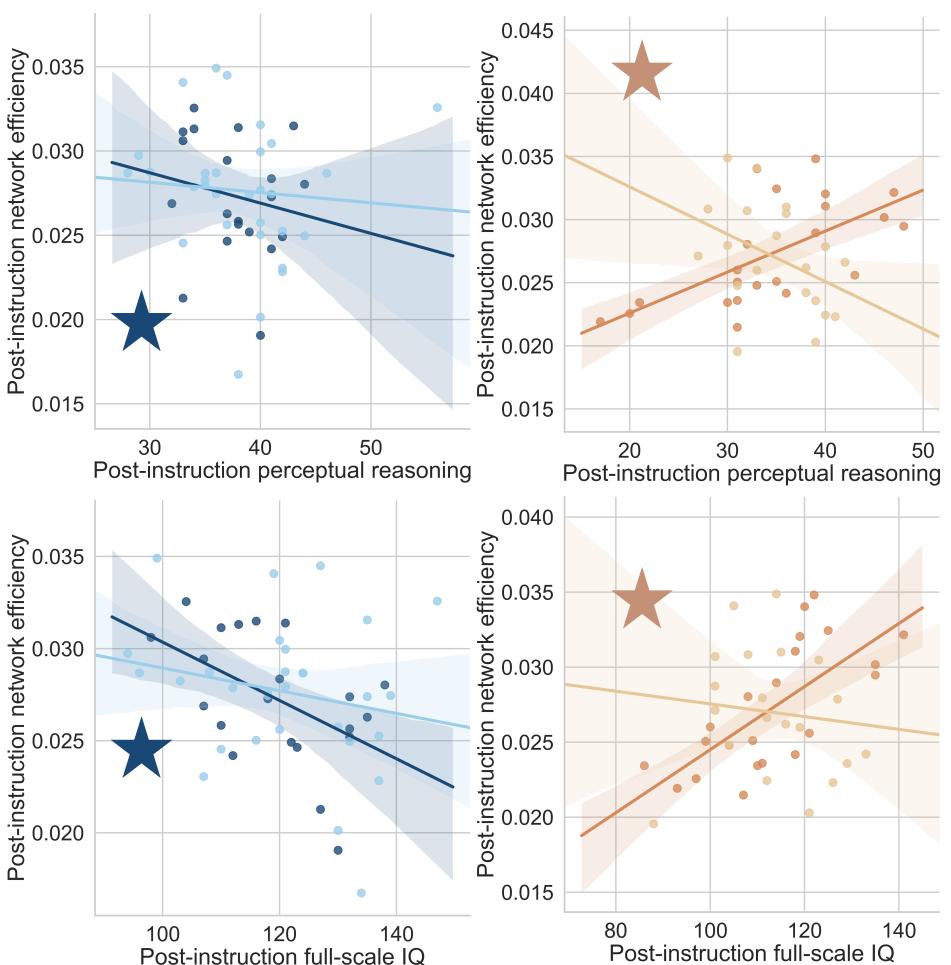
Post-instruction network efficiency, characteristic path length during physics reasoning in the FCI task was also significantly related to post-instruction PRIQ and FSIQ (Fig. 5).

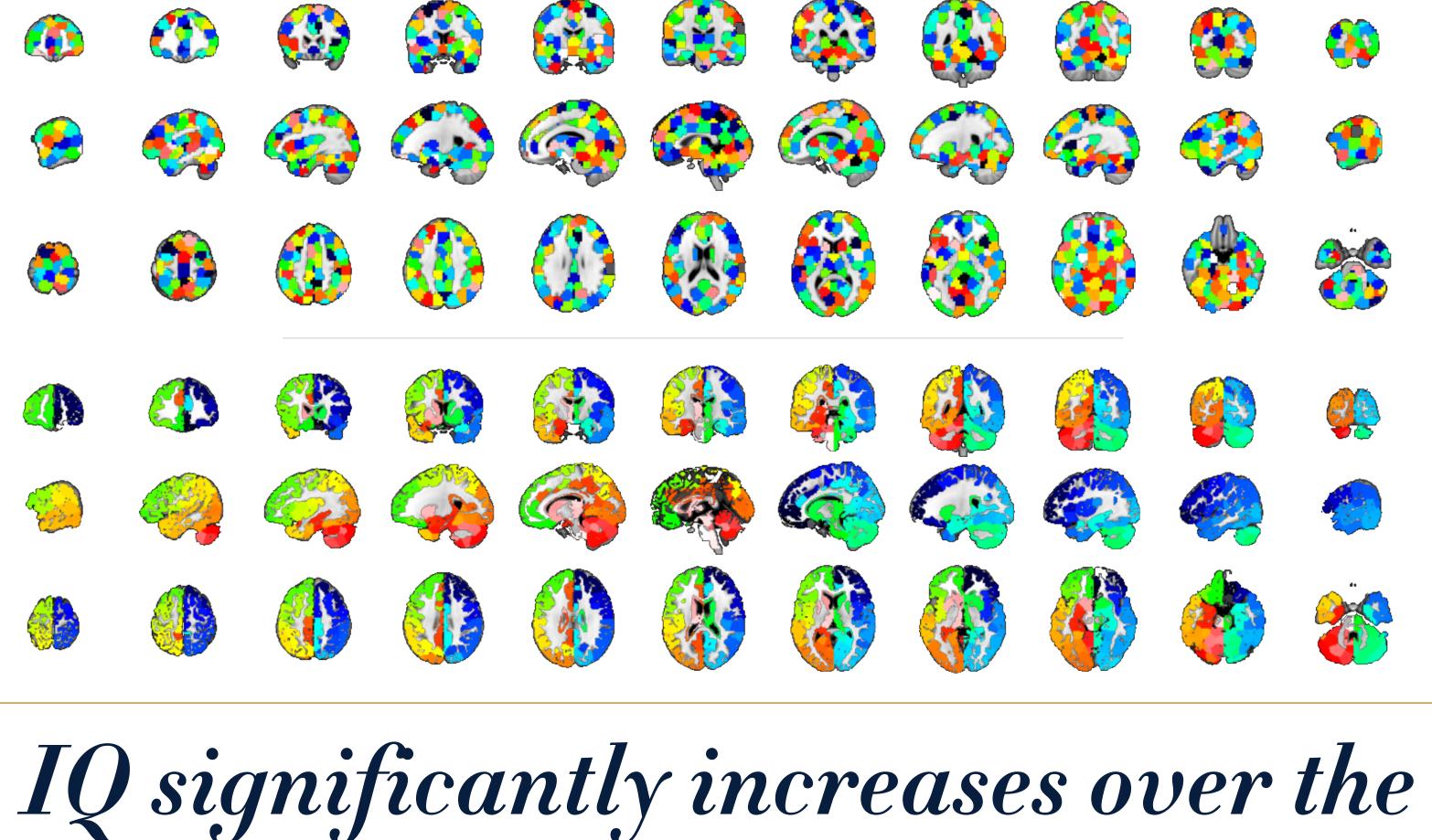
Physics Retrieval

Post-instruction characteristic path length and efficiency are related to post-instruction PRIQ and FSIQ (Fig. 4: efficiency shown, relationship is the same for path length), but the direction of these relationships is different for female and male students, and depending on class type in which they were enrolled. No trends in general retrieval.

Figure 4. Perceptual reasoning (top), full IQ (bottom) and efficiency during physics retrieval in male and female students

PRIQ (top) and FSIQ (bottom) are significantly related to network efficiency and characteristic path length (not shown), with significant interactions between IQ, sex, and class type (modeling vs. lecture)





course of a physics class

- Although IQ is assumed to be stable, full-scale IQ (FSIQ) scores in both male and female students significantly increase over the course of the class ($t_{male} = 11.3$, $t_{female} = 7.8$, p < 0.001) Only Perceptual Reasoning (PRIQ; $t_{male} = 3.7$, $t_{female} = 2.3$) & Processing Speed ($t_{male} = 4.6$, $t_{female} = 4.3$) subscores significantly increased (p < 0.001)
- No significant differences in changes in IQ score between male and female students

- Direction of IQ-brain relationship
 depends on sex and class
- **Topology differs between the sexes and classes**, too, differently per class and sex.

Post-instruction network topology is also related to **changes in both PRIQ and FSIQ** over the course of the semester (in the same directions seen in Fig. 5)

- Direction of IQ-brain relationship
 depends on sex and class
- No relationship between sex, class, or sexXclass and topology.
 No such trends seen in control condition.

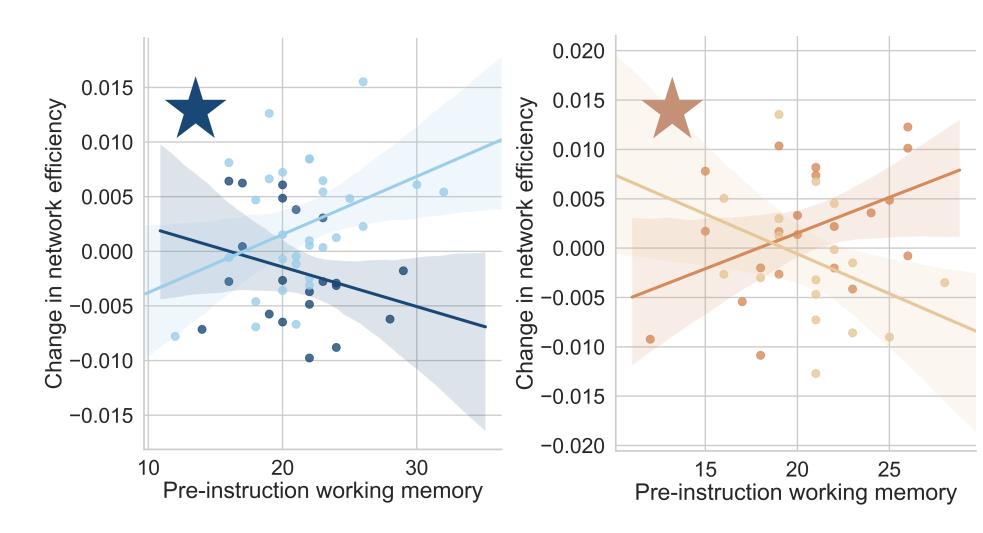
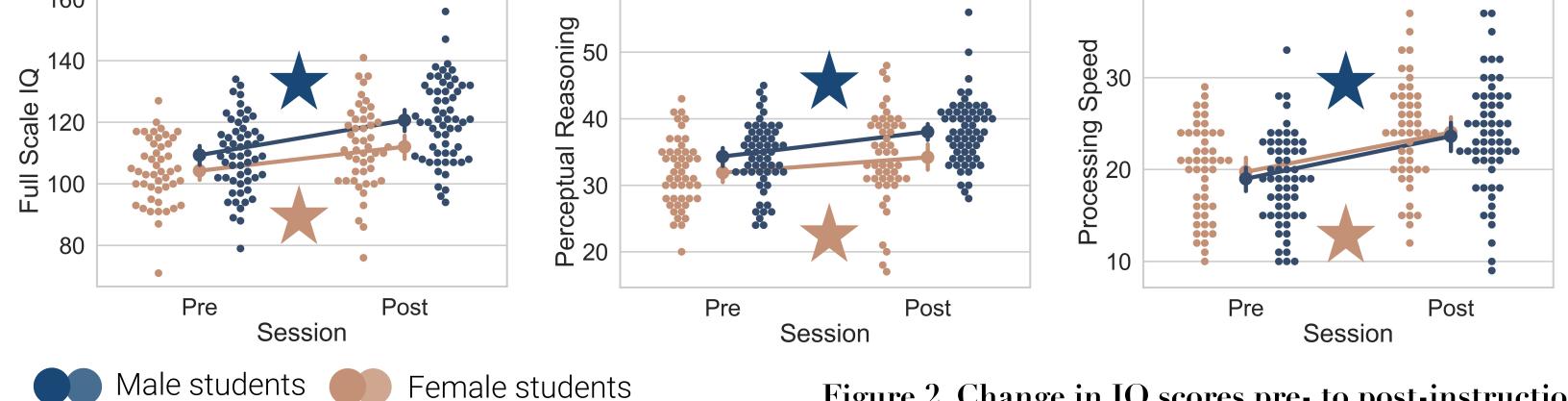


Figure 5. Perceptual reasoning, full IQ and efficiency during physics reasoning in male and female students PRIQ (top) and FSIQ (bottom) are significantly related to network efficiency and characteristic path length (not shown), with significant interactions between IQ, sex, and class type (modeling vs. lecture)

Transitive Inference

Changes in network topology over the course of the semester were significantly related to preinstruction working memory scores Pre-instruction topology 6). (Fig. condition is during the control significantly verbal related to comprehension and this relationship differs with sex and class type, too.



Female students Figure 2. Change in IQ scores pre- to post-instruction Male and female students' changes in Full Scale IQ and it s subscores across the course of the semester.

Only full-scale IQ increases in male students were related to course type (modeling/lecture), when controlling for age and number of years at university (F = 4.083, p(F) = 0.0113)

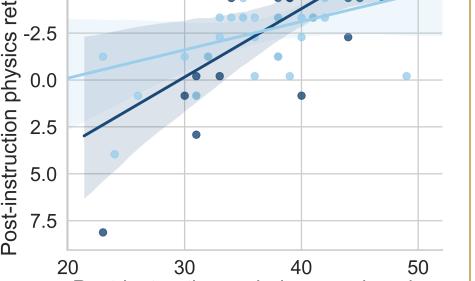
IQ neither predicts nor reflects learning in a physics class

- Neither pre-instruction nor changes in IQ scores were related to course grades or changes in task accuracy
- **Post-instruction Verbal Comprehension** is associated with **post-instruction physics retrieval accuracy** in **male students** (p < 0.01), after controlling for age, years at university, and class type.

Figure 3. Verbal comprehension & physics retrieval accuracy in male students

Male students' changes in Full Scale IQ and its subscores across the course of the semester, in lecture and modeling classes.

Lecture class Modeling class



v -7.5

2 -5.0

Post-instruction verbal comprehension

Figure 6. Working memory and efficiency during general reasoning in male and female students WMIQ significantly related to network efficiency and characteristic path length (not shown), with significant interactions between IQ, sex, and class type (modeling vs. lecture)

Conclusions

- IQ significantly increases over the course of a semester of calculus-based physics, driven by increases in perceptual reasoning and processing speed, for all students.
- IQ is unrelated to outcomes measures, making it a poor predictor of success in a physics course, with one exception: post-instruction verbal comprehension predicts post FCI accuracy.
- Perceptual reasoning and full-scale IQ are related to brain network topology during physicsrelated tasks post-instruction and these relationships differ based on sex and course type.
- o Task-based topology is more related to IQ than resting-state, modularity was not related to IQ.

References

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Funding

This work was funded in part by the National Science Foundation (NSF REAL DRL-1420627).

