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INDIVIDUAL DIFFERENCES IN BILINGUAL EXPERIENCE MODULATE EXECUTIVE CONTROL NETWORK AND PERFORMANCE: BEHAVIORAL AND STRUCTURAL NEUROIMAGING EVIDENCE^g

Dual/multiple language use has been suggested to affect human cognition and neural substrates. Nevertheless, considerable variability emerges concerning replicability of such effects, likely originating in the common practice of reducing the spectrum of bilingualism to a dichotomy of presence vs. absence (i.e., bi- vs. monolingualism), thus diluting the role of interindividual variability in bilingual experience in modulating neuroplastic and cognitive changes. To address this, we operationalized the main bilingual experience factors as continuous variables, investigating their effects on executive control (EC) performance and neural substrate deploying a Flanker task and structural MRI. Higher L2 proficiency predicted better executive performance. Moreover, neuroimaging results indicated that bilingualism-related neuroplasticity may peak at a certain stage of bilingual experience and eventually revert, possibly following functional specialization. Indeed, experienced bilinguals optimized behavioral performance independently of volumetric variations in executive areas. We conclude that individual differences in bilingual experience modulate bilingualism's cognitive and neural consequences.

JEL Classification: Z.

Keywords: bilingualism, bilingual experience factors, executive control, structural MRI, region-based-morphometry

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1. Introduction

Existing evidence suggests that, in a bilingual brain, two language representations are simultaneously active and that they compete for selection during language use (e.g., Kaushanskaya & Marian, 2007; Kroll, Bobb & Hoshino, 2014; Novitskiy, Myachykov & Shtyrov, 2019). Thus, bilinguals need to exercise constant control over their two languages in order to constrain competition and successfully carry out communication. In other words, cognitive mechanisms of selection, inhibition and switching must be continuously activated in order to be able to manage the potential interference between the two language codes. To accomplish such a task, bilinguals rely on a set of abilities that goes by the name of *language control* (Abutalebi & Green, 2007; Green & Abutalebi, 2013). Bilingual language control relies on a neural network comprising cingulo-fronto-parietal cortical areas as well as subcortical structures, largely overlapping with the general-domain executive control network (Abutalebi & Green, 2016). It is argued that, due to the overlap between the mechanisms and brain areas implied in bilingual language control and general executive control, the continuous training experienced by bilingual users may eventually impact their general-domain executive functioning, both at the cognitive and at the neural level (e.g., Bialystok, 2017; Kroll, Dussias, Bice & Perotti 2015). Indeed, several studies reported volumetric changes in areas associated with bilingual language control, with highly proficient bilinguals showing greater grey matter density than age-matched monolinguals (for a review see Li, Legault & Litcofsky, 2014). Similar bilingualism-induced neuroplastic changes have been reported for the white matter tracts connecting key areas of the language control network, such as the inferior and superior longitudinal fasciculi and the inferior fronto-occipital fasciculus (e.g., Gold, Johnson & Powell, 2013; Luk, Bialystok, Craik & Grady, 2011a). These and similar studies have been taken to suggest specific structural brain changes in the executive network as a result of multiple language use.

In similar vein, bilingual experience has also been argued to affect executive control at the behavioral level, with several studies reporting highly proficient bilinguals outperforming monolinguals on a number of executive tasks, across different age groups (for a review, see Bialystok, Craik & Luk, 2012). Nevertheless, considerable variability between the results can be observed in bilingual research. Importantly, several studies failed to replicate findings of enhanced executive performance in bilinguals (e.g., Paap, Johnson & Sawi, 2015) leading some to question the existence of any beneficial effect of bilingualism on cognition (e.g., Lehtonen, Soveri, Laine, Järvenpää, de Bruin & Antfolk, 2018). Such conflicting evidence has been mainly attributed to inconsistencies in how bilingualism is defined (e.g., Luk & Bialystok, 2013; Mishra, 2015; Surrain & Luk, 2019). Indeed, although inter-individual variability is intrinsic to such a multifaceted phenomenon as bilingualism, the tendency to reduce diverse linguistic profiles to a dichotomous categorization (bilinguals vs. monolinguals) still persists in the literature. This approach often neglects important differences within heterogeneous groups and downplays the role of individual differences in profiles of bilinguals in modulating neuroplastic and cognitive changes (e.g., Luk and Bialystok, 2013; Bialystok, 2016). In other words, neglecting the fact that bilingualism is a continuous rather than a binary variable, may make study designs insensitive to its influence on other neurocognitive functions.

At the same time, existing research indicates that individual differences in bilingual experience factors (BEFs) may play an important role in modulating the effects of dual language use on cognition and neuroplasticity. Regarding the impact of individual BEF differences on executive control at the behavioral level, previous studies mainly focused on the role of second language (L2) proficiency. When comparing bilinguals with differing levels of L2 proficiency, higher L2 proficiency has generally been found to be correlate with a greater improvement in executive control. This result has been reported as better conflict resolution performance in a lateralized attention network task (Tao, Marzecová, Taft, Asanowicz & Wodniecka, 2011), in

both canonical (Iluz-Cohen & Armon-Lotem, 2013) and oculomotor (Singh & Mishra, 2012; 2013) versions of the Stroop task, in reduced conflict effect in a Flanker task (Luk, De Sa & Bialystok, 2011b; Sorge, Toplak & Bialystok, 2017; Novitskiy, Myachykov & Shtyrov, in press) and better endogenous attention disengagement in the Posner's cueing paradigm (Mishra, Hilchey, Singh & Klein, 2012). The rationale behind relating higher L2 proficiency to better executive control follows from the observation that bilinguals who are highly fluent in L2 experience a stronger cross-linguistic interplay between their two language systems. It has indeed been reported that highly proficient but not low proficient bilinguals automatically activate L2 lexical items when processing L1-specific targets (Blumenfeld & Marian, 2007). Higher L2 proficiency has also been linked to stronger parallel activation of lexicons (e.g., Van Hell & Dijkstra, 2002), increasing overlap in the time course of L1 and L2 lexicon activation (Guo & Peng, 2006) as well as unconscious activation of translation equivalents in the non-target language (e.g., Wu & Thierry, 2010).

Some studies also investigated the impact of individual- and group-level BEF differences on neuroplasticity of the language/executive control network areas. In a landmark study, Mechelli et al. (2004) reported that increase in grey matter volume (GMV) in bilingual users is positively associated with L2 proficiency and negatively – with L2 age of acquisition (AoA). Such results have been replicated in studies deploying different tasks as proxies of L2 proficiency. Grogan, Green, Ali, Crinion and Price (2009), for example, reported a positive association between GMV of the bilateral caudate nucleus and performance on a phonemic fluency task. The same group (Grogan et al., 2012) also showed a positive correlation between left inferior frontal gyrus GMV and scores in lexical decision and verbal fluency tasks in L2. Similarly, Pliatsikas, Johnstone and Marinis (2014) reported increases in GMV of the cerebellum associated with higher speed of processing of regular verb morphology in L2. Abutalebi et al. (2014) reported an association between increases in GMV of the left temporal pole and

performance in an L2 picture naming task. Finally, Hervais-Adelman, Egorova and Golestani (2018) reported neuroplastic changes in a sample of multilingual speakers in the shape and volume of bilateral caudate nuclei predicted by a composite score accounting for AoA and proficiency of each of the languages spoken by the participants. Similar results also emerged concerning L2 exposure and immersion (i.e., the length of residence in an L2-speaking country). For instance, in their abovementioned 2014 study, Pliatsikas and colleagues reported a positive association between L2 exposure and GMV of the bilateral putamen. The same group reported intriguing results in a more recent study (Pliatsikas, DeLuca, Moschopoulou & Saddy, 2017) investigating the effect of immersion on grey matter density of subcortical regions of the bilingual language control network. Increased GMV, as compared to monolinguals, was registered in highly-immersed bilinguals in the bilateral putamen, right thalamus and bilateral globus pallidus, whereas higher GMV in the left caudate nucleus were reported for the low-immersion group, suggesting that L2 exposure modulates the impact of bilingualism on the brain.

Taken together, these findings highlight the importance of considering the role of individual BEF differences in fostering neuroplastic and cognitive changes in bilinguals. It is also worth noting that, beside the tendency to prefer group comparisons to evaluation of continuous language background variables, research on the cognitive consequences of bilingualism still tends to focus on early/balanced over late/unbalanced bilinguals. This “sampling bias” disregards the fact that late/unbalanced bilinguals represent a steadily increasing portion of the world’s bilingual population due to globalization and global migration processes, most expressed in the ever-increasing diffusion of English as an L2 in the globalized society. Another methodological issue in bilingual research has been the relative scarcity of studies investigating the relationship between behavioral and neural consequences of bilingualism resulting in lack of specificity. As suggested by Del Maschio and colleagues (2018) in one of the

few studies adopting such practice, investigating the relationship between bilingualism-induced neuroplastic changes and behavioral outcomes might inform us on *qualitative*, in addition to *quantitative*, changes on cognition resulting from bilingual experience (see section 4 for further discussion).

Given these premises and in line with recent trends in bilingual research (e.g., Del Maschio et al., 2019; DeLuca, Rothman, Bialystok & Pliatsikas, 2019; Hervais-Adelman et al., 2018), we operationalized the main aspects of bilingual experience as continuous variables, investigating the effects of individual differences in BEFs on the executive performance and neural substrate of a sample of late, unbalanced, Russian-English bilinguals. In this group, we estimated individual differences in bilingual experience on a range of indicators, obtained a measure of executive performance behaviorally using a Flanker task, and investigated the relationship between these measures and GMV of the executive control network, obtained using region-based morphometry.

2. Materials and methods

2.1 Participants

Twenty-two Russian-English participants (9 males; mean age = 22.95, SD \pm 4.38) were recruited mostly from the population of students of the HSE Department of Psychology. All participants acquired English as an L2 formally through instruction at school, although at different ages. All participants were right-handed, as established by the Edinburgh Handedness Inventory scale (Oldfield, 1971). No participant had a history of neurologic or psychiatric illnesses. Socio-demographic variables – age, educational attainment and socio-economic status (SES) – were collected for all participants using the MacArthur Scale of Subjective Social Status (<https://macses.ucsf.edu/research/psychosocial/subjective.php#measurement>). The annual household income bands, here used as a proxy of SES, were adapted to Russian Federation

standards based on the European Social Survey 2016

(https://www.europeansocialsurvey.org/data/country.html?c=russian_federation). Participants' fluid intelligence was also assessed, using a subset of the Raven's Standard Progressive Matrices for adults (Raven, Raven & Court, 2000). Details on demographic and cognitive measures are reported in Table 1. The study was approved by the local research ethics committee, and written informed consent was obtained from all participants.

2.2 Bilingual-experience-factors: L2 age of acquisition, exposure, and proficiency

L2 AoA and exposure were measured using the Language Experience and Proficiency Questionnaire (LEAP-Q), an established tool for assessing language background of bi- and multilingual populations (Marian, Blumenfeld & Kaushanskaya, 2007). The LEAP-Q version was implemented in the NBS Presentation® software (Version 18.1, Neurobehavioral Systems, Inc., Berkeley, CA, www.neurobs.com) such that each question of the questionnaire appeared as one computer screen. The AoA and daily exposure to English measurements were self-rated by participants. To collect an objective measure of English proficiency, we deployed a computerized custom-design L2-to-L1 translation task, implemented in the same version of the NBS Presentation® software as detailed above. The translation task, including 146 English words of differing frequencies, consisted in an unspeeded forced choice in which the participant was presented with an English word and three Russian translation alternatives, all in one column. The participant had to select the correct translation by pressing 1, 2, or 3 on the keyboard. Feedback was provided at the end of each trial for both correct and incorrect responses. Details of the linguistic measures are reported in Table 1.

<Table 1 here>

2.3 Behavioral assessment

All participants performed a version of the Flanker task (Fan, McCandliss, Sommer, Raz & Posner, 2002). Participants were seated in an electrically shielded and acoustically dampened chamber. Experimental stimuli were presented on a 75 cm-diagonal computer screen. Target trials proceeded from the initial fixation point presented at the center of the screen for 400 ms, followed by a row of five horizontal black lines with arrow heads pointing to the left or to the right for 1700 ms. Participants were instructed to detect and signal the direction of the central target arrow by pressing the left or right arrow button on the keyboard as fast as possible. Targets appeared with additional arrows flanked to the same direction as the target arrow (→→→→→) (i.e., congruent condition), with surrounding arrows flanked to the opposite direction of the target (←←→←←) (i.e., incongruent condition), or with surrounding neutral lines (-- -- → -- --) (i.e., neutral condition). Congruent flankers cue the correct response and are typically associated with faster and more accurate task performance; incongruent flankers provide conflicting visual information with regard to the correct response and generally yield performance decline associated with lower accuracy and increasing RT; the neutral condition supports neither the correct nor the incorrect response. Congruent, incongruent, and neutral trials were presented in a pseudo-randomized order during three runs with 96 items (32 for each condition) per run. Prior to the experiment, participants had a practice run consisting of 24 pseudo-randomized trials.

2.4 MRI data acquisition and preprocessing

T1-weighted images were acquired with a Philips Intera 1.5T MRI scanner, using the following parameters: TR = 25 ms, TE = 4.6 ms; flip angle = 30, FOV = 240x240, resolution = 1x1x1 mm, matrix = 256, TA = 5.35 min, mode = 3DFFE, number of slices = 191. The Computational Anatomy Toolbox (CAT12, r1113, <http://dbm.neuro.uni-jena.de/cat/>) within SPM12 (v6906) was used to obtain total amount of GMV within pre-defined brain regions comprising the bilingual language control network (Abutalebi & Green, 2016), by performing

region-based morphometry. Images were first visually inspected to check for gross field distortions and movement artifacts, 2 participants had to be discarded for this reason. For each image, the origin was manually set to correspond to the AC-PC (anterior commissure-posterior commissure) line. The following two-steps procedure was used for GMV extraction. In the first step, raw structural images were segmented into Grey Matter (GM), white matter (WM) and cerebrospinal fluid (CSF) images. The segmentation routine implemented in CAT12 utilizes an adaptive Maximum A Posterior (aMAP) technique that reduces the need for *a priori* information about tissue probabilities (see Rajapakse, Giedd & Rapoport, 1997) and also accounts for local variations and inhomogeneity of GM intensity (Dahnke, Ziegler & Gaser, 2012). Following aMAP segmentation, CAT12 also carries out a Partial Volume Estimation (PVE) of mixed tissue-classes (GM-WM and GM-CSF) (Tohka, Zijdenbos & Evans, 2004) that results in a more accurate segmentation by estimating the fraction of pure tissue of each type within each voxel. The segmentation routine was further improved by using a spatial-adaptive non-local means (SANLM) denoising filter in a pre-segmentation step (Manjón, Coupé, Martí-Bonmatí, Collins & Robles, 2010). Following segmentation, the brains of all of the participants were registered to the ICBM (International Consortium for Brain Mapping) European brain space template by affine regularization. In the final step, GMV values were extracted from the following regions of interest (ROI) part of the bilingual language control network (see figure 1) (Abutalebi & Green, 2016): (1) Left and right anterior cingulate cortex (LACC, RACC); (2) left and right caudate nucleus (LCAU, RCAU); (3) left and right prefrontal cortex (LPFC, RPFC); (4) left and right inferior parietal lobule (LIPL, RIPL). The extraction was performed using an in-built CAT12 function allowing for the estimation of GMV in non-normalized native space using maximum tissue probability labels derived from the Neuromorphometrics Atlas (2012) (<http://Neuromorphometrics.com/>). To control for individual differences in brain sizes, Total Intracranial Volume (TIV) was calculated for each participant by summing the native space global volumes of GM, WM and CSF.

<Figure 1 here>

2.5 Statistical analyses

In order to conduct analysis of the Flanker task's RTs, we removed error trials, false start trials with RTs below 100 ms, and outlier trials, i.e., those falling outside 2 standard deviations from the individual subjects' mean RT values. Neutral condition trials were discarded, as we aimed to investigate the conflict effect, a measure of inhibitory control computed as the RT difference between incongruent and congruent trials. Subsequently, to check for homogeneity in SES, intelligence, and education, we tested for outliers in our sample, using a 2 standard deviations threshold. No outliers were detected for these measures. In addition, to control for eventual confounding effects of such factors on executive control performance, we tested a linear mixed effects model including Flanker's RT as the dependent variable, together with main effects of SES, intelligence and education as fixed factors and crossed random effects of trial and participant. No significant effects emerged for any of the predictors.

Effects of individual differences in BEFs on executive behavioral performance

To examine the effects of the BEF differences on executive behavioral performance, linear mixed effects analyses were conducted using the *lme4* (Bates, Maechler, Bolker & Walker, 2014) and the *lmerTest* (Kuznetsova, Brockhoff & Christensen, 2017) packages in R (R Core Team, 2014). The choice of mixed effects modeling, alongside other advantages, allowed us to increase the number of the data points by assessing executive performance on a trial-by-trial level. Consequently, as the standard Flanker conflict effect, computed as the increase in mean RTs in the incongruent (vs. congruent) condition, could not be deployed, we selected the contribution of task condition (congruent vs. incongruent) to modulating the effect of our predictors as a measure of conflict. The full model included Flanker's RT as the dependent

variable, alongside L2 AoA, L2 exposure, L2 proficiency and task condition, as well as the interactions between these predictors, as fixed factors as well as crossed random effects of trial and participant. Subsequently, the best fitting model was obtained by deploying the *step* function implemented in the *lmerTest* package (Kuznetsova et al., 2017), which allows to perform automatized step-down model-building.

Effects of individual differences in BEFs on the language/executive control network

To investigate the role of BEF differences on bilingualism-related neuroplasticity in the language/executive control network, we used the *lm* function implemented in the *stats* package (R Core Team, 2014) to fit linear models with the GMV of each ROI as the dependent variable, and L2 AoA, exposure and proficiency as predictors. TIV was used as a covariate in all the models to control for individual differences in brain size. Subsequently, the best fitting model was obtained by deploying the *step* function implemented in the *stats* package (R Core Team, 2014), again via an automatized backwards stepwise search.

Relationships between individual differences in BEFs and GMV in modulating executive control performance

Lastly, we conducted linear mixed effects analyses to assess whether individual differences in BEFs and GMV interact in affecting executive performance. To avoid the risk of overfitting, and the difficulties in interpreting the results, related with testing a model including 5 continuous variables as predictors, we decided to compute a bilingual index (BI) taking into account all relevant variables, i.e. L2 AoA, exposure and proficiency. BI was built using the following formula:

$$BI = \frac{(a * L2 \text{ exposure}) * (b * L2 \text{ proficiency})}{(c * L2 \text{ AoA})}$$

To obtain the values of coefficients a , b and c , we tested a linear mixed effects model including Flanker's RT as the dependent variable, main effects of L2 exposure, L2 proficiency, L2 AoA and task condition, and the interactions between task condition and each of the factors as predictors, together with the crossed random effects of trial and participant. The coefficients assigned to a , b , and c were obtained from the absolute value of the interactions' estimates in the model: $a = \text{L2 exposure} * \text{task condition} = |-3.029|$; $b = \text{L2 proficiency} * \text{task condition} = |-8.205|$; $c = \text{L2 AoA} * \text{task condition} = |3.468|$. Note that the interactions containing L2 exposure and L2 proficiency showed an inverse relationship with RT (i.e., increasing levels of such factors predicted a trend towards better executive performance), while the interaction containing L2 AoA revealed the opposite pattern. These trends are in line with the intuitive prediction that increasing L2 exposure and proficiency should contribute positively to the BI, whereas increasing L2 AoA should contribute negatively to the BI. Thus, we inserted L2 exposure and L2 proficiency at the numerator, and L2 AoA at the denominator, in our equation. By adopting the approach just described, we aimed at taking into account the contribution of each BEF in modulating the inhibitory control ability of our participants. Once we computed the BI, we fitted a linear mixed effects model for each ROI, including Flanker's RT as the dependent variable, fixed effects of task condition, BI, GMV, and their interactions as predictors, and crossed random effects of trial and participant.

3. Results

Effects of individual differences in BEFs on executive behavioral performance

Mean Flanker RTs and accuracy measures are reported in Table 2. The stepwise model selection procedure returned the best fitting model that included main effects of task condition and L2 proficiency as well as the interaction between these two factors. As shown in Figure 1, a significant task condition by L2 proficiency interaction was registered in the RT data, with

higher L2 proficiency scores predicting better executive performance (i.e., lower RT, differentially in the incongruent condition) (F value = -3.926, $\text{Pr}(> F) = 8.79\text{e-}05$).

<Table 2 here>

<Figure 2 here>

Effects of individual differences in BEFs on the language/executive control network

We registered a reliable effect of BEFs on GMV of bilateral ACC. This effect also exhibited a trend towards significance in the bilateral PFC. For all four ROIs, the stepwise model selection identified as the best fit a model including main effects of L2 exposure and L2 proficiency, together with their interaction, as predictors of GMV. A crossed L2 proficiency*L2 exposure interaction reliably predicted GMV in the left ACC (t value = -3.016, $\text{Pr}(> |t|) = .00869$) and right ACC (t value = -3.791, $\text{Pr}(> |t|) = .001777$), approaching significance in the left PFC (t value = -2.087, $\text{Pr}(> |t|) = .054318$) and right PFC (t value = -2.109, $\text{Pr}(> |t|) = .052204$). As shown in Figure 2, higher proficiency was associated with increases in GMV only at lower levels of exposure, and, vice-versa, higher exposure was associated with increases in GMV only at lower levels of proficiency. When both of the predictors reached high values, a decreasing trend in GMV emerged.

<Figure 3 here>

Relationships between individual differences in BEFs and GMV in modulating executive control performance

A reliable interaction between individual differences in BEFs and GMV on executive performance emerged for the bilateral caudate and the left PFC. For all these ROI the best fitting model coincided with the full model. In each of the three ROI, a comparable pattern emerged, with BI significantly interacting with GMV and task condition in predicting Flanker RTs. The interaction plot shown in Figure 3 shows that variations in GMV predicted incongruent trials' (i.e., those tapping on inhibitory control) RT at lower, but not at higher levels of BI. In other words, executive performance appeared to be unrelated to GMV of the left caudate (F value = 4.3636, $\text{Pr}(> F) = .036787$), RCAU (F value = 4.2238, $\text{Pr}(> F) = .03993$), and left PFC (F value = 4.9149, $\text{Pr}(> F) = .0266913$) at higher levels of BI. Overall, higher levels of BI predicted better executive performance (i.e., lower RT). Lastly, in the face of increasing GMV, performance levels tended to overlap irrespectively of the level of BI.

<Figure 4 here>

4. Discussion

The challenge of categorizing such a complex multifaceted phenomenon as bilingualism requires going beyond a group-comparison approach. In the present study, we operationalized three main aspects of bilingual experience, namely AoA, exposure, and proficiency in the L2 as continuous variables, in the attempt to elucidate the consequences of dual language use for cognition. We tested the contribution of these factors on executive control, both at the behavioral and at the neural levels. Increase in L2 proficiency reliably predicted better executive control performance. Moreover, the interplay between L2 exposure and proficiency was shown to affect

bilingualism-induced neuroplastic changes, which increased with increasing L2 fluency and use albeit plateauing and eventually regressing at further increasing levels of bilingual competence and exposure. Furthermore, our results indicate that a BI taking into account all three BEFs modulated the relationship between GMV of the language/executive control network and behavioral executive performance. Highly expert bilinguals showed no relationship between performance levels and volumetric changes in their neural substrate, suggesting that they would be able to optimize executive performance even in the face of lower GMV. Below, we discuss these findings in more detail.

Effects of individual differences in BEFs on executive behavioral performance

In line with previous findings, our data indicate that increasing L2 proficiency is associated with better executive control performance in bilinguals. Similar results have been reported across different executive functions tasks including the ANT (Tao et al., 2011), the Stroop task (Iluz-Cohen & Armon-Lotem, 2013; Singh & Mishra, 2012; 2013), the Posner's cueing paradigm (Mishra et al., 2012), and the Flanker task (Luk et al., 2011b; Sorge et al., 2017; Novitskiy et al., in press). As already discussed in the introduction, higher fluency in the L2 would cause bilinguals to experience higher cross-linguistic influence from their two linguistic systems (e.g., Van Hell & Dijkstra, 2002, Wu & Thierry, 2010). Thus, as a result of increased control demand, high-proficient bilinguals would develop a better ability to resolve conflict. At the same time, behavioral bilingualism studies have produced a number of inconsistent outcomes, even when deploying continuous measures of BEFs, with studies failing to firmly establish an effect of L2 proficiency on executive performance, either partially (i.e., in one of several executive tasks) or completely (e.g., Becker, Schubert, Strobach, Gallinat & Kühn, 2016; Dong & Xie, 2014; Rosselli, Ardila, Lalwani & Vélez-Urbe, 2016; Xie, 2018). Dong and Xie (2014), for instance, found no effect of varying L2 proficiency on the level of performance on the same Flanker task deployed in the present study. This inconsistency in the available

behavioral results highlights that testing behavior may not be sufficient to get an insight into the effects of BEFs on executive control. We argue that a better insight can be obtained by investigating the consequences of differences in the bilingual experience for the neural substrate, as also exemplified by the brain morphometry results of the current investigation.

Effects of individual differences in BEFs on the language/executive control network

Individual differences in L2 exposure and proficiency predicted GMV in the bilateral ACC, with a similar trend (although only approaching statistical significance) emerging for the bilateral PFC. In the bilingual brain, the ACC is assumed to underlie cross-linguistic conflict resolution (Abutalebi & Green, 2016), while the PFC is thought to support target response selection and non-target response inhibition during language control and language switching (Abutalebi & Green, 2016). In our investigation, L2 exposure and proficiency showed a crossed interaction, suggesting that a positive association between one predictor and GMV was present only when the value of the other predictor was low. For medium values of L2 exposure or proficiency, GMV volumes remained stable with the other BEF varying. When both predictor values further increased, reaching high levels in both BEFs, a decreasing trend in GMV emerged. This pattern suggests that bilingualism-related neuroplasticity could plateau at a certain stage in bilingual experience, with neuroplastic changes ceasing or possibly even reverting with increasing experience. Our results are in line with Pliatsikas's Dynamic Restructuring Model (DRM; Pliatsikas, 2019). Based on a comprehensive review of the literature on neuroplasticity in bilinguals, L2 learners and simultaneous interpreters, the DRM attempts to formulate a time course for bilingualism-induced structural adaptation. Regarding cortical structures related to language control, as the ACC and PFC, the model predicts volumetric increases in the early stage of L2 acquisition, which would reflect increased effort imposed by controlling competing activations in the two languages. Such effects would be especially palpable in sequential bilinguals, i.e., individuals that started learning and using their L2 later than their native L1 (as

our participants), with the volumetric increases expected to peak and subsequently disappear as language control becomes less effortful with increasing exposure and proficiency. The DRM explains the trajectory of bilingualism-induced neuroplasticity as pruning: bilinguals would develop extra connections to accommodate the increased language control effort, to subsequently eliminate the supernumerary connections after the more efficient ones are identified. This process would also account for the resilience to age-related neurodegeneration observed in aging (see Perani & Abutalebi, 2015): the efficient connections surviving this pruning phase would also be the ones that survive age-related deterioration.

We argue that our results may reflect the mechanisms posited by the DRM, yet we refrain from making conclusive statements in this regard, as the present investigation includes structural, but no functional imaging measurements. Nevertheless, the aforementioned effects of BEF differences – and their interplay with GMV differences (as described below) – in modulating executive performance suggest that the consequences of bilingualism for cognition may evolve with bilingual experience: structural increases might eventually be replaced by enhancements in cognitive efficiency. The results of the conjunct analysis on behavioral and neuroimaging data presented hereafter might better inform us on the processes underlying this posited structure-to-function shift.

Relationships between individual differences in BEFs and GMV in modulating executive control performance

As pointed out in the introduction section, by combining analyses at both behavioral and neural levels, we aimed to gain insight into the *qualitative*, in addition to the *quantitative*, changes induced by bilingual experience on cognition. Arguably our most interesting finding illuminates the role of BEFs in modulating the relationship between the executive neural substrate and the associated behavioral outcomes. For the bilateral caudate and the left prefrontal

cortex, indeed, at higher levels of bilingual experience (i.e., higher BI), executive performance was not affected by the differences in GMV (note also that, in line with our behavioral analyses' results, higher BI levels predicted overall better performance). Although the scope of the structural MRI analysis does not allow an "online" investigation of the underlying processes, we speculate that the phenomenon reported here might constitute the source of the enhanced cognitive reserve observed in senior bilinguals (see Perani & Abutalebi, 2015). Cognitive reserve has been defined as the discrepancy between the severity of age-related deterioration and the resulting level of cognitive impairment (Stern, 2009). Our findings indicate that more experienced bilinguals are able to maintain optimal task performance even in the face of decreasing GMV. In the context of senescence, one might argue that expert senior bilinguals might still be able to optimize their behavioral performance even in the face of age-related brain atrophy. Corroborating evidence comes from the study by Del Maschio et al.'s (2018), which showed a comparable pattern of results for bilingual, but not for monolingual aging individuals. Cognitive reserve is thought to arise from enhanced flexibility and efficiency of an individual's brain networks, enabling optimal task performance even in the face of age-related deterioration (Stern, 2009). Such improved neural efficiency and flexibility, while returning "dividends" in senescence, are believed to arise from lifelong experiential factors that originate in early life. Studies comparing bilinguals and monolinguals well before senescence support this interpretation. Indeed, bilinguals have been reported to make a more efficient use of executive control areas showing lesser brain activation while outperforming monolingual peers at the behavioral level (e.g., Abutalebi et al., 2012). They were also shown to activate different/more extensive brain networks during executive task performance (e.g., Luk, Anderson, Craik, Grady & Bialystok, 2010). The phenomenon observed in the present investigation might therefore constitute the prerequisite to the protection against age-related cognitive decline observed in aging bilinguals, although further investigation is required to provide conclusive evidence in this direction.

Finally, the pattern illustrated in Figure 4 suggests that participants' performance converged at an optimal level in the face of increasing GMV and irrespectively of the level of BI. This result provides insight into the potential reasons behind inconsistencies in bilingual research. Indeed, it is well known that the existing evidence regarding bilingualism-related effects on executive performance is rather mixed, particularly in the young adult age group. Valian (2015) points to a possible cause of this inconsistency: young adults are usually engaged in many cognitively challenging activities (e.g., video-gaming, social interactions, etc.) that could equal the cognitive challenges imposed by learning and controlling two languages. It might thus be difficult to observe cognitive consequences of bilingual experience in this age group without disentangling the contribution of bilingualism from other factors. Our result supports this claim: when neuroplastic changes – induced by any experiential factor, not just bilingualism – reach sufficiently high levels, eventual consequences of bilingualism on executive performance may become unnoticeable.

5. Conclusion

Overall, the reported research shows that individual differences in BEFs play an important role in modulating the consequences of bilingualism for executive functioning, both at the behavioral and at the neural level. The present investigation highlights the importance of treating the multifaceted phenomenon of bilingualism as a continuous spectrum, departing from group comparisons to achieve better consistency in the results and shed light on a still hazy picture. We also advocate the need for the field to shift towards simultaneous investigations of the behavioral and neural consequences of bilingual experience, which may provide additional meaningful insight on the interplay between bilingualism and cognition.

References

Abutalebi, J., & Green, D. (2007). Bilingual language production: The neurocognition of language representation and control. *Journal of neurolinguistics*, *20*, 242-275.

Abutalebi, J., Della Rosa, P. A., Green, D. W., Hernandez, M., Scifo, P., Keim, R., Cappa, S.F., & Costa, A. (2012). Bilingualism tunes the anterior cingulate cortex for conflict monitoring. *Cerebral Cortex*, *22*, 2076–2086.

Abutalebi, J., Canini, M., Della Rosa, P. A., Sheung, L. P., Green, D. W., & Weekes, B. S. (2014). Bilingualism protects anterior temporal lobe integrity in aging. *Neurobiology of aging*, *35*, 2126-2133.

Abutalebi, J., & Green, D. W. (2016). Neuroimaging of language control in bilinguals: neural adaptation and reserve. *Bilingualism: Language and cognition*, *19*, 689-698.

Bates, D., Maechler, M., Bolker, B., & Walker, S. (2014). lme4: Linear Mixed-Effects Models Using Eigen and S4. Available online at: <http://cran.r-project.org/package=lme4>

Becker, M., Schubert, T., Strobach, T., Gallinat, J., & Kühn, S. (2016). Simultaneous interpreters vs. professional multilingual controls: Group differences in cognitive control as well as brain structure and function. *Neuroimage*, *134*, 250-260.

Bialystok, E., Craik, F. I., & Luk, G. (2012). Bilingualism: consequences for mind and brain. *Trends in cognitive sciences*, *16*, 240-250.

Bialystok, E. (2016). The signal and the noise: Finding the pattern in human behavior. *Linguistic Approaches to Bilingualism*, *6*, 517-534.

Bialystok, E. (2017). The bilingual adaptation: How minds accommodate experience. *Psychological bulletin*, *143*, 233.

Blumenfeld, H. K., & Marian, V. (2007). Constraints on parallel activation in bilingual spoken language processing: Examining proficiency and lexical status using eye-tracking. *Language and cognitive processes*, 22, 633-660.

Dahnke, R., Ziegler, G., & Gaser, C. (2012). Local adaptive segmentation. *HBM 2012*. <http://www.neuro.unijena.de/hbm2012/HBM2012-Dahnke02.pdf>

Del Maschio, N., Sulpizio, S., Gallo, F., Fedeli, D., Weekes, B. S., & Abutalebi, J. (2018). Neuroplasticity across the lifespan and aging effects in bilinguals and monolinguals. *Brain and cognition*, 125, 118-126.

Del Maschio, N., Sulpizio, S., Toti, M., Caprioglio, C., Del Mauro, G., Fedeli, D., & Abutalebi, J. (2019). Second language use rather than second language knowledge relates to changes in white matter microstructure. *Journal of Cultural Cognitive Science*, 1-11.

DeLuca, V., Rothman, J., Bialystok, E., & Pliatsikas, C. (2019). Redefining bilingualism as a spectrum of experiences that differentially affects brain structure and function. *Proceedings of the National Academy of Sciences*, 116, 7565-7574.

Dong, Y., & Xie, Z. (2014). Contributions of second language proficiency and interpreting experience to cognitive control differences among young adult bilinguals. *Journal of Cognitive Psychology*, 26, 506-519.

ESS Round 8: European Social Survey Round 8 Data (2016). Data file edition 2.1. NSD - Norwegian Centre for Research Data, Norway – Data Archive and distributor of ESS data for ESS ERIC. [doi:10.21338/NSD-ESS8-2016](https://doi.org/10.21338/NSD-ESS8-2016).

Fan, J., McCandliss, B. D., Sommer, T., Raz, A., & Posner, M. I. (2002). Testing the efficiency and independence of attentional networks. *Journal of cognitive neuroscience*, 14, 340-347.

Gold, B. T., Johnson, N. F., & Powell, D. K. (2013). Lifelong bilingualism contributes to cognitive reserve against white matter integrity declines in aging. *Neuropsychologia*, *51*, 2841-2846.

Green, D. W., & Abutalebi, J. (2013). Language control in bilinguals: The adaptive control hypothesis. *Journal of Cognitive Psychology*, *25*, 515-530.

Grogan, A., Green, D. W., Ali, N., Crinion, J. T., & Price, C. J. (2009). Structural correlates of semantic and phonemic fluency ability in first and second languages. *Cerebral cortex*, *19*, 2690-2698.

Grogan, A., Jones, Ö. P., Ali, N., Crinion, J., Orabona, S., Mechias, M. L., ... & Price, C. J. (2012). Structural correlates for lexical efficiency and number of languages in non-native speakers of English. *Neuropsychologia*, *50*, 1347-1352.

Guo, T., & Peng, D. (2006). Event-related potential evidence for parallel activation of two languages in bilingual speech production. *NeuroReport*, *17*, 1757-1760.

Hervais-Adelman, A., Egorova, N., & Golestani, N. (2018). Beyond bilingualism: multilingual experience correlates with caudate volume. *Brain Structure and Function*, *223*, 3495-3502.

<http://dbm.neuro.uni-jena.de/cat/>

<http://Neuromorphometrics.com/>

<https://macses.ucsf.edu/research/psychosocial/subjective.php#measurement>

<https://www.neurobs.com>

Iluz-Cohen, P., & Armon-Lotem, S. (2013). Language proficiency and executive control in bilingual children. *Bilingualism: Language and Cognition*, *16*, 884-899.

Kaushanskaya, M., & Marian, V. (2007). Bilingual language processing and interference in bilinguals: Evidence from eye tracking and picture naming. *Language Learning*, *57*, 119–163.

Kroll, J.F., Bobb, S.C., & Hoshino, N. (2014). Two languages in mind: bilingualism as a tool to investigate language, cognition, and the brain. *Current Directions in Psychological Science*, *23*, 159-163.

Kroll, J. F., Dussias, P. E., Bice, K., & Perrotti, L. (2015). Bilingualism, mind, and brain. *Annual Review of Linguistics*, *1*, 377-394.

Kuznetsova, A., Brockhoff, P. B., & Christensen, R. H. B. (2017). lmerTest package: tests in linear mixed effects models. *Journal of Statistical Software*, *82*(13).

Lehtonen, M., Soveri, A., Laine, A., Järvenpää, J., de Bruin, A., & Antfolk, J. (2018). Is bilingualism associated with enhanced executive functioning in adults? A meta-analytic review. *Psychological bulletin*, *144*, 394-425.

Li, P., Legault, J., & Litcofsky, K. A. (2014). Neuroplasticity as a function of second language learning: anatomical changes in the human brain. *Cortex*, *58*, 301-324.

Lüdtke D (2018). ggeffects: Tidy Data Frames of Marginal Effects from Regression Models. *Journal of Open Source Software*, *3*(26), 772. doi: [10.21105/joss.00772](https://doi.org/10.21105/joss.00772)

Luk, G., Anderson, J. A., Craik, F. I., Grady, C., & Bialystok, E. (2010). Distinct neural correlates for two types of inhibition in bilinguals: Response inhibition versus interference suppression. *Brain and cognition*, *74*(3), 347-357.

Luk, G., Bialystok, E., Craik, F. I., & Grady, C. L. (2011a). Lifelong bilingualism maintains white matter integrity in older adults. *Journal of Neuroscience*, *31*, 16808-16813.

Luk, G., De Sa, E., & Bialystok, E. (2011b). Is there a relation between onset age of bilingualism and enhancement of cognitive control?. *Bilingualism: Language and cognition*, *14*, 588-595.

Luk, G., & Bialystok, E. (2013). Bilingualism is not a categorical variable: Interaction between language proficiency and usage. *Journal of Cognitive Psychology*, *25*, 605-621.

Manjón, J. V., Coupé, P., Martí-Bonmatí, L., Collins, D. L., & Robles, M. (2010). Adaptive non-local means denoising of MR images with spatially varying noise levels. *Journal of Magnetic Resonance Imaging*, *31*, 192-203.

Marian, V., Blumenfeld, H. K., & Kaushanskaya, M. (2007). The Language Experience and Proficiency Questionnaire (LEAP-Q): Assessing language profiles in bilinguals and multilinguals. *Journal of Speech, Language, and Hearing Research*.

Mechelli, A., Crinion, J. T., Noppeney, U., O'doherty, J., Ashburner, J., Frackowiak, R. S., & Price, C. J. (2004). Neurolinguistics: structural plasticity in the bilingual brain. *Nature*, *431*, 757-757.

Mishra, R. K., Hilchey, M. D., Singh, N., & Klein, R. M. (2012). On the time course of exogenous cueing effects in bilinguals: higher proficiency in a second language is associated with more rapid endogenous disengagement. *The Quarterly journal of experimental psychology*, *65*, 1502-1510.

Mishra, R. K. (2015). Let's not forget about language proficiency and cultural variations while linking bilingualism to executive control. *Bilingualism: Language and Cognition*, *18*, 39-40.

Novitskiy, N., Myachykov, A., & Shtyrov, Y. (2019). Crosslinguistic interplay between semantics and phonology in late bilinguals: neurophysiological evidence. *Bilingualism: Language and Cognition*, *22*, 209-227.

Novitskiy, N., Myachykov, A., & Shtyrov, Y. (in press). Conflict resolution ability in late bilinguals improves with increased second-language proficiency: ANT task evidence. *Frontiers in Psychology*.

Oldfield, R. C. (1971). The assessment and analysis of handedness: the Edinburgh inventory. *Neuropsychologia*, 9, 97-113.

Paap, K. R., Johnson, H. A., & Sawi, O. (2015). Bilingual advantages in executive functioning either do not exist or are restricted to very specific and undetermined circumstances. *Cortex*, 69, 265-278.

Perani, D., & Abutalebi, J. (2015). Bilingualism, dementia, cognitive and neural reserve. *Current opinion in neurology*, 28, 618-625.

Pliatsikas, C., Johnstone, T., & Marinis, T. (2014). Grey matter volume in the cerebellum is related to the processing of grammatical rules in a second language: a structural voxel-based morphometry study. *The Cerebellum*, 13, 55-63.

Pliatsikas, C., DeLuca, V., Moschopoulou, E., & Saddy, J. D. (2017). Immersive bilingualism reshapes the core of the brain. *Brain Structure and Function*, 222, 1785-1795.

Pliatsikas, C. (2019). Understanding structural plasticity in the bilingual brain: The Dynamic Restructuring Model. *Bilingualism: Language and Cognition*, 1-13.

R Core Team. (2014). *R: A Language and Environment for Statistical Computing*. Vienna: R Foundation for Statistical Computing. Available online at: <http://www.R-project.org/>

Rajapakse, J. C., Giedd, J. N., & Rapoport, J. L. (1997). Statistical approach to segmentation of single-channel cerebral MR images. *IEEE transactions on medical imaging*, 16, 176-186.

Raven, J., Raven, J. C., & Court, J. H. (2000). Manual for Raven's progressive matrices and vocabulary scales. Section 3: The standard progressive matrices. Oxford, UK: Oxford Psychologists Press; San Antonio, TX: The Psychological Corporation.

Rosselli, M., Ardila, A., Lalwani, L. N., & Vélez-Urbe, I. (2016). The effect of language proficiency on executive functions in balanced and unbalanced Spanish–English bilinguals. *Bilingualism: Language and Cognition*, *19*, 489-503.

Singh, N., & Mishra, R. K. (2012). Does language proficiency modulate oculomotor control? Evidence from Hindi–English bilinguals. *Bilingualism: Language and Cognition*, *15*, 771-781.

Singh, N., & Mishra, R. K. (2013). Second language proficiency modulates conflict-monitoring in an oculomotor Stroop task: Evidence from Hindi-English bilinguals. *Frontiers in psychology*, *4*, 322.

Sorge, G. B., Toplak, M. E., & Bialystok, E. (2017). Interactions between levels of attention ability and levels of bilingualism in children's executive functioning. *Developmental science*, *20*, e12408.

Stern, Y. (2009). Cognitive reserve. *Neuropsychologia*, *47*, 2015-2028.

Surrain, S., & Luk, G. (2019). Describing bilinguals: A systematic review of labels and descriptions used in the literature between 2005–2015. *Bilingualism: Language and Cognition*, *22*, 1-15.

Tao, L., Marzecová, A., Taft, M., Asanowicz, D., & Wodniecka, Z. (2011). The efficiency of attentional networks in early and late bilinguals: the role of age of acquisition. *Frontiers in psychology*, *2*, 123.

Tohka, J., Zijdenbos, A., & Evans, A. (2004). Fast and robust parameter estimation for statistical partial volume models in brain MRI. *Neuroimage*, 23, 84-97.

Valian, V. (2015). Bilingualism and cognition. *Bilingualism: Language and Cognition*, 18(1), 3-24.

Van Hell, J. G., & Dijkstra, T. (2002). Foreign language knowledge can influence native language performance in exclusively native contexts. *Psychonomic bulletin & review*, 9, 780-789.

Wu, Y. J., & Thierry, G. (2010). Chinese–English bilinguals reading English hear Chinese. *Journal of Neuroscience*, 30, 7646-7651.

Xie, Z. (2018). The Influence of Second Language (L2) Proficiency on Cognitive Control Among Young Adult Unbalanced Chinese-English Bilinguals. *Frontiers in psychology*, 9, 412.

Table 1

Descriptive statistics of socio-demographic, cognitive, and linguistic measures. Mean, standard deviation (SD), and range for each measure are reported.

	<i>Mean (SD)</i>	<i>Range</i>
<i>Age (years)</i>	22.95 (4.38)	18-32
<i>Education (years)</i>	14.09 (2.52)	11-17
<i>SES (score)</i>	3.04 (0.65)	2-4
<i>Raven's Matrices (score)</i>	8.36 (0.85)	7-10
<i>L2 AoA (years)</i>	8.09 (4.01)	4-19
<i>L2 exposure (daily percentage)</i>	16.84 (15.14)	1-69
<i>Translation task (score)</i>	126.82 (13.38)	87-145

Table 2

Participants' mean RT and accuracy (with standard deviations and range) in the Flanker task.

	<i>Mean (SD)</i>	<i>Range</i>
<i>Accuracy (percentage)</i>	97.68 (1.91)	92-100
<i>Congruent condition RT (ms)</i>	525.42 (60.57)	427.83- 653.04
<i>Incongruent condition RT (ms)</i>	646.34 (80.15)	529.4- 832.75

Figure 1

3D-rendered representation of regions of interest in the language control/executive control network. PFC = prefrontal cortex (note that this region is only partially represented for illustrative purposes); CAU = caudate nucleus; ACC = anterior cingulate cortex; IPL = inferior parietal lobule.

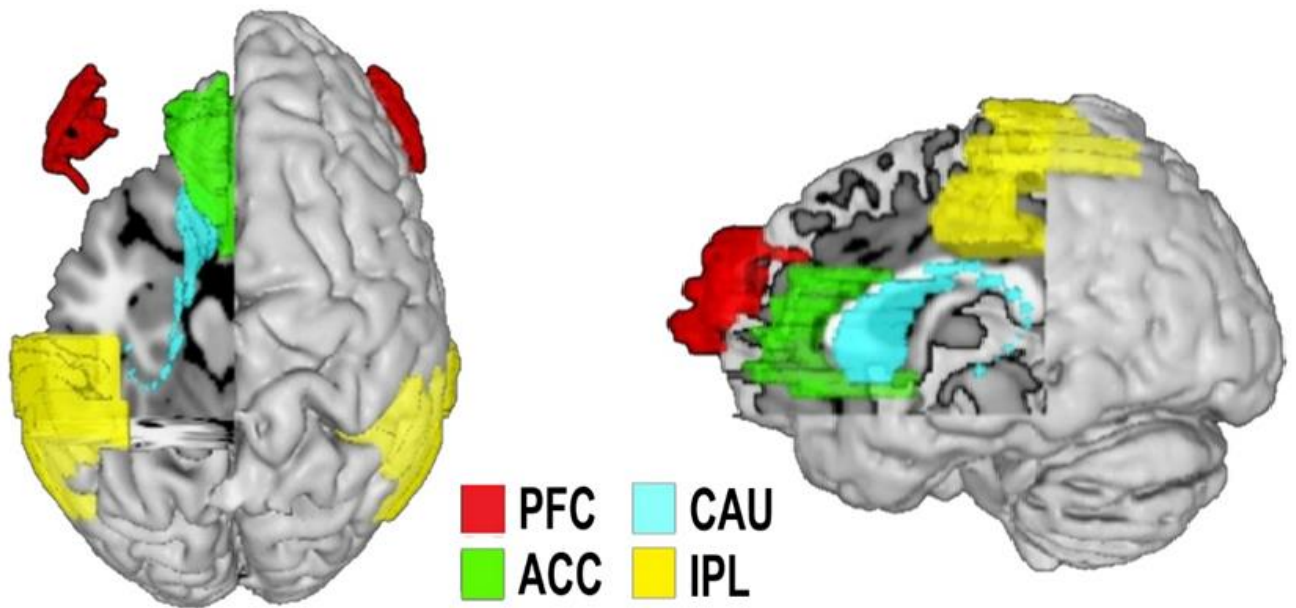


Figure 2

Interaction plot for the L2 proficiency*task condition interaction predicting Flanker's RT (in ms). Increasing levels of L2 proficiency predict lower RT, i.e., better executive performance.

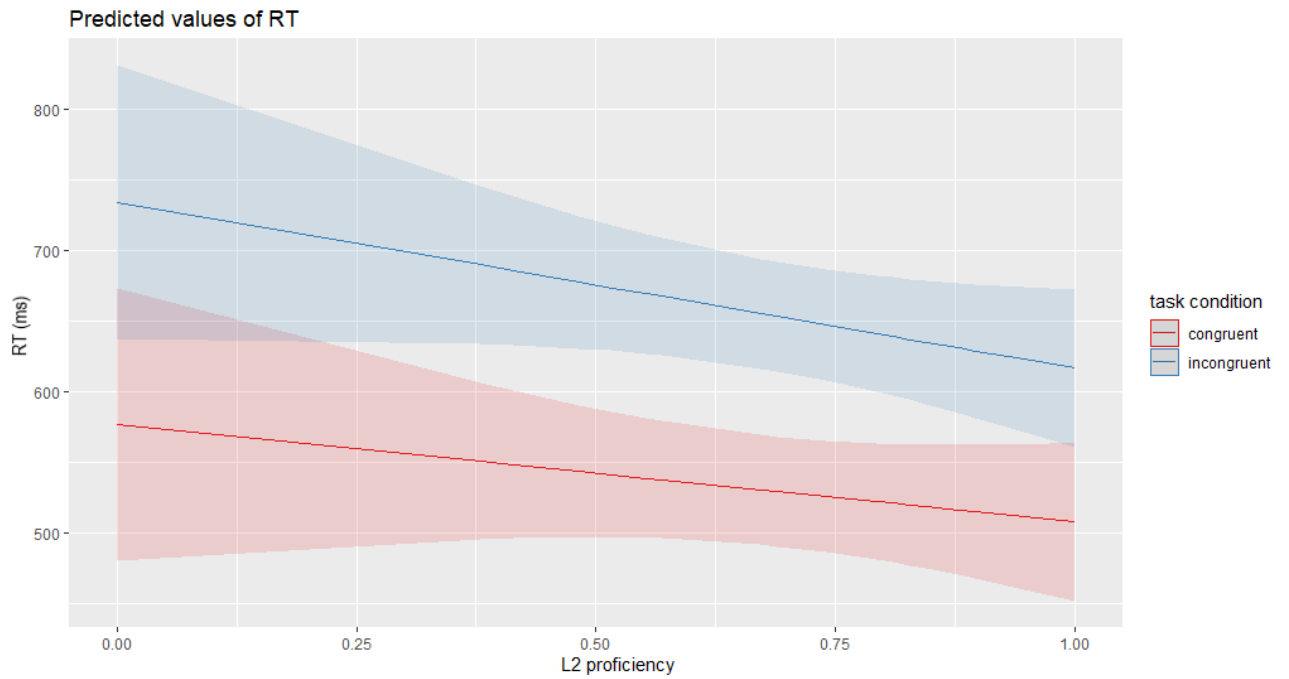


Figure 3

Interaction plot for the crossed L2 proficiency*L2 exposure interaction predicting left anterior cingulate cortex GMV (in cm³). Higher L2 proficiency predicts increases in GMV only at lower levels of L2 exposure. For medium levels of L2 exposure, variations in L2 proficiency do not affect GMV. For high levels of L2 exposure, increases in L2 proficiency predict reductions in GMV. Comparable results emerged for right anterior cingulate cortex, left prefrontal cortex and right prefrontal cortex. Normalized scores are reported for L2 proficiency and L2 exposure. Data was plotted with the sjPlot and ggeffects packages in R. Note that such packages, when a numeric vector is specified as a grouping structure, automatically select representative values for that vector (see Lüdtke, 2018 for more information).

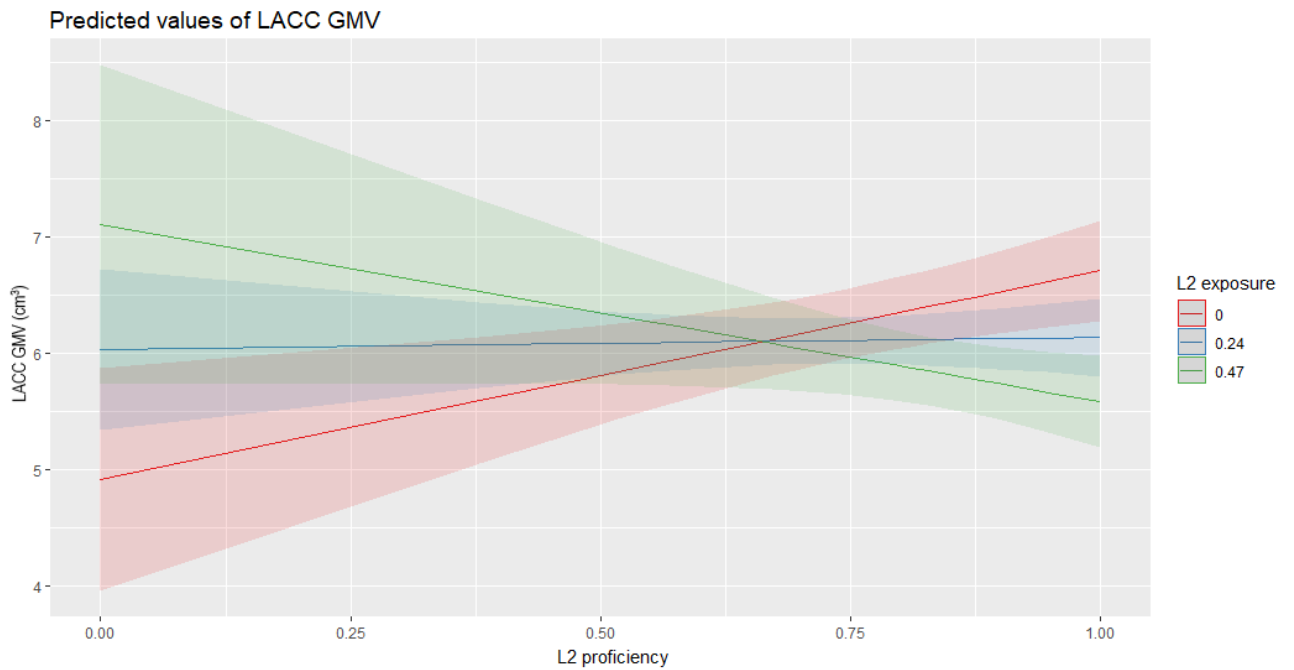
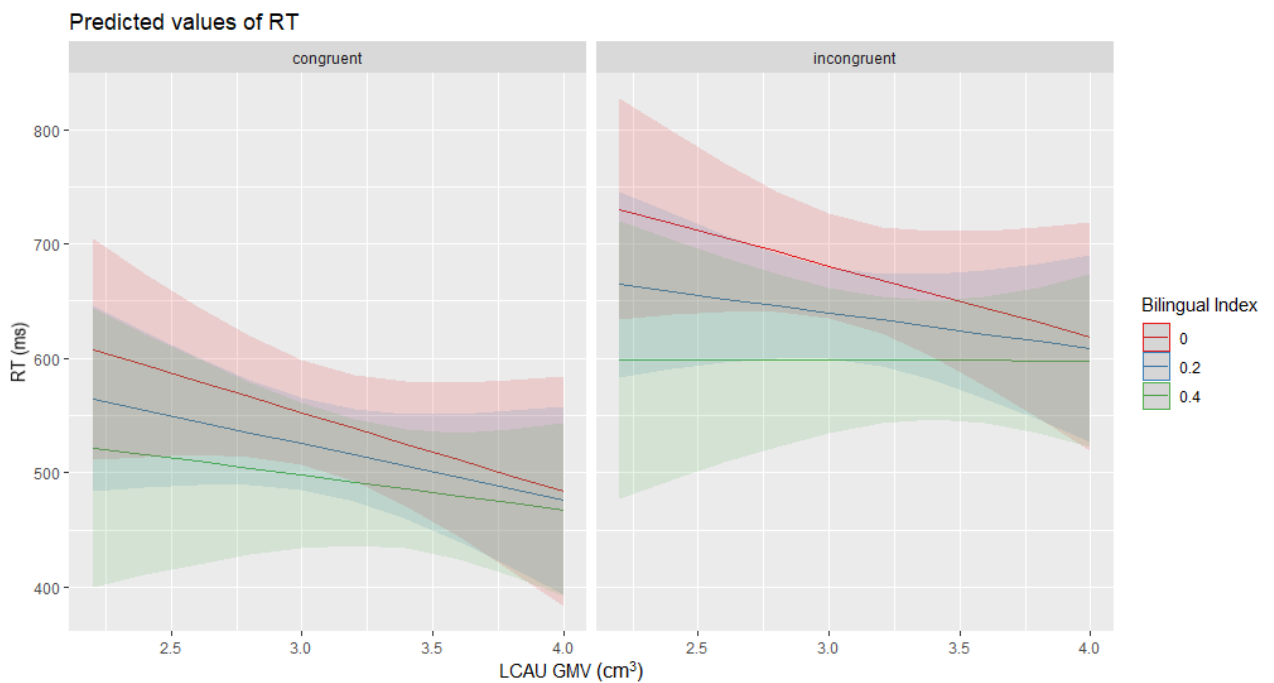


Figure 4

Interaction plot for the bilingual index*task condition*GMV interaction predicting Flanker's RT (in ms), for the left caudate nucleus. Increases in GMV (in cm³) predict lower incongruent RT only at low scores of bilingual index. At increasing levels of bilingual index, variations in GMV do not affect executive performance. Comparable results emerged for the right caudate nucleus and left prefrontal cortex. Normalized scores are reported for the bilingual index. Data was plotted with the sjPlot and ggeffects packages in R. Note that such packages, when a numeric vector is specified as a grouping structure, automatically select representative values for that vector (see Lüdtke, 2018 for more information).



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