Abstract

This study tested the hypothesis that mid-life intellectual, physical and social activities contribute to cognitive reserve (CR). Two hundred and five individuals (196 with MRI) aged 66-88 from the Cambridge Centre for Ageing and Neuroscience (www.cam-can.com) were studied, with cognitive ability and structural brain health measured as fluid IQ and total grey matter volume, respectively. Mid-life activities were measured using the Lifetime of Experiences Questionnaire.

Multivariable linear regression found that mid-life activities (MA) made a unique contribution to late-life cognitive ability independent of education, occupation and late-life activities.

Crucially, MA moderated the relationship between late-life cognitive ability and brain health, with the cognitive ability of people with higher MA less dependent on their brain structure, consistent with the concept of CR.

In conclusion, mid-life activities contribute uniquely to CR. The modifiability of these activities has implications for public health initiatives aimed at dementia prevention.

Word count: 148
1. Introduction
   Participants, materials and analyses

The concept of cognitive reserve is used to explain why some individuals maintain cognitive ability despite impaired brain health as a consequence of ageing and diseases such as Alzheimer’s disease (Stern 2012, Nilsson & Lövdén 2018). Crucially, the CR concept encompasses the notion that late life cognitive activity is influenced by factors occurring earlier in life. Determination of contributors to cognitive reserve is therefore important for “successful” ageing and prevention of dementia. While epidemiological evidence suggests that education and occupation contribute to cognitive reserve (Richards and Deary 2005), there is increasing interest in the additional contribution of other activities undertaken in midlife, given their potential modifiability. This interest is amplified by evidence that mid-life activities of a social or intellectual nature are associated with higher late-life cognitive ability, after adjusting for childhood cognitive ability (Gow et al. 2017), and by a recent review concluding that low levels of physical and social activity in adulthood represent key risk factors for dementia (Livingston et al. 2017).

Rigorous definitions of cognitive reserve not only predict that lifestyle factors will relate to late-life cognitive ability, but also that such factors will moderate the relationship between cognitive ability and brain structure. Specifically, the cognitive ability of individuals with high cognitive reserve should be less dependent on brain structure than those with low cognitive reserve, possibly as a result of compensatory functional network reorganisation (Stern 2017, see also Nilsson & Lövdén 2018). The present study therefore asked whether mid-life activities (MAs) contribute to cognitive reserve by testing two hypotheses: i) MAs contribute to late-life cognitive ability independent of early-life education, mid-life occupation and late-life activities, and ii) MAs moderate the relationship between cognitive ability and brain structure, such that the relationship is weaker in people with who had engaged in more MAs.

Two hundred and five individuals (93 female) aged 66-88 years were selected from the Cambridge Centre for Ageing and Neuroscience (Cam-CAN, www.cam-can.com, Shafto et al 2014) cohort (see Supplementary Material for further description of this sample). Cognitive ability was measured using the Cattell Culture Fair test of fluid intelligence (1971) and lifestyle activities by the Lifetime of Experiences Questionnaire (LEQ, Valenzuela and Sachdev 2007),
modified for UK participants.

The LEQ measures a broad range of cognitively-stimulating experiences and activities during three life phases: youth, 13-29 years; mid-life, 30-64 years; and late-life, 65 years onwards. Within each phase, activities are subdivided into specific or nonspecific. Specific activities are those that are considered to be undertaken primarily within one particular life phase, such as education or working occupation. In contrast, nonspecific activities such as socializing and playing sports are those can be undertaken at any age, and so are applicable to any life phase. The youth specific score (YS, or education) was derived from the UK’s National Career Service categories, multiplied by number of years at each category. The mid-life specific score (MS, or occupation) was based on the standard occupational classification codes from the UK Office of National Statistics, summed across seven mid-life periods. The late-life specific score (LS, or post-retirement activities) reflected social and intellectual activities such as travel or participation in volunteer organisations. The current LEQ does not cater for some “specific” activities being undertaken at other life stages (eg education in mid- or late life) and so their potential contribution to CR could not be evaluated in the current study.

Scores for non-specific activities in youth (YA), mid-life (MA) and late-life (LA) were summed across seven questions about social, intellectual and physical activities. These addressed participation in i) travel, ii) social outings, iii) playing a musical instrument, iv) artistic pastimes, v) physical activity (mild, moderate, vigorous), vi) reading, vii) speaking a second language. Each of the six resulting scores (two types of activity, specific and non-specific, across the three life phases) was scaled to a score from 0-10.

T1- and T2-weighted 1mm isotropic MRI scans were available for 196 participants. Brain structure was measured in terms of total grey matter volume (TGM, see Taylor et al. 2017 and Supplementary Materials). Two participants with outlying adjusted TGM values were removed. The analysis used linear regression via the “lm” function in R 3.5.0 (R Core Team, 2016) to relate Cattell scores to the six LEQ scores above, plus age and sex. Data and analysis scripts are available on the Open Science Framework here: https://osf.io/32gme/, which includes individual scores for the 13 LEQ questions that comprise the MA sum score.
2. Results

The covariance and correlation of Cattell, LEQ scores and Age are shown in a Supplementary Table (Supplementary Materials). All LEQ scores were significantly positively correlated with each other, and with Cattell. Multivariable regression of late-life cognitive ability on the LEQ scores, with age and sex as covariates, showed a strong overall association (adjusted $R^2=0.355$, $F(8,196)=15.0$, $p<.001$). In addition to the expected negative effect of age, the coefficients in Table 1 revealed a unique, positive contribution of YS (i.e. education), replicating previous studies such as Richards and Deary (2005). More interestingly, midlife nonspecific activities (MA) also made an independent positive contribution after adjustment for all other factors. No other LEQ-based category, including the current late-life activities being performed by the individuals (reflected in both LS and LA), made an independent contribution. After adjusting for age and sex, separate regression of MA on cognitive ability revealed an effect size of $R^2=16.0$, comparable to that for YS (education, $R^2=15.8\%$).

*Insert Table 1 around here*

To examine whether MAs contributed not just to late-life cognitive ability but also to cognitive reserve, rigorously defined, we determined whether MAs moderated the relationship between late-life cognitive ability and brain structure (as indexed by TGM). We first adjusted Cattell and TGM scores for 1) Education (YS score), 2) Total Intracranial Volume (TIV) to correct for inter-individual differences in head size, 3) Age and 4) Sex. Linear regression showed an interaction between TGM and MA in predicting Cattell (normalized interaction coefficient = -0.722, standard error = 0.331, $p=0.030$). This moderating term was negative, meaning that the relationship between cognitive ability and brain health diminished with higher MA, as predicted by cognitive reserve theory. This effect is visualized in Figure 1, where a median split is used to divide participants into low and high MA groups: a more positive slope can be seen in the low MA group than in the high MA group.
3. Discussion

This study tested the hypothesis that lifestyle activities in mid-life contribute to the cognitive reserve that supports cognitive ability in late-life. Consistent with this, we found that general mid-life activities (MA) make an independent contribution to late-life cognition over and above age, sex, education, occupation and current (late-life) activities, replicating the findings of Gow et al (2017). However, a rigorous determination of cognitive reserve requires further evidence that it moderates the association between a brain state and cognitive outcome (e.g., Brayne et al 2010). Importantly, we found such a moderation by MA on the relationship between total grey matter volume and cognitive ability, such that cognitive ability in those older individuals who had been involved in rich and varied lifestyle activities in mid-life were less dependent on their current structural brain health.

This evidence that midlife activities contribute to cognitive reserve may have ramifications for the primary prevention of dementia. These activities reflect lifestyle choices, and are therefore amenable to modification. Our observation that the impact of MA was independent of educational attainment and occupational status, suggests that a public health initiative aimed at boosting cognitive reserve via enhancement of MA is generalizable to the entire adult population. The importance of such initiatives for primary prevention of dementia is underscored by the total failure to date to identify interventions for secondary prevention of dementia.

This study does not address the relative contribution to cognitive reserve of the various intellectual, social and physical components of the LEQ midlife activities score. This is because the LEQ non-specific scores are a composite across 13 questions that are not designed to separate these different components. However, other studies shed some light on this issue. While some studies have suggested that physical activity reduces future dementia risk (Richards et al 2003, Middleton et al 2010), the UK Whitehall II study found no association between physical activity and subsequent 15 year cognitive decline (Sabia et al 2017). Similarly, Gow et al (2017) found that mid-life intellectual and social activities, but not
physical activity, were associated with late-life cognitive health. Other studies have also highlighted the benefits of intellectual and social activity (Akbaraly et al. 2009, Köhncke et al 2016). Future work, complementing the LEQ with specific measures for each of these MA components, will be needed to clarify this issue.

Another key objective for future research is identification of the biological mechanisms by which MA exert an effect on late life cognitive ability. On a molecular level, it is possible that the effect is mediated via BDNF-induced synaptogenesis and neurogenesis. There is evidence that the BDNF val66met polymorphism moderates the relationship between CR and cognition (Ward et al. 2015), and high levels of BDNF expression in postmortem brain are associated with slower rates of cognitive decline (Buchman et al. 2016). At systems level, it has long been speculated that a CR effect may be mediated via alterations in functional connectivity. Task-free fMRI studies have identified CR-related changes in network topology (Marques et al. 2016), while there is emerging evidence from task-related fMRI work that CR may be associated with a functional network that is task-invariant (Stern 2017).

Complementing these whole brain approaches, others have identified regional connectivity changes associated with CR (Franzmeier et al. 2017).

There are important limitations of this study. First, LEQ measures are self-report, raising the possibility that more cognitively healthy older people remember more lifetime activities. Second, a limitation of cross-sectional studies is the possibility of reverse causation, namely that a higher cognitive ability throughout life triggers the pursuit of more beneficial lifestyle activities, rather than vice versa. Third, the study cohort did not undergo testing for AD biomarkers and so the potential confounding effect of AD pathology could not be addressed. Finally, LEQ scores were highly positively correlated, so we cannot infer that other variables (e.g., occupation) play no role in late-life cognitive ability. These limitations could be addressed by intervention studies targeting specific aspects of mid-life activity, with later life cognitive ability as the outcome measure, though such studies would require a timeframe of 20-30 years.

In conclusion, our findings suggest that lifestyle activities in mid-life can contribute to cognitive reserve and support late-life cognition. The potential modifiability of these activities has
important implications for public health initiatives aimed at reducing the risk of dementia.

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References


Cattell, RB (1971) Abilities: their Structure, Growth, and Action (Houghton-Mifflin)


Supplementary Materials

Table

<table>
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<th></th>
<th>Cattell</th>
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Variance (diagonal), covariance (upper triangle) and correlation (lower triangle) of Cattell, LEQ scores and Age. Last Column shows min/max range. YS = youth specific activities, YA = nonspecific youth activities, MS = mid-life specific activities; MA = nonspecific mid-life activities, LS = late-life specific activities; LA = nonspecific late-life activities.

*p < 0.05.

CamCAN sample

As part of Phase 2 of the Cambridge Centre for Ageing and Neuroscience (Cam-CAN), data on a wide range of lifestyle, cognitive and neural tests was collected from a healthy, population-based human adult sample, described in more detail in (Shafto et al., 2014). Exclusion criteria include low Mini Mental State Exam (MMSE) (24 or lower), poor hearing (failing to hear 35dB at 1000 Hz in either ear), poor vision (below 20/50 on Snellen test), poor English knowledge (non-native or non-bilingual English speakers), self-reported substance
abuse, and serious health conditions that affect participation (e.g. self-reported major psychiatric conditions, current chemo/radiotherapy, or a history of stroke). Participants whose Primary Care Physician felt are inappropriate to include did not take part in the study. We also excluded people with MRI contraindications including disallowed implants, pacemakers, recent surgery or any previous brain surgery, current pregnancy, facial- or very recent tattoos, or a history of multiple seizures or fits) as well as comfort-related contraindications (e.g. claustrophobia or self-reported inability to lie still for an hour). Ethical approval for the study was obtained from the Cambridgeshire 2 (now East of England-Cambridge Central) Research Ethics Committee. Participants gave full informed consent. The raw data can be acquired through the Cam-CAN data portal (https://camcan-archive.mrc-cbu.cam.ac.uk/dataaccess/index.php)

*MRI acquisition and analysis*

MRI data were acquired using a Siemens 3T TIM TRIO (Siemens, Erlangen, Germany) with a 32-channel head coil at the MRC Cognition & Brain Sciences Unit (CBU), Cambridge, UK. Anatomical images were acquired with a resolution of 1mm³ isotropic using a T1-weighted MPRAGE sequence (TR: 2250ms; TE: 2.98ms; TI: 900ms; 190Hz; flip angle: 9°; FOV: 256 × 240 × 192mm; GRAPPA acceleration factor: 2; acquisition time of 4 minutes and 32 seconds), and a 1mm³ isotropic T2-weighted SPACE sequence (TR: 2800ms; TE: 408ms; flip angle: 9°; FOV: 256 × 256 × 192mm; GRAPPA acceleration factor: 2; acquisition time of 4 minutes and 32 seconds).

MRI data were analysed using the SPM12 software (www.fil.ion.ucl.ac.uk/spm), implemented in the AA 4.2 batching software (https://github.com/rhodricusack/automaticanalysis; for complete description of data and pipelines, see Taylor et al., 2017). The T1 and T2 images were initially coregistered to the MNI template using a rigid-body transformation, and then combined in multimodal segmentation in order to segment the brain into 6 tissue classes: GM, WM, cerebrospinal fluid (CSF), bone, soft tissue, and residual noise. The GM images were then submitted to diffeomorphic registration (DARTEL) to create group template images,
which was then affine-transformed to the MNI template. To accommodate changes in volume from these transformations, the GM images were modulated by the Jacobean of the deformations to produce estimates in MNI space of the original local GM volume. These were then summed across voxels to estimate the total GM, WM and CSF, which were themselves summed to estimate the total intracranial volume (TIV) (see Malone et al., 2015).