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The Berliner Multi-Facet Personality Inventory: An extensive measure of Big Five

2 personality

Victor Rouco^{1,2}, Anja Cengia³, & Matthias Ziegler³

¹ Universitat de Barcelona

² Institut de Neurociències de Barcelona

 3 Humboldt Universität zu Berlin

Author Note

- Add complete departmental affiliations for each author here. Each new line herein must be indented, like this line.
- Enter author note here.
- 11 Correspondence concerning this article should be addressed to Victor Rouco, Postal 12 address. E-mail: victorrouco@ub.edu

13 Abstract

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1. Introduction

Over the last decades, the Five Factor Model as well as the Big Five model have 20 become widely accepted models for describing general attributes of personality. Often the terms are even used synonymously, which is why we will refer to the Big Five from here on. 22 The Big Five is a hierarchical model which describes human individual differences in personality at the dispositional level: one of the most basic, universal, biologically-influenced 24 and stable layers of human inter-individual differences in behavior, cognition and feeling (D. 25 P. McAdams & Pals, 2006). Its hierarchical conception is relevant to acknowledge behavior from the most specific (nuances) to the most broad differences in temperament and character 27 (dimensions), through a varying number of mid-level personality characteristics (facets). 28 Most of the research concerning criterion validity of the Big Five inventories has focused on the covariation between the Big Five dimensions and relevant external outcomes. However, specific dispositional characteristics captured on the facet level might be of extreme utility to 31 provide more complex descriptions of individuality and to predict life outcomes to a major 32 extent (Lounsbury, Sundstrom, Loveland, & Gibson, 2002; S. V. Paunonen & Ashton, 2001; Ziegler et al., 2014). Unfortunately, the number and nature of the facets below the Big Five is far from being consensual. In fact, different facet level models have been proposed 35 (XXXX). One potential reason for this could be that many facet level models were developed after a questionnaire version without such a level had been published. Thus, the facets were developed as an elaboration or extension to an existing domain measure. While this has many theoretical advantages it also has the disadvantage of potentially limiting the search space of possible facets. In this work we aim at maximizing this search space and present a personality questionnaire which is broad at the facet level, open-access, and measurement 41 invariant across two different cultures.

1.2. A short history of the Big Five

Francis Galton is credited as being the one who proposed the fundamental lexical
hypothesis as a ground from where to describe interpersonal differences in personality. The
hypothesis states that every apprehended characteristic in the realm of personality should
have its place in the natural language, a corollary derived from this first statement is that
the essential features must represent a unique word in the lexical universe of this language.
Galton (1884) himself, and later Allport and Odbert (1936) and still later Norman (1967),
used English dictionaries for a systematic collection of all adjectives which could be related
to human personality characteristics. Using exploratory factor analyses on self and other
ratings, five broad factors could repeatedly be extracted from the data.

Cattell was one of the first researchers who systematically applied exploratory factor
analysis in order to explore personality structure. He inspected the correlation structure of
the items in the word lists of his predecessors, finding 16 oblique personality factors,
including one factor specifically for intelligence, these factors form the 16-PF. These 16
factors were the primary factors in a hierarchical structure for Cattell (coetany to L.L.
Thurstone and undoubtedly influenced by him). Cattell himself viewed personality as a
hierarchical structure, containing three layers (Cattell, 1956). The second order factors
resemble the Big Five dimensions (Digman, 1990).

Different researchers followed Cattell in the study of dispositional traits of personality.
One of the most influential models was Eysenck's Big Three. Grounded on a strong
biological basis, Eysenck's theory supposed a link between temperament and personality. Its
structural proposal concerned at first two big factors, named Neuroticism vs. Emotional
stability and Extraversion vs. Introversion. These two dimensions were later joined by a third
factor that Eysenck called Psychoticism. This label was criticized by others who suggested
that a more appropriate term would be psychopathy (Digman, 1990). Eysenck's big two are

still "alive" today in the Big Five, and his third factor, psychoticism, can be operationalized as other dimensions within the Big Five: Agreeableness, Conscientiousness and Openness.

A large number of studies have focused on the problem of personality structure
resulting in a five factor solution (Borgatta, 1964; Fiske, 1949; Norman, 1967; Tupes &
Christal, 1961). Possibly the two most widely cited works relating to the foundations of the
Big Five are those by L. R. Goldberg et al. (2006) and P. T. Costa and McCrae (1995).

Goldberg can be seen as one of the first who extended research concerning the Big Five,
while McRae and Costa's importance rests on popularizing the terminology (OCEAN) and
the development of one of the most used tools to assess personality based on the Big Five:
the NEO-PI. The Big Five dimensions are labeled as follows: I) Extraversion vs. Introversion.

II) Agreeableness or Friendliness. III) Conscientiousness or Achievement or Will. IV)
Emotional Stability vs. Neuroticism, and V) Openness or Intellect or Culture.

One of the most important features of the Big Five is the fact that it could be
replicated in different languages. Research is available in Japanese, Vietnamese, German,
Spanish, Greek, and many more (???; Schmitt et al., 2007). This finding suggests that the
way human beings construe personality is at some point universal and that its basic features
are retained within the Big Five. Another essential characteristic relies on its hierarchical
nature. The five domains are useful to retain the big picture of personality, maximize the
situation consistency and reliably assess difficult subjects such as children. Nonetheless, each
dimension is conceptualized as a latent construct formed by more specific narrow factors
called facets, which in turn are useful to depict the impact of personality characteristics into
specific behaviors and concrete life outcomes.

The Big Five has proven to be a valid theoretical and empirical model to predict relevant life outcomes. Research such as Ozer and Benet-Martínez (2006) or B. W. Roberts, Kuncel, Shiner, Caspi, and Goldberg (2007) has shown that scores for the Big Five dimensions (and their related facets) are able to explain outcomes such as academic and

work performance, health, personality disorders, political attitudes and many more. The
empirical findings linking Big Five measures to life outcomes have reinforced the concurrent
validity of the test scores interpretations. At the same time, the broad nature of the domains
has spurned research into the more fine-grained lower order structure of facets.

98 1.3. Facet Structures

There are a number of models that include a facet structure below the five broad
domains. The most widely known model is the one suggested by P. T. Costa and McCrae
(1995), the NEO-PI-R model. Other popular models have been suggested for the Big Five
Inventory 2 (BFI-2; Christopher J Soto & John, 2016), the IPIP (L. R. Goldberg et al.,
2006), and the HEXACO model (K. Lee & Ashton, 2016), which assumes six broad domains.

Table 1 gives an overview of these different models listing their facets per domain as well as
some information regarding their psychometric properties.

< Table 1 >

As shown in *Table 1*, there are different possibilities of facets forming the domains. 107 However, there is still a degree of overlap between the facets covered by the different 108 instruments. Christopher J. Soto and John (2009) inspected the convergences between the 109 NEO-PI-R and the first version of the BFI, suggesting that two constructs per domain were 110 measured at the facet level by both inventories. The constructs defined by Christopher J. Soto and John (2009) were: Altruism and Compliance for Agreeableness; Anxiety and 112 Depression for Neuroticism; Order and Self-Discipline for Conscientiousness; Assertiveness 113 and Activity for Extraversion; and Aesthetics and Ideas for Openness. The convergence holds 114 for the four instruments listed in Table 1, as these ten constructs are covered within the 115 facets for every instrument. Some of the constructs are explicitly covered at the facet level 116

(e.g. Anxiety); meanwhile others are mainly covered by the four instruments, although 117 sometimes implicitly (e.g. Liveliness in HEXACO resembles the "core" construct Activity, 118 present in all other instruments). The reverse is not always true, not every facet within the 119 four instruments is covered by the constructs proposed by Christopher J. Soto and John 120 (2009). As an example we find Self-Consciousness, a Neuroticism facet defined by the 121 NEO-PI-R and the IPIP-NEO-120, which is not intrinsically tapping at either Anxiety or 122 Depression. The same authors asserted in a later work (Christopher J Soto & John, 2016) 123 that the Big Five domains "can be conceptualized and assessed more broadly or more 124 narrowly", either focusing in a central facet or in a set of peripheral facets, depending the 125 research interest. 126

The mid-level layer between domains and facets has also been explored by DeYoung, 127 Quilty, and Peterson (2007). Their work has focused in the biological consistency of the 128 NEO-PI-R set of facets, thereby proposing a two factor source of variance for each facet of 129 the inventory. In line with their proposal, Agreeableness would be composed by Compassion 130 and Politeness; Neuroticism by Volatility and Withdrawal; Conscientiousness by 131 Industriousness and Orderliness; Extraversion by Enthusiasm and Assertiveness; and 132 Openness by Intellect and Openness. Both Christopher J. Soto and John (2009) and 133 DeYoung et al. (2007) proposals have many points in common. Maybe the labels Volatility 134 and Withdrawal for Neuroticism can be suspicious of a different content than Anxiety and 135 Depression, but when inspected at the item level it is revealed that they are tapping the 136 same components respectively (DeYoung et al. (2007); for item specification). 137

The nomological network commonly assumed in Big Five questionnaires is drawn from nuances through facets to domains, from more specific to more general. Relying on domains to explain and predict behavior can benefit from ease of interpretability. However, predictions for specific contexts can be enhanced if a more specific set of traits is used. On the other hand, using nuances to predict behavior might yield even stronger predictive

ability (Seeboth & Mõttus, 2018), but as the number of predictors grows the interpretations
become more complex. Facets are on a middle ground between nuances and domains, in a
compromise between specificity and sensitivity in the bandwidth-fidelity dilemma. This
narrow aggregation both satisfies the specificity of predictions to concrete situations and
environments and also enhances the ease of interpretability when summarizing individual
personality characteristics.

Personality measured at the facet level has found to be a strong predictor of a large 149 number of outcomes. Satisfaction with life (SWL) is one of them. Neuroticism and Extraversion were recognised as the most important personality dimensions in the prediction 151 of subjective satisfaction (Diener, Oishi, & Lucas, 2003; Schimmack, Diener, & Oishi, 2002). 152 Lately, Schimmack, Oishi, Furr, and Funder (2004) observed that the analysis at the facet 153 level outperform the analysis at the domain level. They observed that Depression and 154 Positive Emotions / Cheerfulness explained SWL above and beyond the dimensions they 155 belong to, reaching to a 30% of explained variability of SWL. Correlations in the Schimmack 156 et al. (2004) study ranged in a longitudinal design from r = -.57 to r = -.49 for the first and 157 from r = .51 to r = .38 for the second and third. We hypothesize that the facets Confidence 158 (N2) and Positive attitude (E4) developed in our instrument will behave similarly 159

Another relevant outcome that has shown to be best predicted with personality at the 160 facet level is academic achievement. The relation of Conscientiousness with academic 161 performance has gained a stable empirical evidence, with correlations ranging from r = .20162 to r = .45 depending in sample specifity (Chamorro-Premuzic & Furnham, 2003; De Fruyt & Mervielde, 1996; Lievens, Coetsier, De Fruyt, & De Maeseneer, 2002; Noftle & Robins, 2007; O'Connor & Paunonen, 2007; S. V. Paunonen & Ashton, 2001; Poropat, 2009, 2014; D. 165 Watson & Watson, 2002). De Fruyt and Mervielde (1996) hypothesized that volitional facets 166 of Conscientiousness would be more proped to exhibit strong relations with academic 167 achievement. In this line, there is a collection of research which points at relations of GPA 168

scores with facets such as Achievement-striving (Chamorro-Premuzic & Furnham, 2003; 169 O'Connor & Paunonen, 2007 r ranging from .15 to .39; D. Watson & Watson, 2002 r = .39) 170 or Work drive (Lounsbury et al., 2002, r = .12). Nonetheless, also other Conscientiousness 171 facets more related to duties or moral driveness have been found to predict significantly GPA 172 scores, like Self-discipline (O'Connor & Paunonen, 2007, r ranging from .18 to .25; D. 173 Watson & Watson, 2002, r = .36) or Dutifulness (Chamorro-Premuzic & Furnham, 2003; 174 O'Connor & Paunonen, 2007, r ranging from .25 to .38). Moreover, narrow level analysis 175 seems to improve the predictive power of personality on academic performance, adding 176 about 10% of explained variance (Lounsbury, Steel, Loveland, & Gibson, 2004; O'Connor & 177 Paunonen, 2007; Ziegler, Danay, Schölmerich, & Bühner, 2010). Openness... 178

At the facet level, O'Connor and Paunonen (2007) found that Achievement-striving, 179 Self-discipline and Dutifulness were the best predictors of academic performance, ranging 180 from r = .15 to r = .39, from r = .18 to r = .46, and from r = .25 to r = .46 respectively. 181 In the inventory presented here, the facets Persistence (C2), Self-discipline (C3) and 182 Carefulnes (C6) are the most related with the previous. Our second hypothesis would be to 183 replicate this findings. Openness is also supposed to play an important role in predicting 184 academic achievement, however evidence has been elusive at the dimension level (Ziegler et 185 al., 2010). Openness to ideas has been found to be a facilitator towards better GPA, wih 186 correlations being found r = .22 in Dollinger and Orf (1991) and r = .15 in Ziegler et al. 187 (2010). We expect to find similar correlations with Open mindedness (O3) in our third 188 hypothesis. Often different facets within the same domain can have effects in opposite 189 directions, partially canceling out the predictive ability of the dimensions. This could be the 190 case of Openness to experience, were facet-level analysis can be unveiling. Openness to fantasy has found to yield a negative relation to GPA r = -.22 for men, whereas Aesthetics 192 yielded a similar relationship with GPA for women r = -.19 (De Fruyt & Mervielde, 1996). 193 In our case we expect that *Creativity* (O1) yield negative correlations with GPA for both 194 sexes, and the overall effect of Opennes on academic achievement being nearly zero, this will 195

be our fourth hypothesis.

Likewise, personality has proven to be a powerful predictor of laboral and educational 197 abseentism (Judge, Martocchio, & Thoresen, 1997; Salgado, 2002). Research has highlighted 198 the predictive power of personality test over the so-called integrity test when predicting 199 absences (Ones, Viswesvaran, & Schmidt, 2003). Again, most research has focused on the 200 dimensional level, although some researchers suggested that personality assessed at a 201 narrower level would improve the predictive ability of the models (Lounsbury et al., 2004; 202 Salgado, 2002). Nonetheless, few studies have explored this relationship to our knowledge, 203 being Lounsbury et al. (2004) and Judge et al. (1997) the most prominent. Judge et al. 204 (1997) reported no predictive gain when examining personality at the facet level for the 205 NEO-PI-R composites of Extraversion and Conscientiousness, whereas Lounsbury et al. 206 (2004) found a modest predictive gain of Work drive over the Big Five dimensions. Therefore, 207 and despite the conceptual expectation of facets maximizing the predictive ability of 208 personality on absecutism, evidence has manifested in favour of a dimension level analysis. 209 Our sixth hypothesis of the criterion validity section will be that *Persistence* (C2) will add a 210 significative proportion of variance explained in the model which includes the five dimensions, in line with Lounsbury et al. (2004). Furthermore, this study would help to 212 envision which facets could be involved in the relation between personality and school abseentism, as an exploratory research question. SAY SOMETHING ABOUT 214 Chamorro-Premuzic and Furnham (2003) 215

As described above, facet measures often yield scores that have stronger test-criterion correlations than their respective domain scores. However, facet scores have also been shown to be related to personality disorders. Thus, the combination of a higher fidelity along with the potential clinical relevance of facet scores might open up unique advantages for clinical research.

1.4. The Big Five and Personality Disorders

Personality disorders are steadily shifting from a categorical definition into a continua 222 conceptualization within the clinical realm. This process is not new for personality science 223 history, as the subject itself moved from a qualitatively distinct set of definitions, called 224 types, into a subset of continuous domains in which both normality and extreme tendencies were moving along, named traits. In fact, the new version of the Diagnostic and Statistical Manual of mental disorders, DSM-V, now proposes two different ways of assessing personality disorders: 1) A descriptive model of personality disorders in section II which 228 mimics the former model of assessing personality disorders and; 2) A novel trait model that 220 follows research on the personality scientific domain (In section III), which conceptualizes 230 personality disorders as extreme tendencies located in the continuum of the Big Five domains 231 and facets (American Psychiatric Association, 2013; T. A. Widiger & Mullins-Sweatt, 2009) 232

This paradigm shift in clinical assessment of personality has led to the construction of 233 the Personality Disorder Inventory (PID-5; R. F. Krueger, Derringer, Markon, Watson, & 234 Skodol, 2012), a 25-facet and five-dimension self-report inventory, with an informant-report 235 version (K. E. Markon, Quilty, Bagby, & Krueger, 2013). These five dimensions mirror the 236 Big Five domains, although with a focus on the maladaptative end of the continuum,: I) 237 Detachment (Big Five's Introversion), II) Antagonism (absence of Big Five's Agreeableness), 238 III) Disinhibition (absence of Big Five's Conscientiousness), IV) Negative affect (Big Five's 239 Neuroticism) and V) Psychoticism (Absence of Big Five's Openness). The PID-5 has shown satisfying evidences of criterion validity (... summary). However, the number of facets per domain on the PID-5 is limited.

In line with what has been stated previously for academic achievement, the
examination of facets may result in an enhancement of the specificity of assessment when
looking at the nature of PDs (L. A. Clark, 2005; Samuel & Widiger, 2008). This

improvement of specificity resulted in a predictive gain ranging from 3% to 16% when comparing facets to domains predicting PD in a study by Reynolds and Clark (2001). 247 Furthermore, the use of facets may be of extreme utility for those PD whose personality 248 profile is less clear at the domain level. As Saulsman and Page (2004) pointed out, 249 Schizotypal and Obsessive-Compulsive disorders are examples of PD which are not well 250 covered by Big Five domains. A reason for it may be found in a pattern inconsistency of 251 facets within the same dimension or in a lack of coverage for essential characteristics of the 252 PD. For example, aberrant cognitions are essential characteristics of schizotypal disorder and 253 are not covered by some instrument's facets like the NEO-PI-R (Samuel & Widiger, 2008; 254 Saulsman & Page, 2004). Likewise, the expected high scores on warmth and low scores on 255 assertiveness could mask the effects of extraversion when predicting Dependent Personality 256 Disorder, following the theoretical correspondence between PD and Big Five facets proposed by Costa Jr. and Widiger (1994). Moreover, the PID-5 has prompted the elaboration of a 258 number of Five Factor Model Personality Disorders (FFMPD) scales to maximize the facet coverage in relation to specific PDs (R. M. Bagby & Widiger, 2018). 260

Facet analysis and dedicated Big Five questionnaires have been used to solve issues like those mentioned in the last paragraph. We propose to base such research on a broader facet basis. To this end we suggest a general instrument to cover a broad number of facets which could aim for fine grained assessments.

$_{65}$ This study

We present in this paper an instrument for personality assessment which aims to cover the need for an internationally usable, open source, and differentiated measure at the facet level. Two studies are presented, for each one inspects the factor structure of the instrument in a different sample drawn from a different culture (American vs. German). Measurement invariance across samples will be examined. Internal consistency and test-criterion

correlations will be estimated for the scores. To sum up, the aim for this research project was to provide an instrument that can be used in non-clinical but also in clinical research which emphasizes the facet level of the Big Five.

274 Methods

Two different studies are presented in this work. The first study uses a sample drawn 275 from the USA bachelor student population. The aim was to detect and confirm a model that maximizes the facet space below the Big Five domains. Exploratory factor analysis (EFA) 277 was used to identify the number of facets per domain. A confirmatory factor analysis (CFA) 278 per facet was specified in order to confirm the item - facet relationship. Reliability measures 279 for the facets and test-criterion correlations will be computed to achieve evidences of 280 reliability and criterion validity. Finally, an exploratory structural equation model (ESEM) 281 was applied to test a full model in which the facets serve as indicators of the Big Five 282 domains. ESEM has gained reputation in the personality field, where the independent cluster 283 model may not capture the complexity of the constructs measured (Marsh et al., 2010). 284

The second sample is a sample representative for the German speaking population of
Germany, Austria and Switzerland. The aim for the second study was to replicate the
structure found in study one, plus assess the degree of measurement invariance of the
proposed model.

Study 1 - US-American Sample

Participants. The sample consisted of 722 American undergraduate students

(59.30% male) who participated voluntarily. The mean age was 21.60 years (SD = 5.90).

Students were emailed a link to a computerized assessment battery that included the IPIP

items as well as several other tests not reported in this paper. The data set was randomly

split into two equally sized samples. Both samples were matched in relation to missing values, outliers and extreme values. In Sample 1 the mean age was 21.80 years (SD= 6.30), in Sample 2 the mean age was 21.50 years (SD=5.60).

297 Measures

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Items from the International Personality Item Pool (IPIP). Altogether, 525 items from the International Personality Item Pool (IPIP) were used to measure

Neuroticism, Extraversion, Openness (to experience), Agreeableness and Conscientiousness.

The IPIP is an open source database of personality items, which was launched in 1996, and contains over 2000 items (L. R. Goldberg et al., 2006). Participants were asked to rate themselves on typical behaviors or reactions on a 5-point Likert scale, ranging from 1 ("Not all like me") to 5 ("Very much like me").

The item selection was part of a different project and the procedure has been explained in detail in the appendix of a study by MacCann, Duckworth, and Roberts (2009). That study also contains part of the sample used here. However, the current data set contains more participants.

Satisfaction With Life (SWL). Measured with a 5 item composite defined in

Diener, Emmons, Larsen, and Griffin (1985), answered in a 7 point Likert-type scale ranging

from 1 (strongly disagree) to 7 (strongly agree). The items are: a) "In most ways my life is

close to ideal", b) "The conditions of my life are excellent", c) "I am satisfied with my life",

d) "So far I have gotten the important things in my life", and e) "If I could live my life over,

I would change almost nothing". Psychometric properties have been reported excellent. (In

which moment have SWLS been measured? just at the same time point than personality?)

GPA. Grade Point Averages measured in high school, university, and in cc.

317 **SAT.**

318 **ACT.**

Absences from class. As a behavioral measure absence from class was asked to report from subjects. ABS2 ABS4 what are the differences?

$\mathbf{2.1.3.}$ Procedure

EFA with subsample 1. To determine the number of possible facets per domain 322 Velicer (1976) Minimum Average Partial (MAP) method and Horn (1965) parallel analysis 323 (PA) were employed for every domain. Based on these results an Exploratory Factor 324 Analysis (EFA) was calculated for each domain via Mplus using a geomin rotation (Quelle) 325 and a Maximum Likelihood estimator (ML). The decision for the preferred number of facets 326 per domain was based partly on comparing model fits (CFI, RMSEA, SRMR). More 327 importantly though was the interpretability of the facet solution. To this end, facet solutions 328 from other personality measures were looked and compared to the found facet structure. If 329 there were important parts missing to present the domain with regards to content, new 330 facets were added a posteriori.

CFA and ESEM with subsample 1. To confirm the structure of facets the EFAs delivered, multiple CFAs were calculated via Mplus. In a first step, measurement models were estimated for each of the facets. To obtain balance between the facets, the items were reduced to five per facet based on item content and loading pattern. In a second step, the estimations for the measurement models on facet levels were repeated via CFA. For both steps estimators were WLSMV (Weighted Least Squares adjusted for Means and Variances). Aim was to ensure an optimal breadth and sufficient reliability. In a final model, all five domain structural models were integrated using ESEM (Asparouhov & Muthén, 2009).

Marsh et al. (2010) could show that ESEM fits personality data better and results in

substantially more differentiated factors than CFA. All facets were allowed to load on all 341 domains. If there would show up facets that do not significantly load on the intended 342 domain, this facets would get eliminated subsequently. The estimators used were ML 343 !WLSMV?, factor scores from the facet CFAs were used as indicators and the rotation was 344 oblique (using Geomin). Model fit was determined based on the guide lines by Hu and 345 Bentler (1999) as well as Beauducel and Wittmann (2005). Consequently, to consider a good 346 fit of a proposed model, the Comparative Fit Index (CFI) should be at or over .95, the 347 Standardized Root Mean Squared Residual (SRMR) smaller than .08 and the Root Mean Square Error of Approximation (RMSEA) smaller than .06. For the ESEM models we 349 compared our results with the findings by Marsh et al. (2010). 350

Reliability

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Chronbach's α and McDonald's ω will be calculated for each facet to provide measures 352 of internal consistency. 353

Criterion validity evidence. To examine the nomological structure of the facets and domains to external constructs, a set of linear models and correlations were fitted. We 355 describe in this section the methods used to test the set of hypothesis described in the 356 introduction.

To explore the first hypothesis we fitted a linear model that included N2 and E4 as 358 predictors and SWL as citerion. Furthermore we fitted a stepwise regression using all the 359 facets to find other important relations. In a third model we compared the power of E4 and 360 N2 predicting SWL against their respective domains, Extraversion and Neuroticism. 361

The second to sixth hypothesis involves academic achievement as criterion variable. 362 The second and third hypothesis will be explored via inspection of the correlation matrix 363 with the Big Five dimensions. The fourth hypothesis will be explored by means of a stepwise

regression were all the Conscientiousness facets will form the initial set of predictors. The
fifth hypothesis will be explored with a simple regression of O3 on high school grades. The
sixth hypothesis will be explored by inspecting the correlation of the facets forming
Openness with academic achievement.

The seventh hypothesis involves the relations between educational absentism and personality. To explore whether C2 would add predictive power to the five dimensions on predictive absences, we will fit a two step regression. Finally, a stepwise regression of all facets on school absentism will be fitted to explore predictions at the facet level.

Results

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Results of EFA. In *Table 2* model fits for the chosen facet model for each domain are shown, as well as Eigenvalues and results from MAP and PA test. To ensure the homogeneity of the facets and to reduce the risk of cross domain loadings, items with factor loadings less than .30 were eliminated. This was only done when item content was also judged as being non-central to the domain in question (Ziegler et al., 2014).

< Table 2 here>

According to the exploratory model, Agreeableness consists of eight facets after two facets were eliminated due to weakly loading and inconsistent items. The remaining facets were named Appreciation, Integrity, Low Competitiveness, Readiness to Give Feedback, Search for Support, Good Faith, Genuineness and Altruism.

Conscientiousness consists of nine facets after one facet with item factor loadings less
than .30 was excluded, they are: Dominance, Persistence, Self-discipline, Task planning,
Goal orientation, Carefulness, Orderliness, Wish to work to capacity and Productivity.

Extraversion is formed by nine facets. A new facet (Energy) was added in order to tap
better the physical component of Extraversion, which was missing in the eight facet solution.
The facets are Sociability, Readiness to take risks, Wish for affiliation, Positive attitude,
Forcefulness, Communicativeness, Humor, Conviviality and Energy.

Neuroticism (interpreted here as emotional stability) consists of seven facets. One facet
was dropped due to poor interpretability, and was therefore not included in the subsequent
analyses. The final set of facets are named Equanimity, Confidence, Carefreeness, Mental
balance, Drive, Emotional robustness and Self-attention.

Openness to experience comprises nine facets. One facet was identified as a method
factor and eliminated, because it solely contained negatively formulated items and no
coherent underlying trait could be identified. Furthermore another facet (Intellect) was
added, because the remaining facets lacked an intellectual content. The facets of Openness
are named Creativity, Wish for variety, Open-mindedness, Interest in reading, Artistic
interests, Wish to analyze, Willingness to learn, Sensitivity and Intellect.

The items to each facet are listed in the appendix (A).

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Results of CFA and ESEM. All measurement models for the facets fitted well, results are summarized in *Table 3*. In this table both models with five items only and models with all items are presented with their respective model fit. The 5-item facets normally outperform the models including all items regarding model fit.

< Table 3 here caption="Model fit for each facet")>

The ESEM of the final model with all five domains yielded an acceptable fit (Marsh et al., 2010): CFI = .87, RMSEA = .072, SRMR = .036. As it can be seen in *Table 4* nearly all facets loaded significantly on their intended domain. Some cross loadings emerged as is

410 typical for ESEM procedures.

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411 < Table 4 here caption="ESEM factor scores")>
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Reliability. Reliabilities for the 5 item facets were calculated with α and ω estimates.

Agreeableness showed a mean α of 0.68, and a mean ω of 0.69. Conscientiousness' mean $\alpha = 0.68$, and mean $\omega = 0.70$. Openness' mean $\alpha = 0.76$, and mean $\omega = 0.77$. Neuroticism mean $\alpha = 0.68$, and mean $\omega = 0.69$. TExtraversion's mean $\alpha = 0.72$, and mean $\omega = 0.74$.

Criterion validity evidence. Our first hypothesis concerning external criteria involved Confidence (N2) and Positive attitude (E4), as the most important predictors of SWL. Our results, in line with Schimmack et al. (2004), showed that both facets were the most salient predictors of the factor in a stepwise regression, and that they explained 0.34% of the factor. The final model reached with the stepwise procedure explained 0.41% of the factor variance. N2 and E4 outperformed Neuroticism and Extraversion, who failed to be significant predictors in a stepwise model which included these four independent variables.

Our second hypothesis was set to replicate the findings of the meta-analysis by 423 O'Connor and Paunonen (2007) with our instrument. In the mentioned study 424 Conscientiousness showed the higher correlations with school grades with a corrected $\rho = .24$, 425 and the facets Achievement striving, Self-discipline and Dutifulnes were the most important 426 predictors at a narrower level. Our results are in line with those reported by O'Connor and 427 Paunonen (2007). Conscientiousness correlated r = 0.23 with high school grades. At the 428 facet level, C2 correlated r = 0.12, C3 r = 0.13 and C6 r = 0.16 with high school grades. 429 On the other hand, Openness r = 0.15 and Agreebleness r = 0.20 at the dimension level; 430 and Goal orientation (C5) r = 0.22 at the facet level also showed high correlations with 431 grades, a result not found in O'Connor and Paunonen (2007). 432

The fourth hypothesis involves *Open mindedness* (O3) as an Opennesss facilitator

towards achieving higher grades. We have found a correlation of r = 0.14 between this facet and high school grades, in line with the findings of Ziegler et al. (2010). Furthermore, we expected some inverse correlations between the facets of Openness, and as a result a near zero correlation with the dimension. We haven't found any Openness facet with a significant inverse correlation with high school grades, and as indicated previously our results don't support a zero correlation of the dimension with high school grades.

Another relevant outcome that has shown to be best predicted with personality at the 440 facet level is academic achievement. The relation of Conscientiousness with school grades has gained a stable empirical evidence, widely inspected in the meta-analysis by O'Connor and 442 Paunonen (2007) were the mean ρ was .24. At the facet level, O'Connor and Paunonen 443 (2007) found that Achievement-striving, Self-discipline and Dutifulness were the best 444 predictors of academic performance, ranging from r = .15 to r = .39, from r = .18 to r = .18445 .46, and from r = .25 to r = .46 respectively. In the inventory presented here, the facets 446 Persistence (C2), Self-discipline (C3) and Carefulnes (C6) are the most related with the 447 previous. Our second hypothesis would be to replicate this findings. Openness is also 448 supposed to play an important role in predicting academic achievement, however evidence 449 has been elusive at the dimension level (Ziegler et al., 2010). Openness to ideas has been 450 found to be a facilitator towards better GPA, wih correlations being found r = .22 in 451 Dollinger and Orf (1991) and r = .15 in Ziegler et al. (2010). We expect to find similar 452 correlations with Open mindedness (O3) in our third hypothesis. Often different facets 453 within the same domain can have effects in opposite directions, partially canceling out the 454 predictive ability of the dimensions. This could be the case of Openness to experience, were facet-level analysis can be unveiling. Openness to fantasy has found to yield a negative relation to GPA r = -.22 for men, whereas Aesthetics yielded a similar relationship with 457 GPA for women r = -.19 (De Fruyt & Mervielde, 1996). In our case we expect that 458 Creativity (O1) yield negative correlations with GPA for both sexes, and the overall effect of 459 Opennes on academic achievement being nearly zero, this will be our fourth hypothesis.

The second to sixth hypothesis involves academic performance as criterion variable.

The second and third hypothesis will be explored via inspection of the correlation matrix

with the Big Five dimensions. The fourth hypothesis will be explored by means of a stepwise

regression were all the Conscientiousness facets will form the initial set of predictors. The

fifth hypothesis will be explored with a simple regression of O3 on high school grades. The

sixth hypothesis will be explored by inspecting the correlation of the facets forming

Openness with academic achievement.

Bivariate correlations of the facets with the external criteria are shown in $Table\ 5$. The facets N2 (Confidence, r=0.53) and E4 (Positive attitude, r=0.49) show the highest correlations with life satisfaction respectively. The SWL scale was also correlated with facets such as N4 (Mental balance, r=0.25) and N5 (Drive, r=0.27) in the Neuroticism / Emotional stability realm. C5 (Goal orientation, r=0.28) and C2 (Persistence, r=0.27) in Conscientiousness. E1 (Sociability, r=0.26) and E9 (Energy, r=0.25) in the Extraversion domain. O9 (Intellect, r=0.24) and O3 (Open-mindedness, r=0.22) in Openness and A6 (Good faith, r=0.25) in the Agreeableness domain.

Regarding the academic performance criteria (GPA, SAT and ACT), the Openness facets O4 (Interest in reading), O6 (Wish to analyze) and O9 (Intellect) yield the most consistent correlations. Interestingly, O4 correlates directly with the reading subtests of SAT (r = 0.25) and ACT (r = 0.2) and shows a negligible relationship with the math (r = 0.06 / 0.08) subscales.

Furthermore, the global GPA scale (assuming cc is for global) correlated with several Conscientiousness and Agreeableness facets. C9 (Productivity, r = 0.4), C5 (Goal orientation, r = 0.38) and C4 (Task planning, r = 0.35) yield the higher correlations with overall GPA respectively in the Conscientiousness domain, and A1 (Appreciation, r = 0.32) in the Agreeableness domain.

Concerning the behavioral criteria controlled in this study, absence from class, the meaningful correlations unsurprisingly loaded within the facets at the Conscientiousness domain. Particularly the facets C4 (Task planning, r = -0.22 / -0.26), C7 (C6: Carefulness, r = -0.21 / -0.21) and C9 (Productivity, r = -0.21 / -0.24) were the most robust indicators of this criteria.

491 < Table 5 here caption="Criterion correlations" >

492 Study 2 – German Sample

Participants. The representative sample consisted of 387 German speakers (49.10% male) with a mean age of 45.60 years (SD = 17.50). The data was collected in a test center.

Measures. The five items per facet derived from Study 1 were translated and
back-translated by bilingual experts, creating a German version of the measure used there.
The translated items can be found in appendix B.

498 Procedure

Step 1 – Examining the structure. To check the facet structure Study 1
delivered, multiple confirmatory factor analyses were calculated via Mplus following an
analogue procedure to Study 1. First, measurement models were estimated for all facets,
using WLSMV as the estimator. Model fit was determined based on the guide lines
mentioned above. In a final model, all five domain structural models were integrated using
ESEM.

Step 2 – Testing for measurement invariance. In a next step, measurement invariance between German and US samples was examined. We followed the procedure

suggested by Sass (2011) and tested configural, factorial and strong factorial invariance. The 507 cutoffs suggested by Chen (2007) were applied to compare model fits. According to this 508 configural measurement invariance can be assumed when the same item is associated with 509 the same factor in each domain, while the factor loadings can differ. If the factor loadings of 510 each item would not differ between the samples, factorial measurement invariance can be 511 assumed. Strong factorial measurement invariance can be assumed when on top of that the 512 intercepts of each item are equal. The limit to factorial measurement invariance was set to Δ 513 CFI < .01, Δ RMSEA < .015 and Δ SRMR < .03, at which the limit to strong factorial 514 measurement invariance was set to Δ CFI < .01, Δ RMSEA < .015, Δ SRMR < .01 as 515 suggested by Chen (2007). 516

${f Results}$

Results of CFA. The measurement models of the American sample were replicated for the reduced number of items per facet. Model fits can also be seen in *Table 3*. The ESEM with all five domains showed a relatively good fit to the data with CFI = .82, RMSEA = .078, SRMR = .044. *Table 6* shows the ESEM factor loadings for the German sample. All facets loaded significantly on their intended domain.

Results of MI. For analyzing measurement invariance the latest facet model
structure (with additional facets) was taken. The results are shown in Table 7. Configural
measurement invariance could be shown for the facets Appreciation of others,
Superiority/Grandiosity, Need to be liked, Crybabiness, Manipulation, Altruism (facets of
Agreeableness), Perseverance, Task Planning, Goal-orientation/Achievement striving,
Preferred Load, Procrastination (facets of Conscientiousness), Assertiveness,
Sociability/Gregariousness, Activity (facets of Extraversion), Irritability, Self-serving
Attention (facets of Neuroticism), Self-attributed Inginuity, Openness to actions and

activities, Openmindedness/Judgement, Love of Learning, Openness to feelings and Intellect (facets of Openness).

Factorial measurement invariance could be shown for the facets Meanness, Trust

(facets of Agreeableness), Control of others, Lack of (Self-) Control, Deliberation/Caution,

Lack of Tidiness/Order (facets of Conscientiousness), Sensation Seeking, Reclusiveness,

Emotionality, Humor (facets of Extraversion), Depression, Anxiety, Self-assuredness,

Lethargia, Sentimentality (facets of Neuroticism), Openness to reading, Openness to arts and

Need for cognition (facets of Openness).

The only facet with strong factorial measurement invariance was Shyness, a facet of Extraversion

Discussion

We have presented in this work an open-access instrument for personality assessment within the Big Five framework, which showed evidences of factorial validity in two different cultures and maximized the space set of facets encompassed. With a modest number of items (202) by comparison with the most influential Big Five inventories presented in *Table* 1, we have reached to a large set of facets which mostly show a robust factorial validity in both studies, as shown in *Table* 3.

The Big Five solution has been recognized as the most replicable model for personality inventories, reaching a hallmark of consensus in personality science for the last decades.

However, some researchers have pointed out that while the Big Five has repeatedely been found when fitting EFA to personality data, its replicability under CFA procedures has been more elusive (R. R. McCrae, Zonderman, Costa, Bond, & Paunonen, 1996). The constriction of the common independent cluster solution, where cross-loadings are restricted to zero, may

suppose a rather strong assumption for personality trait inventories (Marsh et al., 2010).

The idea of facets, or habits, being influenced by more than one domain can definitely make
some sense. ESEM helps overcoming this assumption and provides a measure about how well
the Big Five solution adjusts to the data. Using this procedure, the degree of integration of
our proposed set of facets to the Big Five factor solution has been solid enough according to
the cut-off values proposed by Marsh et al. (2010). The number of significant cross-loadings
in the ESEM models has not been large either, advocating a good discriminant validity.

The instrument presented in this work covers all the "core" facets proposed by 561 Christopher J. Soto and John (2009), either directly or indirectly. The *Energy* construct in 562 Extraversion is literally covered by a three-item facet in our instrument, whereas the 563 Assertiveness construct has been tapped by items belonging to the Wish for affiliation, Communicativeness and Conviviality facets. Altruism is directly reflected in a five-item facet, 565 while the Compliance construct is reflected by our Good faith facet. The Order and Self-discipline constructs proposed by Christopher J. Soto and John (2009) are mirrored by dedicated facets in our instrument. The Anxiety and Depression constructs are mirrored by the facets Mental balance and Emotional robustness, respectively. For the Openess dimension, the Aesthetic contruct is covered by our facet Artistic interest, while the Ideas 570 construct has been reflected by both the Open-mindedness and the Wish to analyze facets. The two-per-facet components proposed by DeYoung et al. (2007) were also being tapped by the set of facets in our inventory. 573

The instrument covers most of facets proposed by the most influential Big Five measures as seen in *Table 1*. The most salient differences are related to the HEXACO model, which entails a six factor solution with a slightly different theoretical conceptualization (K. Lee & Ashton, 2006). Most notably

Although these facets are not being covered directly in our inventory, components of facets from distinct domains in our model retain a glimpse of the missing facets. This

underlies the importance of allowing cross-loadings for trait personality data. Let's use the 580 example of *Patience*, a facet proposed in the HEXACO model for the Agreeableness domain 581 which is not covered in our instrument, nor in the other three Big Five inventories which 582 have been revised. Although patience, there is a notion of a patient trait within the 583 Self-discipline facet in the Conscientiousness domain, specially with items such as "I rush 584 into things" or "I act impulsively when something is bothering me" (See appendix A). In 585 fact, Self-discipline has important cross-loadings with Agreeableness in both samples ($\lambda =$ 586 .256 in the USA sample and $\lambda = .341$ in the german sample). 587

In addition we included even more facets.

589

In addition, evidences for external criteria validity were attained.

We have collected some criterion validity evidences. Like bla bla bla. Nonetheless the multi - facetted nature of the instrument makes forthcoming evidences for criterion and predictive validity promising.

One limitation is the sample used. Students are not a representative population of society and results may not be generalized.

Future directions are to provide a tool with the subset of items for public use. Gather community sample, from more cultures and test the extent of the universality of the instrument. And use the instrument to predict important life outcomes so the links between specific behaviors and facets become richer.

Remove this page. This is used to include the tables' references into the bibliography. 599 Brick and Lewis (2014); Gaughan, Miller, and Lynam (2012); Leone, Chirumbolo, and 600 Desimoni (2012); Mcabee, Oswald, and Connelly (2014); Gaughan, Miller, Pryor, and Lynam 601 (2009); Noftle and Shaver (2006); R. M. Bagby, Taylor, and Parker (1994); Schimmack, Furr, 602 and Funder (1999); Wakabayashi, Baron-Cohen, and Wheelwright (2006); Shaver and 603 Brennan (1992); Ruiz, Pincus, and Dickinson (2003); Mccrae, Kurtz, Yamagata, and 604 Terracciano (2011); Rosander, Bäckström, and Stenberg (2011); K. K. McAdams and 605 Donnellan (2009); Siddiqui (2011); Hagger-Johnson and Whiteman (2007) 606

Ziegler et al. (2014)

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References

Allport, G. W., & Odbert, H. S. (1936). Trait-names: A psycho-lexical study. *Psychological Monographs*, 47(1), i–171. doi:10.1037/h0093360

- American Psychiatric Association. (2013). Diagnostic and statistical manual of mental disorders (5th ed.).
- Asparouhov, T., & Muthén, B. (2009). Exploratory structural equation modeling (Vol. 16, pp. 397–438). doi:10.1080/10705510903008204
- Bagby, R. M., & Widiger, T. A. (2018). Five factor model personality disorder scales: An introduction to a special section on assessment of maladaptive variants of the five factor model. *Psychological Assessment*, 30(1), 1–9. doi:10.1037/pas0000523
- Bagby, R. M., Taylor, G. J., & Parker, J. D. (1994). The twenty-item Toronto Alexithymia scale-II. Convergent, discriminant, and concurrent validity. *Journal of Psychosomatic Research*, 38(1), 33–40. doi:10.1016/0022-3999(94)90006-X
- Beauducel, A., & Wittmann, W. (2005). Simulation study on fit indices in confirmatory

 factor analyses based on data with slightly distorted simple structure. Structural

 Equation Modeling, 12, 41–75. doi:10.1207/s15328007sem1201
- Borgatta, E. (1964). The structure of personality characteristics. *Behavioral Science*, 9(1), 8–17. doi:10.1007/BF01358190
- Brick, C., & Lewis, G. J. (2014). Unearthing the "Green" Personality: Core Traits Predict Environmentally Friendly Behavior. *Environment and Behavior*, 48(5), 635–658. doi:10.1177/0013916514554695
- ⁶²⁹ Cattell, R. B. (1956). Second-order personality factors in the questionnaire realm. *Journal*

```
of Consulting Psychology, 20(6), 411–418. doi:10.1037/h0047239
```

- 631 Chamorro-Premuzic, T., & Furnham, A. (2003). Personality predicts academic performance:
- Evidence from two longitudinal university samples.
- doi:10.1016/S0092-6566(02)00578-0
- 634 Chen, F. F. (2007). Sensitivity of goodness of fit indexes to lack of measurement invariance.
- 635 Structural Equation Modeling, 14(3), 464–504. doi:10.1080/10705510701301834
- 636 Clark, L. A. (2005). Temperament as a unifying basis for personality and psychopathology.
- Journal of Abnormal Psychology, 114(4), 505–521. doi:10.1037/0021-843X.114.4.505
- 638 Costa Jr., P. T., & Widiger, T. A. (1994). A description of the DSM-III-R and DSM-IV
- personality disorders with the five-factor model of personality. Personality Disorders
- and the Five-Factor Model of Personality., (January), 41–56. doi:10.1037/10140-003
- ⁶⁴¹ Costa, P. T., & McCrae, R. R. (1995). Domains and facets: hierarchical personality
- assessment using the revised NEO personality inventory. Journal of Personality
- Assessment, 64(1), 21-50. doi:10.1207/s15327752jpa6401_2
- De Fruyt, F., & Mervielde, I. (1996). Personality and interests as predictors of educational
- streaming and achievement. European Journal of Personality, 10(5), 405–425.
- doi:10.1002/(SICI)1099-0984(199612)10:5<405::AID-PER255>3.0.CO;2-M
- 647 DeYoung, C. G., Quilty, L. C., & Peterson, J. B. (2007). Between Facets and Domains: 10
- Aspects of the Big Five. Journal of Personality and Social Psychology, 93(5),
- 880-896. doi:10.1037/0022-3514.93.5.880
- 650 Diener, E., Emmons, R. A., Larsen, R. J., & Griffin, S. (1985). The Satisfaction With Life
- Scale. Journal of Personality, 49(1), 71–75. doi:10.1207/s15327752jpa4901
- Diener, E., Oishi, S., & Lucas, R. E. (2003). Personality, culture, and subjective well-being.

```
doi:10.1146/annurev.psych.54.101601.145056
```

- Digman, J. M. (1990). Personality Structure: Emergence of the Five-Factor Model. Annual

 Review of Psychology, 41(1), 417–440. doi:10.1146/annurev.ps.41.020190.002221
- Dollinger, S. J., & Orf, L. A. (1991). Personality and performance in "personality":
- 657 Conscientiousness and openness. Journal of Research in Personality, 25(3), 276–284.
- doi:10.1016/0092-6566(91)90020-Q
- Fiske, D. W. (1949). Consistency of the factorial structures of personality ratings from
- different sources. Journal of Abnormal and Social Psychology, 44(3), 329–344.
- doi:10.1037/h0057198
- 662 Galton, F. (1884). The Measurement of Character. doi:10.1037/11352-058
- Gaughan, E. T., Miller, J. D., & Lynam, D. R. (2012). Examining the Utility of General
- Models of Personality in the Study of Psychopathy: A Comparison of the
- HEXACO-PI-R and NEO PI-R. Journal of Personality Disorders, 26(4), 513–523.
- doi:10.1521/pedi.2012.26.4.513
- Gaughan, E. T., Miller, J. D., Pryor, L. R., & Lynam, D. R. (2009). Comparing two
- alternative measures of general personality in the assessment of psychopathy: A test
- of the NEO PI-R and the MPQ. Journal of Personality, 77(4), 965–995.
- doi:10.1111/j.1467-6494.2009.00571.x
- Goldberg, L. R., Johnson, J. A., Eber, H. W., Hogan, R., Ashton, M. C., Cloninger, C. R., &
- Gough, H. G. (2006). The international personality item pool and the future of
- public-domain personality measures. Journal of Research in Personality, 40(1),
- 84–96. doi:10.1016/j.jrp.2005.08.007
- Hagger-Johnson, G. E., & Whiteman, M. C. (2007). Conscientiousness facets and health

```
behaviors: A latent variable modeling approach. Personality and Individual

Differences, 43(5), 1235–1245. doi:10.1016/j.paid.2007.03.014
```

- Horn, J. L. (1965). A rationale and test for the number of factors in factor analysis. Psychometrika, 30(2), 179–185. doi:10.1007/BF02289447
- Hu, L. T., & Bentler, P. M. (1999). Cutoff criteria for fit indexes in covariance structure
 analysis: Conventional criteria versus new alternatives. Structural Equation Modeling,
 6(1), 1–55. doi:10.1080/10705519909540118
- Judge, T. A., Martocchio, J. J., & Thoresen, C. J. (1997). Five-Factor Model of Personality
 and Employee Absense. Journal of Applied Psychology, 82(5), 11. Retrieved from
 c:{\%}5CDocuments and Settings{\%}5Ce8902872{\%}5CDesktop{\%}5Cdata
 disk{\%}5CLibrary{\%}5CCURRENT{\%}5CEndNote{\%}5CCATALOGUED +
 LINKED{\%}5CJudgeetal1997.pdf
- Krueger, R. F., Derringer, J., Markon, K. E., Watson, D., & Skodol, A. E. (2012). Initial
 construction of a maladaptive personality trait model and inventory for DSM 5
 Initial construction of a maladaptive personality trait model and inventory for DSM-5.
 Psychological Medicine, 42(09), 1872–1890. doi:10.1017/S0033291711002674
- Lee, K., & Ashton, M. C. (2006). Further assessment of the HEXACO personality inventory:

 Two new facet scales and an observer report form. *Psychological Assessment*, 18(2),

 182–191. doi:10.1037/1040-3590.18.2.182
- Lee, K., & Ashton, M. C. (2016). Psychometric Properties of the HEXACO-100.

 Assessment, 1-15. doi:10.1177/1073191116659134
- Leone, L., Chirumbolo, A., & Desimoni, M. (2012). The impact of the HEXACO personality model in predicting socio-political attitudes: The moderating role of interest in politics. Personality and Individual Differences, 52(3), 416–421.

doi:10.1016/j.paid.2011.10.049

- Lievens, F., Coetsier, P., De Fruyt, F., & De Maeseneer, J. (2002). Medical students'

 personality characteristics and academic performance: A five-factor model perspective.

 Medical Education, 36 (11), 1050–1056. doi:10.1046/j.1365-2923.2002.01328.x
- Lounsbury, J. W., Steel, R. P., Loveland, J. M., & Gibson, L. W. (2004). An investigation of personality traits in relation to adolescent school absenteeism. *Journal of Youth and Adolescence*, 33(5), 457–466. doi:10.1023/B:JOYO.0000037637.20329.97
- Lounsbury, J. W., Sundstrom, E., Loveland, J. L., & Gibson, L. W. (2002). Broad versus

 narrow personality traits in predicting academic performance of adolescents. *Learning*and Individual Differences, 14(1), 67–77. doi:10.1016/j.lindif.2003.08.001
- MacCann, C., Duckworth, A. L., & Roberts, R. D. (2009). Empirical identification of the
 major facets of Conscientiousness. Learning and Individual Differences, 19(4),
 451–458. doi:10.1016/j.lindif.2009.03.007
- Markon, K. E., Quilty, L. C., Bagby, R. M., & Krueger, R. F. (2013). The Development and
 Psychometric Properties of an Informant-Report Form of the Personality Inventory
 for DSM-5 (PID-5). Assessment, 20(3), 370–383. doi:10.1177/1073191113486513
- Marsh, H. W., Lüdtke, O., Muthén, B., Asparouhov, T., Morin, A. J., Trautwein, U., &
 Nagengast, B. (2010). A New Look at the Big Five Factor Structure Through
 Exploratory Structural Equation Modeling. Psychological Assessment, 22(3), 471–491.
 doi:10.1037/a0019227
- Mcabee, S. T., Oswald, F. L., & Connelly, B. S. (2014). Bifactor Models of Personality and
 College Student Performance: A Broad Versus Narrow View. European Journal of

```
Personality, 28(6), 604-619. doi:10.1002/per.1975
```

- McAdams, D. P., & Pals, J. L. (2006). A new Big Five: Fundamental principles for an integrative science of personality. *American Psychologist*, 61(3), 204–217.

 doi:10.1037/0003-066X.61.3.204
- McAdams, K. K., & Donnellan, M. B. (2009). Facets of personality and drinking in first-year college students. *Personality and Individual Differences*, 46(2), 207–212.

 doi:10.1016/j.paid.2008.09.028
- Mccrae, R. R., Kurtz, J. E., Yamagata, S., & Terracciano, A. (2011). Internal consistency,
 retest reliability and their implications for personality Scale Validity. *Personality and*Social Psychological Bulletin, 15(1), 28–50. doi:10.1177/1088868310366253.Internal
- McCrae, R. R., Zonderman, A. B., Costa, P. T., Bond, M. H., & Paunonen, S. V. (1996).
 Evaluating replicability of factors in the tevised NEO personality inventory:
 Confirmatory factor analysis versus procrustes rotation. Journal of Personality and
 Social Psychology, 70(3), 552–566. Retrieved from http://www.sciencedirect.com/
 science/article/B6X01-46SGF6X-B/2/cfbcc79b23f57818759b3ae2b7f949b5
- Noftle, E. E., & Robins, R. W. (2007). Personality Predictors of Academic Outcomes: Big

 Five Correlates of GPA and SAT Scores. Journal of Personality and Social

 Psychology, 93(1), 116–130. doi:10.1037/0022-3514.93.1.116
- Noftle, E. E., & Shaver, P. R. (2006). Attachment dimensions and the big five personality traits: Associations and comparative ability to predict relationship quality. *Journal*of Research in Personality, 40(2), 179–208. doi:10.1016/j.jrp.2004.11.003
- Norman, W. T. (1967). 2800 Personality Trait Descriptors Normative Operating

- Characteristics for a University Population, 1–279.
- Ones, D. S., Viswesvaran, C., & Schmidt, F. L. (2003). Personality and absenteeism: a meta
- analysis of integrity tests. European Journal of Personality, 17(S1), S19–S38.
- doi:10.1002/per.487
- Ozer, D. J., & Benet-Martínez, V. (2006). Personality and the Prediction of Consequential
- Outcomes. Annual Review of Psychology, 57(1), 401–421.
- doi:10.1146/annurev.psych.57.102904.190127
- O'Connor, M. C., & Paunonen, S. V. (2007). Big Five personality predictors of
- post-secondary academic performance. Personality and Individual Differences, 43(5),
- 971–990. doi:10.1016/j.paid.2007.03.017
- Paunonen, S. V., & Ashton, M. C. (2001). Big Five Predictors of Academic Achievement.
- Journal of Research in Personality, 35(1), 78–90. doi:10.1006/jrpe.2000.2309
- Poropat, A. E. (2009). A Meta-Analysis of the Five-Factor Model of Personality and
- Academic Performance. Psychological Bulletin, 135(2), 322–338.
- doi:10.1037/a0014996
- Poropat, A. E. (2014). A meta-analysis of adult-rated child personality and academic
- performance in primary education. British Journal of Educational Psychology, 84(2),
- 761 239–252. doi:10.1111/bjep.12019
- Reynolds, S. K., & Clark, L. A. (2001). Predicting dimensions of personality disorder from
- domains and facets of the Five-Factor Model. Journal of Personality, 69(2), 199–222.
- doi:10.1111/1467-6494.00142
- Roberts, B. W., Kuncel, N. R., Shiner, R., Caspi, A., & Goldberg, L. R. (2007). The Power
- of Personality. Perspectives on Psychological Science, 2(4), 313–345.

```
doi:10.1111/j.1745-6916.2007.00047.x
```

- Rosander, P., Bäckström, M., & Stenberg, G. (2011). Personality traits and general intelligence as predictors of academic performance: A structural equation modelling approach. Learning and Individual Differences, 21(5), 590–596.

 doi:10.1016/j.lindif.2011.04.004
- Ruiz, M. A., Pincus, A. L., & Dickinson, K. A. (2003). NEO PI-R predictors of alcohol use and alcohol-related problems. *Journal of Personality Assessment*, 81(3), 265–270. doi:10.1207/S15327752JPA8103
- Salgado, J. F. (2002). The Big Five Personality Dimensions and Counterproductive
 Behaviors. International Journal of Selection and Assessment, 10(1&2), 117–125.
 doi:10.1111/1468-2389.00198
- Samuel, D. B., & Widiger, T. A. (2008). A meta-analytic review of the relationships between
 the five-factor model and DSM-IV-TR personality disorders: A facet level analysis.

 Clinical Psychology Review, 28(8), 1326–1342. doi:10.1016/j.cpr.2008.07.002
- Sass, D. A. (2011). Testing measurement invariance and comparing latent factor means within a confirmatory factor analysis framework. *Journal of Psychoeducational*Assessment, 29(4), 347–363. doi:10.1177/0734282911406661
- Saulsman, L. M., & Page, A. C. (2004). The five-factor model and personality disorder empirical literature: A meta-analytic review. *Clinical Psychology Review*, 23(8), 1055–1085. doi:10.1016/j.cpr.2002.09.001
- Schimmack, U., Diener, E., & Oishi, S. (2002). Life-satisfaction is a momentary judgment and a stable personality characteristic: The use of chronically accessible and stable

```
sources. Journal of Personality, 70(3), 345–384. doi:10.1111/1467-6494.05008
789
```

- Schimmack, U., Furr, R. M., & Funder, D. C. (1999). Personality and Life Satisfaction: A 790 Facet-Level Analysis, 1062–1075. doi:10.1177/0146167204264292
- Schimmack, U., Oishi, S., Furr, R. M., & Funder, D. C. (2004). Personality and life
- satisfaction: A facet-level analysis. Personality and Social Psychology Bulletin, 30(8),
- 1062–1075. doi:10.1177/0146167204264292 794

791

- Schmitt, D. P., Allik, J., McCrae, R. R., Benet-Martínez, V., Alcalay, L., Ault, L., . . . 795
- Sharan, M. B. (2007). The geographic distribution of Big Five personality traits: 796
- Patterns and profiles of human self-description across 56 nations. Journal of 797
- Cross-Cultural Psychology. doi:10.1177/0022022106297299 798
- Seeboth, A., & Mõttus, R. (2018). Successful explanations start with accurate descriptions:
- Questionnaire items as personality markers for more accurate prediction and mapping 800
- of life outcomes. Journal of Personality. doi:10.17605/OSF.IO/U65GB 801
- Shaver, P. R., & Brennan, K. A. (1992). Attachment Styles and the "Big Five" Personality
- Traits: Their Connections With Each Other and With Romantic Relationship
- Outcomes. Society for Personality; Social Psychology. 804
- Siddiqui, K. (2011). Personality influences Mobile Phone usage. *Interdisciplinary Journal of* 805
- \dots , (1981), 554–563. Retrieved from 806
- $http://papers.ssrn.com/abstract=2468985{\\%}0Ahttp://scholar.google.com/scholar?hl=en{\&}btnowledge abstract=2468985{\%}outperformance abstract=2468985{\%}$ 807
- Soto, C. J., & John, O. P. (2009). Ten facet scales for the Big Five Inventory: Convergence
- with NEO PI-R facets, self-peer agreement, and discriminant validity. Journal of 809
- Research in Personality, 43(1), 84–90. doi:10.1016/j.jrp.2008.10.002 810
- Soto, C. J., & John, O. P. (2016). The Next Big Five Inventory (BFI-2): Developing and

```
Assessing a Hierarchical Model With 15 Facets to Enhance Bandwidth ... The Next
```

- Big Five Inventory (BFI-2): Developing and Assessing a Hierarchical Model With 15
- Facets to Enhance Bandwidth, Fidelit, 113(June), 117–143.
- doi:10.1037/pspp0000096
- Tupes, E. C., & Christal, R. E. (1961). Recurrent person-
- ality factors based on trait rating. Lackland Air Force Base, TX: USAF. Retrieved from
- https://ejwl.idm.oclc.org/login?url=http://search.ebscohost.com/login.aspx?direct=true{\&}db=sih
- live
- Velicer, W. F. (1976). Determining the number of components from the matrix of partial
- correlations. Psychometrika, 41(3).
- Wakabayashi, A., Baron-Cohen, S., & Wheelwright, S. (2006). Are autistic traits an
- independent personality dimension? A study of the Autism-Spectrum Quotient (AQ)
- and the NEO-PI-R. Personality and Individual Differences, 41(5), 873–883.
- doi:10.1016/j.paid.2006.04.003
- Watson, D., & Watson, D. (2002). General and Specific Traits of Personality and Their
- Relation to Sleep and Academic Performance. Journal of Personality, 70(2), 177–206.
- doi:10.1111/1467-6494.05002
- Widiger, T. A., & Mullins-Sweatt, S. N. (2009). Five-Factor Model of Personality Disorder:
- A Proposal for DSM-V. Annual Review of Clinical Psychology, 5(1), 197–220.
- doi:10.1146/annurev.clinpsy.032408.153542
- Ziegler, M., Bensch, D., Maaß, U., Schult, V., Vogel, M., & Bühner, M. (2014). Big Five
- facets as predictor of job training performance: The role of specific job demands.
- Learning and Individual Differences, 29, 1-7. doi:10.1016/j.lindif.2013.10.008
- ⁸³⁵ Ziegler, M., Danay, E., Schölmerich, F., & Bühner, M. (2010). Predicting Academic Success

with the Big 5 Rated from Different Points of View: Self-Rated, Other Rated and Faked. European Journal of Personality, 24 (July 2010), 341–355. doi:10.1002/per

Table captions