Verbal aptitude hurts children's economic decision making accuracy

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Abstract

The impact of children's decision making increases with age and has relatively increased through time. Although a lot is known about cognitive development, less is known about how this development impacts decision accuracy in economic situations. This study builds on revealed preference theory to study the impact of cognitive aptitude on economic decision making accuracy and explores the intervening role of decision heuristics. In a study (n=100) where children from three age groups had to make choices between combinations of products, we found that decision accuracy was lower for kindergarteners than for children from the third and sixth grade, replicating and validating older findings. We found that one aspect of cognitive aptitude, namely verbal aptitude, hurts rather than helps decision accuracy. Further explorations suggested that this relation was due to the decreased use of the "more is better" heuristic, a child's preference for options with many units, which decreased with increasing verbal aptitude but increased rational decision making. We discuss the implications of the negative effect of verbal aptitude on economic decision making accuracy.

Keywords : Revealed preference, intelligence, accurate decision making, economic decision making, verbal aptitude, children's decision accuracy

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Children are important economic decision makers. Already at early ages, their impact on household consumption decisions seems substantial (e.g., Calvert, 2008; Cherchye, De Rock, and Vermeulen, 2012). Parents might buy more expensive items or engage in different activities than they usually would, simply because they want only the best for their babies and young children. Children might also actively influence household decisions by asking their parents for items, like candy during a shopping trip or activities during leisure time. Also as children grow older and increasingly start making their own decisions (while at the same time having larger budgets to spend and more decision autonomy), they become an even more desirable audience to target for marketers and important for policy makers to take into account (Dauphin, El Lahga, Fortin, and Lacroix, 2011). However, so far relatively little is known about decision processes in children, or about whether and to what extent the economic decisions that children make could be regarded as accurate. These are important questions to ask because only knowledge of how children make decisions can serve as a tool to help these very children make good decisions and become competent (adult) decision makers.

The present paper intends to shed more light on the question as to whether children are capable of making accurate economic decisions, and specifically, which factors influence these decisions. We empirically study decisions of children of various ages (i.e., 5, 8, and 12-year old) to gain insight in how (accurate) decision-making unfolds in children. In addition, we consider individual differences in various components of cognitive aptitude (i.e., verbal aptitude, mathematical aptitude, creative aptitude) to gauge the importance of each of these components for accurate economic decision-making in children (within age groups). We also explore the role of decision heuristics in the possible relation between cognitive aptitudes and decision accuracy (e.g., Morsanyi and Handley, 2008). We rely on an objective measure of decision accuracy based on economic theory by calculating the percentage of budget children waste due to making suboptimal decisions.

ACCURATE DECISION-MAKING OF CHILDREN OF DIFFERENT AGES

Age-related changes in cognitive abilities are closely linked to how economic knowledge and decision-making accuracy unfolds in children (Roedder John, 1999). Piaget was one of the first to develop a framework capturing shifts in cognitive abilities, and proposed a distinction between a sensorimotor (birth to two years), a preoperational (two to seven years), a concrete

operational (seven to eleven years), and a formal operational (eleven through adulthood) stage of cognitive development (Ginsburg and Oppenheimer, 1988). Though preoperational children develop symbolic thought, they will not easily move away from immediate perceptual properties of stimuli still. This approach to the world disappears in concrete operational children, who do not accept perception as a reality but think about whatever they encounter in a more thoughtful and abstract way. Unlike preoperational children, concrete operational children can already consider multiple dimensions of a stimulus at once. In the formal operational stage of cognitive functioning, children adopt even more complex thoughts about concrete as well as hypothetical stimuli and situations, and move towards adult thought patterns (Ginsburg and Oppenheimer, 1988).

This framework fits nicely with the categorization proposed by Roedder (1981) that is particularly relevant when focusing on economic decision-making. Specifically, decision makers younger than seven (preoperational) have been described as ‘limited processors’, who often have difficulties to store and retrieve information even when they are prompted to do so. They can be expected to make consumption decisions based on one single salient attribute such as color. Decision makers older than seven but younger than eleven (concrete operational) have been labeled ‘cued processors’, suggesting that they are able to store and retrieve information, but need explicit cues before being able to do so. They can be expected to be relatively thoughtful in their decisions, and consider more than one salient attribute (price may become relevant too), while also employing a decision strategy that fits the task environment (sweetness is desirable for candy but not for soup). Finally, decision makers over twelve (formal operational) or so-called ‘strategic processors’ use various strategies for storing and retrieving information, and are quite successful at it. Their decisions are made in an adaptive manner, and depend on the decision situation (Roedder John, 1999).

There is empirical support for the idea that younger children rely on fewer product attributes or dimensions when comparing products and making decisions (Bahn, 1986; Capon and Kuhn, 1980; Jansen, Van Duijvenvoorde, and Huizinga, 2012). In one study for instance, children of various ages (i.e. kindergartners, fourth graders, eighth graders, college students) were shown notebooks varying on four dimensions (e.g., shape), and rated their liking for each notebook. Afterwards, they also indicated their liking for each dimension. Whereas kindergartners did not even succeed in incorporating their preferences for one dimension in the overall ratings, older

children could do this, and late adolescents integrated two or even more dimensions (Capon and Kuhn, 1980).

There is also research hinting at the fact that the ability to adapt strategies to demands of specific decision environments in a flexible way increases with age. Specifically, with age, decision-makers restrict their search to smaller and more promising subsets of available information, and switch from highly demanding to less demanding decision strategies as the complexity (number of available alternatives and attributes) of the decision environment increases (Gregan-Paxton and John, 1997; Payne, Bettman, and Johnson, 1993). One important factor in this developmental process is the growing sensitivity to decision making costs like search costs (Gregan-Paxton and John, 1997). More recent demonstrations showed that the sensitivity to probability in preference formation increases from virtually zero in pre-schoolers to solid in adults, and that there are intermediate stages among primary school children (Betsch, Lehmann; Jekel, Lindow, and Gloeckner, 2018).

Harbaugh, Krause, and Berry (2001) used a revealed preference methodology to investigate emerging rationality of children. Their experiment consisted of different tests of transitivity as one of the basic requirements for rationality, and provided a measure of the size of rationality violations. Specifically, children of various ages were asked to select one of several product combinations consisting of varying quantities of chips and fruit juice, and they made such choices multiple times. The authors observed that second graders displayed more transitivity violations than sixth graders and undergraduates did, but they found no difference between the latter two age groups. A comparable experimental procedure had been used earlier by Sippel (1997), who studied similar transitivity violations in college students by giving them choices between different consumption goods using different budget sets.

Taken together, results of previous studies looking at various indicators of decision-making accuracy of children suggest that decision-making accuracy increases with age. In what follows, we explore the role of within age cognitive aptitude variation in decision-making accuracy of children.

COGNITIVE APTITUDE AND ACCURATE DECISION-MAKING

It can be expected that next to age, other individual difference factors will also have an influence on decision making accuracy of children. One study hinted at the idea that reliance on

long-term and working memory, and possibly also inhibitory control processes influence advantageous decision making of children (Van Duijvenvoorde, Jansen, Bredman, and Huizenga, 2012). In line with these findings, Weller and colleagues (2012) not only observed structural similarities in how preadolescent children (10- to 11-year olds), late adolescents (18- to 19-year olds) and adults make decisions, but they also found that individual differences in effortful control were associated with decision-making accuracy across all ages. Effortful control has long been associated with many executive functions originating in the prefrontal cortex of the human brain, and has been shown to be related to intellectual achievement in childhood (Blair and Razza, 2007) and adolescence (Boisvert, Stadler, Vaske, Wright, and Nelson, 2013).

Also creativity or the ability to generate new ideas or come up with multiple solutions to one problem has been proposed as an aspect of cognitive aptitude that is associated with decision making (Guilford, 1982; Sligh, Conners, and Roskos-Ewoldson, 2005). Silvia (2008) provided evidence for this link in a sample of fifth-graders. Levin, Bossard, Gaeth, and Yan (2014) focused on various individual differences besides age differences in understanding decision processes of children. They found that measures such as numeracy help explain age-related and individual differences in decision making accuracy, in particular for decisions involving risk. Also verbal skills have been shown to play a role in children's decision making and problem solving abilities (Kytta, Aunio, Lepola, and Hautamaki, 2014). Specifically, young children's (ages 4 – 7) ability to solve problems was positively influenced by the extent of their vocabulary (cf. also Bjork and Bowyer-Crane, 2013). Taken together, these findings suggest that besides age, also cognitive aptitude plays an important role in children's decision making accuracy. In the present paper, we will take on the systematic study of the influence of both age and cognitive aptitude variations within age on accurate economic decision making of children.

ASSESSING DECISION-MAKING ACCURACY THROUGH REVEALED PREFERENCES

In economics, decision accuracy is operationalized as the maximization of utility given a certain budget. Utility maximization implies preference consistency across choices (Choi, Fisman, Gale, and Kariv, 2007). We will rely on so-called revealed preference tests to study children's decision-making accuracy (with revealed preferences referring to the idea that choice behavior reveals underlying preferences). Following revealed preference theory (Samuelson, 1938), a chosen set of consumer goods x_i is revealed preferred over some other set x_j at the prevailing price

regime p_i if and only if x_i was chosen over x_t when both options were available. In this simple set-up this is equivalent to x_i being more expensive than x_t when using the price regime p_i (since then both bundles (i.e. the chosen set of consumer goods) could indeed have been bought in situation i). If the individual (as a utility maximizer) always chooses the best bundle he/she can get, then, if x_i is revealed preferred to x_t , he/she must never choose x_t when x_i is also available (i.e. x_i is not strictly within the budget set when x_t is chosen). This requirement is called the Weak Axiom of Revealed Preference (WARP).

Varian (1982) formulated the General Axiom of Revealed Preferences (GARP), which makes use of transitivity of preferences. A chosen bundle of goods x_i is “indirectly revealed preferred” over some other bundle x_t , if and only if there exists a sequence of bundles $x_j, x_k, \dots, x_s$ such that x_i is directly preferred over x_j , x_j is directly preferred over x_k , $\dots$, and x_s is directly preferred over x_t . Similar to before, according to the GARP, if a bundle x_i is indirectly revealed preferred to x_t , then x_t should never be chosen when x_i is also available (i.e. x_i is not strictly within the budget set when x_t is chosen). Varian proved that GARP is a sufficient and necessary condition for decision-makers' choices to be consistent with utility maximization.

Let us consider the following example to understand the logic behind GARP and illustrate a violation. Suppose that in the first situation the prices and budget allow the child to choose between plate A with two grapes and one cookie and a plate B with one grape and two cookies. Assume that the child chooses plate A, while the prices indicate that grapes are more expensive than cookies. In the next choice situation the cookies become more expensive than grapes and the budget remains the same, but the child can still choose between the same two plates. Assume it chooses now plate B. This is a violation of WARP, since in both choice situations the child could have chosen a bundle that is strictly cheaper and that it prefers more. In our example the child thus failed to maximize the utility. We consider such a violation as a lack of **decision accuracy**.

EXPLORING COMPONENTS OF DECISION MAKING (IN)ACCURACY

Our choice data offers a unique opportunity to increase our insights in the mechanisms behind decision (in)accuracy. Based on the extensive choice data (9 choices per child, see below for more discussion), we set out to identify and quantify a number of decision heuristics that children may be using while making their choices, which may enhance or hinder rational decision making as defined in revealed preference theory. We distinguished five relevant heuristics from the literature

that were quantifiable based on the choice data. Note that these heuristics are not fostering accuracy or inaccuracy per se but present particular ways to make the decision. Note also that we do not propose specific a priori hypotheses as to the possible mediational role of these heuristics between cognitive aptitude and decision accuracy. This part of the study is meant to be only explorative. First, decision makers may simplify their choice by using the *compromise rule*: When in doubt about which option to choose, decision makers may prefer the choice set that offers all the available options (Novemsky, Dhar, Schwarz, and Simonson, 2007). A second simplifying heuristic is based on the focusing effect, where decision makers pick one salient feature of a product to make their decision (Legrenzi, Girotto, and Johnson-Laird, 1993). In our context, this is arguable the number of items each choice option offers (i.e. *More is better*). Another heuristic that children may use is the balancing rule. When choosing multiple products, decision makers may solve difficult trade-offs by balancing between important dimensions (Dhar and Simonson, 1999). This can occur within choice situations (e.g. choosing a juice and a dry snack rather than two of your favorite snacks to balance the consumption experience, *balancing within choices*, which is the third heuristic in our setting) or across choices (e.g. “I just picked the plate with biscuits in round 3, so now I choose the plate with mandarins”, *balancing across choices*, the fourth heuristic in our setting). Notice that *balancing within choices* trades off between product dimensions (e.g. dry vs juicy, healthy vs unhealthy) whereas the *compromise* rule implies a selection of the middle option, which is the plate with all products in our setting. A final heuristic makes the decision maker focus on the products (what) rather than on their characteristics (e.g. color or amount) and therefore pick the option with their preferred products. This heuristic has been identified as the “*take the best*” rule (Gigerenzer, Hoffrage, and Kleinbölting, 1991) or attitude-based (as opposed to attribute-based) decision making (Payne, Bettman, and Johnson, 1988). Note that this heuristic distinguishes itself from the *More is better* heuristic because the focus of ‘*Take the best*’ is on the product whereas that of *more is better* is on the amount.

THE PRESENT STUDY

Research approach

Selection of age categories. We created an incentive-compatible decision situation and rely on revealed preference tests to study accurate decision making of children aged five, eight, and twelve. These ages coincide with the average ages that children enter the preoperational, concrete

operational, and formal operational stages of decision-making as proposed by Roedder (1981), respectively.

Measurement of cognitive aptitude. We obtained information on cognitive aptitude. The literature on cognitive aptitude is not particularly coherently built around one theory but sampling from a diverse set of sources (Grigorenko, Jarvin and Sternberg, 2002, testing the triarchic theory of human intelligence; Johnson and Bouchard, 2005; comparing three other theories given a large set of tests), we selected mathematical, verbal, and creative ability as a justifiable set of aptitudes. These are three indices (1) that come back in a variety of models, (2) that have been linked to decision accuracy according to our literature review above, and (3) that can conveniently be measured by means of teacher ratings. Teachers' assessments of cognitive aptitude have proven to be a convenient yet reliable source of information on children's cognitive characteristics (Hoge and Coladarci, 1989; Lonqvist, Vainikainen, and Verkasalo, 2012).

Specifically, we asked teachers to rate creative aptitude, verbal aptitude, and mathematical aptitude for each child. Based on the findings by Begeny, Eckert, Montarello and Storie (2008), we measured both a relative and an absolute measurement to increase reliability. The relative item asked to position each child in terms of each of the three relevant aptitudes in comparison to his or her peers the teacher had taught by means of an 8-category rating scale (with "excellent" = top 2%; "very good" = top 10%; "good" = top 25%; "average" = top 50%; "less than average" = bottom 50%; "bad" = bottom 25%; "very bad" = bottom 10%; "terrible" = bottom 2 %). The absolute item asked to rate each of the three relevant skills of each child in comparison to his or her average peers, and ranged from 1 (very weak) to 10 (very strong). The Spearman intra-class correlations between the two indices were $r_{\text{verbal}} = .909$; $r_{\text{mathematical}} = .924$, and $r_{\text{creative}} = .899$. The high internal consistency for the three components of cognitive aptitude allowed us to construct a composite index for the three components by first transforming scores on the first item (so that these scores also ranged from 1 to 10), and then averaging the transformed scores on the first item with the original scores on the second item.

Designing the decision situation. We use two instantiations of decision accuracy based on the revealed preference logic, GARP and Afriat's index, to assess decision accuracy of children varying in age (Van Bruggen and Heufer, 2017). The GARP index is the dichotomous index indicating whether or not the series of choices that an individual makes are consistent with GARP. Because this index reflects whether or not each decision maker passes the rationality test, we will

call this the **pass rate** in the remainder. In addition, we will use Afriat's index which is a quantitative measure of decision accuracy that varies between 0 (complete inaccuracy) and 1 (complete accuracy, no GARP violations, see Afriat (1973) and Varian (1990) for precise formal definitions).

A convenient feature of the Afriat index is that it can be interpreted as a percentage of budget wasted. Formally, to compute the Afriat index for observation t , we need to find all bundles of goods that are indirectly revealed preferred to the bundle x_t . One of these bundles is the bundle x_t itself. For each of these bundles we compute the cost at prices p_t and we compute the minimum of all these costs. The Afriat index is then the ratio of this minimum and the available budget at observation t . Since the minimum is at most equal to the available budget, the index is bounded between 0 and 1 by default, with 1 indicating that no money is wasted since there are no preferred bundles at a cheaper cost available. The lower the index, the more budget is wasted. In the analyses we will use Afriat's index (from 0-1) and interpret it as budget waste (which corresponds to the reverse variable).

To calculate the index we rely on a choice task in which children have to choose between seven different combinations of three products on nine different occasions. The prices of the product combinations are kept constant within each choice occasion, but vary across choice occasions. This procedure is adapted from a more complex purchase procedure (see also Harbaugh, et al., 2001) which allows participants to select product combinations from a continuous budget set defined by given prices and budgets (i.e., participants get to select their own combinations of products). Yet our procedure has the advantage that it is easier for young children than the classic procedure where decision makers have to allocate their budget across products, because it does not require calculations in spending the available budget. However, at the same time this procedure still allows us to assess the loss of decision accuracy in a realistic choice setting with budget restrictions and different price regimes.

Specifically, children were presented with the nine choice sets. In each set, they could choose one of seven plates. Each plate displayed a product combination consisting of a given quantity of grapes (units of 10 grams, coinciding with one grape), mandarins (units of 12.5 grams, coinciding with one part) and letter biscuits (units of 5 grams, coinciding with one biscuit). To make the choice easy for children while keeping its diagnosticity for our measure of decision accuracy high, we included the three 'extreme' combinations with all budget spent on a single product, one plate

with a combination of all three options, and three plates with only two of the three products. This constellation also helped children to have an immediate overview of the choice options provided.

Figure 1

An illustration of the experimental set-up

Legend. Figure 1 displays two examples of the nine choice sets as illustrations of the experimental set up: the second choice set and the fifth choice set (left and right panel, resp.). As we asked the children to choose one plate per set, the price of these 7 plates can be considered identical. Table A2 shows the proportions of the products given the price regimes that are displayed in Table A1. Note that the quantities in the setup have been multiplied by a factor 4 (across all 63 plates) to yield substantial and integer numbers. Figure 1 illustrates that (consistent with Tables A1 and B1) the price of grapes is higher than that of letter biscuits in choice set 2 because plate 1 has fewer grapes than plate 3 has letter biscuits. In choice set 5, this situation is reversed. A selection of plate 1 in choice set would reveal a strong preference for grapes. If the same child would not choose plate 3 in choice set 5, it would display limited decision accuracy (lots of budget waste). Our set of nine choices allowed us to assess decision accuracy in a fine-grained way (see research approach section – decision accuracy, for more details).

The prices were not made explicit but served to construct the combinations in the choice sets. Because the prices of the seven plates were kept constant within each choice set, choosing one's favorite boils down to optimizing the utility the decision maker gets, given the available budget. As discussed in detail in Cosaert and Demuynck (2015), the given selection of prices resulted in a powerful experimental design. More precisely, for randomly generated data the pass rate of GARP is only 0.13, which will turn out to be significantly smaller than the pass rate of the children in our study. Appendix A presents the implicit prices that correspond with the plates and the associated choice sets. Figure 1 shows how the choice situation looked like for the participants.

Method

Participants. Participants were 100 children (39 kindergartners, aged 5 (37) or 6 (2); 31 third graders; aged 8 (27) or 9 (4), and 30 sixth graders, aged 11 (5) or 12 (25)) attending four different schools in Belgium, with a mean age of approximately 8 years. The initial number of participants was 108 but the experiment failed for eight children (interrupted or failed compliance). Each child received written permission both from its parents and the school prior to participation. The selected schools are situated in a relatively rural area with a predominantly Caucasian population (98% in our sample). It is important to note that primary education is free and heterogeneous in terms of economic background in Belgium. Note that Cosaert and Demuynck (2015) used this data set as an illustration of discrete choice sets (ignoring the cognitive aptitude measurements) and that Bruyneel, Cherchye, Cosaert, De Rock, and Dewitte (2017) used data from the same children who subsequently made similar but collective decisions in pairs.

Procedure. Children were welcomed in a separate room in their schools one at a time. They were informed that they would be asked to make nine sequential product choices, and that one of these nine choices would be randomly selected and given to them for real at the end of the study. In order to familiarize them with the products they could choose from (i.e., grapes, mandarins, and letter biscuits), we gave them the chance to taste these products before they had to make their actual choices. It was made clear to them that these were the actual products they would be making choices about later.

Subsequently, children were presented with the first of nine choice sets. In each set, they could choose one of seven plates as described in the *research approach* section. After the children had chosen a plate, the experimenter showed the next choice set (i.e., again consisting of

seven plates, each showing a given quantity of grapes, mandarins, and letter biscuits), and stressed that the next choice was as important as the previous one, and that both choices were independent of each other. This process was repeated nine times. An important advantage of our procedure is that children in our study did not face explicit prices and budgets when selecting consumption bundles, yet their choices could be interpreted as defined under implicit prices and budgets. This procedure was similar to the one of Harbaugh et al. (2001), who argued that selecting consumption bundles from a continuous budget set (defined by given prices and budget) is too complicated for young children as it requires abstract mathematical reasoning.

At the end of the study, the children drew one of nine cards that displayed a number ranging from one to nine corresponding to the choices they made in the respective choice sets. They then received their respective consumption combinations.

For explorative reasons, we obtained age (in years) and gender of all children included in the study. We also collected information on the number of older siblings in each child's family. We needed this information to control for possible influences of exposure to older siblings on children's decision making capabilities (e.g., Roedder John, 1999; Zajonc and Mullally, 1997).

Results

Descriptives. Children chose substantial amounts of all products and displayed heterogeneity in their choices (Table 1), which confirms that the products were sufficiently attractive to the sample and offers a useful basis to assess choice accuracy. Appendix B shows the full distribution of choices across grades and products. Table 1 provides an overview of budget share information (across all choice situations) and pass rate on the rationality test for all children, as well as for each of the three age groups separately ($n = 100$). As we did not observe any effects of gender and number of older siblings on the other relevant variables that we analyzed, we collapsed across gender and number of siblings for the remainder of the analyses. We report the full model with these two variables included in the model in Appendix C (Table C1).

Decision Accuracy. Note that in our discrete setting we only offer a subset of product combinations which does not always include the most optimal combination (because the children were not allowed to make their own combination of goods). As discussed in Cosaert and Demuyne (2015) this does not interfere with our pass rate, but it does imply that our calculations

of Afriat's index are only a lower bound. The pass rate does not have this weakness but is a binary variable and thus less sensitive.

Table 1

Average budget shares (and standard deviations) of the three product categories, and pass rate on the rationality test, by grade

	Grapes	Mandarins	Letter biscuits	Pass rate
Kindergarten	.273 ^a (.204)	.174 (.180)	.553 (.284)	0.31
Third grade	.312 (.117)	.235 (.173)	.452 (.208)	0.48
Sixth grade	.378 (.159)	.347 (.179)	.275 (.178)	0.53
Total	.317 (.172)	.245 (.190)	.438 (.258)	0.43

^a The figures represent the share of the budget (summing to 1 in rows) that individuals spent on the three products, broken down per age category. We took the average within each specific age category.

Table 2

Budget waste (Afriat's index) for the three grades

	N	Afriat's index ^a	Stdev	Min	Max
Kindergarten	39	.604	.382	.111	1
Third grade	31	.737	.321	.11	1
Sixth grade	30	.747	.362	.111	1
All children	100	.688	.361	.11	1

^a Lower values represent higher budget waste, with 1 representing perfect consistency

Table 2 ($n = 100$) provides an overview of the Afriat indices capturing budget wasted in suboptimal choices for our sample including all children, as well as for each subsample including each of the three age groups separately. To compute these indices, we rely on the (implicit) prices and budgets underlying the construction of our experimental choice sets. To recall, lower values indicate that a larger part of the budget is wasted due to irrational choices. The means of the Afriat indices indicate that the youngest children wasted on average 39.6 % of their budget due to suboptimal choices, whereas this percentage decreased to 26.3% in the third grade and 25.3% in the sixth grade. A two-sample Wilcoxon rank sum test revealed that the difference between kindergarteners and the other two groups was significant ($z = 2.005$, $p = 0.045$). The same test comparing kindergarteners to the two other groups individually went in the same direction in both comparisons but did not reach conventional levels of significance (difference between kindergarteners and third-graders: $z = 1.575$, $p = 0.110$; and between kindergarteners and sixth-graders: $z = 1.814$, $p = 0.070$). The difference in budget waste between third- and sixth-graders was negligible and not significant ($z = 0.668$). These results indicate that our participants wasted a significant proportion of their budgets selecting suboptimal options, and kindergartners did so to a larger extent than third and sixth graders did. Figure 2 shows the distribution of the Afriat index and shows that the age effect reflects the migration from the very low scores to scores close to optimal decision making (Afriat index of one).

Test of the hypotheses about the association between cognitive aptitude and decision accuracy. Next, we analyzed the association between the individual differences in the three components of cognitive aptitude (i.e., mathematical, verbal, creative aptitude) and budget waste in a series of regression analyses. Because the teacher failed to assess cognitive aptitude for one child, one observation was not included in the present analysis (and all subsequent analyses involving cognitive aptitude). To assess divergent validity, we first analyzed the correlations between the three components. Table 3 ($n = 99$) shows that the correlation between verbal and mathematical aptitude is high and that the correlation between these two on the one hand and creative aptitude on the other hand, are moderate. The low correlations with age show that the teachers succeeded in assessing the aptitude while ignoring age effects. Table 3 also shows that there are no reliable relations between cognitive aptitude and decision accuracy (as calculated with spearman correlations, because decision accuracy was not normally distributed).

Figure 2*The distribution of the Afriat index*

Legend. The upper panel shows the distribution (in percentages) of the Afriat index in the whole sample. The lower panel shows the distribution broken down per age group.

Table 3*Correlations between the three components of cognitive aptitude, age, and decision accuracy.*

	Age	Verbal aptitude	Mathemati cal aptitude	Creative aptitude	Afriat's index
Age	1.000				
Verbal aptitude	-0.040	1.000			
Mathematical aptitude	0.076	0.771**	1.000		
Creative aptitude	0.065	0.481**	0.489**	1.000	
Afriat's index ^a	0.19*	-0.15	0.00	0.08	1.000
Pass/fail ^a	0.17*	-0.11	0.03	0.14	0.81**

^a This row displays Pearson correlations because of the bimodality of this variable. The rest of the table displays Spearman correlations.

* $p < .10$. ** $p < .01$.

We then explored the role of the three components of cognitive aptitude on decision accuracy (budget efficiency) in a series of regressions. Because the distribution of the Afriat indices was far from normal (Shapiro-Wilk $W = 0.74$, $p < .0001$, see also Figure 2), we used the fractional logit model, which uses the empirical distribution of the findings (Papke and Wooldridge, 1996, proc GLIMMIX empirical in SAS®)¹. The parameter estimates of the parallel OLM² models that we conducted were close to identical for all conducted analyses. However, the GLIMMIX model led to more pronounced significance tests than OLM as the standard errors were lower in the former model. We also assessed the multi-collinearity (with OLM) because verbal and mathematical aptitude were highly correlated. In none of the models tested (see Appendix C, Table C2) did the variation inflation index exceed 2.68, which can be considered as acceptable, and suggests that multi-collinearity does not play a role in our main findings. In the first pair of regressions, we regressed the Afriat index on the child's age and the three components of cognitive aptitude. In model 1, the child's grade was included as a set of two dummies: third grade (yes or no) and sixth

¹ GLIMMIX, or "Generalized Linear MIXed Model", is a regression model where the empirical distribution of the criterion is used rather than a normal approximation. This is particularly appropriate in cases where the distribution strongly deviates from the normal distribution, as in our case (see Fig. 1).

² OLM or Ordered Logistic Regression is a Logistic Regression where the criterion/dependent variable is ordered.

grade (yes or no). In model 2, we substitute the age (which showed some variation within grade) for the grade variable to optimize the used information. In the second group of regressions, we regressed the binary pass/fail rate of the revealed preference test in an analogous set of two regressions using Probit models.

The results suggest that decision accuracy is negatively related to verbal aptitude and that there is an indication that it is positively related to creative aptitude. The age effect that the non-parametric analysis had revealed when third and sixth graders were compared to kindergarteners, was also apparent here, though not significantly so. We also note that we tested the interactions between age and the three components of cognitive aptitude, but that none of these approached significance, possibly due to a lack of power. The Appendix shows the regression results.

Exploring if decision heuristics mediate the relation between age and cognitive aptitude on the one hand and decision accuracy on the other hand. We finally analyzed the role that decision heuristics discussed above played in the relations we observed. We first quantified the heuristics based on the choice data per child, then assessed (1) their sensitivity to age and cognitive aptitude, (2) their correlation with decision accuracy, and (3) if they were logical candidates for mediation, the extent to which they mediated the relation between age and cognitive aptitude on the one hand and decision accuracy on the other hand, thereby following the process logic as outlined by Hayes, 2018)

The *compromise rule* was quantified as whether or not all 3 options are in a choice. Children who chose all three products in the nine choice situations obtained the maximum score of 9 on this heuristic. The *More is better* rule was quantified as the number of times the child selected the option with the largest number of food items she could get, irrespective of what this food item was. A child that chose the plate with the largest number of items for nine times obtained the maximum score of 9 for this heuristic. *Balancing within choice* was quantified as the number of times a child selected an option with letter biscuits as well as exactly one of the fruits in every choice. Again, a child that chose such a combination for the nine choice situations obtained the maximum score of 9. *Balancing across choices* was quantified as the number of times a child shifted from an option where biscuits were the most numerous product to one where either mandarins or grapes were the most numerous option. Because the child could shift only 8 times between the nine choice situations, the maximum score for each child was 8 for this heuristic. Note that the balancing

dimension could be related to healthy/unhealthy but also to juicy/dry, but this is not relevant for our purposes. The *take the best* heuristic (Gigerenzer, Hoffrage, and Kleinbölting, 1991) was quantified as the number of products that was never chosen across the nine choice situations for each individual. This variable range from 0-2 per child. For instance, a child that always took the mandarins irrespective of the price obtained the maximum score of 2 on this heuristic. A child that in contrast was very sensitive to the prices and varied products, obtained the minimum score of 0. Note that in our dataset this number is confounded with strong intra-individual differences in preferences. The *Take the best* heuristic was hard to define on the level of the choice because identifying what is the best builds on the child's other choices. To avoid interdependencies, we decided to quantify this heuristic on the product level.

Table 4

Descriptive statistics for the five heuristics and their correlation with the two indices of decision accuracy

Heuristic ^a	<i>M (Sd)</i>	<i>Range</i>	<i>Afriat</i>	<i>Pass</i>	<i>compro mise</i>	<i>More is better</i>	<i>Balancing within choice</i>	<i>Balancing across choice</i>
<i>Take the best</i>	0.31 (0.58)	0 (>50%)-2	0.09	0.09	-.34**	.13	-.19	-.21*
<i>Compromise</i>	1.36 (2.16)	0 (>25%)-9	0.22*	0.30**	1.0	-.45**	-.22*	-.18
<i>More is Better</i>	2.07 (0.84)	0(>10%)-8	0.24*	0.23*		1.0	-.28**	-.09
<i>Balancing within choices</i>	1.86 (0.88)	0(>25%)-8	-0.12	-0.21			1.0	.26**
<i>Balancing across choices</i>	2.66 (1.84)	0(>10%)-7	-0.67**	-0.46**				1.0

^a *n* = 100

* *p* < .05. ***p* < .01

Table 4 shows the descriptive statistics for these heuristics (number of observations, average across the children, and the range), their inter-correlations and their correlations to the two decision accuracy indices with child as the unit of observation. Table 5 shows how they are associated with the predictors in our core model. Overall, we found that the heuristics were sufficiently distinct

from each other and from the decision accuracy indices. We also found that the *compromise* and the *more is better* heuristics were positively associated to decision accuracy, whereas *balancing across choices* was negatively associated to decision accuracy. (The latter is not surprising because shifting across choices is the hallmark of choice inconsistency). As to the predictors, we found that the *Take the best* heuristic reduced with age, whereas the use of the *compromise* rule increased with age. Note that the possible range for the *Take the best* heuristic is lower than for the other heuristics because it can only be defined across choices. Of the three cognitive aptitude indices, language was the only one that was related to heuristic use. Specifically, verbal aptitude was positively correlated with the use of *compromise* and *balancing across choices*, but negatively to *more is better*, attesting to its important role in decision making.

These initial analyses revealed two candidate statistical mediators underlying the negative relationship between verbal aptitude and decision accuracy: the use of *more is better* and of *balancing across choices*. Specifically, *more is better* was negatively related to verbal aptitude and positively to decision accuracy, and *balancing across choices* was positively related to verbal aptitude and negatively to *decision accuracy*. As the use of the *compromise* heuristic was *positively* related to both verbal aptitude and decision accuracy, it could not mediate the *negative* relation between verbal aptitude and decision accuracy. However, as the *compromise* heuristic was positively related to both age and decision accuracy, it is a candidate mediator for the relation between age and decision accuracy.

To test mediation, we start from our basis model described above and include the use of either *more is better* and *balancing across choices* as predictors in the model (one by one). Using the process macro for SAS® (Hayes, 2018), we found that including *more is better* reduced the strength of the negative relationship between verbal aptitude and decision accuracy to insignificance. The bootstraps testing procedure of the Process macro (n=10000) revealed that *more is better* marginally ($p < .10$) mediated the negative relation between verbal aptitude and decision accuracy. Although the association between verbal aptitude and decision accuracy also shrunk when *balancing across choices* was included in the model, the bootstrap test showed no significance ($p < .15$). So we conclude that verbal aptitude reduced the use of *more is better*, which in its turn reduced decision accuracy.

Table 5*Regressions of the five heuristics on the predictors of the model and gender*

	<i>Take the best</i>	<i>Compromise</i>	<i>More is better</i>	<i>Balancing within choices</i>	<i>Balancing across choices</i>
<i>predictor</i>					
<i>Third grade</i>	-0.24* ^b (0.14)	0.81 (0.50)	0.28 (0.41)	0.38 (0.39)	0.45 (0.44)
<i>Sixth grade</i>	-0.38** (0.14)	1.70*** (0.51)	-0.53 (0.41)	-0.18 (0.40)	-0.73 (0.45)
<i>Gender</i> ^a	0.13 (0.12)	-0.20 (0.43)	0.48 (0.35)	-0.02 (0.33)	-0.02 (0.38)
<i>Verbal</i>	-0.05 (0.06)	0.48** (0.21)	-0.53** (0.17)	-0.02 (0.16)	0.31* (0.19)
<i>Mathematics</i>	0.05 (0.06)	-0.27 (0.20)	0.22 (0.16)	0.11 (0.16)	-0.29 (0.18)
<i>Creative</i>	-0.01 (0.05)	0.11 (0.17)	0.10 (0.14)	-0.11 (0.13)	-0.02 (0.15)

^a girls= 0; boys = 1^b Beta's and standard errors between brackets. Significant betas in bold.* $p < 0.10$. ** $p < 0.05$. *** $p < 0.01$

We tested the use of the *compromise* heuristic as a possible statistical mediator between age and decision accuracy following the same logic. Table 5 shows that adding the *compromise* heuristic to the model reduced the relation between age and decision accuracy, and that the use of the *compromise* heuristic indeed marginally significantly mediated the link between age and decision accuracy ($p < .10$). Older children use the *compromise* option more frequently and this in its turn increased decision accuracy. We report on the details of these mediation analyses in the Appendix.

GENERAL DISCUSSION

The present paper set out to investigate to what extent children of various ages (i.e., 5, 8, and 12-year old) are capable of making accurate economic decisions, and which components of cognitive aptitude (i.e., verbal aptitude, mathematical aptitude, creative aptitude) influence these decisions.

We designed an empirical study largely based on the procedure used by Harbaugh et al. (2001) requiring children to select bundles of products from choice sets consisting of seven combinations of up to three products. We relied on two direct measures of decision accuracy: the Afriat index, which quantifies the percentage of budget wasted in suboptimal decisions, and whether or not the child passed the economic rationality test. We explored the role of decision heuristics as potential mediating factor between cognitive aptitude and decision accuracy.

Summary of findings

This paper reports three findings: A (replicated) age effect, a (novel) correlation between cognitive aptitude and accurate economic decision making, and a (explorative) mediational role of decision heuristics.

Age related difference in decision accuracy. We found that 5-year olds wasted a considerable proportion of their budgets (i.e., approximately 40%) in suboptimal choices, whereas this number decreased for 8-year olds and 12 year-olds (i.e., to approximately 25%). We did not find any differences between the latter two age groups. This finding largely resonates with earlier findings (Harbaugh, Krause and Berry, 2001) and consumer developmental theories (Roedder John, 1999), implying that decision quality improves with age. At the same time this finding validates the specificities of our procedure.

The role of within age cognitive aptitude variation in decision accuracy. The second major finding pertains to the role of cognitive aptitude variability (within age groups) in accurate economic decision making. Regarding mathematical aptitude, we found no relationship with decision accuracy. This lack of relation is consistent with Harbaugh, Krause and Berry (2001), the study which is methodologically closest to ours, who also found none. Going beyond Harbaugh, Krause and Berry's (2001), we also looked at influences of verbal and creative aptitude on age-related differences in decision making. We showed that creative aptitude, the capacity to combine techniques to solve new problems, was (weakly) positively related to economic decision accuracy. We found this evidence only when we analyzed the pass/fail Garp index but not when we used the continuous Afriat index. So this finding remains to be verified. Although we are not aware of prior research linking creative aptitude to (economic) decision quality, this finding is in line with the general finding that cognitive aptitude benefits economic decision accuracy (e.g. Ball, Mann and Stamm, 1994).

More remarkable was our finding that verbal aptitude was *negatively* related to economic decision quality (again controlling for the effect of age). Because this finding was not anticipated, we first further scrutinized the validity of this finding. Note that the measures of verbal and mathematical aptitude are strongly positively related (Table 1), which is consistent with prior literature showing that verbal aptitude plays an important role in mathematical performance (Durand, Hulme, Larkin, and Snowling, 2005; Pimperton and Nation, 2010; Pina, Fuentes, Castillo, and Diamantopoulou, 2014). For instance, Pina et al. observed a relationship between verbal working memory and complex arithmetic problems, as well as an association between language and knowledge of quantitative concepts and arithmetic ability. This correlation thus validates our measure of verbal aptitude. Further, additional analyses showed that the remarkable dissociation between verbal and creative aptitude that we observed when pass/fail rates were regressed on the predictors is robust against further controls or with varying subsets of the predictors. In other words, increasing levels of cognitive aptitude seem to have a double-faced effect on economic decision quality. Increasing its creative component benefits decision quality, whereas the verbal component, while strongly co-varying with the mathematical component, seems to distort economic decision making.

Exploration of the role of decision heuristics. We then explored the possible role of decision heuristics, without having clear a priori expectations. Based on the literature, we were able to identify five decision heuristics that we could quantify based on our rich choice data set. *Balancing within choice* (i.e. the preferable selection of plates with both biscuits and fruit on) was linked to neither cognitive aptitude nor decision accuracy. The use of the *take the best* heuristic (take the plate with your preferred product(s) on) decreased with age but was not associated with decision accuracy. The three other heuristics showed associations with both a predictor and the criterion, and were therefore candidates to mediate the relation between cognitive aptitude and decision accuracy.

The use of the *compromise* heuristic (the preferable selection of plates with all products on) was positively associated with age, consistent with prior findings (e.g. Jansen et al., 2012) and with decision accuracy. The data suggested that the *compromise* heuristic indeed (marginally) mediated the age effect on decision accuracy. No other heuristics mediated the relation between age and decision accuracy. The use of the *balancing across choices* heuristic (i.e. the tendency to shift between a tasty plate and a healthy plate across the choices settings) increased with verbal

aptitude but had a negative effect on decision accuracy. The mediational role of this heuristic was not significant, although the (probably less accurate) linear analysis suggested mediation. The use of the *more is better* heuristic (i.e. the preference for plates with big quantities) showed the opposite pattern: it reduced with increasing verbal aptitude, whereas it increased decision accuracy. The data also showed that the *more is better* indeed (marginally significantly) mediated the negative relation between verbal aptitude and decision accuracy. We speculate about a possible explanation for these patterns of findings below.

Implications and contributions

Our findings on the age effects add to the literature on consumer socialization. One aspect of consumer socialization involves the development of shopping skills, including comparisons between attributes like prices, volumes, and sizes (Roedder John, 1999). Making such comparisons between (implicit) prices and product volumes is precisely what was asked of the children participating in our study. The Afriat index we used to calculate the percentage of budget participants wasted in suboptimal decisions showed expected age trends in our sample, in that 5-year olds wasted a significantly larger proportion of their budgets than older children did. We did not find age differences between 8-year olds and 12-year olds, who still wasted a considerable percentage of their budgets (approximately 25%) in suboptimal choices. Figure 1 illustrates that the average budget waste was due rather to a minority of participants who displayed low decision accuracy, which is shrinking with age. As such, even our oldest participants displayed considerably more suboptimal behavior than what has been reported in previous studies using similar procedures but including older samples. For instance, one study using a very comparable procedure, requiring undergraduates to select consumption bundles consisting of three goods from a continuous budget set, reported more accurate choice making than we observed in the current study (Bruyneel, Cherchye and De Rock, 2012; although these authors reported different accuracy indices). It thus seems that children aged 12 are not yet as accurate decision makers as young adults are.

This conclusion is however partly at odds with the results of Harbaugh, Krause and Berry (2001), who observed levels of rationality in sixth graders that were very similar to the rationality levels they observed in undergraduates. One potential explanation for these seemingly discrepant findings is that Harbaugh, Krause and Berry (2001) investigated a choice setting including two products only, which is presumably less complex than our choice setting including three products.

One could argue that adding products makes consumption decisions more difficult, and that this effect is more pronounced for younger consumers. It would be interesting to manipulate complexity of choice settings experimentally in one study, and investigate the effect of levels of complexity on decision making accuracy across age groups directly. Our finding that the use of the compromise heuristic may mediate the age effect on decision accuracy is consistent with the argument in terms of complexity. As discussed earlier (Jansen et al., 2012), older children can integrate more dimensions of products, which would easily lead to a more nuanced choice behavior, of which the compromise heuristic is one possible instantiation.

Our findings on the effect of cognitive aptitude add to the literature on the factors determining decision accuracy. We are not the first to show that not all components of cognitive aptitude have an equally strong impact on decisions. Previous research has indicated the importance of incorporating and contrasting diverse measures tapping into different components of cognitive aptitude when studying decision making, as doing so should provide more nuanced insights in which components of cognitive aptitude relate most strongly to which types of decision problems (e.g., Shamosh and Gray, 2008). What do we know about the influence of verbal aptitude? Szucs, Devine, Soltesz, Nobes, and Gabriel (2014) looked at how various cognitive abilities were linked to mathematical accuracy of nine-year-olds. They found support for a network theory of mathematical accuracy (in which important processing nodes were phonological processing, verbal knowledge, visuo-spatial short-term and working memory, spatial ability, and general executive functioning) for primary school children, and concluded that studies should consider the complexity of processes underlying mathematical abilities, and not just focus on a single or a few explanatory variables. In a meta-analysis looking at delay discounting in adults, verbal skills seemed more important predictors of decisions than nonverbal skills (Shamosh and Gray 2008).

What sets our situation apart from this earlier work showing that verbal aptitude helps decision making? Szucs, Devine, Soltesz, Nobes, and Gabriel (2014)'s findings suggest that verbal aptitude helps people incorporate the future in their decision making. Other work (Stok, de Vet, de Ridder, and de Wit, 2012; de Vet, Stok, de Ridder, Brunso, Baban, and Gaspar, 2014) shows that children use complicated (verbal) compensation rules to regulate food intake. We speculate that the use of such verbal rules is not necessarily beneficial for decision quality because the rules and the complexity that they entail may interfere with internally consistent choices. We suggest

that the mediating role of the decision rules for which we find some evidence, is consistent with this idea. First, the use of rules incorporating the future (like “don’t eat too much sugar”, “think about your future health”) that we speculate relies on verbal aptitude may interfere with a straightforward heuristic like the *more is better* heuristic, because such rules may push a decision maker away from her genuine preferences on some occasions (e.g. when letter biscuits are the most numerous options available). This tendency may result in budget waste. Indeed, Stamos, Bruyneel, Cherchye, De Rock and Dewitte (2018), using the same methodology, showed that varying mindset (i.e. cognition versus intuition) increased the budget waste as calculated across all decisions, but did not affect decision accuracy per se.

A similar reasoning can be applied to a heuristic like the *balancing across choices heuristic*, which is a good example of a verbal rule that incorporates the future (via the goals that people want to achieve), possibly at the cost of consistency in preferences. Some of the strategies that young adolescents have been reported to use (Stok, de Vet, de Ridder, and de Wit, 2012), namely *compensation* (e.g. “treating myself to something tasty in the weekend when I ate healthy the whole week”) or *planning* (e.g. “planning in advance how much I will eat of something (like a bag of chips)”) come close to the idea of balancing across choices in the present study (e.g. “I picked the grapes on one occasion, so I can now take the letter cookies”). We speculate that the construction of such rules relies on verbal aptitude and may help self-regulation, but may also distort economic decision making because it drives individuals away from their preferences, and does so in inconsistent ways. One example of such a rule that may distort decision accuracy, which has been documented among adult consumers, is the *licensing effect* (Kahn and Dhar, 2006). It reflects the phenomenon that when consumers have made substantial progress to reach a specific goal (e.g. they have eaten a carrot, which is consistent with their health goal) they tend to subsequently relax their strivings (and for instance consume chocolate). Because the value of the choice options may be affected on each choice occasion due to the effects of different goals, these patterns may be hard to rationalize and lead to budget waste (Stamos, et al., 2018). Further research (with more statistical power) is needed to validate these findings, and to assess whether the other heuristics, which could rely to a larger or smaller extent on verbal aptitude, play a role in decision accuracy.

We also want to revisit the lack of difference between 8- and 12 year olds in our study in terms of budget waste. Our literature review suggests that children of these two age groups are in

different cognitive phases. Also in light of the finding that late adolescents display high levels of economic decision accuracy (Bruyneel, et al., 2012), this may suggest that the use of verbal rules needs fine-tuning when it emerges around this age. Prior research indeed documented similar age-related stagnations or dips. Smith, Xiao and Bechara (2012) used the Iowa Gambling Task, and looked at both affective decision making capabilities and a battery of established cognitive neuropsychological assessments in their sample. They concluded that decision making abilities progress in a J-shaped curve in this task, which is in contrast to the typical linear development of executive functions. More developmentally naïve children performed better on the Iowa Gambling Task than older early adolescents. Performance again became advantageous toward the end of the teenage years. Besides underlining the importance of tapping into various abilities underlying decision making, this study also shows the significance of tracking the development of these various abilities and resulting decision making outcomes in children over time.

Limitations and suggestions for future research

In this work, we built on a measurement technique from economics to evaluate internal consistency in decision making (Varian, 1982). This measurement assumes that there is a stable internal preference, at least during the observation period when the decisions are made. Only under this assumption can we assume that deviations reflect decision inaccuracy. However, an alternative possibility is that preferences are to some extent volatile, and hence that behavioral variability in choices reflects adaptive behavioral variation rather than preference inconsistency (e.g. learning, Regenwetter and Davis-Stover, 2012). Indeed, during play, (typically) developing children show a good deal of behavioral variability in their choices, which can hardly be considered as 'budget waste' (Bancroft, Thompson, Peters, Dozier, and Harper, 2016). This could (partially) explain the age gradient we observed. We must acknowledge that the econometric method we used cannot strictly differentiate between decision inaccuracy and preference volatility, although the fact that the decisions were made in a short time span should reduce this concern to some extent.

The specific set of products is another limitation of this study because it has a certain profile of health, taste, and meaning to the children, and some of the findings may be connected to these product specificities. Replications with other objects is therefore desirable. In addition, although we did our best to simplify the decision situation to make it suitable for kindergarteners, it may still have been challenging to them to make choices and to understand that every choice was

independent and equally important. We do not have manipulation checks to assess to what extent children actually looked at all seven options during each decision. One solution would be to implement more time between each choice, although such a change would substantially increase the cost of the study.

Another limitation relates to the measurement of cognitive aptitude. Although teacher ratings are considered to be reliable (Begeny, Eckert, Montarello and Storie, 2008; Hoge and Coladarci, 1989), they may be more convergent when it comes to different components of cognitive aptitude than the underlying reality, which would reduce the power of our measurements. More complicated dimensions of cognitive aptitude (like working memory capacity or executive functioning) can hardly be assessed by means of teacher ratings. Using validated tests of cognitive aptitude would be a more valid way to measure the different components of cognitive aptitude but would again make the study much more expensive.

Our focus on the validity of assessing decision accuracy came at a cost: The validity of the (exploratively added) indices of decision heuristics may have relatively suffered. For one, the choice situations were not designed with an eye to differentiating the heuristics optimally. To give just one example, the operational definitions of the *compromise* (i.e. count the number of times that all products are chosen) and the *balancing-within-choices* heuristic (i.e. count the number of times two products have been chosen, one biscuit and one fruit) are artificially differentiated here, given that choosing two products may also be driven by a compromise mindset, and choosing all products may also be driven by a balancing mindset to some extent. Further, the relationship between the same quantification of the *compromise* and the *more is better* heuristic must be negatively related in our design because they rule each other out to some extent. Future research may want to focus on differentiating the heuristics better, perhaps at the cost of the validity of the assessment of the decision accuracy. The secondary focus on the heuristics also leads to the fact that they have been measured on different scales with different sensitivities and ranges (i.e. the level of the single choice, the level of all choices, or level of the transition between choices). This prevented us from reliably profiling individual decision makers with respect to their dominant choice or their decision making heuristics. Future studies may want to design the choice situations such that individual choice makers can be clustered into decision making profiles depending on how well their behavior fits one profile. These profiles may then be used to gain more fine-grained

insights into which decision heuristics underlie the relation between age, aptitude, and decision accuracy.

Besides procedural improvements and replication tests, it may be interesting to get a better view on the nature and source of the inaccuracies. Figure 1 shows an interesting dichotomous pattern and it may be interesting to learn what drives the inaccuracies of the group scoring low on decision accuracy. Future research may also focus in more detail on the mechanism through which verbal aptitude may hinder economic decision making and assess its scope and impact in daily life. We speculate that the specification of behavioral rules may distort decision accuracy because rules are not used consistently in subsequent decision situations. We find some evidence but we need a more systematic study with more power to assess this more thoroughly. Does this distortive effect cease when people reach adulthood or does it persist? The already discussed licensing effect suggests that this effect may persist (Khan and Dhar, 2006). And if it does, what could or should be done to mitigate its negative effects on decision quality?

From a developmental perspective, it may also be interesting to find out whether the sources of the errors change over time. We see that the error rates stagnate between the age of 8 and 12, but as these children are obviously evolving cognitively during this time span, the error rate may be driven by different mechanisms or strategies, the adoption of which may entail temporary drawbacks (Smith et al., 2012).

Conclusion

This paper showed how economic decision making accuracy can be measured using thorough measurement techniques, that it improves with age although the trajectory is less straightforward than one may expect, and that verbal aptitude seems to hurt rather than help decision quality. We invite future research to use the powerful method we used here, dissect our findings further, and investigate the broader implication of the role of verbal aptitude in economic decision making.

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Appendix A. The price regimes and the choice sets**Table A1***The nine different price regimes that were used in the experimental set-up.*

Prices		
1 unit of grapes	1 unit of mandarins	1 unit of letter biscuits
8	4	1
8	3	2
9	3	1
1	8	4
2	8	3
1	9	3
4	1	8
3	2	8
3	1	9

Table A2

The nine different choice sets, each consisting of seven different plates that were used in the experimental set-up.

Quantities			Quantities		
Grapes	Mandarins	Biscuits	Grapes	Mandarins	Biscuits
Choice 1			Choice 6		
1.5	0	0	2	0.5	1.33
0	3	0	3	0.75	0
0	0	12	0	0.75	2
0.5	1	4	3	0	2
0.75	0	6	Choice 6		
0.75	1.5	0	0	1.33	0
0	1.5	6	0	0	4
Choice 2			12	0	0
1.5	0	0	4	0.44	1.33
0	4	0	6	0.66	0
0	0	6	0	0.66	2
0.5	1.33	2	6	0	2
0.75	0	3	Choice 7		
0.75	2	0	0	0	1.5
0	2	3	3	0	0
Choice 3			0	12	0
1.33	0	0	1	4	0.5
0	4	0	0	6	0.75
0	0	12	1.5	0	0.75
0.44	1.33	4	1.5	6	0
0.66	0	6	Choice 8		
0.66	2	0	0	0	1.5
0	2	6	4	0	0
Choice 4			0	6	0
0	1.5	0	1.33	2	0.5
0	0	3	0	3	0.75
12	0	0	2	0	0.75
4	0.5	1	2	3	0
6	0.75	0	Choice 9		
0	0.75	1.5	0	0	1.33
6	0	1.5	4	0	0
Choice 5			0	12	0
0	1.5	0	1.33	4	0.44
0	0	4	0	6	0.66
6	0	0	2	0	0.66
			2	6	0

Appendix B. The distribution of amounts chosen across grades and products

A larger proportion of children from kindergarten chose no grapes (18%), no mandarin (31%) or only biscuits (13%) than among the older children (resp. 7%, 19%, and 3% for third graders, and 0%, 10%, and 0% for sixth graders). If anything, this makes it easier for kindergarteners to display consistency in their choices.

Figure B1

The proportion of children of the three age categories choosing the three specific goods.

Appendix C. Regression analyses testing the association between cognitive aptitude and age on the one hand and decision quality

Table C1

<i>predictor</i>	Afriat's index		Binary pass/fail rate	
	Model 1	Model 2	Model 1	Model 2
Mathematical	0.03 ^b (0.03)	0.04 (0.03)	0.10 (0.13)	0.11 (0.13)
Verbal	-0.07** (0.03)	-0.07*** (0.03)	-0.28* (0.15)	-0.28* (0.15)
Creative	0.02 (0.02)	0.02 (0.02)	0.19* (0.11)	0.19* (0.11)
Third grade ^a	0.12 (0.08)		0.41 (0.32)	
Sixth grade ^a	0.12 (0.09)		0.47 (0.32)	
Age		0.02 (0.01)		0.06 (0.05)

^a The benchmark category in model 1 is 'kindergarten'.

^b Betas with standard errors between brackets

* $p < .10$. ** $p < .05$. *** $p < .01$

Appendix D. Robustness tests**Table D1**

The association between cognitive aptitude and decision accuracy controlling for gender and number of older siblings

<i>Predictor</i>	Afriat's index		Binary pass/fail rate	
	Model 1	Model 2	Model 1	Model 2
Mathematical	0.03 (0.03) ^a	0.04 (0.03)	0.10 (0.14)	0.27 (0.17)
Verbal	-0.07** (0.03)	-0.07*** (0.03)	-0.28* (0.15)	-0.51** (0.20)
Creative	0.02 (0.02)	0.03 (0.02)	0.22* (0.11)	0.32** (0.13)
Third grade	0.12 (0.08)	0.14 (0.08)	0.42 (0.32)	0.68 (0.55)
Sixth grade	0.12 (0.09)	0.14 (0.09)	0.45 (0.33)	0.56 (0.34)
Gender	-0.01 (0.08)		0.19 (0.28)	
Number of older siblings		0.04 (0.04)		-0.02 (0.15)
N	99	92	99	92

^a Beta's and standard errors between brackets.

* $p < 0.10$. ** $p < 0.05$. *** $p < 0.01$

Table D2

The association between cognitive aptitude and decision accuracy controlling for gender and number of older siblings

	Afriat's index			
<i>predictor</i>	Model 1	Model 2	Model 3	Model 4
Age	0.02 (0.01)	0.02 (0.01)	0.02 (0.01)	0.02 (0.01)
Verbal		-0.04* (0.02)		-0.05* (0.02)
Creative			-0.00 (0.02)	0.02 (0.02)

* $p < 0.05$

<i>criterion</i>	Pass/fail rate			
<i>predictor</i>	Model 1	Model 2	Model 3	Model 4
Age	0.08* (0.05)	0.08* (0.05)	0.07 (0.05)	0.07 (0.05)
Verbal		-0.09 (0.08)		-0.19* (0.10)
Creative			0.09 (0.09)	0.20* (0.11)

* $p < 0.10$.

Appendix E. The mediation tests

Table E1

The mediation test (using Process) of the use of heuristics underlying the association between cognitive aptitude and grade on the one hand and decision accuracy on the other hand

	Model 2 ^a	Afriat's index		
		Model 2 with "more is better"	Model 2 with "balancing across choices"	Model 2 with "compromise"
<i>Mathematical</i>	0.04 (0.03)	0.02 (0.03)	0.00 (0.03)	0.05 (0.03)
<i>Verbal</i>	-0.07* (0.04)	-0.05 (0.04)	-0.04 (0.03)	-0.09** (0.04)
<i>Creative</i>	0.02 (0.03)	0.02 (0.03)	0.01 (0.02)	0.01 (0.03)
<i>Age</i>	0.02 (0.01)	0.02 (0.01)	0.00 (0.01)	0.01 (0.01)
<i>More is better</i>		0.04** (0.02)		
<i>Balancing across choices</i>			-0.13*** (0.02)	
<i>Compromise</i>				0.04** (0.02)
<i>Bootstrap test</i>		0.02* (0.01)	-0.04 (0.03)	0.01* (0.005)

^a We used model 2 because one variable for age group made it more efficient to test and display the mediation between age group and decision accuracy by *compromise*. The conclusions are identical when model 1 is used.

^b Betas with standard errors between brackets. We remark that the betas and the corresponding significance tests are biased because Hayes Process procedure assumes normality, which is violated. Only the bootstrap test of mediation is unbiased.

* $p < .10$. ** $p < .05$. *** $p < .01$