Neurocognitive profiles in a community sample of children & adolescents: sociodemographic, behavioral and emotional correlates

Marc Barakat†, Wael Shamseddeen†, Elissa Hanna2, Leyla Akoury Dirani1, David Brent3, Giovanna Porta3 and Fadi T. Maalouf1*

Abstract

Objectives We aim to investigate the association between cognitive profiles of children and adolescents, classified using latent class analysis and emotional, behavioral and sociodemographic factors in a community-based sample of children and adolescents.

Methods The sample consists of 161 participants recruited from a nationally representative household sample of 1517 children and adolescents who participated in a national mental health survey. Participants and their parents completed the following scales: Strengths and Difficulties Questionnaire, Mood and Feelings Questionnaire, and Screen for Child Anxiety Related Emotional Disorders. Participants were then administered a battery of cognitive tests from the Cambridge Automated Neuropsychological Assessment Battery as well as the Wechsler Nonverbal Scale of Ability.

Results Latent class analysis showed a 2-class model that was stable and had good entropy (0.837). Class 1 included 58 participants (36.0%), and Class 2 included 103 participants (64.0%). Participants belonging to Class 1 performed better than those in Class 2 on all cognitive tasks. Chances for belonging to a specific Class varied based on different sociodemographic, behavioral and emotional factors: older participants with higher Intellectual Quotient, higher parental educational level, lower reported hyperactivity and better-reported mood were more likely to belong to Class 1.

Conclusion Our results show a tight correlation between different emotional, behavioral and sociodemographic variables on one hand and cognitive functioning in children and adolescents on the other hand. This supports the continuous need for an integrative approach when building norms for cognitive tests that account at the least for all these variables.

Keywords Neurocognitive profile, IQ, Behavioral correlates, Emotional correlates, Children, Adolescents

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Background
Emotions and cognition have long been considered as 2 independent systems [37]. However, current research has shown that they influence one another, and even share underlying brain structures, such as the amygdala [50]. Brooks et al. [9] investigated the impact of depression on individuals’ attention, reaction times, and memory. Results showed that children and adolescents with depression performed worse than healthy participants on measures of verbal memory, visual memory, reaction time, and tasks of attention. A similar negative impact of depression on attention was also reported in adults by Sommerfeldt et al. [44] who showed that the presence of another disorder alongside depression exacerbates cognitive dysfunction. Furthermore, intellectual abilities were reported to be possibly influenced by emotions, with lower IQ scores being observed with severe symptoms of depression [6].

Recent studies investigated the ability of some cognitive measures to predict psychological distress later in life. For example, Kertz et al. [26] found that deficits in attention shifting during preschool years were linked to increased severity of anxiety three and half years later, and to increased depression severity five and a half years later, whereas deficits in inhibition were predictive of greater depression severity and increased anxiety several years later. Likewise, Gale et al. [19] showed that adolescents who had slow reaction times were at increased risk for depression and anxiety during adulthood. Finally, in a recent longitudinal study, Piek et al. [38] suggested that the development of motor skills during infancy and early childhood could predict anxiety and depressive symptoms at the ages 6 to 12 years.

Other studies have explored the clinical value of cognitive profiles in the course of psychiatric disorders. For example, our team has shown that deficits in executive functioning could be state markers rather than trait markers of depression [32], whereas sustained attention impairment seemed to be specific to patients with bipolar disorder [33]. F. Maalouf et al. [30] also showed that adolescents with depression who exhibited lower impulsivity and less difficulty in inhibition attention showed greater improvement in symptoms after a six-week treatment with a selective serotonin reuptake inhibitor (SSRI).

In addition to this established correlation between emotions and cognition, one prominent contributor to cognitive development are sociodemographic factors. In fact, cognitive abilities are known to change with age mostly by improving over time, before they start to decline in late adulthood and old age. Sikora-Wachowicz et al. [42] examined age-related differences in short-term memory. Their results showed that humans become more prone to false memories as they age. Fandakova et al. [14] showed a similar effect of age on short term memory, with better results seen in young adults and teenagers compared to old adults and children. Amundsen et al. [3] also showed a significant age effect on visual-spatial working memory when comparing ninth graders’ performance on a “Visual Test for Learning” to that of younger children. On the other hand, Tao et al. [45] sought to study how attention, another cognitive process, develops among school-aged children. The study results showed that children aged between 9 and 12 years performed better on tasks measuring focused attention, alertness, attentional switching, supervisory attention, divided attention, sustained attention, and spatial attention compared to children aged between 7 and 8 years. These findings suggest that working memory and executive function improve with age in children.

Gender is another factor that was shown to possibly underly cognitive differences between individuals. For example, Rodríguez-Negro et al. [41] found gender differences in motor abilities of children between the ages of 6 and 8 years, with girls showing better balance skills and boys better motor precision and motor agility skills. On the other hand, Riley et al. [39] showed gender-related differences in sustained attention when comparing the performance of participants in countries where gender inequality pervades to that of others where no gender inequality was found. In a meta-analysis conducted by Voyer et al. [51] on gender differences in visual-spatial working memory in participants between the ages of 3 and 86, males were found to have better performance than females, but only in participants aged between 13 and 17 years.

Some of the previous results seemed to suggest that gender effect could be influenced by other moderating factors such as age. In fact, Kalyanshetti and Vastrad [25] examined the effect of both gender and age on tactile, auditory, and visual reaction times in children (between 8 and 12 years old) and adults (between 18 and 25 years old) of both genders. The results indicated that boys were significantly faster than girls, however, in adults, females and males did not differ on reaction times. Moreover, when comparing children and adults of both genders, adults (18 to 25 years old) exhibited faster reaction times than children of ages 8 to 12 years old. This study suggested that age could be a possible moderating factor between gender and reaction time. Likewise, Kokštejn et al. [27] examined the association among age, gender, and motor abilities in preschoolers between the ages of 3 and 6 years. Girls of ages 3 and 4 years old were found to have better fine motor and balance skills compared to boys of the same age; in contrast, six-year-old boys outperformed girls of the same age in catching and aiming. Moreover, when the authors controlled for age, they
found a significant gender-related difference in motor abilities. Cognitive abilities, such as motor abilities, working memory, and sustained attention, are also believed to be correlated with the socioeconomic status. For example, the development of motor abilities in school-aged children (6 to 10 years old) was found to increase as the socioeconomic status (SES) of the children's families increased [15], Aiman et al. [2] found that children who came from a low SES had lower fine motor performance compared to those who came from higher SES. Furthermore, Ferreira et al. [15] showed that better home environment quality was related to better development of motor abilities in children. However, they found that age moderated the relationship between SES, home environment, and motor abilities; younger children seemed to be impacted by SES more than older children. Similarly, Da Rosa Piccolo et al. [12] showed that children with higher SES had better performance on working memory and executive function tasks, but SES had a stronger influence on the performance of children who were younger than nine years old. Last, Lawson et al. [28] investigated the association between SES and executive functions across childhood, adolescence, and young adulthood. They found that participants of lower SES performed worse than those of higher SES. SES has also been shown to influence the neural systems of selective attention [52]. In conclusion, available evidence suggests that higher SES positively impacts the development of motor abilities, memory, and attention.

In summary, studies seem to support a significant correlation between cognitive performance and emotional disorders such as anxiety and depression in children and adolescents. However, none of these studies has explored the correlation between cognitive performance and reported symptoms of anxiety and depression in a non-clinical community-based population while controlling for the effects of different sociodemographic factors on cognition, such as age, gender, IQ, and parental SES.

This is the first study to examine the existing cognitive profiles in a community-based sample of children and adolescents, and to investigate the emotional symptoms as well as behavioral and sociodemographic factors associated with those cognitive profiles.

Methods
Sampling and recruitment

This study is part of a larger research project that was conducted by our team. A nationally representative household sample of 1517 children and adolescents living in Lebanon completed this study between February 2018 and November 2018 following recruitment using probability sampling as part of the Psychopathology in Children and Adolescents Study (PALS) [31] or the detailed methodology including sampling techniques. After completing the main study protocol, participants aged between 5 and 17 years 11 months were given the option to participate in the present study. A total of 161 agreed to participate and sat for the testing around within 1–3 months following the original interview. As presented in Table 1, compared to those who refused, participants were more likely to come from families of lower income ($p = 0.02$). With respect to father’s education, participants’ fathers were more likely to have elementary education or less, while non participants’ father were more likely to have high school education or less, but both groups were similar with respect to university education (Table 1). On the other hand, there was no difference between participants and non-participants with respect to age, sex, maternal education, paternal employment, maternal employment, or history of any chronic illness (Table 1). The two groups also differed on depression scores but there were no differences between the two groups on the other measures of emotional and behavioral symptoms.

The study was conducted in accordance with Helsinki Declaration. It was approved by the local Internal Review Board. Written informed consent was obtained from parents of children and child assent were obtained. Computerized cognitive tasks that take about 40–50 min to complete [40] as well as a short standardized assessment of intellectual abilities were administered.

Instruments and measures

General information sheet

A general information sheet was first completed by a parent/legal guardian who provided information on basic demographics including family structure, family income and parental level of education.

Emotions and behavior screening

The Strengths and Difficulties Questionnaire (SDQ)- parent and child version The SDQ is a self-report that was used to screen for emotional, conduct, and hyperactivity difficulties in children aged 5 to 7 years and for conduct and hyperactivity symptoms in participants aged 8 to 17 years [20]. We used an Arabic version of the SDQ that was previously validated [53].

The Mood and Feelings Questionnaire MFQ The MFQ is a 33 item self-report measure of depressive symptoms among children/adolescents aged 8 to 18 [11]. The previously validated Arabic version along with previously established cutoff scores in parents and children were used [46].
The SCARED screens for anxiety disorders in children and adolescents [7]. SCARED was completed by children and adolescents aged 8 to 17 and one parent/guardian and previously established local cut-off scores were used [24].

Cognitive tasks
A battery of CANTAB tasks was then administered by a trained clinical psychologist. The tests were done on an IPAD where software recorded the performance of each participant and data was later extracted and linked to the participants. Participants also underwent the Weschler

<table>
<thead>
<tr>
<th>Table 1</th>
<th>Socio-demographics, emotional and behavioral analysis</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Cognitive Testing</strong></td>
<td></td>
</tr>
<tr>
<td>Not included N = 1356</td>
<td>Included N = 161</td>
</tr>
<tr>
<td>Sex</td>
<td></td>
</tr>
<tr>
<td>Male</td>
<td>627 (49.6)</td>
</tr>
<tr>
<td>Female</td>
<td>684 (50.4)</td>
</tr>
<tr>
<td>Income</td>
<td></td>
</tr>
<tr>
<td>up to $1000</td>
<td>629 (51.6)</td>
</tr>
<tr>
<td>$1000-$2000</td>
<td>448 (36.8)</td>
</tr>
<tr>
<td>more than $2000</td>
<td>141 (11.6)</td>
</tr>
<tr>
<td>Father education</td>
<td></td>
</tr>
<tr>
<td>Elementary or less</td>
<td>319 (23.8)</td>
</tr>
<tr>
<td>High school or less</td>
<td>815 (60.8)</td>
</tr>
<tr>
<td>University or more</td>
<td>207 (15.4)</td>
</tr>
<tr>
<td>Mother education</td>
<td></td>
</tr>
<tr>
<td>Elementary or less</td>
<td>182 (13.5)</td>
</tr>
<tr>
<td>High school or less</td>
<td>923 (68.7)</td>
</tr>
<tr>
<td>University or more</td>
<td>239 (17.8)</td>
</tr>
<tr>
<td>Father employment</td>
<td></td>
</tr>
<tr>
<td>Employed</td>
<td>1236 (91.2)</td>
</tr>
<tr>
<td>Mother employment</td>
<td></td>
</tr>
<tr>
<td>Unemployed</td>
<td>971 (72.2)</td>
</tr>
<tr>
<td>Employed</td>
<td>374 (27.8)</td>
</tr>
<tr>
<td>Any chronic physical illness&lt;sup&gt;a&lt;/sup&gt;</td>
<td></td>
</tr>
<tr>
<td>Mean ± SD</td>
<td>Mean ± SD</td>
</tr>
<tr>
<td>Age</td>
<td>10.53 ± 3.28</td>
</tr>
<tr>
<td>SDQ emotional Child&lt;sup&gt;b&lt;/sup&gt;</td>
<td>1.94 ± 2.11</td>
</tr>
<tr>
<td>SDQ emotional Parent&lt;sup&gt;c&lt;/sup&gt;</td>
<td>1.80 ± 1.96</td>
</tr>
<tr>
<td>SDQ hyperactivity Child&lt;sup&gt;d&lt;/sup&gt;</td>
<td>2.77 ± 2.15</td>
</tr>
<tr>
<td>SDQ hyperactivity Parent&lt;sup&gt;e&lt;/sup&gt;</td>
<td>3.30 ± 2.35</td>
</tr>
<tr>
<td>MFQ Child&lt;sup&gt;d&lt;/sup&gt;</td>
<td>6.17 ± 7.93</td>
</tr>
<tr>
<td>MFQ Parent&lt;sup&gt;d&lt;/sup&gt;</td>
<td>4.72 ± 7.09</td>
</tr>
<tr>
<td>SCARED Child&lt;sup&gt;d&lt;/sup&gt;</td>
<td>13.78 ± 12.39</td>
</tr>
<tr>
<td>SCARED Parent&lt;sup&gt;d&lt;/sup&gt;</td>
<td>10.27 ± 10.26</td>
</tr>
</tbody>
</table>

<sup>a</sup> Any chronic disorder: Arthritis, asthma, cancer, epilepsy, immunodeficiency, inflammatory bowel disease, heart problems, liver or kidney disease, migraine headaches, anemia, sensory impairment, obesity, and another other chronic physical illness

<sup>b</sup> N = 437 Not included & 56 included

<sup>c</sup> N = 1236 Not included & 148 included

<sup>d</sup> N = 1038 Not included & 130 Included

Screen for Child Anxiety Related Emotional Disorders (SCARED) The SCARED screens for anxiety disorders in children and adolescents [7]. SCARED was completed by children and adolescents aged 8 to 17 and one parent/guardian and previously established local cut-off scores were used [24].
Non-Verbal Abilities Assessment (WNV) test. The CAT- 
TAB battery of tests was done in 40–50 min whereas the 
WNV test took an additional 20 min.

The Motor Screening test (MOT)  MOT was adminis-
tered at the beginning of the battery as an introduction to 
the touch screen for the participant. It screens for visual, 
movement, and comprehension difficulties. Participants 
are instructed to touch a cross that appears in different 
locations on the screen. Mean latency in milliseconds 
of correct responses (MOTML) is the main outcome 
measure.

Rapid Visual Processing Task (RVP)  RVP assesses sus-
tained attention. Single digits in random order on the 
screen. Participants are instructed to respond to three 
sequences of three digits “2–4–6,” “3–5–7”; and “4–6–8”. 
RVP A and RVP Probability of Hit (RVPPH) were our 
main outcome variables as they measured how good the 
subject was at detecting target sequences. A higher score 
reflected a better performance.

Delayed Matching to Sample (DMS)  DMS assesses 
short-term memory by presenting participants with a 
complex pattern on a computer screen (the sample) fol-
lowed by four patterns only one of which is identical to 
the first pattern. In some trials, the sample and choice 
items are presented consecutively, while in others there 
is a 0, 4, or 12-s delay between the sample and the choice 
items. Participants are instructed to touch the pattern on 
the screen that matches the sample. Our main outcome 
measure was the percentage of correct answers for all 
delays.

Stop Signal Task (SST)  The SST serves as an indicator 
of one’s ability to inhibit responses. In this task, a white 
ring appears on a black screen, followed by a visual arrow 
(pointing left or right) after a 500 ms delay. Participants 
are instructed to touch the right-hand press pad for 
right-pointing arrows and the left-hand press pad for left-
pointing arrows. However, they are required to withhold 
their response if a beep (auditory signal) is heard. The 
primary outcome variable, SSTSSRT (Stop Signal Reac-
tion Time), quantifies the estimated time at which an 
individual can successfully inhibit responses 50% of the 
time.

Reaction Time (RTI)  Reaction Time was employed 
to evaluate both motor and mental response speeds, 
encompassing movement time, reaction time, response 
accuracy, and impulsivity. In the task, participants are 
required to choose and maintain a button press at the 
screen’s bottom. Circles, one in the simple mode and 
five in the five-choice mode, are presented above. In 
each instance, a yellow dot emerges in one of the circles, 
prompting the participant to swiftly release the bottom 
button and select the circle containing the dot. The pri-
mary outcome measures include the median duration for 
a subject to release the response button and the median 
time taken for a subject to release the response button 
and choose the target stimulus.

Paired Associates Learning (PAL)  Paired Associates 
Learning assesses visual memory and the ability to 
acquire new information. In this task, boxes are presented 
on the screen and are randomly “opened.” Within one or 
more of these boxes, a pattern is revealed. Subsequently, 
the patterns are individually displayed in the center of 
the screen, and the participant is tasked with identifying 
the box where the pattern was originally located. In the 
event of an error, the boxes are opened sequentially as a 
reminder of the pattern locations. The primary outcome 
is the number of times a subject correctly identifies the 
box on their initial attempt while recalling the pattern 
locations.

Spatial Working Memory (SWM)  The Spatial Working 
Memory task assesses the retention and manipulation of 
visuospatial information, serving as a metric for strategy 
and working memory errors. The test initiates with the 
display of colored squares on the screen. Participants, 
through a process of elimination, are required to identify 
one yellow ‘token’ in each set of boxes and utilize them 
to fill an empty column on the right side of the screen. 
The number of boxes can progressively increase, up to a 
maximum of 12, for participants to search. Primary out-
come measures include the number of instances subjects 
incorrectly revisit a box where a token was previously 
found and the number of instances subjects initiate a new 
search pattern from the same starting box as before. Con-
sistently starting from the same point suggests the use of 
a planned strategy in locating the tokens.

Wechsler Nonverbal Scale of Ability (WNV)  The WNV 
is an assessment tool designed to measure intellectual 
abilities in culturally and linguistically diverse groups, 
employing nonverbal content. The battery comprises 
two forms, each featuring four subtests. The first form, 
applicable to participants aged 4 to 7 years, includes 
matrices, coding, object assembly, and recognition. The 
second form, suitable for examinees aged 8 to 17 years, 
consists of matrices, coding, spatial span, and picture 
arrangement.

A condensed version allows the use of matrices and 
recognition subtests to calculate the Full-Scale Score for
the younger group, and matrices and picture arrangement subtests for the older group. Raw scores obtained in each subtest are then converted into scale scores using norms specific to each age group.

Data analysis
In order to classify participants based on their performance on cognitive tests, latent class analysis (LCA) was performed using the scores of the tasks’ outcome measures. The analysis was initially performed using 1-class model (BIC = 4446.86), then the number of classes was increased to find the model with the best fit to the data, as determined by the Bayesian information criteria (BIC = 4446.86 for 1-class model, 4295.84 for 2-class model, and 4187.29 for 3-class model), and Lo-Mendell-Rubin (LMR) adjusted likelihood ratio tests (p = 0.01 for 2-class model, and 0.13 for 3-class model. To be able to describe the cognitive profiles of those classes, they were compared with respect to the 12 cognitive measures using an independent sample t-test.

Classes were also compared with respect to demographics and parental education and employment using Pearson Chi-square or Fisher exact test and with respect to SDQ, SCARED, and MFQ scores using independent sample t-tests. Finally, variables significantly associated with “cognitive class” were entered into a multivariate binary logistic regression with the cognitive class being the outcome. This analysis was performed for all age groups using SDQ scores, and for patients above age 8 using MFQ and SCARED scores.

The latent class analysis was performed using Mplus, while the rest of the statistical analysis was performed using SPSS. P-values less than 0.05 were considered to be statistically significant.

Results
The sample consisted of 161 participants with 52.8% (N=85) being males. Age was between 5 years and 7-years-11-months for 7 participants (4.3%), between 8 years and 11-years-11-months for 86 participants (53.4%), and between 12 years and 17-years-11-months for 68 participants (42.2%). Baseline data are presented in Table 1.

Latent class analysis
Latent class analysis showed that the model that best fits the data was the 2-class model was stable and had good entropy (0.837). In addition, compared to the 1-class solution, the 2-class model had lower BIC (4295.84 vs. 4446.86) and significant Vuong-Lo-Mendell-Rubin and Lo-Mendell-Rubin Adjusted Likelihood Ratio Tests (p’s = 0.01).

The first class (Class 1) included 58 participants (36.0%), and the second class (Class 2) included 103 participants (64.0%). Class 1 had significantly higher raw scores on tasks measuring correct responses (i.e. DMS Percent Correct (All Delays), RVP A, RVP Probability of Hit, and PAL First Attempt Memory Score compared to the second class), and lower raw scores on tasks measuring speed (i.e. SST Stop Signal Reaction Time, MOT Mean Latency, RTI Median Five Choice Movement Time, RTI Median Five Choice Reaction Time, RTI Simple Median Movement Time, RTI Simple Median Reaction Time, SWM Between Errors, and SWM Strategy (6–8 Boxes)) (Table 2). In summary, these results suggest better performance of participants belonging to Class 1 on all tasks.

Factors associated with classes
As reflected in Table 2, there was no significant difference between the two classes with respect to sex (p = 0.59). On the other hand, there was a significant association with age, with Class 1 being significantly older (12.5 ± 2.87 years) than Class 2 (10.31 ± 2.88 years; p = 0.001). In addition, Class 1 was associated with higher parental education (p = 0.05) and parental income (p = 0.01) as compared to Class 2. Finally, participants in Class 1 were found to have a higher mean IQ score compared to class two (p < 0.001).

MFQ and SCARED reports were available for participants who were at least 8 years old. There was no significant difference between the two groups with respect to child report on MFQ and SCARED. On the other hand, Class 1 was significantly associated with lower parental report on the MFQ (p = 0.00) and SCARED as compared to Class 2 (p = 0.029) (Table 3).

Multivariate logistic regression
In the multivariate logistic regression (Table 4), odds of belonging to Class 2 compared to Class 1 were lower with increased age (OR = 0.74, 95% CI = 0.63–0.86), and in children of parents with high school educational level compared to those of parents with elementary educational level (OR = 0.09, 95% CI = 0.01–0.51). In addition, there was a borderline significant association with SDQ hyperactivity with odds of belonging to Class 2 being 1.2 times higher for higher “SDQ hyperactivity score as per parent report”.

As MFQ and SCARED instruments were administered to children above age 8, the above analysis was repeated among this age group only (Table 5). There were no changes in the results presented in Tables 2 and 3. With respect to the multivariate logistic regression (Table 5), age group, highest parental education level, and MFQ score as per parent report were
Table 2  Classes versus neuropsychological testing

<table>
<thead>
<tr>
<th></th>
<th>Class 1</th>
<th>Class 2</th>
<th>t-test</th>
<th>df</th>
<th>p-value</th>
</tr>
</thead>
<tbody>
<tr>
<td>SST Stop Signal Reaction Time&lt;sup&gt;a&lt;/sup&gt;</td>
<td>309.97 ± 79.35</td>
<td>338.61 ± 75.62</td>
<td>-2.16</td>
<td>141</td>
<td>0.032</td>
</tr>
<tr>
<td>MOT Mean Latency&lt;sup&gt;a&lt;/sup&gt;</td>
<td>764.85 ± 151.52</td>
<td>960.97 ± 335.11</td>
<td>-5.09</td>
<td>153.34</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>RTI Median FiveChoice Movement Time&lt;sup&gt;a&lt;/sup&gt;</td>
<td>300.22 ± 86.62</td>
<td>348.48 ± 78.2</td>
<td>-3.61</td>
<td>159</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>RTI Median FiveChoice Reaction Time&lt;sup&gt;a&lt;/sup&gt;</td>
<td>413.47 ± 69.95</td>
<td>467.55 ± 109.21</td>
<td>-3.82</td>
<td>156.31</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>RTI Simple Median Movement Time&lt;sup&gt;a&lt;/sup&gt;</td>
<td>263.89 ± 93.92</td>
<td>321.44 ± 97.48</td>
<td>-3.64</td>
<td>159</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>RTI Simple Median Reaction Time&lt;sup&gt;a&lt;/sup&gt;</td>
<td>360.82 ± 46.41</td>
<td>414.07 ± 98.15</td>
<td>-4.66</td>
<td>155.27</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>SWM Between Errors&lt;sup&gt;a&lt;/sup&gt;</td>
<td>6.34 ± 5.21</td>
<td>21.2 ± 4.83</td>
<td>-18.18</td>
<td>158</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>SWM Strategy (6–8 Boxes)&lt;sup&gt;a&lt;/sup&gt;</td>
<td>7.21 ± 1.78</td>
<td>9.78 ± 1.36</td>
<td>-9.53</td>
<td>95.10</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>DMS Percent Correct (All Delays)&lt;sup&gt;b&lt;/sup&gt;</td>
<td>82.21 ± 12.07</td>
<td>71.77 ± 16.38</td>
<td>4.55</td>
<td>144.73</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>RVP A&lt;sup&gt;b&lt;/sup&gt;</td>
<td>0.84 ± 0.05</td>
<td>0.79 ± 0.07</td>
<td>4.57</td>
<td>145</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>RVP Probability of Hit&lt;sup&gt;b&lt;/sup&gt;</td>
<td>0.45 ± 0.14</td>
<td>0.37 ± 0.17</td>
<td>3.06</td>
<td>145</td>
<td>0.003</td>
</tr>
<tr>
<td>PAL First Attempt Memory Score&lt;sup&gt;b&lt;/sup&gt;</td>
<td>14.19 ± 3.67</td>
<td>12.09 ± 3.86</td>
<td>3.37</td>
<td>159</td>
<td>&lt;0.001</td>
</tr>
</tbody>
</table>

SST Stop Signal Task, MOT Motor Screening Task, RTI Reaction Time, SWM Spatial Working Memory, DMS Delayed Matching to Sample; RVP: Rapid Visual Information Processing
<sup>a</sup> A lower score reflected a better performance
<sup>b</sup> A higher score reflected a better performance

Table 3  Classes versus demographics and psychiatric measures

<table>
<thead>
<tr>
<th></th>
<th>Class 1</th>
<th>Class 2</th>
<th>χ²</th>
<th>df</th>
<th>p-value</th>
</tr>
</thead>
<tbody>
<tr>
<td>Gender</td>
<td></td>
<td></td>
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<td></td>
</tr>
<tr>
<td>Male</td>
<td>29 (50.0)</td>
<td>56 (54.4)</td>
<td>0.28</td>
<td>1</td>
<td>0.59</td>
</tr>
<tr>
<td>Female</td>
<td>29 (50.0)</td>
<td>44 (45.6)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Mother</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Unemployed</td>
<td>36 (62.1)</td>
<td>72 (69.9)</td>
<td>1.03</td>
<td>1</td>
<td>0.31</td>
</tr>
<tr>
<td>employed</td>
<td>22 (37.9)</td>
<td>31 (30.1)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Highest Parental Education</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Elementary</td>
<td>2 (3.5)</td>
<td>19 (18.6)</td>
<td>FET = 7.21</td>
<td>-</td>
<td>0.05</td>
</tr>
<tr>
<td>Highschool</td>
<td>41 (71.9)</td>
<td>62 (60.8)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>University</td>
<td>14 (24.6)</td>
<td>21 (20.6)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Income</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>up to $1000</td>
<td>26 (53.1)</td>
<td>67 (72.8)</td>
<td>6.30</td>
<td>2</td>
<td>0.01</td>
</tr>
<tr>
<td>$1000–$2000</td>
<td>13 (26.5)</td>
<td>17 (18.5)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>more than $2000</td>
<td>10 (20.4)</td>
<td>8 (8.7)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Age</td>
<td>12.5 ± 2.87</td>
<td>10.31 ± 2.88</td>
<td>4.64</td>
<td>159</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>SDQ emotional Child&lt;sup&gt;a&lt;/sup&gt;</td>
<td>1.83 ± 2.01</td>
<td>2.78 ± 2.32</td>
<td>-1.80</td>
<td>66</td>
<td>0.08</td>
</tr>
<tr>
<td>SDQ emotional Parent</td>
<td>1.66 ± 1.65</td>
<td>2.76 ± 2.49</td>
<td>-3.36</td>
<td>154.85</td>
<td>.001</td>
</tr>
<tr>
<td>SDQ hyperactivity Child&lt;sup&gt;a&lt;/sup&gt;</td>
<td>3.11 ± 2.3</td>
<td>3.91 ± 2.52</td>
<td>-1.36</td>
<td>66</td>
<td>0.18</td>
</tr>
<tr>
<td>SDQ hyperactivity Parent</td>
<td>2.91 ± 2.5</td>
<td>4.38 ± 2.7</td>
<td>-3.39</td>
<td>159</td>
<td>.001</td>
</tr>
<tr>
<td>MFQ Child&lt;sup&gt;a&lt;/sup&gt;</td>
<td>6.4 ± 7.52</td>
<td>8.17 ± 9.23</td>
<td>-1.23</td>
<td>152</td>
<td>0.219</td>
</tr>
<tr>
<td>MFQ Parent&lt;sup&gt;b&lt;/sup&gt;</td>
<td>4.17 ± 4.34</td>
<td>9.04 ± 10.98</td>
<td>-3.87</td>
<td>135.38</td>
<td>0.000</td>
</tr>
<tr>
<td>SCARED Child&lt;sup&gt;a&lt;/sup&gt;</td>
<td>15.52 ± 11.22</td>
<td>18.56 ± 12.96</td>
<td>-1.48</td>
<td>152</td>
<td>0.140</td>
</tr>
<tr>
<td>SCARED Parent&lt;sup&gt;b&lt;/sup&gt;</td>
<td>11.98 ± 8.79</td>
<td>15.8 ± 12.59</td>
<td>-2.21</td>
<td>148.75</td>
<td>0.029</td>
</tr>
<tr>
<td>IQ</td>
<td>105.38 ± 13.25</td>
<td>93.09 ± 14.46</td>
<td>5.33</td>
<td>159</td>
<td>&lt;0.001</td>
</tr>
</tbody>
</table>

SDQ Strengths and Difficulties Questionnaire, MFQ Mood and Feelings Questionnaire, SCARED Screen for Child Anxiety Related Emotional Disorders, IQ Intelligence Quotient
<sup>a</sup> available for 68 participants
<sup>b</sup> available for 154 participants who were at least 8 years old
independently associated with the cognitive class. The odds of being in class 2 compared to Class 1 decreased with each one-year increase in age (OR = 0.72, 95% CI = 0.61–0.85). In addition, participants with parents with at least high-school education were 0.1 times as likely to belong to Class 2 as compared to participants whose parents had elementary education. Finally, the odds of belonging to Classe 2 were 1.1 times higher for higher MFQ-parents scores.

### Table 4
Multivariate logistic regression model among all participants (N = 140)

<table>
<thead>
<tr>
<th></th>
<th>p-value</th>
<th>OR</th>
<th>95% C.I. for OR</th>
<th>Lower</th>
<th>Upper</th>
</tr>
</thead>
<tbody>
<tr>
<td>Sex</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Male</td>
<td>0.32</td>
<td>1.00</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Female</td>
<td>0.001</td>
<td>0.74</td>
<td>0.63</td>
<td>0.86</td>
<td></td>
</tr>
<tr>
<td>Age Group</td>
<td>&lt;0.001</td>
<td>0.63</td>
<td>0.26</td>
<td>2.27</td>
<td></td>
</tr>
<tr>
<td>Income</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>up to $1000</td>
<td>0.06</td>
<td>0.76</td>
<td>0.26</td>
<td>1.52</td>
<td></td>
</tr>
<tr>
<td>$1000-$2000</td>
<td>0.06</td>
<td>0.26</td>
<td>0.06</td>
<td>1.05</td>
<td></td>
</tr>
<tr>
<td>more than $2000</td>
<td>0.06</td>
<td>0.26</td>
<td>0.06</td>
<td>1.05</td>
<td></td>
</tr>
<tr>
<td>Highest Parental Education</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Elementary</td>
<td>0.01</td>
<td>0.09</td>
<td>0.01</td>
<td>0.51</td>
<td></td>
</tr>
<tr>
<td>Highschool</td>
<td>0.12</td>
<td>0.21</td>
<td>0.03</td>
<td>1.51</td>
<td></td>
</tr>
<tr>
<td>University</td>
<td>0.10</td>
<td>1.21</td>
<td>0.96</td>
<td>1.51</td>
<td></td>
</tr>
<tr>
<td>SDQ emotional parent report</td>
<td>0.05</td>
<td>0.18</td>
<td>0.10</td>
<td>1.41</td>
<td></td>
</tr>
<tr>
<td>SDQ hyperactivity parent report</td>
<td>0.05</td>
<td>0.18</td>
<td>0.10</td>
<td>1.41</td>
<td></td>
</tr>
</tbody>
</table>

### Table 5
Logistic regression model among participants who were at least 8 years old (N = 134)

<table>
<thead>
<tr>
<th></th>
<th>p-value</th>
<th>OR</th>
<th>95% C.I. for OR</th>
<th>Lower</th>
<th>Upper</th>
</tr>
</thead>
<tbody>
<tr>
<td>Sex</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Male</td>
<td>0.26</td>
<td>0.61</td>
<td>0.26</td>
<td>1.42</td>
<td></td>
</tr>
<tr>
<td>Female</td>
<td>0.01</td>
<td>0.72</td>
<td>0.61</td>
<td>0.85</td>
<td></td>
</tr>
<tr>
<td>Age Group</td>
<td>&lt;0.01</td>
<td>0.43</td>
<td>0.22</td>
<td>0.93</td>
<td></td>
</tr>
<tr>
<td>Income</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>up to $1000</td>
<td>0.12</td>
<td>0.26</td>
<td>0.06</td>
<td>1.08</td>
<td></td>
</tr>
<tr>
<td>$1000-$2000</td>
<td>0.06</td>
<td>0.26</td>
<td>0.06</td>
<td>1.08</td>
<td></td>
</tr>
<tr>
<td>more than $2000</td>
<td>0.06</td>
<td>0.26</td>
<td>0.06</td>
<td>1.08</td>
<td></td>
</tr>
<tr>
<td>Highest Parental Education</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Elementary</td>
<td>0.09</td>
<td>0.19</td>
<td>0.03</td>
<td>1.29</td>
<td></td>
</tr>
<tr>
<td>Highschool</td>
<td>0.09</td>
<td>0.19</td>
<td>0.03</td>
<td>1.29</td>
<td></td>
</tr>
<tr>
<td>University</td>
<td>0.03</td>
<td>0.11</td>
<td>0.10</td>
<td>1.21</td>
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</tr>
</tbody>
</table>

### Discussion
This is the first study to investigate the cognitive profiles of a community-based sample of children and adolescents through latent class analysis and examine their association with emotional symptoms and sociodemographic factors.

Results showed that there are 2 different “cognitive classes” within a community sample where participants belonging to Class 1 showed significantly better performance on all tests. Specifically, their raw scores were higher on tasks measuring precision and correct answers and were lower on tasks based on a speed component. Since all of the tasks rely mainly on executive abilities, it could be argued that participants who belong to Class 1 have better prefrontal cortex functioning, or possibly a higher maturity of this brain area. Both the ventromedial prefrontal cortex (vmPFC) and the dorsolateral prefrontal cortex (dlPFC) are believed to allow better executive functions such as planning and abstract reasoning, as well as inhibition and self-control [22, 35]. This neurofunctional hypothesis seems to be supported by this study’s outcome.

Our results showed that the odds of belonging to Class 1 were higher for older participants while controlling for all other variables. This significant age effect falls in line with previous studies that showed a better cognitive performance in older children, especially on executive tasks such as working memory or different attention tasks [3, 14, 42, 45]. Such abilities are known to develop with age, which explains why older children are expected to outperform younger ones. Another expected result that supports executive functioning differences between groups was the predictive value of IQ in determining group distribution: children with higher IQ scores had more chances to belong to Class 1 even after controlling for age. In fact, higher IQ is believed to reflect higher overall cognitive functioning and more specifically better executive functioning. For example, Arffa [4] showed that gifted children outperformed other children on executive tests. Also, Der and Deary [13] examined the association between intelligence and reaction time. Their results indicated a significant positive correlation that seemed to increase with age. Finally, Smits-Engelsman and Hill [43] showed that poor motor abilities in children were associated with a lower IQ.

Our results also showed that children with low hyperactivity scores on the SDQ had higher chances of belonging to Class 1, when controlling for their age, IQ, or SES aspects. This result comes as no surprise, since hyperactivity, which is usually controlled by the frontal lobes, has been reported to affect several cognitive abilities, including executive ones. For example, Bolfer et al. [8] looked at reaction time in children...
with attention deficit/hyperactivity disorder (ADHD). Results showed that male participants between the ages of 9 and 12 with ADHD had slower reaction times compared to the control group. Norvilitis [36] explored executive functioning in ADHD and showed that children with ADHD-Combined exhibit impairments in cognitive functions such as in planning, sustained attention, and spatial working memory. Also, Mokobane et al. [34] reported that school-aged children with ADHD had poor fine motor abilities.

Higher family income and educational status are known to be predictors of better cognitive abilities [12, 15]. They are also usually expected to vary in the same direction, with higher family income typically found in communities with higher educational levels [10]. As expected, our results showed a significant effect for parental education in predicting cognitive abilities in our sample, and the odds of belonging to Class 2 was lower for children whose parents have a high school educational level. However, our results did not show that family income predicted cognitive profiles. This result is not compatible with previously published studies [12, 28] that showed a positive correlation between high SES and cognitive performance. Nevertheless, this discrepancy between our results and the literature could stem from two aspects: In previous studies, educational level and family income are referred to as SES, which makes it harder to determine the exact contribution of parental education and/or income in that concept. Moreover, and despite analyzing the effect of age in some studies, none of the previous studies controlled for additional variables, such as IQ, that can be affected by low SES [16, 49] and explain the variability of the cognitive outcome. The same argument could be made to explain the absence of a significant predictive value for gender in our study. In previous studies, cognitive differences were reported between males and females [51], and this effect seemed to be moderated by age. However, none of the previous studies pertaining to this aspect controlled for sociodemographic factors or IQ.

When controlling for all factors in the general sample, we found no significant effect for SDQ emotional symptoms in predicting participants’ belonging to the different “cognitive classes”. However, in the logistic regression model in which we only included participants who were at least 8 years old, higher parental reports of depressive symptoms as measured by the MFQ were significantly associated with belonging to Class 2 and hence, with having poorer performance on executive function tasks. This is in line with findings showing that depressive symptoms in community samples are associated with impaired performance on executive function tasks [1], a finding that seems to be age-specific since depressive symptoms have been shown to have a larger effect on neuropsychological functioning in early adolescence relative to later adolescence [1].

In clinical samples, depressive symptom severity and earlier age of onset of depressive disorders have also been associated with worse executive functions in youth [32]. In addition, executive functioning has been shown to predict prognosis in this population. Specifically, worse executive function predicted a lack of improvement in youth treated for depression [30] and later worse socio-occupational outcome among outpatient youth with depression in a cohort study [29].

Neural circuits that subserve executive functioning have emerged as important biological markers in depression in youth. Indeed, adolescents with depression showed reduced activation in the right dorsolateral and anterior cingulate gyrus when performing relevant tasks compared to healthy controls [21] and reduced connectivity between the prefrontal cortex and neural circuits that respond to positive and negative emotional information [18, 48].

The association between mood symptoms and executive functioning is complex and bi-directional. While we were able to show a correlation in our present cross-sectional study, we cannot confirm whether poor executive functions are a predictor or a consequence of long-lasting depressive symptoms on a vulnerable developing brain. Future studies should consider examining the association of neurocognitive deficits and mood symptoms in a developmental context and a longitudinal fashion [17].

Our findings should be considered in light of the following limitations: Our sample may not be representative of the population and our results may not be generalizable to other contexts knowing that not all participants who were recruited for the national survey opted to participate in the present study. This may have led to a self-selection bias. In addition, our data examined the presence of emotional and behavioral symptoms based on reports from participants and their parents and we did not examine clinical diagnoses per se. Hence, the association with the cognitive classes needs to be understood in that context although subclinical symptoms predict the presence of disorders [5]. Lastly, an important limitation is the fact that self-reports were not completed on the day of testing which wasn’t possible for logistics reasons. This time lag would not affect sociodemographic factors which typically do not change over a short period of time but may have impacted emotional symptoms which could have worsened or improved by the time of testing.

Conclusion

In conclusion, this study is the first to examine sociodemographic factors and emotional symptoms at play in neurocognitive performance in a community-based
sample. Our results suggest that different factors correlate with cognitive abilities and most of which are rarely considered in clinical settings. In fact, while most of the cognitive assessment tools typically control for age and/or sex, they rarely do so for other pertinent factors such as SES. This study also sheds the light on an existing gap in several clinical tools, and strongly advocates for an integrative approach when building norms for cognitive tests.

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Authors’ contributions
MB, WS and GP analyzed the data, MB, WS and FM interpreted the data and drafted the manuscript, all authors reviewed and revised the manuscript.

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Availability of data and materials
The datasets used and/or analyzed during the current study are available from the corresponding author upon reasonable request.

Declarations

Ethics approval and consent to participate
This study was approved by the American University of Beirut-Institutional Review Board.

Consent for publication
Not applicable.

Competing interests
Dr. Brent reports the following: Royalties from Guilford Press, UpToDate, and eRT; consultant, Healthwise, on Scientifics Board of Klingenstein Third Generation Foundation and the American Foundation for Suicide Prevention, Fund- eRT; consultant, Healthwise, on Scientifics Board of Klingenstein Third Genera-

References

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