Better Open Than Intellectual: The Benefits of Investment Personality Traits for Learning

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Abstract
The investment theory of adult intelligence posits that individual differences in knowledge attainment result from people’s differences in cognitive ability and their propensity to apply and invest that ability, which is referred to as investment personality traits. Here, we differentiated intellectual (i.e., intellectual curiosity) and nonintellectual investment (i.e., openness to experience), and we tested their respective predictive validity for knowledge attainment in four independent lab-based studies (overall \( N = 649 \)). Openness to experience was positively associated with knowledge attainment across all four studies, and this effect was by and large independent of cognitive ability. By contrast, intellectual curiosity was not related to knowledge attainment. The findings suggest that openness to experience, rather than intellectual curiosity, is the investment personality trait that broadly benefits learning and adult intelligence.

Keywords
investment personality traits, openness, intellectual curiosity, intelligence, learning

Received July 14, 2017; revision accepted October 20, 2017

Introduction
The investment theory of adult intelligence proposes that individual differences in knowledge attainment result from people’s cognitive ability and their propensity to apply and invest their ability over time (Ackerman, 1996; Cattell, 1943). The interplay of ability and investment has been referred to as “the hungry mind” that is thought to drive learning behaviors and achievement (von Stumm, Hell, & Chamorro-Premuzic, 2011). To date, the predictive validity of investment traits for knowledge attainment has been inferred from their positive associations with markers of adult intellect, mainly academic performance and measures of crystallized intelligence (e.g., Ackerman & Rolfhus, 1999; Fleischhauer, Enge, Brocke, Strobel, & Strobel, 2010; Poropat, 2009; Sobelet & Salthouse, 2010; von Stumm & Ackerman, 2013; Ziegler, Danay, Heene, Asendorpf, & Bühner, 2012). However, no direct observations of the association between investment traits and the accumulation of knowledge as it occurs in real time in typical learning situations have been reported. That is, earlier studies on the association between investment and knowledge operationalized the latter through content that was acquired outside controlled laboratory settings, for example, crystallized intelligence academic performance, or information learned through personal studies (e.g., Ackerman & Beier, 2006; Ziegler et al., 2012). Here, we tested for the first time to what extent investment traits predicted knowledge attainment across learning tasks that varied in situational constraints and cognitive demands in four independent lab-based studies.

Investment Personality Traits
Investment personality traits “refer to stable individual differences in the tendency to seek out, engage in, enjoy, and continuously pursue opportunities for effortful cognitive activity” (von Stumm, Chamorro-Premuzic, & Ackerman, 2011, p. 225). With that, investment traits are conceptually similar to curiosity, which can “broadly defined as a desire to acquire new knowledge and new sensory experience that motivates exploratory behavior” (Litman & Spielberger, 2003, p. 75). Akin to curiosity (Grossnickle, 2016), the investment trait construct space can be differentiated into two subdomains (von Stumm, Hell, & Chamorro-Premuzic, 2011): the “love for knowledge” and “openness to experience” (Berlyne, 1954; Loewenstein, 1994; McCrae, 1994).
Various brain imaging, behavioral genetic, and psychometric studies have substantiated that intellectual curiosity and openness to experience are related, yet distinct constructs that map the investment trait construct space (e.g., DeYoung, Graziotpene, & Peterson, 2012; DeYoung, Shamosh, Green, Braver, & Gray, 2009; Kaufman et al., 2016; von Stumm & Ackerman, 2013; Wainwright, Wright, Luciano, Geffen, & Martin, 2008).

Love for knowledge, or intellectual curiosity (cf. Mussel, 2010), was initially conceptualized as a temporal motivational state that was situationally aroused and stimulated targeted information seeking (Berlyne, 1954, 1960). Under this model, intellectual curiosity was thought to emerge either when individuals were confronted with information that challenged their beliefs, attitudes, or knowledge, or when they identified a gap in their existing framework of knowledge (Kang et al., 2009; Loewenstein, 1994). Both conditions were understood to prompt explorations of the environment to acquire new information that resolved the conceptual conflict or filled the knowledge gap (Berlyne, 1960).

More recently, intellectual curiosity has become recognized as a stable trait dimension of individual differences in the preference for cognitively challenging or complex tasks and intellectual leisure time pursuits (Mussel, 2013). Here, intellectual curiosity is no longer assumed to vary primarily as a function of the situation but rather because of inherent differences between people in their tendency to purposefully seek out knowledge (Goff & Ackerman, 1992), which is at the center of the investment trait construct space (von Stumm & Ackerman, 2013). However, limiting investment to effortful and purpose-driven knowledge seeking ignores the possible benefits of openness to experiences that are not readily identifiable as “intellectual” for the accumulation of knowledge (Grossnickle, 2016).

If actively seeking out knowledge is the essence of intellectual curiosity, the readiness to cognitively engage with perception, fantasy, aesthetics, and emotions is at the core of openness (DeYoung et al., 2012; Kaufman et al., 2016). Accordingly, openness to experience has been described as the “preference for the new and different in many aspects of life” (McCrae, 1994, p. 257). Although this preference not inherently aimed at knowledge attainment, openness is likely to benefit learning in everyday life, especially in situations that are not readily recognized as opportunities for intellectual pursuits.

Divergent Predictive Validity of Openness and Intellectual Curiosity

The differentiation of the investment trait construct space into openness and intellectual curiosity is further supported by their distinct associations with cognitive ability and academic performance (Soubelet & Salthouse, 2010; von Stumm & Ackerman, 2013). For one, openness is less strongly associated with cognitive ability, especially with measures of fluid intelligence, than intellectual curiosity (Fleischhauer et al., 2010; Hill et al., 2013; Mussel, 2010, 2013; Soubelet & Salthouse; von Stumm & Ackerman, 2013). Accordingly, cognitive ability is likely to confound associations between intellectual curiosity and knowledge to a greater extent than associations between openness and knowledge attainment. For the other, several large-scale meta-analyses reported that openness is only modestly related to academic performance with effect sizes approximating .10 (e.g., McAbee & Oswald, 2013; Poropat, 2009; Vedel, 2014), while meta-analytic estimates for the association between intellectual curiosity and academic performance yielded estimates of .30 (von Stumm & Ackerman, 2013; von Stumm, Hell, & Chamorro-Premuzic, 2011). One study, which was based on a correlation matrix of coefficients assembled from different meta-analyses (von Stumm, Hell, & Chamorro-Premuzic, 2011), even showed that intellectual curiosity was positively, but openness negatively, associated with academic performance after adjusting for the variance that both traits shared.

Overall, these findings suggest that intellectual curiosity is a better predictor of academic performance than openness. However, the previous evidence suffers from two important limitations. For one, associations between openness and academic performance have been studied many more times than those between intellectual curiosity and academic performance. As a consequence, meta-analyses on intellectual curiosity spanned a modest number of effect sizes and thus, are likely to produce biased estimates (cf. von Stumm & Ackerman, 2013). For the other, no study to date has tested the divergent predictive validity of intellectual curiosity and openness for knowledge attainment as it occurs. That is, earlier studies reported associations between investment traits with various markers of previously attained knowledge, like crystallized intelligence tests and academic achievement, or knowledge that was attained outside laboratory settings (e.g., Ackerman & Beier, 2006), but none related investment traits to knowledge in controlled study situations as learning occurred. A laboratory setting allows for comparing the predictive validity of openness and intellectual curiosity for knowledge attainment across differently restrictive learning situations, ranging from mundane experiences to formal education settings.

The Current Studies

In the current research, we sought to provide direct evidence for the divergent validity of intellectual and nonintellectual investment for knowledge attainment. To this end, we conducted four independent studies that each employed a different learning paradigm, to test investment-knowledge links across increasingly restricted learning situations. The first was an informal everyday learning opportunity that invited participants to engage with a website at their own discretion. The second task also resembled an informal learning situation...
but participants were directly instructed to study brief trivia facts. In the third task, participants were offered rewards for learning from short articles, whose content was modestly intellectually challenging. This task resembled more closely formal learning situations than the first and second task. The final learning paradigm simulated study situations as they occur in formal education settings. Participants were prompted to study 2,000-word long, scholarly texts and promised rewards depending on their performance in a subsequent exam. After completing each study’s task, participants answered multiple-choice questions about the tasks’ content to assess their learning success. We then tested to what extent intellectual curiosity and openness predicted knowledge attainment, before and after adjusting for individual differences in cognitive ability. Across studies, we predicted that intellectual curiosity was more strongly associated with knowledge attainment in learning situations that emphasized the opportunity for knowledge attainment, while openness was expected to predict learning in situations that did not demand intellectual engagement. In the following, each study is reported in detail; their respective learning tasks are described in full in the supplementary materials (1 to 4).

**Method**

**Study 1**

Participants were presented with a website about the Plitvice Lakes, a picturesque lake district in Croatia, and asked to engage at their own discretion with the website, its weblinks, and the therein contained information for 5 min. This design constituted an everyday learning situation, for which we expected positive associations for both investment traits with knowledge attainment (i.e., information recall), with larger effect sizes for openness than for intellectual curiosity. Cognitive ability was predicted to marginally attenuate the associations.

**Sample.** A total of 201 participants, including 108 men, 92 women, and one participant who preferred not to indicate gender, were recruited from the metropolitan area of London through online advertisements (e.g., gumtree.co.uk; \( n = 120 \)); flyers (e.g., in cafes in central London; \( n = 41 \)); university research volunteer panels (\( n = 24 \)); and approaching participants in person near the testing sites at two major London universities (\( n = 19 \)). Participants were native English speakers and had lived in the United Kingdom for 10 years or more. They ranged in age from 18 to 75 years with a mean of 33.21 (\( SD = 12.71 \)). The majority of the participants reported to have a university degree (63%) and to earn less than £15,000 per annum (51%).

**Measures**

**Cognitive ability.** Logical reasoning ability was assessed with Raven’s Progressive Matrices (Set E; Raven, 1968) and the lettersets test (Ekstrom, French, & Harman, 1976). For Raven’s, 12 items showed grids of 3 rows × 3 columns each with the lower right hand entry missing. Participants chose from eight alternatives the one that completed the 3 × 3 matrix figure. The test was timed at 4 min. For lettersets, participants identified one mismatching four-letter set, inferring a rule underlying the composition of four other four-letter sets. The test had 12 items and was timed at 4 min.

**Investment.** Openness was assessed by the NEO Five-Factor Inventory (NEO-FFI) scale for Openness to Experience (Costa & McCrae, 1992) that consists of 12 items rated on a 5-point Likert-type scale ranging from strongly disagree to strongly agree. The scale includes three items from the Ideas facet of Openness to Experience, which were here excluded from the openness score (i.e., nonintellectual investment). Intellectual curiosity was assessed with the Need for Cognition scale (Cacioppo & Petty, 1982), which consisted of 18 self-report items rated on a 5-point Likert-type scale ranging from strongly disagree to strongly agree. The scale has been shown elsewhere to have excellent psychometric properties (Cacioppo & Petty, 1982; Cacioppo, Petty, Feinstein, & Jarvis, 1996).

**Information access and recall.** Information recall was assessed by 20 multiple-choice questions on the website information about the Plitvice Lakes, each with five answer options, including one correct response (coded as 1) and “I don’t know” (coded as 0, together with incorrect responses). In addition, the time that participants spent on the Plitvice Lakes main page was recorded, as was the number of pages that participants visited over the course of 5 min (maximum 17).

**Procedure and apparatus.** Testing took place in designated laboratory cubicles on desktop computers at two large London universities between July and December 2013. Participants left all personal items (e.g., mobile phones) with a research assistant outside their cubicle. After completing the cognitive tests, participants were informed that they had a 5-min break, during which they were free to do whatever they preferred. They were then shown an image of the Plitvice Lakes in Croatia (Figure 1), which are part of the UNESCO World Heritage. The Plitvice Lakes were here chosen (a) because they are largely unknown in Britain, thus reducing the likelihood that participants had personal associations with or prior knowledge of the lakes and (b) because their visual appeal was judged as likely to intrigue the people’s interest. Underneath the Plitvice Lakes’ image, five links were shown that could be clicked to access further information. Four included written information on Terrain, Climate, Local Customs, and Flora; each of these also included three additional links leading to further information. The fifth link was a Gallery and showed six different images of the Plitvice Lakes. Overall, a maximum of 17 pages could be accessed. A pilot study (\( N = 9 \)) confirmed that
all the information could be comfortably reviewed in 5 min. The information consisted of real and made-up facts that were simple, memorable, and highly unlikely to be known outside of this study. For example, one link informed about the amount of annual rainfall at the Plitvice Lakes (real), and another described different orchid types of the Plitvice Lakes (fictional). On each page, a “Back” button was shown at the bottom left corner that allowed participants to navigate back to the previous pages. Afterward, participants completed the personality measures and answered multiple-choice questions on the Plitvice Lakes, including two multiple-choice questions for each main page and one for each of the three additional links (i.e., the links within each main information page; overall 4 main links × 2 questions + 3 sublinks × 4 main links × 1 question = 20 questions). Individual testing sessions lasted approximately 30 min; participants received £10 compensation.

**Results.** Study 1 had 80% power to detect a correlation of .2 and 98% to detect a correlation of .3 (based on the analysis sample of n = 184, details below). Table 1 reports the descriptive statistics for this and the following three empirical studies; correlation matrices for each study are in the supplementary materials (5). In all studies, including the current one, scores were normally distributed and internal consistency values (i.e., reliability with Cronbach’s alpha) were satisfactory. Across studies, Cronbach’s alpha values were slightly higher for measures of intellectual curiosity than for openness, reflecting the difference in narrowly assessing an inclination for knowledge pursuit compared with capturing the broader openness to non-intellectual internal and external experiences.

The scores from the cognitive tests correlated at r = .47, were z-transformed and added to a unit-weighted composite of cognitive ability. For 10 participants, irregularities occurred during testing (e.g., participant kept mobile phone despite instructions, or left cubicle when Plitvice Lakes main page was presented); these were excluded from all analyses (N = 191). Seven participants (3.7%) did not engage at all with the Plitvice Lakes page throughout the 5-min break period, while the majority of participants spent on average 3 min browsing 10 out of 17 pages. Participants reported after the study that, when they had not viewed the Plitvice Lakes information, they had napped, dozed, or stared into space. They were excluded from the subsequent analysis (i.e., analysis N = 184).

Openness and intellectual curiosity were positively intercorrelated (r = .42). Linear regression models, adjusted for age, gender, testing location, and recruitment method,
showed that openness was positively associated with the answer scores regarding website content on the Plitvice Lakes ($\beta = .24$, $p = .003$, 95% confidence interval [CI] [.08, .39]) but intellectual curiosity was not ($\beta = .04$, $p = .639$, 95% CI [−.12, .20]), with the model accounting for 6.8% of the variance (i.e., adjusted $R^2$; Figure 2a). Adjusting for cognitive ability ($\beta = .13$, $p = .090$, 95% CI [−.20, .28]) did not change the results, with openness remaining a significant predictor ($\beta = .22$, $p = .006$, 95% CI [.06, .38]), and the overall model accounting for 7.8% of the variance.

Discussion. Study 1 showed that openness rather than intellectual curiosity was associated with knowledge attainment in an unconstrained everyday learning task (i.e., when viewing a website). This finding confirmed only partly our hypotheses, which predicted stronger effects of openness, than of intellectual curiosity, on knowledge attainment but expected significant positive associations for both. The results suggest that intellectual curiosity does not benefit learning in situations that appear not to offer opportunities for intellectual engagement. As predicted, cognitive ability did not confound the relationship between openness and learning. The model explained overall 7.8% of the variance in knowledge attainment, suggesting medium effect sizes for the association between openness and knowledge.

Study 2

By contrast to the learning paradigm used in Study 1, participants were directly prompted to engage with the learning materials in Study 2, and asked to study short texts about trivia facts. Participants were, however, neither told that they were to be examined on the content of trivia facts nor were they incentivized to attain knowledge. Because Study 2 offered an opportunity for intellectual engagement similar to that in Study 1 (i.e. low intellectual engagement), we predicted that openness, but not intellectual curiosity, would benefit learning and that cognitive ability would not affect these associations, in line with the findings from Study 1.

Sample. Overall, 179 participants were recruited, including 131 prospective university students and their parents who were approached during the Open Days at two major London universities and compensated with sweets and crisps, as well as 48 undergraduate students, who participated in exchange for course credits. Participants who spent less than 90 s completing the logical reasoning test (i.e., letter sets) were excluded from all analyses ($N = 9$). The majority of the analysis sample identified as female ($N = 148$ of 170; 83%), and ages ranged from 15 to 65 years ($M = 23.30$, $SD = 11.34$). The majority had had obtained a school leaving certificate as highest educational qualification (77%).
Figure 2. Regression plots for intellectual curiosity and openness on knowledge attainment across Studies 1 through 4. Note. The y-axis shows z-scores for knowledge attainment; the x-axis represent z-scores for the respective investment trait. Regression plots were adjusted for confounders (Study 1: age, gender, recruitment method, testing location; Study 2: age, gender; Study 3: age, gender, native English speaker; Study 4: age, gender, native English speaker, dyslexia, study condition), and openness and intellectual curiosity were entered simultaneously. For all studies, results were unchanged if confounders were dropped from the respective models Study 4a and 4b refer to immediate and delayed recall, respectively.
Measures

Cognitive ability. Participants completed lettersets test from Study 1, but due to a technical error 11 instead of 12 items. In addition, participants completed a short-term memory test (von Stumm, 2016). The test consisted of 10 series of five and seven number pairs or pairs of numbers and letters that were presented for 5 and 10 s, respectively. Participants were then asked to recall in the order that the pairs were shown in within 25 and 30 s, respectively (total n = 58).

Investment. By contrast to Study 1, intellectual curiosity was operationalized by the NEO Personality Inventory—Revised (NEO-PI-R) Ideas facet (Costa & McCrae, 1992), which can be used interchangeably with the Need for Cognition scale (DeYoung et al., 2012; Mussel, 2010; von Stumm & Ackerman, 2013). Openness was again measured by the NEO-FFI scale, excluding three items from the Ideas facet.

Information recall. Participants completed 10 multiple-choice questions that assessed their factual recall and conceptual understanding of the five short articles (two questions per article). Each question had five answer options, including the correct answer and one “I don’t know” option. Correct answers were coded as 1, and all others as 0.

Procedure and apparatus. Testing took place in designated research cubicles on desktop computers at two major London universities between June 2014 and October 2015. Participants left all personal items (e.g., mobile phones) with a research assistant outside their cubicle. Participants read in their own time through the trivia fact articles, which ranged in length between 90 and 119 words. The trivia facts were unlikely to be known outside of this study and included (1a) spotting lying in the face, (b) sign language, (c) English freedom medals, (d) Guinness World of Records most dangerous tree, and (e) placebo and nocebo effects. After the reading, participants completed cognitive and personality tests and then answered the multiple-choice questions.

Results. The study had a power of 75% to detect an association of .2 and a power of 98% to detect a correlation of .3 (based on the analysis sample N = 170). Cognitive test scores correlated at r = .30, which reflects the very different nature of the two administered cognitive tests. A composite of cognitive ability was computed like in Study 1. Openness and intellectual curiosity correlated at r = .45.

A multiple regression model, adjusted for age and gender, showed that openness predicted learning significantly (β = .19, p = .032, 95% CI [.02, .37]) but intellectual curiosity did not (β = -.04, p = .616, 95% CI [-.22, .13]), with the model accounting for 2.2% of the variance in recall (i.e., adjusted R²; Figure 2b). Adjusting for cognitive ability (β = .10, p = .184, 95% CI [-.05, .25]) did not change these results, with openness remaining a significant predictor (β = .19, p = .038, 95% CI [.01, .36]) and the model accounting for 2.7% of the variance.

Discussion. Study 2 confirmed Study 1’s finding that openness, not intellectual curiosity, benefitted knowledge attainment in an everyday learning task. By contrast to Study 1, where participants engaged with the learning materials at their own discretion, participants in Study 2 were directly prompted to study. Thus, openness’ learning-related benefits for knowledge appear not to be restricted to completely unconstrained learning situations but they also emerged in a more guided learning context. Like in Study 1, the association between openness and learning was independent of cognitive ability. The effect size of the association between openness and information recall was comparable to Study 1, but the overall model accounted for far less variance (i.e., 7.8% vs. 2.6%).

Study 3

As in Study 2, participants were directly prompted to engage with the study materials, in this case five texts about film-related topics. By contrast to Studies 1 and 2, participants were informed that they would be examined on the study materials, and they were incentivized to perform well by the chance to win £20 for a top score. Accordingly, the current learning task was more constrained than those in Studies 1 and 2, although the film texts were only modestly cognitively demanding, with little opportunity for intellectual mastery. Thus, we expected to replicate our findings from Studies 1 and 2, with openness emerging as stronger predictor of knowledge attainment compared with intellectual curiosity.

Sample. A total of 130 participants attended lab-based testing sessions, including prospective students and their parents, who were approached at university Open Day events, and undergraduate students. From the overall sample, 32 participants were excluded because they either (a) failed one of three quality control items (N = 24) or (b) spent less than 2 min reading the study materials (N = 8). Thus, the analysis sample consisted of 98 participants aged 16 to 56 years (M = 22.0; SD = 8.4). Just over half reported school leaving certificates as highest educational qualification (56%). Participants were compensated with sweets and crisps (Open Days) and course credits (undergraduates).

Measures

Cognitive ability. The lettersets test from Study 1 and the short-term memory test from Study 2 were administered.

Investment. Openness and intellectual curiosity were assessed as described in Study 2.

Information recall. Participants completed six multiple-choice questions for each of the five film articles, with three assessing factual information recall and three capturing conceptual understanding (i.e., 30 questions overall). Each question had five answer options, including the correct answer...
and one “I don’t know” option. To stop participants from trying to look up the answers elsewhere rather than recalling them from memory, questions were timed according to their complexity and length, allowing between 12 and 28 s for an answer (mode = 20 s). The materials were repeatedly piloted and revised prior to the actual testing phase (N = 22).

**Procedure and apparatus.** Testing took place in designated laboratory spaces on desktop computers at three U.K. universities between July 2015 and March 2016. Participants left all personal items (e.g., mobile phones) with a research assistant outside their cubicle. They were then instructed that they could earn £20 if they correctly answered all later multiple-choice questions about the content of the film articles that they were asked to read. Participants were advised that the questions were timed and that they could and should not cheat on them. Film was chosen as a topic because it attracts wide public interest. The articles were specifically created to ensure that their content was unlikely to be known outside of this study, with topics including (a) general introduction to film, (b) film industry and funding, (c) advances in film technology, (d) film actors and stars, and (e) film awards and prizes. Articles ranged in length from 266 to 356 words (M = 311 words), and they were each presented for a maximum of 120 s. After reading the articles, participants completed the cognitive tests and personality measures and reported their five favorite movies, before answering 30 multiple-choice questions on the articles’ content.

**Results.** Study 3 had a power of 50% to detect a correlation of .2, and a power of 85% to detect a correlation of .3 (based on the analysis sample N = 98). The cognitive ability test scores correlated at r = .41, and a composite was computed as in Study 1. Openness and intellectual curiosity correlated at r = .43. A multiple regression model, adjusted for age, gender, and being a native English speaker versus not, showed that openness was significantly associated with knowledge attainment (β = .27, p = .017, 95% CI [.05,. .50]), but intellectual curiosity was not (β = −.07, p = .554; 95% CI [−.30, .16], Figure 2c). However, the model’s adjusted R² suggested that it accounted only for 1.8% of the variance in recall. After adding cognitive ability, the model accounted for 9.3% of the variance, with both openness (β = .27, p = .014, 95% CI [.06, .49]) and cognitive ability (β = .30, p = .004, 95% CI [.10, .50]) emerging as significant predictors.

**Discussion.** As in Studies 1 and 2, openness significantly predicted knowledge attainment but intellectual curiosity did not. Also confirming the findings of Studies 1 and 2, the association between openness and learning was independent of cognitive ability. That said, cognitive ability was a significant predictor of knowledge attainment and accounted for the largest proportion of variance, by contrast to openness, which only explained a very small amount of variance.

**Study 4**

For our final study, we sought to simulate a formal learning situation that (a) was more restricted than the paradigms used in Studies 1 through 3, (b) offered opportunity for intellectual mastery, and (c) demanded effortful cognitive engagement. Participants attended three lab-based study sessions over the course of 3 weeks. Each week, participants studied an unknown scholarly text of approximately 2,000 words length that focused on history, science, and economics, respectively. Afterward, they completed exam-style multiple-choice questions on the current as well as on the previous weeks’ study topics. Participants were incentivized each week to get as many exam questions as possible right by the chance to win £50 each week for the top score, in addition to the participants’ baseline compensation fee.

In line with previous meta-analyses that showed weak associations between openness and academic performance (McAbee & Oswald, 2013; Poropat, 2009; Vedel, 2014) and stronger links between intellectual curiosity and academic performance (von Stumm & Ackerman, 2013), we expected here that intellectual curiosity rather than openness would predict knowledge attainment. We also hypothesized that cognitive ability would attenuate the association between intellectual curiosity and learning, following on from earlier studies that reported substantial shared variance between intellectual curiosity and intelligence (e.g., Fleischhauer et al., 2010; Hill et al., 2013).

**Sample.** Overall, 233 participants were recruited, of whom 229 completed all measures relevant to this study in Week 1, while 206 participants completed Week 2, and 197 participants attended Week 3. After excluding participants who spent less than 60 s on each week’s learning task, the analysis sample consisted of 180 participants with complete data (44 men, 135 women, 1 who preferred not to say). Ages ranged from 18 to 58 years (M = 26.48, SD = 10.26). Almost 81% of the participants were undergraduate students; the remainder were adult volunteers, who were recruited through newspaper advertisements and flyers in local businesses.

**Measures**

**Cognitive ability.** The lettersets test as described in Study 1 was administered.

**Investment.** Openness and intellectual curiosity were assessed as described in Study 2.

**Information recall.** Each week, participants completed eight exam-style multiple-choice questions on the current week’s text. In addition, they also completed eight exam-style multiple-choice questions on the previous weeks’ texts in Weeks 2 and 3 (i.e., 16 questions overall in Week 2 and 24 questions in Week 3). Half of the questions pertaining to the
previous weeks’ texts had also been administered in the previous week, and half were new questions. For all questions, half assessed factual and half conceptual knowledge conceptual knowledge. Each question had five answer options, one being correct and one being “I don’t know.”

Procedure and apparatus. Testing took place in designated laboratory spaces on desktop computers with speakers at a London university between March 2016 and January 2017. Participants left all personal items (e.g., mobile phones) with a research assistant outside their cubicle. In Week 1, they were randomly allocated to either read or listen to each week’s 2,000 words text, which had been specifically written for this study and featured three different scholarly topics, including history (i.e., the Cuban Missile Crisis), science (i.e., CRISPR), and economics (i.e., the Dotcom bubble). Participants were instructed that they could win £50 if they achieved a top score in a set of exam-style multiple-choice questions that they were to complete after the study period. In the listening condition, participants heard via headphones the 20-min long digital records of the texts that were read out by a professional speaker (i.e., a male priest). In the reading condition, participants viewed the text on screen. Participants received pens and notepaper to take notes on the text or recording, if they wanted, and they were allowed to reread or relisten to the texts as long as they wished. Precautions were taken to avoid that participants copied the texts. After the participants finished studying the current week’s text, they called for a research assistant, who removed the notepaper. In Week 1, participants then completed measures of cognitive ability and personality (see Note 1), before they answered the exam-style multiple-choice questions. In Weeks 2 and 3, participants completed other self-report measures (see Note 1), before answering the multiple-choice questions. After completing Week 3, each participant received £40 compensation, as well as prize money of up to £150 for the highest scorers.

Results. This study had a power of 78% to detect a correlation of .2 and a power of .98% to detect an association at .3 (based on the analysis sample of N = 180). Openness and intellectual curiosity correlated at .42. Knowledge attainment was operationalized as immediate recall (i.e., responses to multiple-choice exam questions pertaining to the current week’s text, n_questions = 24) and delayed recall (i.e., responses to multiple-choice exam questions pertaining to the previous week’s text, n_questions = 16; Table 1).

Linear regression models showed that openness was significantly associated with immediate recall (β = .24, p = .003, 95% CI [.08, .39], Figure 2d), after adjusting age, gender, being a native English speaker versus not, having dyslexia versus not and learning condition (i.e., audio vs. reading). Intellectual curiosity was, however, not significantly associated with immediate recall (β = .11, p = .159, 95% CI [.−.04, .27]), with the model accounting overall for 8.8% of the variance in recall. Cognitive ability significantly predicted immediate recall (β = .25, p < .001, 95% CI [.11, .39]) and attenuated its association with openness (β = .19, p = .015, 95% CI [.04, .34]), with the model accounting overall for 14.6% of the variance.

For delayed recall, openness was also a significant predictor (β = .19, p = .020, 95% CI [.03, .34]) and intellectual curiosity was not (β = .12, p = .130; 95% CI [−.04, .28], Figure 2e), with the model accounting for 7.8% of the variance. After adding cognitive ability (β = .18, p = .014, 95% CI [.04, .32]), the association between openness and delayed recall became nonsignificant (β = .15, p = .056, 95% CI [−.01, .31]), with the model accounting for 10.5% of the variance.

Discussion. Contradicting our hypothesis, intellectual curiosity was not associated with knowledge attainment in this comparatively constrained and cognitively demanding learning task. Instead, openness was again the better predictor of learning success, including both immediate and delayed recall, with corresponding models accounting for around 8% of the variance. That said, cognitive ability also significantly predicted immediate and delayed knowledge attainment, and notably attenuated their association with openness. It therefore appeared in Studies 3 and 4 that cognitive ability, rather than intellectual curiosity, gained predictive strength in the learning task that was highest in situational constraints and cognitive demands.

General Discussion

The hungry mind concept posits that the interplay between cognitive ability and investment personality traits is at the core of the learning process and thus informs knowledge attainment. Here, we differentiated two domains within the investment trait construct space, namely openness and intellectual curiosity (DeYoung et al., 2012, 2009; Grossnickle, 2016; von Stumm & Ackerman, 2013; Wainwright et al., 2008), and we tested their predictive validity for knowledge attainment across four controlled learning tasks. Openness refers to the tendency to cognitively engage with perception, fantasy, aesthetics, and emotions (DeYoung et al., 2012; Kaufman et al., 2016). By contrast, intellectual curiosity captures individual differences in the preference for engaging in mentally challenging tasks and the purposeful pursuit of knowledge (Goff & Ackerman, 1992; Mussel, 2013).

Openness, Not Intellectual Curiosity, Predicts Learning

Across the four studies reported here, intellectual curiosity and openness were intercorrelated with r values ranging from .42 to .45, suggesting up to 20% common variance. Notwithstanding their empirical overlap, we found here reliable evidence for conceptually and operationally differentiating openness from intellectual curiosity. Specifically, we showed for the first time that openness, but not intellectual curiosity, benefited knowledge attainment across learning tasks that varied in situational constraints and cognitive
demands. This finding is surprising, given that previous studies reported substantial positive associations between intellectual curiosity and markers of adult intellect, such as academic performance and measures of crystallized intelligence (Ackerman & Rolfhus, 1999; Fleischhauer et al., 2010; Soubelet & Salthouse, 2010; von Stumm & Ackerman, 2013; von Stumm, Hell, & Chamorro-Premuzic, 2011). Although we might have expected that a preference for intellectual leisure time pursuits and cognitive mastery was unrelated to learning in mundane situations (e.g., website reading, trivia facts; Studies 1 and 2), intellectual curiosity was expected to prompt knowledge attainment in more constrained and challenging learning contexts, which were simulated in Studies 3 and 4 akin to learning situations at school and university. Conversely, a substantial body of empirical evidence has suggested that openness was a weak predictor of learning achievement in formal education (e.g., McAbee & Oswald, 2013; Poropat, 2009; Vedel, 2014; von Stumm, Hell, & Chamorro-Premuzic, 2011), but it emerged here as a stable and strong predictor of knowledge attainment across tasks.

We can only speculate about the reasons for the discrepancy between the current and previous findings. For one, it is possible that earlier meta-analyses overestimated associations between intellectual curiosity and knowledge attainment, because they relied on few, heterogeneous studies (von Stumm & Ackerman, 2013; von Stumm, Hell, & Chamorro-Premuzic, 2011). For the other, meta-analytic associations between openness and knowledge attainment may have been systematically underestimated, because openness is not yet routinely differentiated in research into open and intellectual investment (DeYoung et al., 2012; DeYoung et al., 2009; von Stumm, Hell, & Chamorro-Premuzic, 2011).

A necessary next step in explaining the findings will be identifying the behavioral mechanisms that explain the association between investment personality traits and knowledge attainment. Two previous studies tested whether engaging in cognitively stimulating activities—for example, attending evening classes or visiting museums and theaters—mediated the relationship between investment and knowledge but found no supporting evidence (Soubelet & Salthouse, 2010; von Stumm, 2012). An alternative behavioral mechanism to explain the investment-learning link is the ways in which people construe their learning experiences (Stine-Morrow, 2007). However, the notion of individual differences in construing experiences is difficult to study in terms of directly observable behaviors. That said, it seems plausible that openness benefits learning because it predisposes to perceiving and extracting information across situations, including mundane day-to-day experiences, as well as more cognitively challenging undertakings.

**Investment, Cognitive Ability, and Knowledge**

Across four learning paradigms, the relationship between openness and knowledge attainment was by and large independent of cognitive ability, which was assessed by short tests in the current research, rather than through a comprehensive battery of measures. Thus, it remains possible that the observed associations are confounded by cognitive ability. That said, the predictive validity of cognitive ability for knowledge attainment was notably greater in Studies 3 and 4, which were more constrained and cognitively demanding, than in Studies 1 and 2. Furthermore, cognitive ability attenuated the association between openness and knowledge attainment in Study 4, which had the highest degree of situational constraint. The “strong situation hypothesis” suggests that in situations of greater press—that is, in high-stake settings where learning is hard and directly associated with reward and appraisal—the effect of personality on behavior is weakened, while capacity-related characteristics gain predictive validity (Ackerman, 2013; Cooper & Withey, 2009). In line with this idea, the current research showed that maximum performance measures, like cognitive ability (i.e., what a person can do), played a greater role in the strong situation learning tasks compared with those with fewer situational constraints. However, the association between openness, a typical performance measure (i.e., what a person will do), and knowledge attainment remained stable across learning tasks regardless of situational constraints, although corresponding effect sizes varied. Thus, our findings provide only partial support for the “strong situation hypothesis”: The predictive validity of cognitive ability grew with increasing situational strength, but the effect of openness was not weakened at the same time.

**Strengths and Limitations**

The research reported here has several strengths, including that each study was sufficiently powered, tested adult participants in designated lab spaces, and used reliable measures. The work is also not without weaknesses. First, the assessment of knowledge attainment relied predominantly on information recall while other aspects of learning were not assessed, for example, understanding complex relationships, transferring knowledge and skills across situations, and synthesizing information. That said, for all administered learning tasks, the recall questions were designed to assess both factual and conceptual knowledge of the study materials. Second, the current research only tested short-term associations between investment traits and knowledge attainment, either within one assessment session or across 1 week. Going forward, longitudinal studies must substantiate the hungry mind concept and test the long-term effects of openness on learning (Ackerman & Rolfhus, 1999; Ziegler et al., 2012). Third, this research did not explore the behavioral mechanisms that give rise to the association between investment and learning. Thus, we can only speculate about the causal processes that underlie the hungry mind concept. Finally, the investment traits studied here accounted, with adjustment for covariates, for 1.8% to 8.8% of the variance in knowledge attainment, suggesting
that corresponding associations range in effect sizes from small to medium. Thus, the influence of investment traits in learning may be similar, if not weaker to that of other factors, which were not presently assessed, for example, interest or conscientiousness.

**Conclusion**

The current research empirically supported the hungry mind concept that views investment personality traits and cognitive ability as key determinants of knowledge attainment. Specifically, we showed that open investment benefitted learning across four tasks that varied in situational constraints and cognitive demands. By contrast, intellectual curiosity was not associated with knowledge attainment for any of our learning task. Our findings propose assessing and treating openness and intellectual curiosity as separate entities in future research to identify replicable associations between investment and learning. More importantly, our results also suggest that it is better to be open, rather than intellectual for accumulating knowledge.

**Declaration of Conflicting Interests**

The author declared no potential conflicts of interest with respect to the research, authorship, and/or publication of this article.

**Funding**

The author disclosed receipt of the following financial support for the research, authorship, and/or publication of this article: This research was supported by the Eysenck Memorial Award 2012 and by Grant RFP-15-03 from the Imagination Institute (www.imagination-institute.org), funded by the John Templeton Foundation. The first author is supported by a Jacobs Research Foundation Fellowship 2017-2019. Barnaby Brien, Yordanka Dimova, Ian Hammert, Alan Mak, Hannah Scott, and Abegail Tan helped with the data collection, stimuli design, and technical support.

**Notes**

1. Participants also completed several measures and tasks that are not relevant to the current analyses. These are not reported here.
2. The study conditions were not relevant to the current analyses and are therefore not further discussed in detail.

**Supplement Material**

Supplementary material is available online with this article. Also, all data files and study materials for this manuscript are available from the Open Science Framework on https://osf.io/d85wy/.

**References**


