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3 Title: Assortative Mating and the Evolution of Human Trait Covariation

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99 Key words: Assortative mating, trait covariation, agent-based modeling, cross-cultural studies

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Abstract:

Human traits covary. For instance, a person's standing height predicts their intelligence test scores; and the Big Five personality dimensions correlate under a "general factor of personality." Trait covariation poses a puzzle for evolutionary theories of mind and behavior and myriad hypotheses attempt to explain specific patterns of trait correlation, including shared condition-dependence, life history strategies, and facultative calibration. Here we show that one process, assortative mating for mate value, generates a signature pattern of covariation. We use agent-based models to demonstrate that assortative mating causes the evolution of a positive manifold of desirability, d , such that an individual who is desirable as a mate along any one dimension tends to be desirable across all other dimensions. Further, we use a large cross-cultural sample with $n = 14,478$ from 45 countries around the world to show that this d -factor emerges in human samples, is a cross-cultural universal, and is patterned in a way uniquely consistent with an evolutionary history of assortative mating. Our results suggest that assortative mating can explain the evolution of a broad structure of human trait covariation.

Significance Statement:

Mate choice lies close to reproduction, the engine of biological evolution. Patterns of mate choice consequently have power to direct the course of evolution. Here we provide evidence that one pattern of human mate choice—the tendency for mates to be similar in overall desirability—caused the evolution of a structure of correlations that we call the d factor. Because of this d factor, individuals who are desirable as a mate on one trait dimension tend to be desirable across all other dimensions as well. We combine computer simulations with analysis of a large cross-cultural sample to show that this pattern of trait correlation appears across cultures and is uniquely consistent with a human evolutionary history of assortative mating.

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Introduction

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Humans mate with self-similar partners across a wide array of dimensions. For example, mated partners tend to be improbably similar to one another in terms of education (1), intelligence (2), and physical attractiveness (3). One critical dimension of assortative mating is that for “mate value,” or overall desirability as a mating partner (4). To the extent that all individuals vie for the most consensually desirable partners on the mating market, those highest in mate value tend to have the greatest power of choice and use that power to select high mate value partners (5). Mated partners consequently tend to have correlated mate values (6).

Such assortative mating for mate value creates “cross-character assortment”: correlations between mated partners on otherwise independent traits (7). Consider a scenario in which humans mate assortatively for mate value and mate value is determined by just two preferred characteristics: kindness and intelligence. Here, all else equal, a kind person will be higher in mate value and will tend to attract higher mate value partners. These high mate value partners, relative to randomly chosen partners, are disproportionately likely to be intelligent. Assortative mating for mate value will therefore pair kind people with intelligent partners at above-chance rates. Such cross-character assortment does occur in married couples for specific traits; for instance, physically attractive women tend to marry wealthier men (8).

When mated partners produce offspring, this pattern of mate choice translates into a pattern of inheritance. Offspring inherit correlated traits from their assortatively mated parents because human individual differences tend to be heritable (9). A kind person mated to an intelligent partner will be relatively likely to produce offspring who are both kind and intelligent. This inheritance of correlated traits, iterated across generations, can cause the evolution of trait covariation: traits that are initially distributed randomly with respect to one another in the

152 population gradually become correlated across generations. Consistent with this rationale, human
153 couples do in fact show evidence of assortative mating at both the phenotypic and genetic level
154 (10, 11). Further, trait covariances are frequently mediated by genetic factors, including both
155 pleiotropic and correlated genes (12, 13). Finally, the genetic correlation between height and
156 intelligence appears to emerge due to contributions from both pleiotropic genes and from
157 covariance in underlying genes due to cross-character assortative mating (13).

158 Beyond height and intelligence, a large prior literature explores, either directly or
159 indirectly, correlations between desirable traits (see Table S1 for a non-comprehensive review).
160 Findings from this literature are mixed, with some trait correlations proving robust and others,
161 such as between physical attractiveness and intelligence, finding only infrequent support.
162 However, this literature is also marked by relative homogeneity in participant populations and
163 great heterogeneity in sample sizes, measures, and methods. Furthermore, studies in this
164 literature very rarely test assortative mating as a potential source of trait covariation. In light of
165 this, in the current research we sought to provide three novel contributions.

166 First, in conjunction with computational models of assortative mating, we analyze a new
167 sample of real-world data that is both large and includes participants from around the world.
168 Second, unlike the prior literature, we do not focus on covariation in absolute trait levels but on
169 covariation in desirability—that is, in deviation of trait value from the opposite sex’s
170 preferences. This is a subtle but important distinction. To the degree that one sex’s ideal
171 preference is not maximal on a trait dimension, mate value will be nonlinear with respect to that
172 trait dimension. For example, people most strongly express a preference for mates in the 90th
173 percentile of intelligence, rather than the 99th (14). This preferences makes mate value a non-
174 linear function of intelligence: all else equal, high mate value people will be relatively high on

175 intelligence, whereas moderate mate value people could be close to either the 99th or 75th
176 percentile on intelligence. The trait covariation created by assortative mating for mate value will
177 consequently be nonlinear with respect to these trait dimensions as well. The effects of
178 assortative mating will therefore be clearest when analyzing covariation in terms in desirability,
179 rather than in terms of absolute trait level.

180 Third and finally, while previous work has explored assortative mating's power to
181 construct covariation between two trait dimensions, assortative mating actually predicts the
182 evolution of a broader covariance structure among preferred traits (7). Humans express mate
183 preferences for a wide array of traits (15) and these preferences predict real mate choices (16).
184 When multiple preferences contribute to mate selection, assortative mating for mate value has
185 the potential to produce intercorrelation in desirability across all preferred characteristics. More
186 than just bivariate correlations, what should emerge from assortative mating for mate value
187 across generations is a positive manifold of desirability, which we call d , organized around mate
188 value such that a person who is desirable as a mate along any one preferred dimension tends also
189 to be desirable across all other dimensions.

190 Here we test this hypothesis using agent-based models and a large cross-cultural sample.
191 We first use a series of evolutionary agent-based models to demonstrate that assortative mating
192 causes the evolution of a general factor of desirability, d , within a set of initially uncorrelated
193 traits and to identify a pattern of results diagnostic of this process. Next, we compare data from
194 these simulated populations to a sample of $n = 14,487$ people from 45 countries around the
195 world. We use this cross-cultural sample to show that this d -factor does in fact emerge across
196 human populations and that it is patterned precisely as predicted by our evolutionary agent-based

197 models, suggesting human desirability covariation is partially explained by an evolutionary
198 history of assortative mating.

199 **Results**

200 First, we examine the results of a series of evolutionary agent-based models
201 demonstrating the evolution of desirability covariation in simulated mating markets. Agents
202 within these simulations possess a set of 10 traits, drawn initially from random normal
203 distributions, and 10 corresponding mate preferences. Agents compute attraction to one another
204 as mates by computing the Euclidean distance between their own preferences and each potential
205 mate's traits. Agents next select each other as mates based on these attractions and reproduce
206 with their chosen partners. Agents with trait values closer to a randomly determined optimum
207 value have more "energy." Agents with more energy produce more offspring, introducing a
208 selection pressure favoring agents with trait values—and preferences for trait values—closer to
209 optimum. Offspring inherit traits and preferences from their parents and start the life cycle anew
210 by calculating their attraction to one another. This process repeats for 1,000 generations of
211 simulated evolution. Agents within these simulations mate assortatively for overall "mate value,"
212 the extent to which their traits match the opposite sex's mate preferences overall. Across model
213 runs, the correlation between partner mate values in the final generation of the model was $r_{\text{mean}} =$
214 $.93$, 95% CI $[.93, .94]$, where mate value was calculated as the Euclidean distance between each
215 agent's trait values and the average value preferred for each trait by the opposite sex. This agent-
216 based model therefore allowed us to assess whether an evolutionary history of assortative mating
217 for mate value would construct a d -factor from initially random traits such that agents who are
218 more desirable along any given trait dimension tend to be desirable across all others as well.

219 To determine whether agent populations evolved a *d*-factor, we computed the
220 “desirability” of each agent on each trait dimension as the absolute deviation between that
221 agent’s trait value and the average of all opposite-sex agents’ preferred value for that trait. These
222 desirability values were scaled such that a higher desirability value indicated the agent was closer
223 to the opposite sex’s preferred value on that trait. We then subjected these trait-level desirability
224 values to principal component analysis, extracting just the first principal component at each
225 generation. In the final generation, we additionally saved each agent’s factor score on the *d*-
226 factor and the loading of each trait onto this *d*-factor.

227 Figure 1 shows that the agent populations in the primary agent-based model do in fact
228 evolve a *d*-factor from initially uncorrelated traits. In the first generation of evolution, when
229 agent traits were uncorrelated, the *d*-factor explained just $M = 15.36\%$, 95% CI [15.22%,
230 15.50%] of the variance in trait-level desirability across model runs. However, after 1,000
231 generations of assortative mating, this *d*-factor increased in size to explain $M = 40.58\%$, 95% CI
232 [39.04%, 42.12%] of the variance in trait-level desirability. Although agent traits were initially
233 distributed randomly, by the final generation of the agent-based model, a *d*-factor evolved such
234 that agents that were desirable as a mate on any given trait dimension were likely to be desirable
235 across all other trait dimensions.

236 Evidence that assortative mating produced this *d*-factor comes from two additional
237 effects (Figure 1). First, if the *d*-factor represents a general dimension of desirability as a mate,
238 agent *d*-factor scores should strongly correlate with their overall mate value. Indeed, a multilevel
239 model with agents nested within model runs shows that, in the final generation, agent mate
240 values strongly predict their factor scores on the *d*-factor across model runs, $\beta = .98$, $SE = 0.002$,
241 $p < .001$.

242 Second, if the *d*-factor evolves because of assortative mating's ability to create cross-
243 character assortment, traits that generate stronger cross-character assortment should tend to load
244 more strongly onto the *d*-factor. To test this prediction, we calculated the cross-character
245 assortment generated by each trait dimension as the average correlation between desirability on
246 that dimension and partner desirability across each of the other nine trait dimensions. We then
247 used each trait's cross-character assortment estimate to predict its factor loading onto the *d*-
248 factor. Indeed, in the final generation of the agent-based model, traits that generated stronger
249 cross-character assortment tended to load more strongly onto the *d*-factor across model runs, $\beta =$
250 1.14, $SE = 0.03$, $p < .001$. That is, more than merely correlating with one another, trait-level
251 desirabilities correlate in a systematic way: each trait dimension's loading onto the *d*-factor is
252 proportional to its actual involvement in cross-character assortative mating.

253 Precisely the same pattern of effects emerges in all 45 countries of the human cross-
254 cultural sample (Figure 2). We calculated the desirability of all participants in this sample in the
255 same way as for the agent-based models: the absolute deviation between each participant's trait
256 value and the opposite sex's average preference value. Consistent with prior research (e.g. 14),
257 participants on average expressed high but not maximal preferences on each of the five
258 dimensions; the average preference value across traits and across participants was $M = 5.85$ (SD
259 $= 1.12$) out of a maximum of 7. Accordingly, trait-level desirabilities were strongly but
260 imperfectly correlated with absolute trait values; Table S1 presents the correlations between
261 absolute trait values and desirabilities for both males and females across countries. Participant
262 mate values were calculated as the scaled Euclidean distance between participant traits and
263 opposite-sex preferences.

264 We estimated the d -factor separately for each country by extracting the first principle
265 component from participant trait-level desirabilities in that country. Across all countries, this d -
266 factor explained $r^2_{\text{mean}} = 42.33\%$, 95% CI [40.15%, 44.51%] of the variance in trait-level
267 desirability. As in the primary agent-based model, participants who were more desirable as a
268 mate along any one preference dimension were more likely to be desirable across all preference
269 dimensions.

270 A multilevel model predicting participant factor score from overall mate value, with
271 participants nested within country, showed that, as in the primary agent-based model, participant
272 mate value nearly perfectly predicted d -factor scores, $\beta = 0.99$, $SE = 0.02$, $p < .001$. Finally, a
273 multilevel model predicting d -factor loading from cross-character assortment found a significant
274 relationship, $\beta = .35$, $SE = .05$, $p < .001$, such that, across countries, traits that generated more
275 cross-character assortment also loaded more strongly onto d . This means that desirability
276 dimensions are not only correlated, but that they show a systematic pattern of covariance: in
277 human data, just as in agent-based models of evolution under assortative mating, preferred trait
278 dimensions load onto the d -factor to the degree that they are actually involved in cross-character
279 assortative mating.

280 Several further analyses establish that this pattern of results is robust and diagnostic of an
281 evolutionary history of assortative mating (see supplementary information). First, the pattern of
282 results observed in the primary agent-based model is robust to higher mutation rates and to lower
283 levels of assumed trait heritability. Second, the pattern of results observed in the primary agent-
284 based model and the cross-cultural human sample are diagnostic in that they do not emerge in
285 agent-based models where mate choice is not assortative for mate value. Third, trait covariation
286 could alternatively emerge because different traits are manifestations of a common underlying

287 condition variable (17, 18). However, the pattern of results observed in the primary agent-based
288 model and across countries does not emerge in a simulation in which agent traits are
289 manifestations of an underlying condition factor and mate choice is not assortative for mate
290 value.

291 Two further results suggest this pattern of effects is specifically explained by assortative
292 mating and is not a mathematical inevitability or a byproduct of background trait covariation.
293 First, it is possible that the *d*-factor and mate value relate simply because they are both calculated
294 based on deviations between an individual's traits and the opposite sex's preferences. However,
295 whereas mate value is directly the deviation between a person's traits and the opposite sex's
296 preferences, the *d*-factor is a structure of covariances between deviations across dimensions. Any
297 relationship between mate value and the *d*-factor thus depends on the existence of such a pattern
298 of covariances. Indeed, the pattern of effects observed across countries does not emerge when
299 mate value and the *d*-factor are calculated based on scrambled participant traits that do not share
300 the raw data's correlational structure. This shows that a *d*-factor is not inevitable, but rather
301 depends on the particular covariance structure produced by assortative mating. Second, the *d*-
302 factor could be a byproduct of some other source of covariation with no intrinsic connection to
303 mate value. However, the pattern of results observed in the cross-cultural sample and primary
304 agent based model do not emerge when mate value and *d* are computed based on deviations from
305 random values rather than from the opposite sex's mate preferences. This demonstrates that the
306 *d*-factor is specifically organized around the opposite sex's mate preferences and is not simply a
307 byproduct of independent trait covariation.

308 Finally, this pattern of results could plausibly emerge due to participant self-report biases.
309 For instance, participants who rate themselves more desirably on any one dimension could be

310 more likely to rate themselves and their partner desirably across all dimensions due to positive
311 illusions or other self-report rating biases (19, 20). However, the *d*-factor and the diagnostic
312 pattern of results still emerge in two smaller samples in which participant trait ratings are not
313 derived from participant self-reports.

314 **Discussion**

315 These results document a pattern of desirability covariation, *d*, that emerges across 45
316 countries such that a person who is desirable as a mate on any one trait dimension is more likely
317 to be desirable as a mate across all other trait dimensions. Scores on the *d*-factor are nearly
318 perfectly correlated with individual mate value. Finally, more than merely correlating with one
319 another, desirability dimensions load onto this *d*-factor to the degree that they actually generate
320 cross-character assortment. This pattern of results is precisely the same pattern that emerges in
321 agent-based models of evolution under assortative mating but that does not emerge in models
322 without assortative mating. Overall, this suggests that an evolutionary history of assortative
323 mating has produced a specific pattern of desirability covariation in humans.

324 While promising, this research does have limitations and leaves open some important
325 future directions. First, although we do find evidence of a *d*-factor consistent with an
326 evolutionary history of assortative mating, assortative mating is also clearly not explaining all of
327 the trait covariation in our data. This can be seen, for instance, in the fact that the correlation
328 between *d*-factor loading and cross-character assortment is much weaker in the cross-cultural
329 sample than it is in the agent-based models, indicating other factors are influencing the degree of
330 trait covariation in our data. These other factors likely include measurement variance, condition-
331 dependence processes, facultative calibration, and even direct effects between traits—for
332 instance, intelligence likely directly affects financial prospects by influencing occupational

333 success. Teasing apart the relative contributions of assortative mating and these other sources of
334 observed covariation is a clear next step for future research.

335 A second important future direction is resolving the mixed findings in this broader
336 research area. For instance, although in our samples physical attractiveness, intelligence, and
337 health load onto the *d*-factor in theoretically consistent ways, other studies have failed to find
338 correlations between these traits (e.g. 21, 22). These inconsistencies must be explained. One
339 obvious candidate explanation is difference in measurement. For example, prior studies finding
340 null correlations between physical attractiveness and intelligence have often used standardized
341 intelligence tests whereas our samples exclusively used rated intelligence. It is possible that these
342 measures produce different results because they tap different constructs. For instance, rated
343 intelligence measures might be more likely to show covariance patterns consistent with
344 assortative mating because they more closely tap the folk concept of intelligence that actually
345 drives mate choice. After all, people select mates on the basis of their lay perceptions, and not on
346 the basis of standardized intelligence exams. However, this does open a clear question for future
347 research: if the folk concept of intelligence does not tightly map onto *g*, what precisely does it
348 track? And of course, although the pattern of results found in the primary agent-based models are
349 mirrored in all of our human samples, across self-reports, partner-reports, third-party reports, and
350 combinations therein, it is impossible to completely rule out rater bias as an alternative
351 explanation for our results. Stronger evidence for the existence of a *d*-factor must come from
352 future research exploring desirability covariation independent of subjective ratings entirely.

353 Despite these limitations, the universality and patterning of the *d*-factor in our cross-
354 cultural samples resemble a fingerprint of assortative mating on the evolution of human trait
355 distributions. Previous work has documented the organizing effects of mate choice between

356 specific preferred trait dimensions (13). Our results show that these patterns of desirability
357 covariation emerge not merely between specific traits but rather across all preferred trait
358 dimensions and that these patterns of trait covariation emerge across cultures from around the
359 world. Assortative mating appears to have shaped patterns of inheritance throughout human
360 evolution such that mate value is not distributed randomly across individuals; rather, desired
361 traits covary around an underlying dimension of mate value. This fact contributes to explaining
362 the existence of human trait covariation across domains and highlights the importance of mate
363 choice in broadly understanding human evolution.

364 **Materials and Methods**

365 **Agent-Based Model**

366 We constructed and analyzed an evolutionary agent-based model of a mating market. The
367 primary model generated 200 agents at the start of each model run. Each agent possessed 10
368 traits, initially drawn from random normal distributions centered on $M = 4$ with $SD = 2$. Agents
369 also had 10 corresponding mate preferences; preferences, like traits, were initially drawn from
370 random normal distributions centered on $M = 4$ with $SD = 2$. Each agent was additionally
371 assigned an energy value based on the value of their traits. At the start of each model run, the
372 model selected a random value as optimal for each trait dimension. Each agent earned energy
373 proportional to the absolute deviation between their trait value and optimum value for that trait
374 dimension such that agents who were closer to the optimum value across all traits had more
375 energy. These energy values were used to control reproduction and introduce natural selection
376 into the model. Finally, all agents had a sex: half of all agents were randomly assigned to be
377 female and the remaining half were male.

378 After initialization, agents followed a life cycle in which they computed how attracted
379 they were to one another, selected each other as mates based on these attractions, reproduced
380 with their chosen mates, and then died. This life cycle was repeated for 1,000 generations of
381 evolution.

382 **Attraction.** In the first phase of the life cycle, agents computed how attracted they were
383 to one another based on their mate preferences. Each agent computed their attraction to all
384 opposite-sex agents. Attraction was calculated as the Euclidean distance between the agent's
385 preference vector and each potential mate's trait vector. These distances were then scaled and
386 transformed such that a value of 10 indicated that the potential mate perfectly matched the
387 agent's preferences, whereas a value of 0 indicated that the potential mate was the worst possible
388 fit to the agent's preferences. This attraction algorithm has shown to be a good model of the
389 algorithm used by human mate choice psychology (23).

390 **Mate selection.** The attraction calculation phase produced two matrices: one matrix
391 containing how attractive each male agent found all female agents and another containing how
392 attractive each female agent found all male agents. In the next phase of the life cycle, the model
393 multiplied these attraction matrices together to produce the mutual attraction matrix. Each cell of
394 this matrix represented how mutually attracted all possible agent couples would be. The model
395 next paired the most mutually attracted possible couple and then removed this couple from the
396 mutual attraction matrix. This pairing process iterated until all possible couples were formed.

397 **Reproduction.** Agents next reproduced with their chosen partner. Agent couples
398 reproduced in proportion to the sum of their energy values. In this way, agents who had trait
399 values closer to optimum—and mate preferences for these trait values—were more likely to
400 reproduce each generation. Energy values were scaled prior to reproduction such that highest

401 energy couple in each generation had 10% greater reproductive success on average than the
402 lowest energy couple, yielding a moderate and realistic selection pressure in favor of optimum
403 traits and preferences (24). Each offspring inherited each of their preference and trait values
404 randomly from either parent. A small amount of random normal noise ($M = 0$, $SD = .006$),
405 equivalent to .1% of the total trait range, was added to each inherited trait to simulate mutation.
406 This is intended to simulate the cumulative effects of many, small-impact mutations (e.g. see
407 (25)). Half of all offspring were randomly assigned to be female; the other half were randomly
408 assigned to be male. The number of offspring produced each generation was equal to the starting
409 population size.

410 **Death.** After reproduction, all agents of the parent generation died. Offspring then began
411 the life cycle anew in the next generation. After 1,000 generations of evolution, the model
412 retained the final generation of parent couples. The result for each model run was a final
413 population of $n = 200$ that represented the results of evolution under conditions of assortative
414 mating.

415 **Cross-cultural data**

416 **Participants.** Participants in the cross-cultural sample were $n = 14,487$ individuals
417 (7,961 female) from 45 different countries from all inhabited continents around the world.
418 Participants in each study site were recruited from two sources: roughly half of all participants
419 were recruited from university populations and the remaining half were recruited from
420 community samples. Not all study sites kept records of participant sample source; however,
421 among those sites with records ($n = 6,637$), 47.14% ($n = 3,129$) of participants came from
422 community samples. All participant data was collected in person because online samples tend to
423 be less representative of populations in developing countries (26). Participants were $M = 28.79$

424 years old ($SD = 10.64$) and ages ranged from 18 to 91 ($Mdn = 25$). Most participants ($n = 9,236$,
425 63.75%) reported being in an ongoing, committed, romantic relationship. Of these, 49.26%
426 reported being in a dating relationship, 12.59% were engaged, and 38.14% were married. This
427 study was approved by the University of Texas at Austin Institutional Review Board and by the
428 equivalent at each researcher's home university.

429 **Measures.** All participants reported their mate preferences in an ideal long-term mate,
430 described as a committed, romantic partner, using a 5-item mate preference instrument. This
431 instrument contained five 7-point bipolar adjective scales on which participants rated their ideal
432 partner's standing on five separate traits: intelligence, kindness, health, physical attractiveness,
433 and financial prospects. Each trait was rated between two extremes, for instance, from 1
434 representing "very unkind" to 7 representing "very kind." Participants additionally used the same
435 rating scales to describe their own standing on each of these five traits and to rate their actual
436 long-term partner, if they had one. This mate preference instrument was translated into local
437 languages and back-translated by researchers at each study site.

438 **Data analysis**

439 Data analysis proceeded in several parallel stages for both the agent-based models and
440 the cross-cultural data. First, within each country and each model run, we calculated the average
441 preferences of all males and the average preferences of all females. These preferences were used
442 to compute two values within country and within model run. We first calculated the overall mate
443 value of each agent, each participant, and their partners as the Euclidean distance between that
444 individual's traits and the average preferences of the individual's opposite sex. This mate value
445 estimate is a single summary value that reflects the degree to which each person or agent
446 embodies the preferences of the opposite sex across all dimensions. These distances were scaled

447 such that a value of 10 meant the individual perfectly matched the opposite sex's average
448 preferences and a value of 0 meant the individual provided the worst possible match to the
449 opposite sex's preferences. Prior studies have found that these Euclidean mate values predict
450 both desirability as a mate and power of choice on the mating market (23). For plotting purposes,
451 agent mate values were standardized to a common scale within model runs before producing
452 figures to control for variation in population mate values across model runs.

453 Second, we used average preferences to calculate each agent and each participant's
454 "desirability" on each trait dimension. Desirabilities were calculated as the absolute difference
455 between the individual's trait value for each dimension and the opposite sex's average preference
456 value for that dimension; desirability values were re-scaled such that higher values indicated a
457 closer fit to the opposite sex's preferences. Rather than a single summary variable as for mate
458 value, this yields a vector of values for each agent or participant, with each value reflecting the
459 degree to which that agent or participant matches the opposite sex's mate preference on that
460 specific trait dimension.

461 Next, we subjected these desirability scores to principal component analysis. Principal
462 component analyses were run separately for each run of the agent-based model and for each
463 country within the cross-cultural sample. Male and female desirability scores were additionally
464 analyzed separately because men and women have different mate preferences (15). Each
465 principal component analysis extracted a single principle component, the *d*-factor, from trait-
466 level desirabilities. From these principal component analyses, we saved the total variance in trait-
467 level desirability explained by this *d*-factor (averaged across males and females), the loadings of
468 each desirability dimension onto this *d*-factor, and each participant or each agent's factor score.

469 All data, model script, and analysis script are available on the Open Science Framework

470 (https://osf.io/6g4pq/?view_only=cc6d7e1f76a3474d8c917c544acc1033).

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540 women, and women with higher adiposity. *PLOS ONE* 9(7):e100966.

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Figure Legends

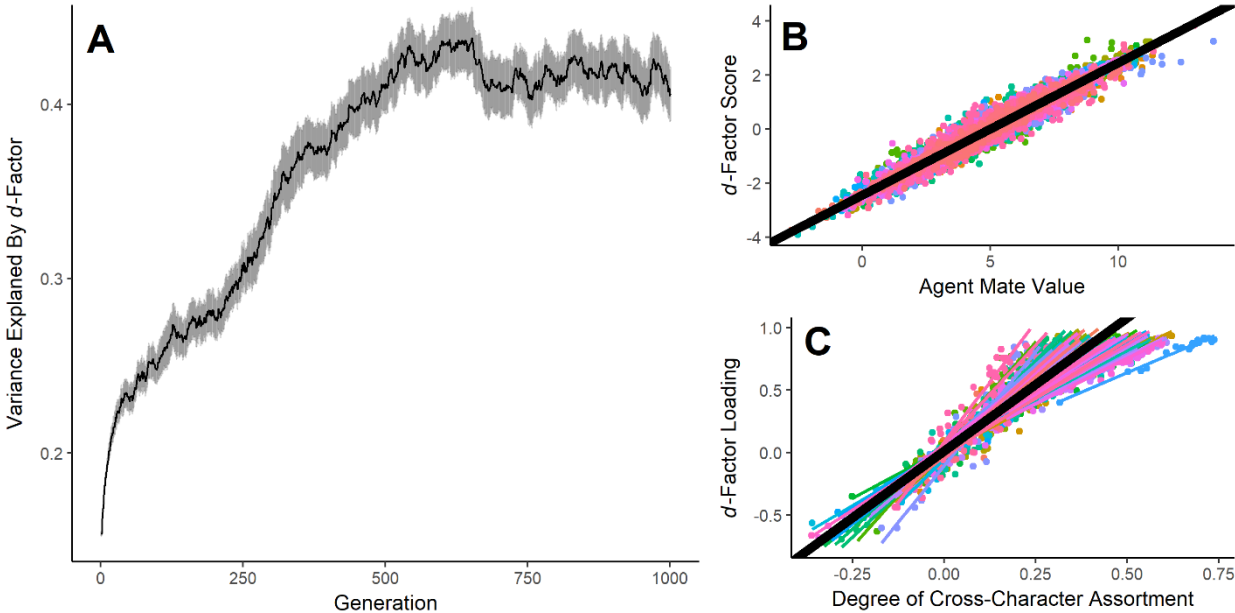
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Figure 1. Results from the primary agent-based model. Agents within this simulation evolve a *d*-factor that explains a moderate portion of trait variation (a). Scores on this *d*-factor strongly predict agent mate value (b). Traits that more strongly predict partner mate value load more strongly onto the *d*-factor (c). Dots represent individual observations; colored lines represent trend lines for individual model runs; black lines represent overall trends across model runs. Different colors correspond to observations from different model runs.

Figure 2. Results from the human cross-cultural sample. Across countries, the *d*-factor explains a moderate amount of variance in trait-level desirability (a). Scores on the *d*-factor are strongly correlated with participant mate value across countries (b). Desirability dimensions that more strongly predict partner mate value tend to load more strongly onto the *d*-factor across countries (c). Dots represent individual observations; colored lines represent trends from individual countries; black lines represent average trends across countries. Different colors correspond to observations from different countries.

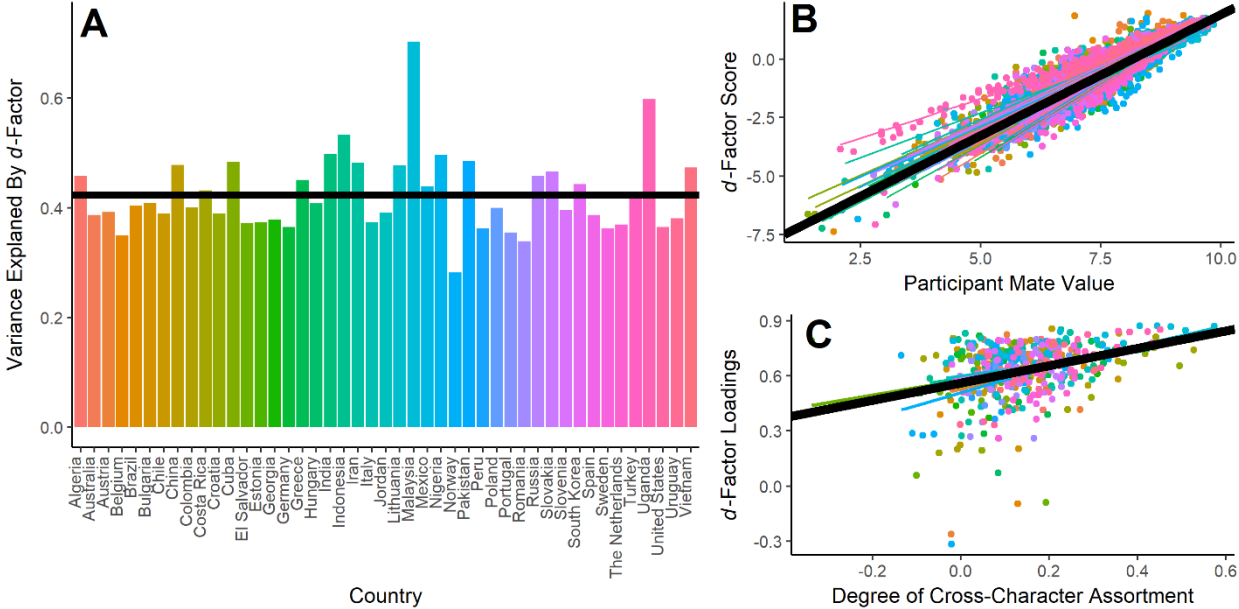
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Figures



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Figure 1



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569 Figure 2

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571 **Supplementary Information**572 **Supplementary Model 1: Manipulating the Mutation Rate in the Primary Agent-Based**573 **Model**

574 To assess the robustness of the results reported in the primary agent-based model, we ran
575 a secondary version of this model in which the mutation rate was increased by factor of 10 to a
576 value of 0.06. The *d*-factor explained less variance in trait-level desirability in the final
577 generation of the modified model compared to the primary model: $M = 26.89\%$, 95% CI
578 [25.62%, 28.15%] relative to $M = 40.58\%$, 95% CI [39.04%, 42.12%]. Nonetheless, the variance
579 explained by the *d*-factor was still significantly greater than the variance explained in the initial
580 generation, $M = 15.25\%$, 95% CI [15.12,15.38]. As in the primary agent-based model, agent
581 mate value strongly predicted agent *d*-factor scores, $\beta = 0.87$, $SE = 0.02$, $p < .001$. Agents who
582 were higher in overall mate value still scored higher on the *d*-factor than agents lower in overall
583 mate value. Traits that generated greater cross-character assortment also loaded more strongly
584 onto the *d*-factor, $\beta = 1.01$, $SE = 0.01$, $p < .001$. Overall, the pattern of effects documented in the
585 primary agent-based model emerge even when the assumed mutation rate is increased by a factor
586 of 10. These results suggest that this pattern of results is robust to assumptions about mutation
587 rates.

588 **Supplementary Model 2: Simulating Imperfect Heritability**

589 One limitation of the primary agent-based model is that it assumes all traits are perfectly
590 heritable: all trait and preference variation is due to inherited factors. To determine whether this
591 simplification spuriously produces the *d*-factor we observed, we created a separate set of models
592 that simulate imperfect heritability. These models are identical to the primary model except for
593 how inherited trait and preference values relate to agent features. In these models, each agent has

594 a set of “manifest” trait and preference values. These manifest values are the sum of an inherited
595 trait or preference value, inherited from parents just as in the primary model, and random normal
596 noise. The standard deviation of this noise was calibrated such that, on average, inherited values
597 explained a fixed percentage of the variance in manifest trait values, simulating imperfect
598 heritability. Agents selected each other as mates on the basis of their manifest traits and
599 preferences, but offspring inherited only their parent’s inherited values, with mutation, and not
600 the noise component of their manifest trait and preference values.

601 We ran two such models with two levels of heritability: 50% heritability, based on a
602 meta-analysis of heritability studies (9), and 75% heritability, chosen to be intermediate between
603 the 50% heritability model and the primary agent-based model. We conducted the same set of
604 analyses on these models in the primary agent-based model, focusing analysis on the manifest
605 values rather than the inherited values. Figure S1 shows that these models produced the same
606 pattern of effects observed in the primary agent-based model and the cross-cultural data,
607 although with effects attenuated proportional to the degree of assumed heritability. In both
608 models, the *d*-factor gradually evolves to explain a larger proportion of the variance in trait-level
609 desirability in the final generation of evolution ($h^2 = .50$: $M = 16.43\%$, 95% CI [16.21%,
610 16.66%]; $h^2 = .75$: $M = 25.01\%$, 95% CI [24.02%, 26.00%]) compared to the first generation of
611 evolution ($h^2 = .50$: $M = 15.25\%$, 95% CI [15.14%, 15.36%]; $h^2 = .75$: $M = 15.31\%$, 95% CI
612 [15.19%, 15.44%]). In both models, *d*-factor scores were strongly correlated with agent mate
613 value: $h^2 = .50$: $\beta = .58$, $SE = 0.03$, $p < .001$, $h^2 = .75$: $\beta = .93$, $SE = 0.01$, $p < .001$. Finally, traits
614 that generated stronger cross-character assortment also tended to load more strongly onto the *d*-
615 factor: $h^2 = .50$: $\beta = .46$, $SE = 0.03$, $p < .001$, $h^2 = .75$: $\beta = 1.02$, $SE = 0.03$, $p < .001$.

616 These supplementary models demonstrate that the *d*-factor and the associated pattern of
617 effects observed in the primary model also emerge in models that assume imperfect heritability.
618 The *d*-factor is smaller, and its relationship with other variables weaker, in models that assume
619 lower heritability—this makes sense, as manifest values have a larger random noise component
620 in these models. However, in all models we nonetheless observe the same diagnostic pattern of
621 effects.

622 **Supplementary Model 3: The *d*-Factor and Random Mate Choice**

623 Figure S2 shows that the pattern of effects found in the primary agent-based model and
624 cross-cultural sample do not appear in a model in which mate choice is random with respect to
625 mate preferences and agents therefore do not mate assortatively for overall mate value. Here the
626 *d*-factor explains only a small proportion of the variance in trait-level desirability in both the first
627 generation, $M = 15.34\%$, 95% CI [15.23%, 15.46%] and the final generation, $M = 16.26\%$, 95%
628 CI [16.10%, 16.43%]. Further, this change from the first to the final generations appears to
629 emerge because of a sudden increase in variance explained within the first few generations
630 followed by no change over the subsequent generations—perhaps due to the initial effects of
631 selection. This contrasts with the continuous, gradual evolution of the *d*-factor observed in the
632 primary models. Agent mate values do predict their *d*-factor scores, but relatively weakly, $\beta =$
633 0.23 , $SE = 0.02$, $p < .001$. Finally and critically, a trait's ability to generate cross-character
634 assortment does not predict its loading on to the *d*-factor when mate choice is not assortative for
635 mate value, $\beta = -0.002$, $SE = 0.03$, $p = .95$. These results indicate mate choice that is not
636 assortative for mate value will not, on its own, cause the evolution of a *d*-factor as observed in
637 the primary agent-based model and the cross-cultural human sample.

638 **Supplementary Model 3: Desirability Covariance as a Function of Common Cause Rather**
639 **Than Assortative Mating**

640 One potential alternative explanation for the covariation in desirability is a common
641 condition variable that influences the development of all traits (17, 18). To assess whether
642 condition-dependence alone would produce the pattern of effects observed in the primary agent-
643 based model, we constructed an alternative model in which agents do not inherit separate trait
644 values from their parents, but rather inherit a single condition variable. The model then creates
645 each agent's 10 trait values by adding random noise to their random condition variable, with a
646 pre-determined standard deviation, such that agent condition variables explain approximately
647 45% of the variance in each of their traits. Agent condition variables can thus be thought of as an
648 underlying factor that determines the development of all agent traits. Agent energy values were
649 determined by their manifest trait values. The optimal trait value was set to a value of 7 for all
650 traits across model runs, rather than being randomly set as in the primary model. This
651 simplification ensured that an increase in condition (e.g. by mutation) simultaneously moved all
652 agent traits closer to optimum. Agents in this model still reproduced in proportion to energy,
653 yielding a selection pressure in favor of higher condition values. Agents in this model paired
654 randomly with respect to preferences and therefore did not mate assortatively for mate value. We
655 allowed such a population of agents to evolve for 1,000 generations in the same way as
656 populations in the primary agent-based model and conducted the same analyses on the final
657 agent populations.

658 Figure S3 shows the results of the condition-dependent agent-based model. Because the
659 condition variable determines all trait values, the d -factor does explain a large proportion of the
660 variance in trait-level desirability in the final generation, $M = 47.94\%$, 95% CI [47.29%,

661 48.58%]. Agent mate value is strongly related to agent *d*-factor scores, $\beta = 0.97$, $SE = 0.002$, $p <$
662 $.001$. However, *d*-factor loadings were independent of cross-character assortment in this model,
663 $\beta = 0.05$, $SE = 0.06$, $p = .37$.

664 Overall, the pattern of desirability covariation produced by condition-dependence is
665 qualitatively different from the pattern of desirability covariation produced by assortative mating.
666 Both processes produce a large *d*-factor that is correlated with agent mate value. But crucially,
667 condition-dependence, in the absence of assortative mating for mate value, does not produce the
668 correlation between *d*-factor loadings and the degree of cross-character assortment generated by
669 each trait. That the human data across cultures more closely approximates the pattern of the
670 primary agent-based model than the condition-dependent model suggests that desirability
671 covariation in humans is better explained by an evolutionary history of assortative mating than
672 by condition-dependence alone.

673 These results of course do not indicate that none of the covariance observed in the cross-
674 cultural sample is explained by condition-dependence; rather, it merely demonstrates that
675 condition-dependence alone is not sufficient to explain the specific pattern of effects we
676 observed. Condition-dependence and assortative mating could and likely do independently
677 contribute to human trait covariation. Furthermore, condition-dependence and assortative mating
678 could be complementary sources of trait covariation. For instance, it is possible that some
679 preferred traits initially signal orthogonal dimensions of condition. However, through assortative
680 mating for mate value, these independent dimensions of condition become correlated with one
681 another as a *d*-factor emerges. So the apparent unidimensionality of condition could in part
682 emerge due to assortative mating.

683 **Supplementary Models 3 and 4: Assessing the Specificity of the *d*-Factor across Cultures**

684 Results of the primary analyses demonstrate a pattern of desirability covariation that
685 occurs in human data from across cultures and that is consistent with the pattern of results
686 observed in agent-based models in which populations evolve under conditions of assortative
687 mating. Here we conduct two tests designed to assess the specificity of this *d*-factor. First, it is
688 possible that the *d*-factor and participant mate value correlate not because of an evolutionary
689 history of assortative mating but merely because both the *d*-factor and participant mate values
690 were calculated using deviations from the opposite sex's preferences. If this were the case, even
691 random trait ratings should produce the pattern of results observed in the cross-cultural human
692 data and the primary agent-based model as long as mate value and the *d*-factor are both
693 calculated using deviations of these random traits from the opposite sex's preferences.

694 To test this alternative hypothesis, we randomly scrambled participant trait ratings from
695 the cross-cultural sample within country, within sex, and within trait. The result was that, for
696 each trait variable, each participant was randomly assigned the trait rating of another same-sex
697 participant from their country. This scrambling maintains the mean, standard deviation, and
698 distribution of each trait variable but removes all above-chance correlations between trait
699 variables. We then used these scrambled trait ratings to compute participant mate values and to
700 estimate the *d*-factor and participant *d*-factor scores. For comparison, we also applied the same
701 scrambling procedure to the trait values of agents within the primary agent-based model. We
702 then applied the same analyses to the scrambled human and agent data as to the primary human
703 and agent-based model samples.

704 Figure S4 shows the results of analyzing the scrambled human data. Across cultures, the
705 *d*-factor based on scrambled human data explained significantly less variance in trait-level
706 desirability ($r^2_{\text{mean}} = 25.29$, 95% CI [24.56%, 26.02%]) than the *d*-factor based on unscrambled

707 human data ($r^2_{\text{mean}} = 42.33$, 95% CI [40.14%, 44.51%]). Participant mate value in the scrambled
708 human data did predict participant d -factor scores, $\beta = 0.37$, $SE = 0.03$, $p < .001$; however, this
709 relationship was weaker than found in the unscrambled human data. Finally, the degree of cross-
710 character assortment generated by a desirability dimension did not relate to that dimension's
711 loading onto the d -factor in the scrambled human data, $\beta = 0.02$, $SE = 0.05$, $p = .72$.

712 Importantly, a similar pattern of results emerges when data from the primary agent-based
713 model are scrambled in the same fashion. Figure S5 presents these results. In the final generation
714 of the agent-based model, the d -factor based on scrambled data explains a significantly smaller
715 proportion of the variance in trait-level desirability: $r^2_{\text{mean}} = 15.37\%$, 95% CI [15.25%, 15.48%]).
716 Scrambled d -factor scores do relate to agent mate values calculated from the same traits, but do
717 so relatively weakly $\beta = 0.23$, $SE = 0.02$, $p < .001$. The loading of each desirability dimension
718 onto the d -factor did not relate to its ability to generate cross-character assortment, $\beta = -0.01$, SE
719 $= 0.03$, $p = .75$.

720 Overall, these results show that, although the d -factor and mate value are both calculated
721 with respect to deviation from the opposite sex's mate preferences, the pattern of effects
722 documented in the primary analyses is not inevitable. When trait-level data is randomized, the d -
723 factor explains a relatively small proportion of the variance in trait-level desirability, factor
724 scores on the d -factor relate only weakly to mate value, and loadings onto the d -factor do not
725 relate to cross-character assortment. The correlational structure created by an evolutionary
726 history of assortative mating is necessary to produce the effects observed in the primary analyses
727 of both agent-based models and the cross-cultural human data.

728 Next we conducted a series of analyses to test an alternative aspect of specificity: that is,
729 could the results of the primary analyses emerge if mate value and the d -factor were calculated

730 with respect to deviations from any trait point or are they specific to deviations from the opposite
731 sex's mate preferences? To assess this in the cross-cultural sample, we first created a random
732 preference vector for each sex within country by pulling five draws from a random uniform
733 distribution constrained between values of 1 and 7. We then calculated each participant's
734 random-point "mate value" as the scaled Euclidean distance between their own traits and this
735 random vector, rather than the opposite sex's preferences. Additionally, we calculated trait-level
736 desirability as the absolute deviation between participant traits and each of these random
737 preferences and estimated the d -factor based on these random-point desirabilities.

738 The results were broadly similar to but nonetheless distinct from the primary analyses
739 (Figure S6). The d -factor calculated with respect to a random point still explained a moderate
740 proportion of trait-level desirability, $r^2_{\text{mean}} = 41.35\%$, 95% CI [39.48%, 43.22%]. Factor scores
741 on the random-point d -factor also correlated strongly with random-point mate value, but the
742 relationship was weaker than in the primary analyses, $\beta = 0.69$, $SE = 0.04$, $p < .001$. Finally,
743 desirability dimensions that generated more cross-character assortment did load more strongly
744 onto the d -factor, $\beta = 0.35$, $SE = 0.05$ $p < .001$. However, this effect was much less consistent
745 across cultures, as evidenced by the larger variance in the random effect of slope across cultures,
746 $s^2 = 7.54$, for the random-point d -factor compared to $s^2 = 0.03$ in the primary analyses. Indeed, in
747 some countries the relationship between factor loading and cross-character assortment was
748 negative for the random-point d -factor.

749 The same pattern of differences emerges when the data from the primary agent-based
750 model are reanalyzed such that agent desirability and mate value are calculated with respect to a
751 random point (Figure S7). As in the cross-cultural sample, the random-point d -factor explains a
752 still-moderate proportion of the variance in trait-level desirability, $r^2_{\text{mean}} = 40.24\%$, 95% CI

753 [38.70%, 41.77%]. Random-point d -factor scores do relate to agent random-point mate values,
754 but the relationship is weaker than in the primary analyses: $\beta = 0.40$, $SE = 0.03$, $p < .001$. Unlike
755 the human cross-cultural data, dimensions that load more strongly onto the random-point d -
756 factor no longer generate more cross-character assortment, $\beta = .13$, $SE = 0.09$, $p = .15$. However,
757 like in the cross-cultural sample, in this model, the variance in the random slope component is
758 much larger ($s^2 = 1.83$) than in the primary analyses ($s^2 = 0.23$) such that the relationship
759 between d -factor loading and cross-character assortment is sometimes negative.

760 Overall, in both the agent-based models and the human cross-cultural sample, a different
761 pattern of effects emerges when the d -factor and mate value are calculated with respect to a
762 random point in the trait space rather than the opposite sex's average mate preferences. The
763 random-point d -factor still explains a moderate proportion of the variance in trait-level
764 desirability; however, scores on this random-point d -factor have a weaker relationship with mate
765 value than does the true d -factor and the relationship between cross-character assortment and
766 factor loadings is less consistent for the true d -factor than the random-point d -factor. That the
767 same difference in pattern of effects emerges in the human cross-cultural data as in the agent-
768 based model data suggests that the d -factor in the cross cultural sample is not arbitrary, but rather
769 is organized around the opposite sex's preferences because of an evolutionary history of
770 assortative mating.

771 **Supplementary Model 5: Desirability Covariance as a Function of Participant Rating** 772 **Biases Rather Than Assortative Mating**

773 The results of the primary agent-based model as well as the primary analysis of the cross-
774 cultural sample document a d -factor that appears patterned as though it evolved through a history
775 of assortative mating. However, participant rating biases could provide an alternative explanation

776 for this *d*-factor. If participants suffer from “positive illusions,” (19, 20) participants who see
777 themselves more positively on any one dimension could be biased to see themselves and their
778 partner desirably across all other dimensions. This participant rating bias, rather than assortative
779 mating, could explain the cross-cultural pattern of covariation between desirabilities across trait
780 dimensions.

781 However, the pattern of effects observed in the cross-cultural human sample emerges in
782 two additional samples that are not limited to participant self-reports. The first is a dyadic sample
783 of newlyweds. The details of this sample have been reported elsewhere (23). Participants were
784 214 people composing 107 newlywed couples. Each participant reported their ideal long-term
785 mate preferences for 40 personality dimensions using a 40-item Big Five questionnaire. For each
786 participant, we also have ratings of each participant’s actual traits on the same 40 dimensions
787 from four sources: participant self-reports, partner ratings, and the ratings of two independent
788 interviewers. We calculated a composite trait rating for each participant on each of the 40
789 dimensions by averaging together these self, partner, and third-party interviewer reports. We
790 then conducted the same analyses on this newlywed sample as on each country from the cross-
791 cultural sample, calculating mate value as the Euclidean distance between the trait composites
792 and the opposite sex’s average references and calculating trait-level desirabilities on each
793 dimension as the absolute difference between each participant’s trait composite and the opposite
794 sex’s average preference on that trait dimension.

795 The *d*-factor explained $r^2 = 24.30\%$ of the variance in trait-level desirability in this
796 newlywed sample. Scores on this *d*-factor were strongly related to participant mate value, just as
797 in the cross-cultural sample and the primary agent-based model, $\beta = 0.92$, $SE = 0.03$, $p < .001$
798 (Figure S8). Finally, traits loaded on to the *d*-factor in this sample to the extent that they

799 generated cross-character assortment, $\beta = .70$, $SE = 0.08$, $p < .001$. Even when participant trait
800 ratings are not based exclusively on self-reports, a d -factor emerges from participant
801 desirabilities patterned exactly as in the primary agent-based model. This strongly suggests that
802 the d -factor is not entirely an artifact of participant self-report biases.

803 The d -factor in this newlywed sample additionally shows the same degree of specificity
804 observed in the primary agent-based model and the cross-cultural sample. To assess the
805 specificity of the newlywed d -factor, we calculated participant desirabilities based on 100
806 random scramblings of participant traits and separately based on 100 random preference points
807 (Figure S9). A d -factor based on scrambled participant traits explains just $r^2_{\text{mean}} = 6.06\%$, 95%
808 CI [6.02%, 6.09%] of the variance in participant desirabilities. As in the primary agent-based
809 model and the cross-cultural sample, scores on this scrambled d -factor relate only weakly to
810 participant mate value, $\beta = 0.15$, $SE = 0.01$, $p < .001$, and loadings onto the d -factor were
811 unrelated to traits' predictive power, $\beta = -0.01$, $SE = 0.02$, $p = .46$.

812 Furthermore, consistent with the primary agent-based model and cross-cultural sample,
813 the newlywed random-point d -factor explained a moderate proportion of the variance in trait-
814 level desirability, $r^2_{\text{mean}} = 22.67\%$, 95% CI [22.48%, 22.86%]). However, scores on this d -factor
815 relate relatively weakly to participant mate value, $\beta = 0.40$, $SE = 0.02$, $p < .001$. Finally, traits do
816 load onto the random-point d -factor to the extent that produce cross-character assortment, $\beta =$
817 $.65$, $SE = 0.05$, $p < .001$; however, this relationship is less consistent than in the primary analysis
818 as evidenced by the variance in the random slope component in the random-point analysis ($s^2 =$
819 25.83) and the occasional negative slopes relating desirability predictive power and d -factor
820 loading (Figure S10).

821 Overall, the results of these specificity tests perfectly match those observed in the
822 primary agent-based model and the cross-cultural sample. A *d*-factor based on scrambled traits
823 does not explain substantial variance in trait-level desirability, scores on this *d*-factor relate only
824 weakly to participant mate value, and the extent to which traits generate cross-character
825 assortment does not relate to how strongly they load onto this scrambled *d*-factor. This indicates
826 that the *d*-factor in the newlywed sample is not tautologous and its emergence does in fact
827 depend on the covariance structure created by assortative mating. This *d*-factor is additionally
828 specific to mate value: a *d*-factor produced based on deviations from a random point, rather than
829 the opposite sex's preferences, does not relate as strongly to participant mate value and shows an
830 inconsistent relationship between cross-character assortment and factor loading.

831 Finally, we conducted the same analyses using composite trait ratings from this sample
832 based only on the third-party interviewer ratings. This removes participant positive illusions or
833 rating biases from the data entirely. These analyses produced the same results as in the cross-
834 cultural sample and the primary agent-based model (Figure S11). The *d*-factor explained $r^2 =$
835 25.52% of the variance in trait-level desirability when trait ratings are based only on interviewer
836 reports. Scores on this *d*-factor were still strongly related to participant mate value, $\beta = 0.95$, SE
837 $= 0.02$, $p < .001$. Finally, traits loaded on to the *d*-factor in this sample to the extent that they
838 generated cross-character assortment, $\beta = 0.62$, $SE = 0.09$, $p < .001$. Overall, results from this
839 newlywed sample, even when stripped of participant's own rating biases, precisely mirror those
840 observed in the primary analyses and suggest that the *d*-factor and its specificity are not
841 byproducts of participant self-report biases.

842 However, one shortcoming of this newlywed sample is that the trait measurements in this
843 sample overlap only partially with the traits included in the human cross-cultural sample. For

844 example, kindness is likely captured by variables such as “warm” and “selfish” and intelligence
845 by “intelligent” and “analytical.” However, in the newlywed sample, we did not analyze
846 variables corresponding to health, wealth, or physical attractiveness. For this reason, we finally
847 analyzed a third human sample. This sample contained data on $n = 382$ people who were
848 members of 191 romantic, heterosexual dyads. Participants were $M = 49.86$ years old ($SD =$
849 14.48) and were in their relationships for $Mdn = 216.7$ months. Participants reported their ideal
850 preferences in a long-term, committed romantic partner on 20 7-point bipolar adjective scales.
851 These scales included the five dimensions collected in the cross-cultural sample in addition to
852 others, including characteristics such as masculinity/femininity, religiosity, and desire for a
853 family. Participants additionally rated themselves and their romantic partner on each of these
854 dimensions. We averaged these self- and partner-reports to create a trait composite for each
855 participant across all 20 dimensions and conducted the same analyses on these 20 dimensions as
856 in the agent-based models, cross-cultural sample, and newlywed sample.

857 Again, the same pattern of effects observed in the primary agent-based models and the
858 cross-cultural sample emerged in this sample even when trait ratings were not based exclusively
859 on self-report. The d -factor explained $r^2 = 25.52\%$ of the variance in trait-level desirability.
860 Scores on the d -factor strongly predicted participant mate value, $\beta = 0.95$, $SE = 0.02$, $p < .001$
861 (Figure S12). Finally, traits loaded on to the d -factor in this sample to the extent that generated
862 cross-character assortment, $\beta = .62$, $SE = 0.09$, $p < .001$. Finally, precisely same pattern of results
863 still emerges in this sample even when analyses are based exclusively on partner-reports—
864 excluding self-reports entirely (Figure S12). The d -factor still explains $r^2 = 25.52\%$ of the
865 variance in trait-level desirability; d -factor scores strongly predict participant mate value ($\beta =$

866 0.95, $SE = 0.02$, $p < .001$); and traits that generate more cross-character assortment loaded more
867 strongly onto the d -factor ($\beta = 0.55$, $SE = 0.14$, $p < .001$).

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Supplementary Figure Legends

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Figure S1. Results from agent-based models in which traits and preferences are imperfectly heritable. The d -factor still evolves to explain an above-chance level of variation in trait-level desirability when heritability is set to 75% (a) or 50% (d). Agent mate value strongly predicts factor scores when heritability is set to 75% (b) or 50% (e). Finally, traits generate greater cross-character assortment still load more strongly onto the d -factor (c: $h^2 = 75\%$; f: $h^2 = 50\%$). Colored lines represent individual model runs; black lines represent average trends across model runs.

Figure S2. Results from the agent-based model in which agents do not mate assortatively for mate value. The d -factor explains only a small proportion of the variance in trait-level desirability (a). Agent mate value only weakly predicts agent scores on the d -factor (b). Finally, traits that generated more cross-character assortment do not load more strongly onto the d -factor (c). Colored lines represent individual model runs; black lines represent average trends across model runs.

Figure S3. Results of an agent-based model wherein traits are determined by an underlying condition variable. The d -factor explains a large proportion of the variance in trait-level desirability (a). Additionally, d -factor scores are strongly related to agent mate value (b), however the relationship is less consistent across model runs than in the primary agent-based model. Finally, the degree to which traits generate greater cross-character assortment is not correlated with loadings onto the d -factor (c).

Figure S4. Results from the cross-cultural data when trait ratings are scrambled so as to remove their correlational structure. The d -factor based on this scrambled data explains a relatively small proportion of the variance in trait-level desirability (a). Factor scores on this scrambled d -factor relate only weakly to participant mate values calculated from the same data (b). Finally, desirabilities generate greater cross-character assortment do not load onto the d -factor more strongly overall (c).

Figure S5. Results from the primary agent-based model when agent trait values are scrambled so as to remove their correlational structure. The d -factor based on this scrambled data explains a relatively small proportion of the variance in trait-level desirability (a). Factor scores on this scrambled d -factor relate only weakly to agent mate values calculated from the same data (b). Finally, desirabilities that generate greater cross-character assortment do not load more strongly onto the d -factor (c).

Figure S6. Results from the cross-cultural sample when mate value and the d -factor are calculated based on deviations from a random point. The random-point d -factor still explains a moderate proportion of the variance in trait-level desirability (a). However, the relationship between d -factor scores and mate value is much weaker than in the primary analyses (b) and the relationship between factor loadings and cross-character assortment is less consistent (c).

Figure S7. Results from primary agent-based model when agent mate value and the d -factor are calculated based on deviations from a random point. The random-point d -factor still explains a

915 moderate proportion of the variance in trait-level desirability (a). However, the relationship
916 between d -factor scores and mate value is much weaker than in the primary model (b) and the
917 relationship between factor loadings and cross-character assortment is less consistent (c).
918

919 **Figure S8.** The relationship between d -factor scores and participant mate value (a) and
920 desirability factor loading and predictive power (b) in the newlywed sample. Even when
921 participant trait ratings are not based exclusively on self-report, mate value strongly predicts d -
922 factor scores and traits load onto the d -factor to the extent that they generate cross-character
923 assortment.
924

925 **Figure S9.** Analysis of the newlywed sample when desirabilities are calculated based on
926 scrambled traits. Scores on this scrambled d -factor do relate significantly but weakly to
927 participant mate values (a). Furthermore, traits that generate stronger cross-character assortment
928 do not tend to load more strongly onto the scrambled d -factor (b).
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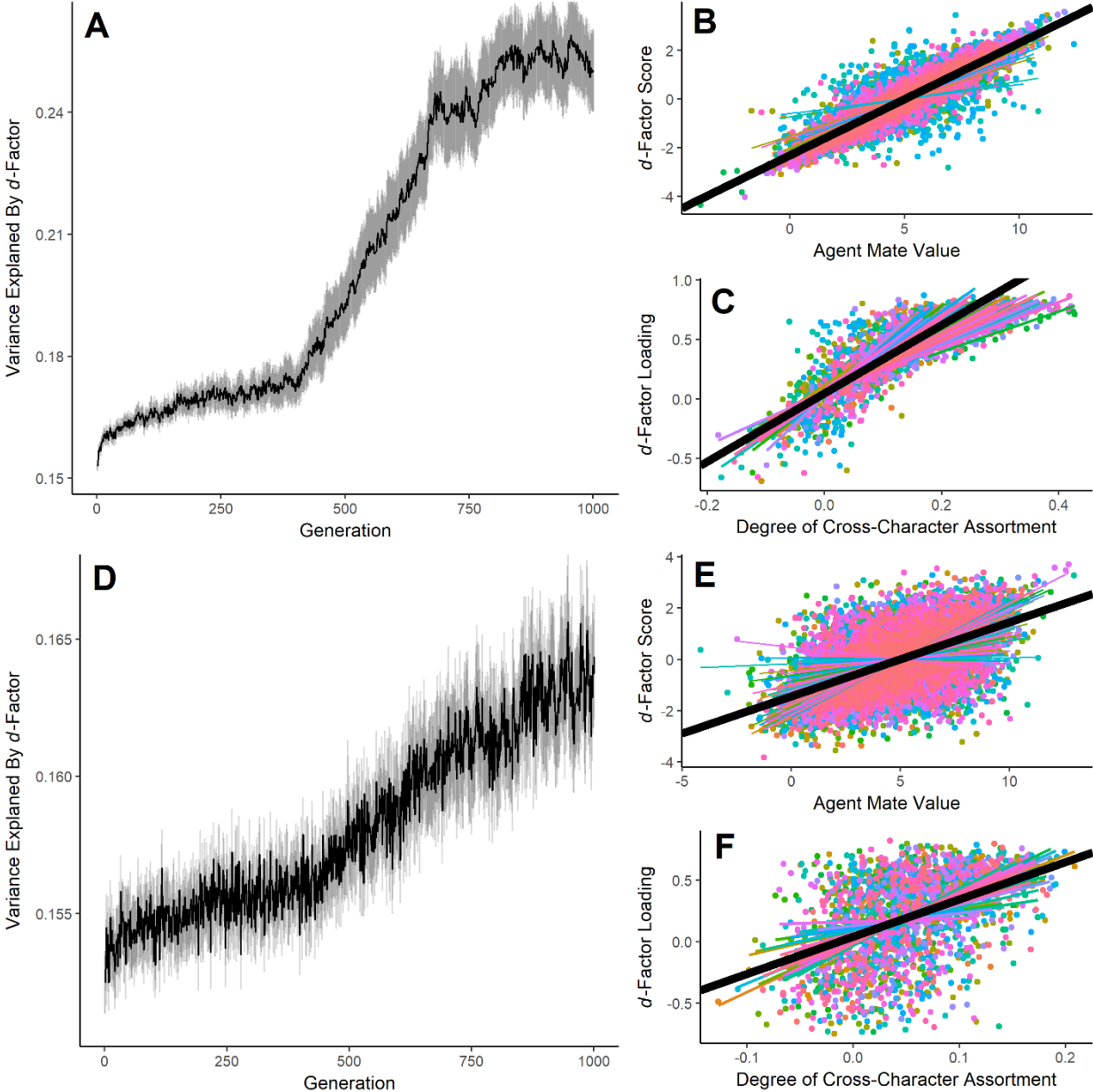
930 **Figure S10.** Analysis of the newlywed sample when desirabilities are calculated with respect to
931 100 random points rather than the opposite sex's mate preferences. Participant mate value does
932 predict scores on this random-point d -factor, but only weakly (a). Furthermore, traits that
933 generate more cross-character assortment do load more strongly onto the random-point d -factor,
934 however this relationship is less consistent than observed in the primary analyses (b).
935

936 **Figure S11.** The relationship between d -factor scores and participant mate value (a) and
937 desirability factor loading and predictive power (b) in the newlywed sample when trait
938 composites are based only on third-party interviewer report. Mate value still strongly predicts d -
939 factor scores and traits load onto the d -factor to the extent that they generate cross-character
940 assortment.
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942 **Figure S12.** The relationship between d -factor scores and participant mate value (a) and
943 desirability factor loading and predictive power (b) in a sample of romantic dyads in which trait
944 ratings are based on composites of self- and partner-report. Mate value still strongly predicts d -
945 factor scores and traits load onto the d -factor to the extent that they predict partner mate value.
946 The same results emerge when trait ratings are based exclusively on partner report (c and d).
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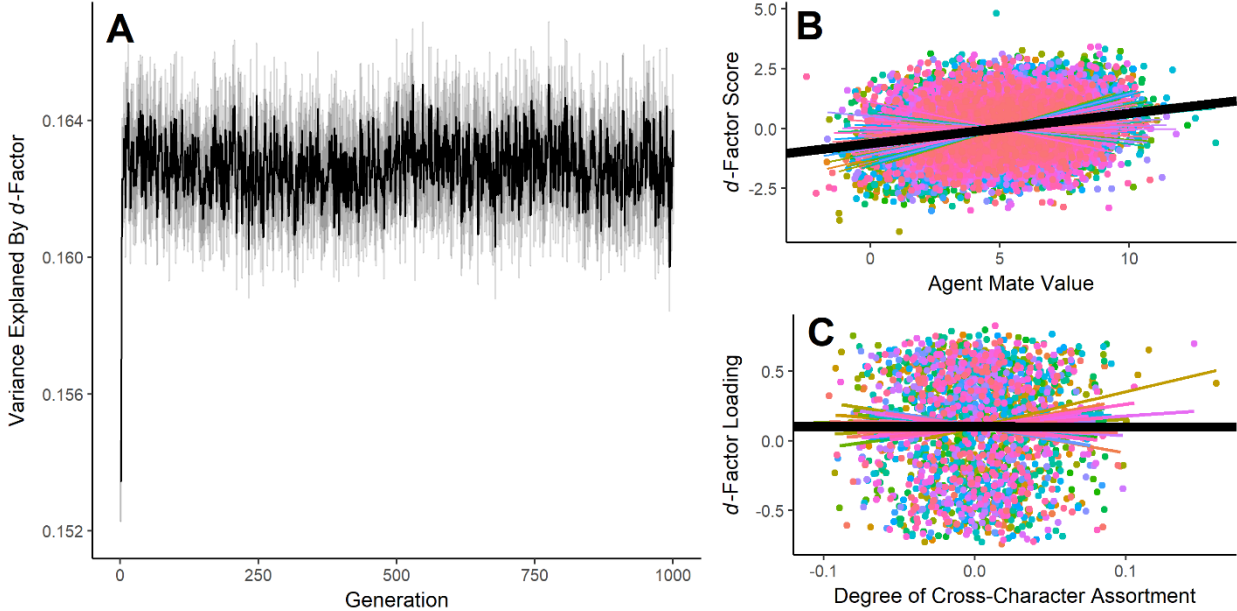
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Supplementary Figures



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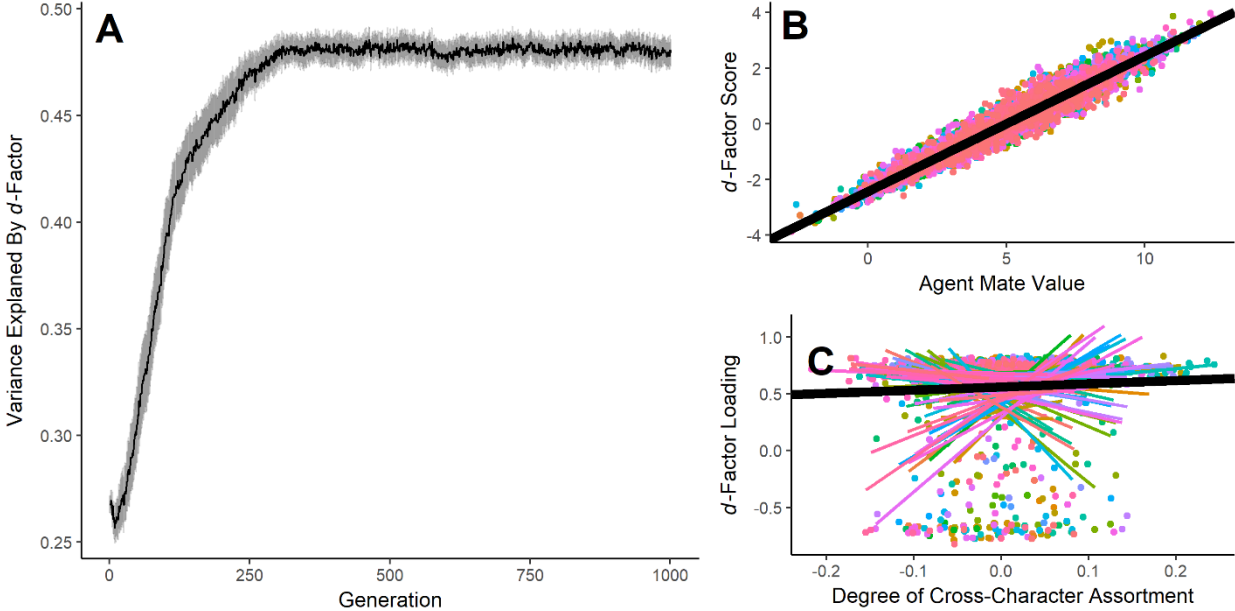
951 Figure S1



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953 Figure S2

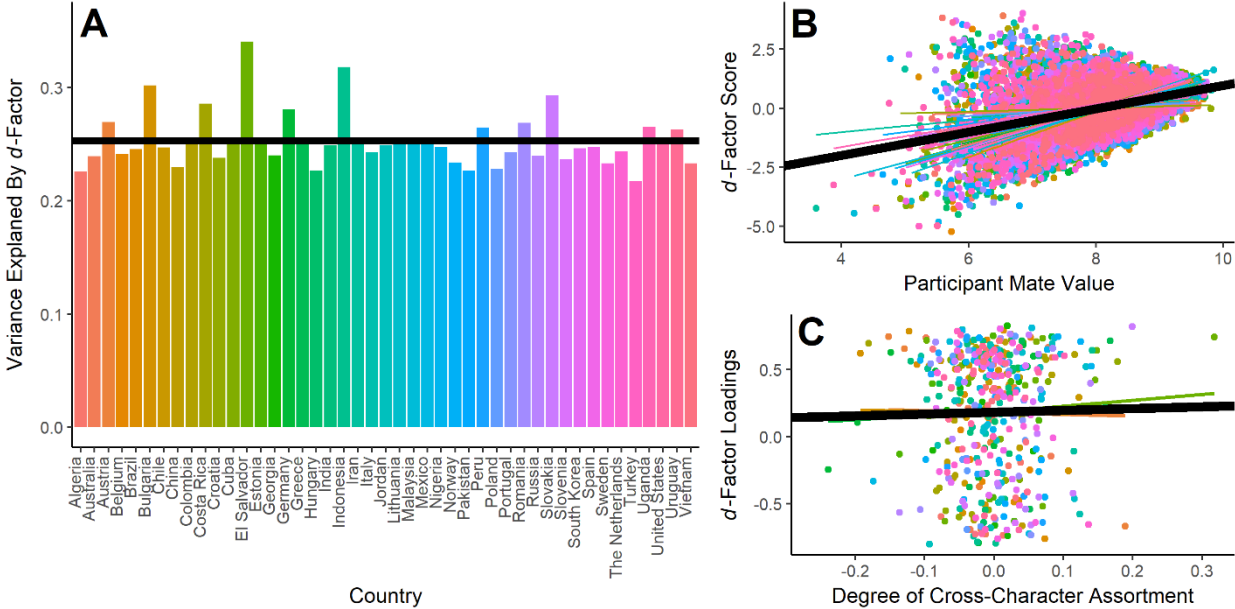
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956 Figure S3

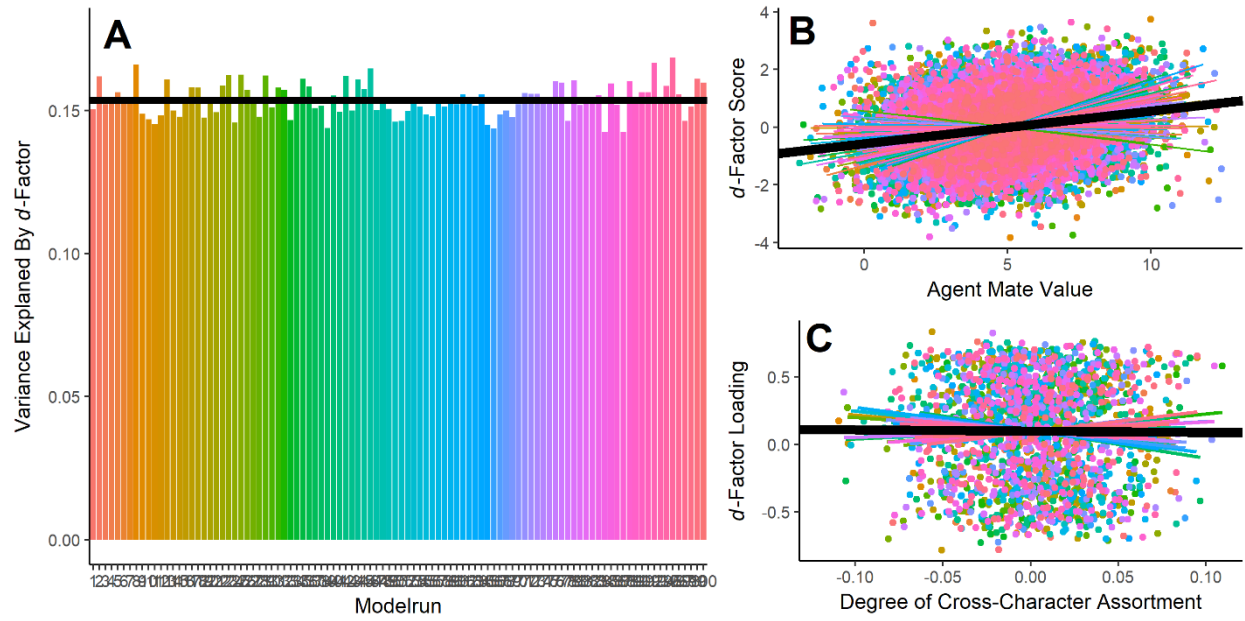
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959 Figure S4

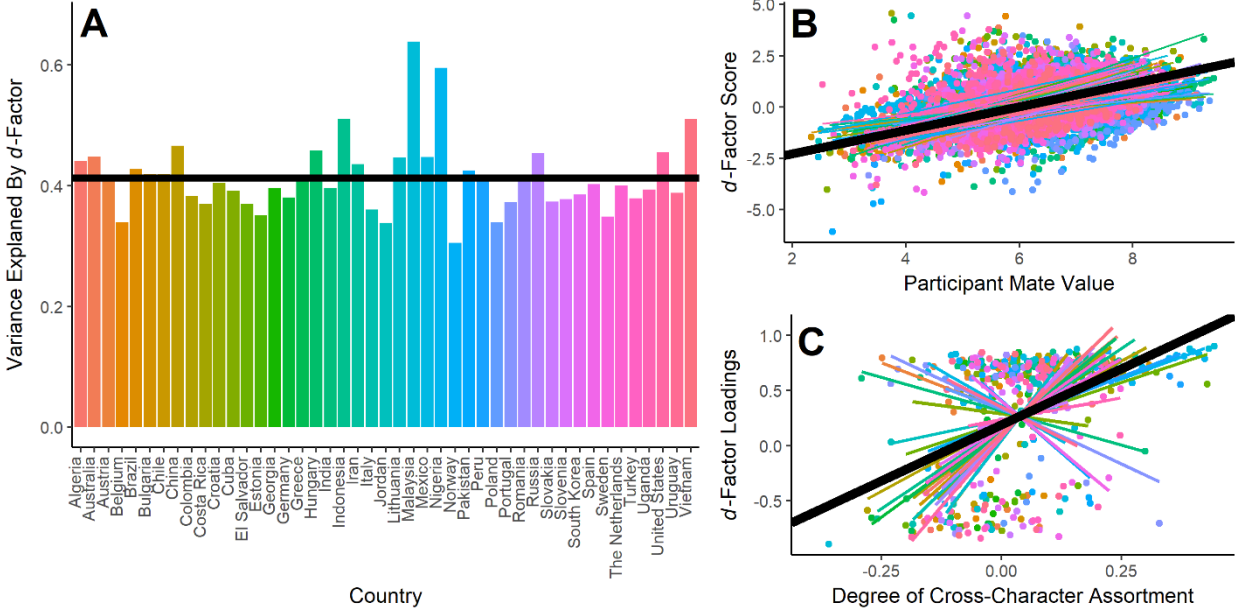
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962 Figure S5

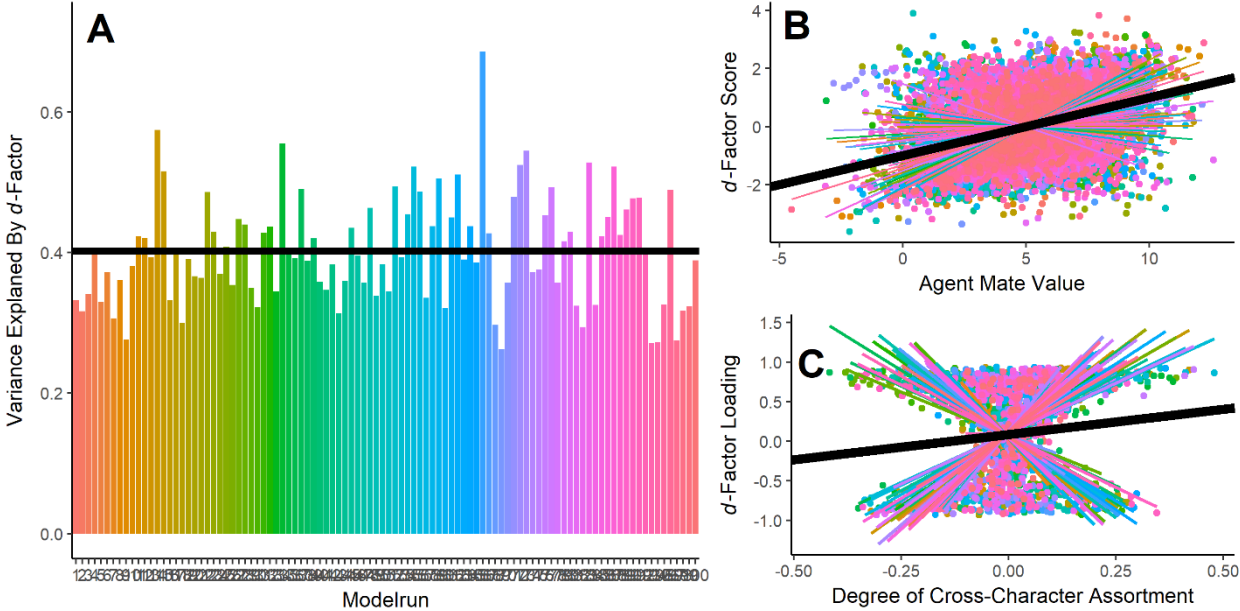
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965 Figure S6

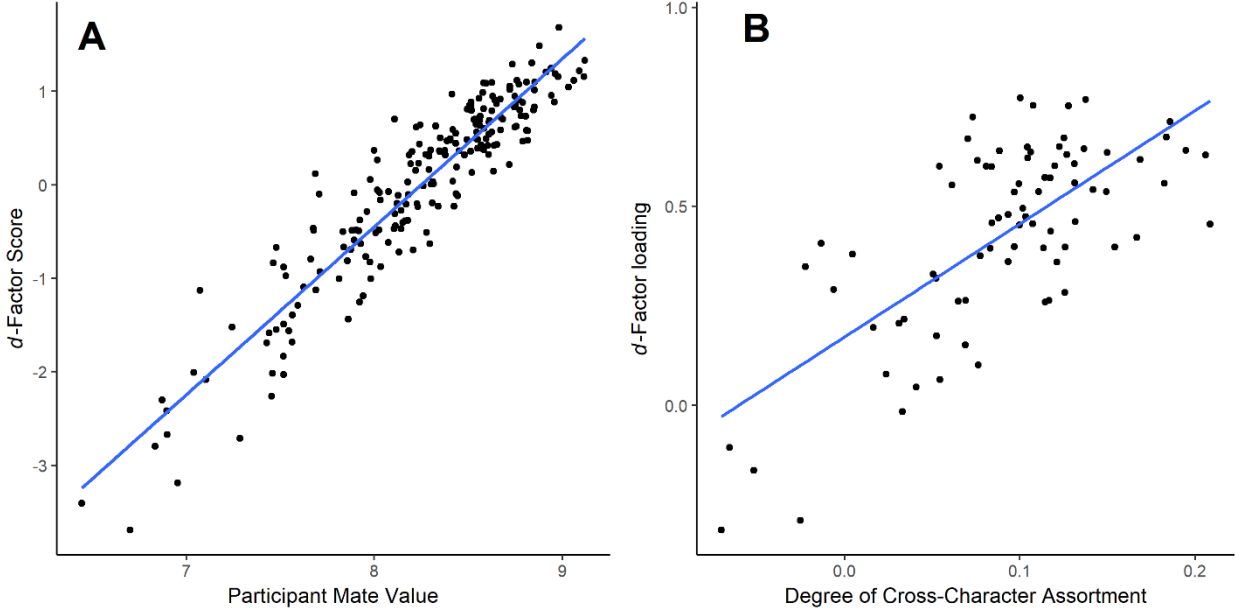
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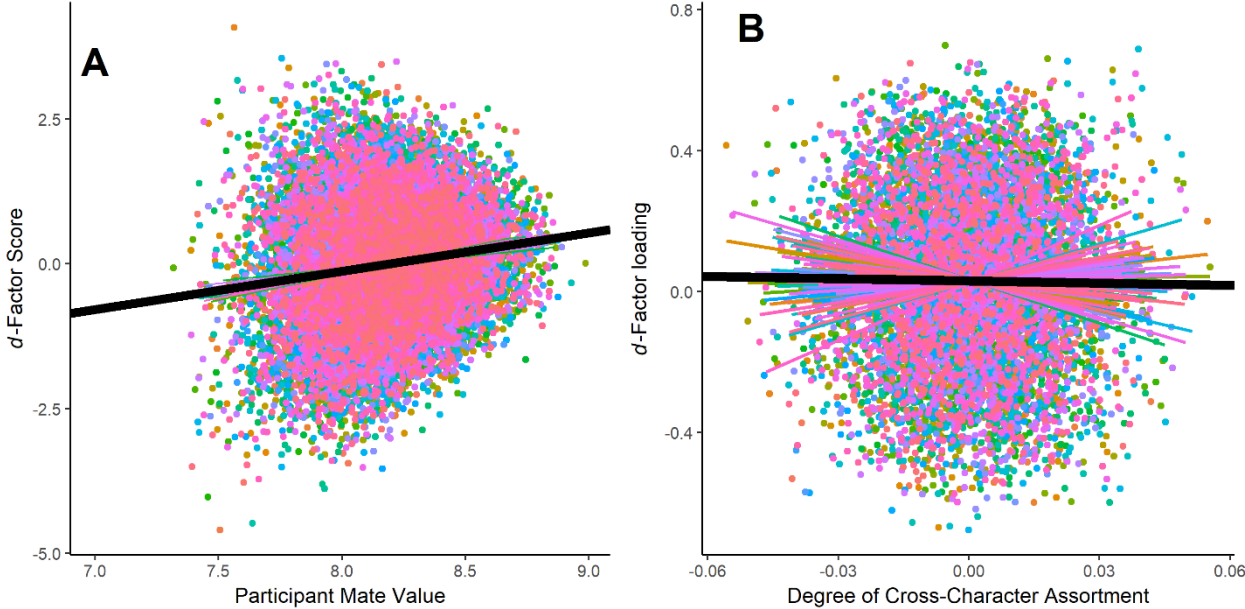
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971 Figure S8

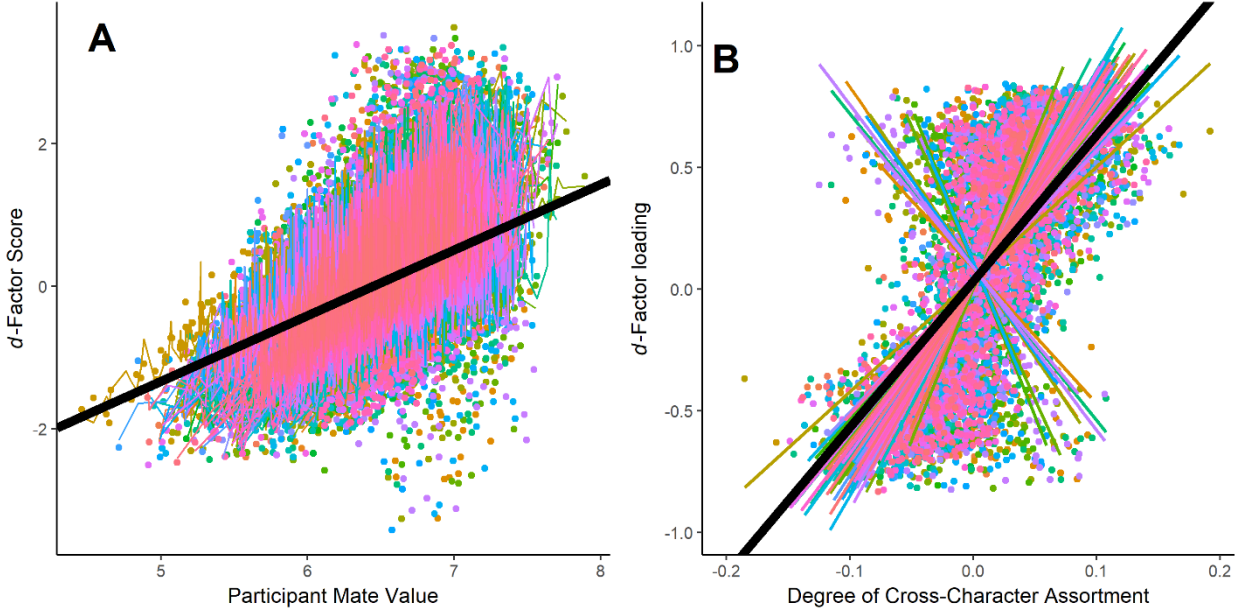
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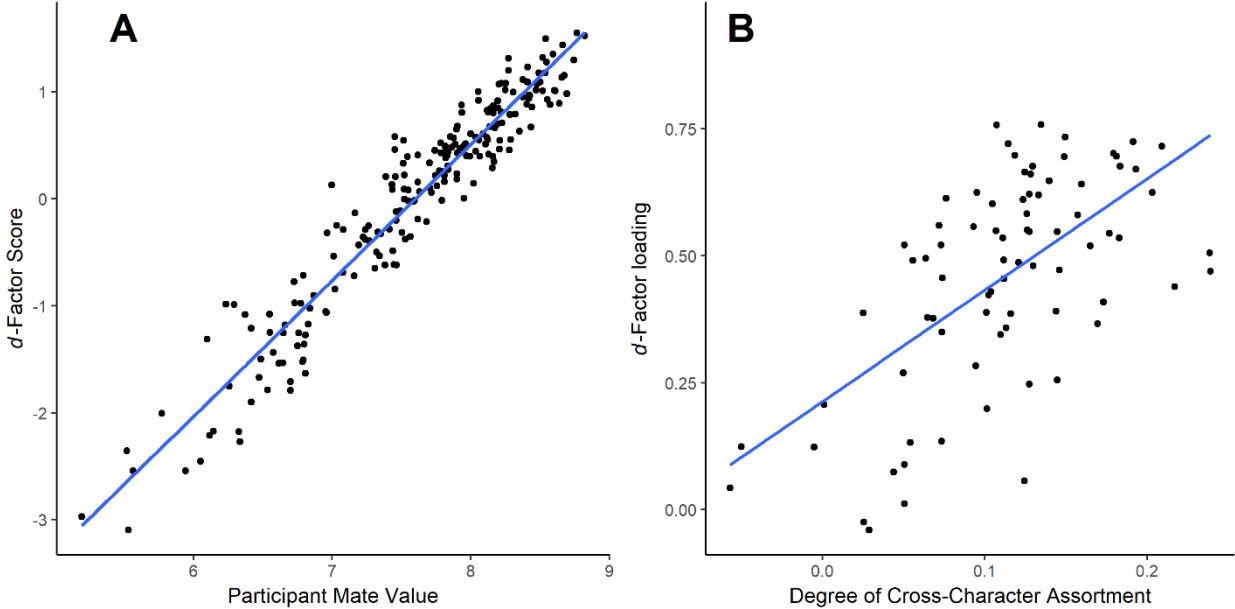
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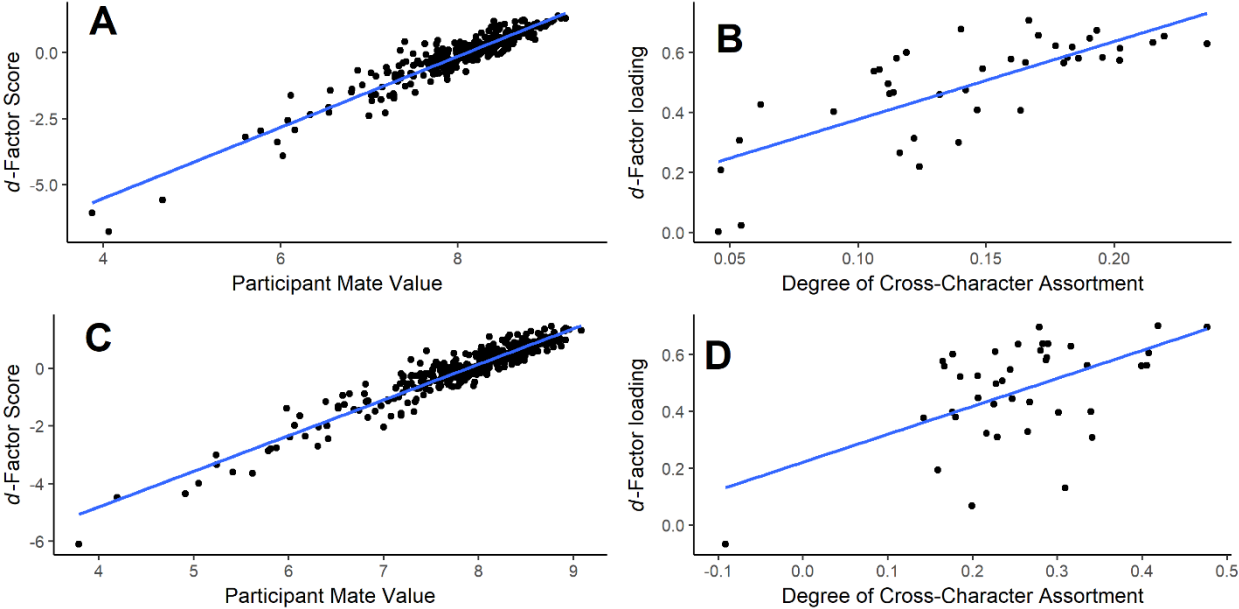
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980 Figure S11

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983 Figure S12

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985 **Table S1**986 *Correlations between putatively desirable traits in the prior literature*

Reference	Desirable Trait Correlation	Correlation Found?
Ashton, M. C., et al. (2000). Fluid intelligence, crystalized intelligence, and the openness/intellect factor. <i>Journal of Research in Personality</i> (34), 198-207.	Fluid intelligence & Openness (+)	YES
Ashton, M. C., et al. (2000). Fluid intelligence, crystalized intelligence, and the openness/intellect factor. <i>Journal of Research in Personality</i> (34), 198-207.	Crystalized intelligence & Openness (+)	YES
Ashton, M. C., et al. (2000). Fluid intelligence, crystalized intelligence, and the openness/intellect factor. <i>Journal of Research in Personality</i> (34), 198-207.	Overall intelligence & Openness (+)	YES
Banks, G. C., et al. (2010). Smarter people are (a bit) more symmetrical: a meta-analysis of the relationship between intelligence and fluctuating asymmetry. <i>Intelligence</i> (38), 393-401.	Intelligence (g-factor) & overall bilateral bodily symmetry (+) (meta-analytic)	YES
Bourdage, J. S., et al. (2007). Big Five and HEXACO model personality correlates of sexuality. <i>Personality and Individual Differences</i> (43), 1506-1516.	Relationship exclusivity & Honesty-Humility (+)	YES
Bourdage, J. S., et al. (2007). Big Five and HEXACO model personality correlates of sexuality. <i>Personality and Individual Differences</i> (43), 1506-1516.	Relationship exclusivity & Conscientiousness (+)	YES
Bourdage, J. S., et al. (2007). Big Five and HEXACO model personality correlates of sexuality. <i>Personality and Individual Differences</i> (43), 1506-1516.	Relationship exclusivity & Agreeableness (+)	YES
Dunkel, C. S. (2013). The general factor of personality and general intelligence: evidence for a substantial association. <i>Intelligence</i> (41), 423-427.	General factor of personality (GFP; substantive correlations among the socially desirable big five personality traits) (all +)	YES
Dunkel, C. S. (2013). The general factor of personality and general intelligence: evidence for a substantial association. <i>Intelligence</i> (41), 423-427.	General factor of personality (GFP) & Intelligence (g-factor) (+)	YES
Escasa, M., et al. (2010). Male traits associated with attractiveness in Conambo, Ecuador. <i>Evolution and Human Behavior</i> (31), 193-200.	Physical attractiveness & Warriorship (+)	YES
Escasa, M., et al. (2010). Male traits associated with attractiveness in Conambo, Ecuador. <i>Evolution and Human Behavior</i> (31), 193-200.	Physical attractiveness & Hunting ability (+)	YES
Escasa, M., et al. (2010). Male traits associated with attractiveness in Conambo, Ecuador. <i>Evolution and Human Behavior</i> (31), 193-200.	Physical attractiveness & Social rank (+)	YES
Feingold, A. (1992). Good-looking people are not what we think. <i>Psychological bulletin</i> , 111(2), 304.	Physical Attractiveness & Mental Ability (+) (meta-analytic)	NO
Figueredo, A. J., et al. (2014). A psychometric assessment of human life history strategy: A meta-analytic construct validation. <i>Evolutionary Behavioral Sciences</i> (8), 148-185.	GFP & Intelligence (g-factor) (+) (meta-analytic)	YES

Figueredo, A. J., et al. (2014). A psychometric assessment of human life history strategy: A meta-analytic construct validation. <i>Evolutionary Behavioral Sciences</i> (8), 148-185.	GFP & emotional intelligence (+) (meta-analytic)	YES
Figueredo, A. J., et al. (2014). A psychometric assessment of human life history strategy: A meta-analytic construct validation. <i>Evolutionary Behavioral Sciences</i> (8), 148-185.	GFP & Cooperativeness (+) (meta-analytic)	YES
Figueredo, A. J., et al. (2014). A psychometric assessment of human life history strategy: A meta-analytic construct validation. <i>Evolutionary Behavioral Sciences</i> (8), 148-185.	GFP & Antagonism (-) (meta-analytic)	YES
Figueredo, A. J., et al. (2014). A psychometric assessment of human life history strategy: A meta-analytic construct validation. <i>Evolutionary Behavioral Sciences</i> (8), 148-185.	Intelligence (g-factor) & emotional intelligence (+) (meta-analytic)	YES
Figueredo, A. J., et al. (2014). A psychometric assessment of human life history strategy: A meta-analytic construct validation. <i>Evolutionary Behavioral Sciences</i> (8), 148-185.	Intelligence (g-factor) & Antagonism (-) (meta-analytic)	YES
Figueredo, A. J., et al. (2014). A psychometric assessment of human life history strategy: A meta-analytic construct validation. <i>Evolutionary Behavioral Sciences</i> (8), 148-185.	Intelligence (g-factor) & Cooperativeness (+) (meta-analytic)	YES
Fink, B., et al. (2005). Facial symmetry and the 'big five' personality factors. <i>Personality and Individual Differences</i> (39), 523-529.	Facial symmetry & Extraversion (+)	YES
Fink, B., et al. (2005). Facial symmetry and the 'big five' personality factors. <i>Personality and Individual Differences</i> (39), 523-529.	Facial symmetry & Neuroticism (-)	YES
Fink, B., et al. (2016). Handgrip strength and the big five personality factors in men and women. <i>Personality and Individual Differences</i> (88), 175-177.	Physical strength & Extraversion (+)	YES
Fink, B., et al. (2016). Handgrip strength and the big five personality factors in men and women. <i>Personality and Individual Differences</i> (88), 175-177.	Physical strength & Neuroticism (-)	YES
Gurven, M., et al. (2013). How universal is the Big Five? Testing the five factor model among forager-farmers in the Bolivian Amazon. <i>Journal of Personality and Social Psychology</i> (104), 354-370.	Prosocial personality & Industriousness (+)	YES
Humphreys, L. G., et al. (1985). Longitudinal correlation analysis of standing height and intelligence. <i>Child Development</i> (56), 1465-1478.	Intelligence (g-factor) & Height (+)	YES
Judge, T. A. & Cable, D. M. (2004). The effect of physical height on workplace success and income: preliminary test of a theoretical model. <i>Journal of Applied Psychology</i> (89), 428-441.	Height & Occupational Advancement (+) (meta-analytic)	YES
Judge, T. A. & Cable, D. M. (2004). The effect of physical height on workplace success and income: preliminary test of a theoretical model. <i>Journal of Applied Psychology</i> (89), 428-441.	Height & Income (+) (meta-analytic)	YES

Judge, T. A., et al. (2009). Does it pay to be smart, attractive, or confident (or all three)? <i>Journal of Applied Psychology</i> (93), 742-755.	Intelligence (g-factor) & Physical attractiveness (+)	YES
Judge, T. A., et al. (2009). Does it pay to be smart, attractive, or confident (or all three)? <i>Journal of Applied Psychology</i> (93), 742-755.	Physical attractiveness & Income (+)	YES
Judge, T. A., et al. (2009). Does it pay to be smart, attractive, or confident (or all three)? <i>Journal of Applied Psychology</i> (93), 742-755.	Intelligence (g-factor) & Income (+)	YES
Judge, T. A., et al. (2009). Does it pay to be smart, attractive, or confident (or all three)? <i>Journal of Applied Psychology</i> (93), 742-755.	Intelligence (g-factor) & Level of education (+)	YES
Keller, M. C., et al. (2013). The genetic correlation between height and IQ: shared genes or assortative mating? <i>PLoS Genetics</i> (9), e1003451.	Intelligence (g-factor) & Height (+)	YES
Kerry, N. & Murray, D. R. (2018). Strong personalities: investigating the relationships between grip strength, self-perceived formidability, and the Big Five personality traits. <i>Personality and Individual Differences</i> (131), 216-221.	Physical strength & Neuroticism (-)	YES
Kerry, N. & Murray, D. R. (2018). Strong personalities: investigating the relationships between grip strength, self-perceived formidability, and the Big Five personality traits. <i>Personality and Individual Differences</i> (131), 216-221.	Physical strength & Extraversion (+ in one sample; null in another)	MIXED
Lukaszewski, A. W. & Roney, J. R. (2011). The origins of extraversion: joint effects of facultative calibration and genetic polymorphism. <i>Personality and Social Psychology Bulletin</i> (37), 409-421.	Physical strength & Physical attractiveness (+)	YES
Lukaszewski, A. W. & Roney, J. R. (2011). The origins of extraversion: joint effects of facultative calibration and genetic polymorphism. <i>Personality and Social Psychology Bulletin</i> (37), 409-421.	Physical attractiveness & Extraversion (+)	YES
Lukaszewski, A. W. & Roney, J. R. (2011). The origins of extraversion: joint effects of facultative calibration and genetic polymorphism. <i>Personality and Social Psychology Bulletin</i> (37), 409-421.	Physical strength & Extraversion (+)	YES
Lukaszewski, A. W. (2013). Testing an adaptationist theory of trait covariation: relative bargaining power as a common calibrator of an interpersonal syndrome. <i>European Journal of Personality</i> (27), 319-410.	Physical strength & Physical attractiveness	YES
Lukaszewski, A. W. (2013). Testing an adaptationist theory of trait covariation: relative bargaining power as a common calibrator of an interpersonal syndrome. <i>European Journal of Personality</i> (27), 319-410.	Physical attractiveness & Extraversion (+)	YES
Lukaszewski, A. W. (2013). Testing an adaptationist theory of trait covariation: relative bargaining power as a common calibrator of an interpersonal syndrome. <i>European Journal of Personality</i> (27), 319-410.	Physical attractiveness & Negative emotionality (-)	YES

Lukaszewski, A. W. (2013). Testing an adaptationist theory of trait covariation: relative bargaining power as a common calibrator of an interpersonal syndrome. <i>European Journal of Personality</i> (27), 319-410.	Physical attractiveness & Secure attachment (+)	YES
Lukaszewski, A. W. (2013). Testing an adaptationist theory of trait covariation: relative bargaining power as a common calibrator of an interpersonal syndrome. <i>European Journal of Personality</i> (27), 319-410.	Physical attractiveness & Attachment anxiety (-)	YES
Lukaszewski, A. W. (2013). Testing an adaptationist theory of trait covariation: relative bargaining power as a common calibrator of an interpersonal syndrome. <i>European Journal of Personality</i> (27), 319-410.	Physical attractiveness & Interpersonal trust (+)	YES
Lukaszewski, A. W. (2013). Testing an adaptationist theory of trait covariation: relative bargaining power as a common calibrator of an interpersonal syndrome. <i>European Journal of Personality</i> (27), 319-410.	Physical strength & Extraversion (+)	YES
Lukaszewski, A. W. (2013). Testing an adaptationist theory of trait covariation: relative bargaining power as a common calibrator of an interpersonal syndrome. <i>European Journal of Personality</i> (27), 319-410.	Physical strength & Negative emotionality (-)	YES
Lukaszewski, A. W. (2013). Testing an adaptationist theory of trait covariation: relative bargaining power as a common calibrator of an interpersonal syndrome. <i>European Journal of Personality</i> (27), 319-410.	Physical strength & Secure attachment (+)	YES
Lukaszewski, A. W. (2013). Testing an adaptationist theory of trait covariation: relative bargaining power as a common calibrator of an interpersonal syndrome. <i>European Journal of Personality</i> (27), 319-410.	Physical strength & Attachment anxiety (-)	YES
Lukaszewski, A. W. (2013). Testing an adaptationist theory of trait covariation: relative bargaining power as a common calibrator of an interpersonal syndrome. <i>European Journal of Personality</i> (27), 319-410.	Physical strength & Interpersonal trust (+)	YES
Manson, J. H. (2015). Life history strategy and the HEXACO personality dimensions. <i>Evolutionary Psychology</i> (13), 48-66.	Extraversion & Investment in kin (+)	YES
Manson, J. H. (2015). Life history strategy and the HEXACO personality dimensions. <i>Evolutionary Psychology</i> (13), 48-66.	Agreeableness & Investment in kin (+)	YES
Manson, J. H. (2015). Life history strategy and the HEXACO personality dimensions. <i>Evolutionary Psychology</i> (13), 48-66.	Altruism & Extraversion (+)	YES
Manson, J. H. (2015). Life history strategy and the HEXACO personality dimensions. <i>Evolutionary Psychology</i> (13), 48-66.	Altruism & Honesty-Humility (+)	YES
Manson, J. H. (2015). Life history strategy and the HEXACO personality dimensions. <i>Evolutionary Psychology</i> (13), 48-66.	Altruism & Conscientiousness (+)	YES

Mathes, E. W. & Kahn, A. (1975). Physical attractiveness, happiness, neuroticism, and self-esteem. <i>The Journal of Psychology</i> (90), 27-30.	Physical attractiveness & Neuroticism (-)	YES
Mathes, E. W. & Kahn, A. (1975). Physical attractiveness, happiness, neuroticism, and self-esteem. <i>The Journal of Psychology</i> (90), 27-30.	Physical attractiveness & self-esteem (+)	YES
Meier, B. P., et al. (2010). Are sociable people more beautiful? A zero-acquaintance analysis of agreeableness, extraversion, and attractiveness. <i>Journal of Research in Personality</i> (44), 293-296.	Physical attractiveness & Agreeableness (+)	YES
Meier, B. P., et al. (2010). Are sociable people more beautiful? A zero-acquaintance analysis of agreeableness, extraversion, and attractiveness. <i>Journal of Research in Personality</i> (44), 293-296.	Physical attractiveness & Extraversion (+)	YES
Meier, B. P., et al. (2010). Are sociable people more beautiful? A zero-acquaintance analysis of agreeableness, extraversion, and attractiveness. <i>Journal of Research in Personality</i> (44), 293-296.	Agreeableness & Extraversion (+)	YES
Mitchem, D. G., et al. (2014). No relationship between intelligence and facial attractiveness in a large, genetically informative sample. <i>Evolution and Human Behavior</i> (36), 240-247.	Intelligence & Facial attractiveness (+)	NO
Morgan, A. B., & Lillienfeld, S. O. (2000). A meta-analytic review of the relationship between antisocial behavior and measures of executive function. <i>Clinical Psychology Review</i> (20), 113-126.	Executive functioning & Antisocial behavior (-)	YES
Ormel, J., et al. (1994). Common mental disorders and disability across cultures. <i>Journal of the American Medical Association</i> (272), 1741-1748	(meta-analytic) Overall physical health & Absence of ICD psychiatric disorders (+)	YES
Ormel, J., et al. (1994). Common mental disorders and disability across cultures. <i>Journal of the American Medical Association</i> (272), 1741-1748	Absence of physical disabilities & Absence of ICD psychiatric disorders (+)	YES
Phalane, K.G., et al. (2017). Facial appearance reveals immunity in African men. <i>Scientific Reports</i> (7), 7443.	Immune function (cytokine response) & Facial attractiveness (+)	YES
Prokosch, M. D., et al. (2005). Intelligence tests with higher g-loadings show higher correlations with body symmetry: Evidence for a general fitness factor mediated by developmental stability. <i>Intelligence</i> (33), 203-213.	Intelligence (g-factor) & overall bilateral bodily symmetry (+)	YES
Rantala, M. J., et al. (2012). Adiposity, compared with masculinity, serves as a more valid cue to immunocompetence in human mate choice. <i>Proceedings of the Royal Society of London B</i> (280), 20122495.	Immune function (antibody response) & Physical attractiveness (+)	YES
Roberts, B. W., et al. (2007). The power of personality: The comparative validity of personality traits, socioeconomic status, and cognitive ability for predicting important life outcomes. <i>Perspectives on Psychological Science</i> (2), 313-345.	Intelligence & Age at death (+)	YES
Roberts, B. W., et al. (2007). The power of personality: The comparative validity of personality traits, socioeconomic status, and cognitive ability for predicting important life outcomes. <i>Perspectives on Psychological Science</i> (2), 313-345.	(meta-analytic) Conscientiousness & Age at death (+)	YES
	(meta-analytic)	

Roberts, B. W., et al. (2007). The power of personality: The comparative validity of personality traits, socioeconomic status, and cognitive ability for predicting important life outcomes. <i>Perspectives on Psychological Science</i> (2), 313-345.	Neuroticism & Age at death (-) (meta-analytic)	YES
Roberts, B. W., et al. (2007). The power of personality: The comparative validity of personality traits, socioeconomic status, and cognitive ability for predicting important life outcomes. <i>Perspectives on Psychological Science</i> (2), 313-345.	Intelligence & Occupational success (+) (meta-analytic)	YES
Roberts, B. W., et al. (2007). The power of personality: The comparative validity of personality traits, socioeconomic status, and cognitive ability for predicting important life outcomes. <i>Perspectives on Psychological Science</i> (2), 313-345.	Positive personality traits & Occupational success (+) (meta-analytic)	YES
Schermer, J. A. & Vernon, P. A. (2010). The correlation between general intelligence (g), a general factor of personality (GFP), and social desirability. <i>Personality and Individual Differences</i> (48), 187-189.	General factor of personality (GFP; substantive correlations among the socially desirable big five personality traits) (all +)	YES
Schermer, J. A. & Vernon, P. A. (2010). The correlation between general intelligence (g), a general factor of personality (GFP), and social desirability. <i>Personality and Individual Differences</i> (48), 187-189.	General factor of personality (GFP) & Intelligence (g-factor) (+)	YES
Schulte, M. J. (2004). Emotional intelligence: not much more than g and personality. <i>Personality and Individual Differences</i> (37). 1059-1068.	Intelligence (g-factor) & Emotional intelligence (+)	YES
Schulte, M. J. (2004). Emotional intelligence: not much more than g and personality. <i>Personality and Individual Differences</i> (37). 1059-1068.	Intelligence (g-factor) & Neuroticism (-)	YES
Schulte, M. J. (2004). Emotional intelligence: not much more than g and personality. <i>Personality and Individual Differences</i> (37). 1059-1068.	Emotional intelligence & Agreeableness (+)	YES
Sell, A., et al. (2017). Cues of upper body strength account for most of the variance in men's bodily attractiveness. <i>Proceedings of the Royal Society of London B</i> (284), 20171819.	Physical strength & Physical (bodily) attractiveness (+)	YES
Shackelford, T.K. & Larsen, R. J. (1999). Facial attractiveness and physical health. <i>Evolution and Human Behavior</i> (20), 71-76.	Cardiovascular health & Facial attractiveness (+)	YES
Shackelford, T.K. & Larsen, R. J. (1999). Facial attractiveness and physical health. <i>Evolution and Human Behavior</i> (20), 71-76.	Respiratory health & Facial attractiveness (+)	YES
Shamosh, N. A. & Gray, J. R. (2008). Delay discounting and intelligence: a meta-analysis. <i>Intelligence</i> (36), 289-305.	Intelligence (g-factor) & delay discounting (i.e. future orientation) (+) (meta-analytic)	YES
Shoup, M. L. & Gallup, G. G., jr. (2008). Men's faces convey information about their bodies and their behavior: what you see is what you get. <i>Evolutionary Psychology</i> (6), 469-479.	Grip strength & Facial attractiveness (+)	YES
Silventonen, K., et al. (2006). Genetic contributions to the association between height and intelligence: evidence from Dutch twin data from childhood to	Intelligence (g-factor) & Height (+)	YES

middle age. <i>Genes, Brain, & Behavior</i> (5), 585-595.		
Sturgis, P., et al. (2010). Does intelligence foster generalized trust? An empirical test using the UK birth cohort studies. <i>Intelligence</i> (38), 45-54.	Intelligence & Interpersonal trust (+)	YES
Sutin, A. R., et al. (2009). Personality and career success: concurrent and longitudinal relations. <i>European Journal of Personality</i> (23), 71-84.	Emotional stability & Income (+)	YES
Sutin, A. R., et al. (2009). Personality and career success: concurrent and longitudinal relations. <i>European Journal of Personality</i> (23), 71-84.	Conscientiousness & Income (+)	YES
Van der Linden, D., et al. (2010). The General Factor of Personality: A meta-analysis of Big Five intercorrelations and a criterion-related validity study. <i>Journal of Research in Personality</i> (44), 315-327.	General factor of personality (GFP; substantive correlations among the socially desirable big five personality traits) (all +) (meta-analytic)	YES
Van der Linden, D., et al. (2012). Overlap between general factors of personality in the big five, giant three, and trait emotional intelligence. <i>Personality and Individual Differences</i> (53), 175-179.	General factor of personality (GFP; substantive correlations among the socially desirable big five personality traits) (all +)	YES
Van der Linden, D., et al. (2012). Overlap between general factors of personality in the big five, giant three, and trait emotional intelligence. <i>Personality and Individual Differences</i> (53), 175-179.	General factor of personality (GFP) & Emotional intelligence (+)	YES
Von Rueden, C. R. et al. (2015). Adaptive personality calibration in a human society: Effects of embodied capital on prosocial traits. <i>Behavioral Ecology</i> (26), 1071-1082.	Physical strength & Prosocial leadership orientation (+)	YES
Von Rueden, C. R. et al. (2015). Adaptive personality calibration in a human society: Effects of embodied capital on prosocial traits. <i>Behavioral Ecology</i> (26), 1071-1082.	Education level & prosocial leadership orientation (+)	YES
Von Rueden, C. R. et al. (2015). Adaptive personality calibration in a human society: Effects of embodied capital on prosocial traits. <i>Behavioral Ecology</i> (26), 1071-1082.	Physical strength & Education level (+)	YES
Weeden, J. & Sabini, J. (2005). Physical attractiveness and health in western societies: a review. <i>Psychological Bulletin</i> (131), 635-653.	Female Physical attractiveness & Female Health (+)	YES
Weeden, J. & Sabini, J. (2005). Physical attractiveness and health in western societies: a review. <i>Psychological Bulletin</i> (131), 635-653.	Male Physical attractiveness & Male Health (+)	NO
Weege, B., et al. (2015). Women's attractiveness perceptions of men's dance movements in relation to self-reported and perceived personality. <i>Evolutionary Psychological Science</i> (1), 223-27.	Dancing ability & Extraversion (+)	YES
Weege, B., et al. (2015). Women's attractiveness perceptions of men's dance movements in relation to self-reported and perceived personality. <i>Evolutionary Psychological Science</i> (1), 223-27.	Dancing ability & Neuroticism (-)	YES
Zebrowitz, L. A., et al. (2002). Looking smart and looking good: Facial cues to intelligence and their origins. <i>Personality and Social Psychology Bulletin</i> (28), 238-249.	Intelligence & Facial attractiveness (+)	YES
Zebrowitz, L. A., et al. (2002). Looking smart and looking good: Facial cues to intelligence and their origins.	Intelligence & Facial symmetry (+)	YES

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238-249.

988 **Table S2**989 *Correlations between traits and desirabilities in cross-cultural sample*

	1	2	3	4	5	6	7	8	9	10
1. Intelligence (T)		0.36	0.31	0.45	0.35	0.70	0.20	0.20	0.32	0.06
2. Kindness (T)	0.39		0.34	0.31	0.25	0.24	0.66	0.25	0.22	0.08
3. Health (T)	0.35	0.39		0.39	0.29	0.19	0.23	0.75	0.28	0.10
4. Phys. Att. (T)	0.45	0.34	0.46		0.44	0.30	0.20	0.28	0.80	0.12
5. Resources (T)	0.37	0.26	0.29	0.42		0.22	0.17	0.22	0.33	0.38
6. Intelligence (D)	0.72	0.29	0.24	0.30	0.27		0.26	0.23	0.36	0.20
7. Kindness (D)	0.27	0.80	0.30	0.26	0.22	0.34		0.29	0.26	0.19
8. Health (D)	0.26	0.30	0.80	0.35	0.24	0.29	0.33		0.31	0.20
9. Phys. Att. (D)	0.25	0.20	0.29	0.63	0.27	0.35	0.28	0.36		0.26
10. Resources (D)	0.19	0.14	0.17	0.23	0.64	0.29	0.20	0.25	0.36	

990 *Note: Data from male participants is below the diagonal; female participants are above the*991 *diagonal. T: absolute trait value; D: desirability.*