





Introduction

Auditory-motor synchronization is a validated method used to address gait [1], upper limb [2], and motor speech [3] rehabilitation in Parkinson's disease and other populations with a motor neurological disorder [4].

Predictable auditory cues optimize movement patterns while reducing the brain's reliance on dopaminergic (DA) responses in the ventral striatum, albeit exact mechanisms for the reduction in DA uptake require further investigation [5]. AIM

This study investigates auditory-motor synchronization using network science methods. HYPOTHESIS

Global efficiency (GE) and clustering coefficient (CC) will be higher during the fingertapping auditory-motor to 1Hz tone task, in comparison to the self-paced tapping task.

Methods

PARTICIPANTS ($n = 14$)	
 Fourteen non-musician healthy adults Age: 22-35 years old (M = 30.4 ± 5) Education: Bachelor (4), Masters (7), PhD (3) Right-handed 	 Sex: 7 females, 7 males Race: Caucasian (8), AA Asian (1), Caribbean (1)
 PROTOCOL Structural T1 fMRI tasks 6-MIN BLOCK DESIGN Self-paced tapping at a referenced 1Hz tempo Finger-tapping to a steady 1Hz auditory beat 	MATERIAL 3-T MRI Siemens Magnete syngo MR D13 fMRI time series used grac images (TR/TE=2000/25 n
 ROIS – Subnetworks [6, 7] [1] Auditory (AUD) [2] Basal ganglia network (BGN) [3] Central executive network (CEN) [4] Dorsal attention network (DAN) [5] Default mode network (DMN) [6] Sensorimotor (MOT) [7] Orbitofrontal cortex (OFC) 	BEHAVIORAL DATA E-prime 3.0 (Psychology to auditory stimuli and visual and a SRBOX captured fin- times.

PRE-PROCESSING

[8] Saliency (SAL)

[9] Visual (VIS)

fMRI data were realigned, slice-time corrected and co-registered to the structural images. Structural images were warped and normalized to MNI template space using SPM8.

NETWORK GENERATION

Pearson's correlation was applied to the time series voxel correlation pairs to produce a crosscorrelation matrix, on which a threshold $S = \log(N)/\log(K)$ was applied to each participants. This dichotomized the data into equivalent node density binary adjacency matrix (Aij) [8]. S = 2.5 was used based on previous work about network fragmentation [9]. *N* = number of network nodes; *K* = average node degree/links

NETWORK ANALYSIS

fficial journal of the European Gastrointestinal Motility Society, 34(4), e14271. https://doi.org/10.1111/nmo.1427

The spatial consistency of nine brain subnetworks as community structures were investigated at the group level using a two-sample Jaccard index permutation test based on the scaled inclusivity (SI) statistics [10].

A mixed-effect multivariate regression model was used to test the hypotheses that changes in global efficiency (GE) and clustering coefficient (CC) were associated with the experimental tasks in the studied subnetworks [11, 12].

First order (direct) and second order (indirect) connectivity analyses were performed to understand the connections from the BGN with other subnetworks [13].

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ALTERED BASAL GANGLIA TOPOLOGY ASSOCIATED WITH AUDITORY-MOTOR SYNCHRONIZATION

^a Music and Health Science Research Collaboratory, University of Toronto, Toronto, Ontario ^b Laboratory for Complex Brain Networks, Wake Forest School of Medicine, Winston-Salem, North Carolina

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1 st & 2 nd Order Connectivity Analysis from the BGN										
		BGN	AUD	CEN	DAN	DMN	OFC	SMN	SN	VIS
1 st order	Self-paced	10.41%	15.67%	8.68%	19.03%	4.88%	12.66%	13.79%	7.44%	7.44%
(direct)	1Hz tone	<mark>21.64%</mark>	4.38%	9.67%	8.74%	15.99%	5.18%	12.57%	11.43%	10.40%
2 nd order (indirect)	Self-paced	13.69%	2.30%	10.01%	27.14%	2.51%	11.84%	13.48%	9.51%	9.51%
	1Hz tone	2.89%	5.72%	12.38%	11.35%	19.81%	2.39%	18.19%	12.61%	14.67%

Percentages represent the distribution of connections for the BGN with other subnetworks.

First order (direct) connections from BGN



ed on first and second order connectivity analyses at the group level.

	Visualizations generated using MRIcron software (ht	ttp://www.nitrc.org/projects/mricron), base
<u>.9</u>	Collaborative Program In Neuroscience (CPIN)	Contact
	Stéphanie K. Lavigne, MA, MT-BC, MTA, NMT	Music and Health Science Researc
he	PhD Candidate, <u>stephanie.lavigne@mail.utoronto.ca</u> in <u>stephanie-lavigne-ma-mt-bc-mta-nmt</u>	90 Wellesley St. W, Toronto

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Clustering	Coefficient
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Wake Forest University School of Medicine

Mixed-Effects Multivariate Regression Model within the BGN					
Finger-tapping to 1Hz tone <u>in comparison</u> to self-paced tapping					
Model Parameters	Network Parameter	Estimate	<i>p</i> -value		
Strength of connection	Global efficiency	0.0236	.0064		
	Clustering coefficient	-0.01663	0.0487		
Connection	Global efficiency	0.1801	.0112		
probability	Clustering coefficient	-0.3513	<.0001		
Confounding variables (5): age, sex, years of education, right-handedness score, race					

Efficiency is a measure of information flow in the brain. The increased BGN global efficiency in the presence of predictable auditory cues during a finger-tapping task may indicate the influence of auditory cueing on higher neural synchronization within and via the basal ganglia subnetwork.

ch Collaboratory Laboratory for Complex Brain Networks sic.utoronto.ca http://lcbn.wakehealth.edu b, ON M5S 1C5 Medical Center Blvd, Winston-Salem, NC 27127

Conclusion