Indirect genetic effects: a key component of the genetic architecture of behaviour

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Behavioural ecology research increasingly focuses on why genetic behavioural variation can persist despite selection. Evolutionary theory predicts that directional selection leads to evolutionary change while depleting standing genetic variation. Nevertheless, evolutionary stasis may occur for traits involved in social interactions. This requires tight negative genetic correlations between direct genetic effects (DGEs) of an individual’s genes on its own phenotype and the indirect genetic effects (IGEs) it has on conspecifics, as this could diminish the amount of genetic variation available to selection to act upon. We tested this prediction using a pedigreed laboratory population of Mediterranean field crickets (Gryllus bimaculatus), in which both exploratory tendency and aggression are heritable. We found that genotypes predisposed to be aggressive (due to DGEs) strongly decreased aggressiveness in opponents (due to IGEs). As a consequence, the variance in total breeding values was reduced to almost zero, implying that IGEs indeed greatly contribute to the occurrence of evolutionary stasis. IGEs were further associated with genetic variation in a non-social behaviour: explorative genotypes elicited most aggression in opponents. These key findings imply that IGEs indeed represent an important overlooked mechanism that can impact evolutionary dynamics of traits under selection.

Behavioural ecologists increasingly focus on studying the adaptive processes maintaining individual differences in behaviour within animal populations. Several adaptive explanations have been proposed for why selection might maintain behavioural variation rather than erode it (reviewed by1–3). For example, frequency dependent selection1, temporal and spatial heterogeneity4, 5, or life-history trade-offs6–8 have all been implied to explain the stable coexistence of different behavioural ‘types’ within populations. It is implicitly assumed that genes carried by focal individuals contribute to behavioural differences, such that directional selection should both erode variance and cause a change (over generations) in mean phenotype9, 10. However, evolutionary theory also predicts that evolutionary stasis may occur despite directional selection in the presence of ‘indirect genetic effects’ (IGEs) generated by social interactions11–15. This key insight has largely been ignored in behavioural ecology theory explaining individual variation in behaviour, despite the fact that many behavioural traits are expressed as part of social interactions.

Quantitative genetic theory implies that social interactions can have major evolutionary repercussions, particularly when an individual’s phenotype is affected by the genotypes of conspecifics: these effects are called IGEs12, 13, 15. IGEs can greatly influence evolutionary processes when they are correlated with the direct genetic effects (DGEs) of an individual’s genotype on its own phenotype. For example, in mussel cultures, individuals genetically predisposed to grow quickly in competitive situations are also genetically predisposed to reduce growth in others by depriving them of feeding opportunities16. The resulting negative genetic correlation between DGEs and IGEs can impose major evolutionary constraints, by effectively reducing the amount of variation in total breeding value of a trait within a population17–19. The presence of IGEs may thus lead to evolutionary stasis in the phenotype, implying that directional selection does not necessarily lead to evolutionary change. Interestingly, positive genetic correlations between DGEs and IGEs are predicted to instead speed up the response to directional selection relative to expectations from classic evolutionary theory (e.g. refs 14 and 15). For example, a positive covariance between DGEs and IGEs on aggression in a study of deer mice (Peromyscus maniculatus) implies that this trait can evolve

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Results

Sources of variation in single traits. Exploration behaviour was significantly repeatable ($r = 0.45$) and heritable ($h^2 = 0.28$) (see also Santostefano et al. under review). Aggressiveness was also significantly repeatable ($r = 0.17$) and heritable ($h^2 = 0.05$), while it additionally harboured a significant opponent identity effect ($r = 0.11$) (see also Santostefano et al. under review; estimates re-printed in Table 1). Here we expanded upon these analyses by estimating IGEs on aggression and testing for their correlation with DGEs. Doing so, demonstrated that this opponent effect harboured a small, but significant, amount of genetic variation for focal aggression ($V_{IGE} = 0.026, SE = 0.017$) (Model 6, Table 1). In other words, there was genetic variation not just in the tendency of individuals to be aggressive, but also in the level of aggressiveness they elicited in their social context. Furthermore, the genetic correlation between DGEs and IGEs for aggression was strong and negative ($r = -0.83, SE = 0.37$) (Model 7, Table 1). AIC model comparison to simpler models also provided strongest support for this final model (Model 7, Table S1). In other words, individuals genetically predisposed towards expressing higher levels of aggression as a focal were also predisposed to suppress aggressiveness in their opponents. As a consequence of this tight negative genetic correlation, the estimated total heritable variation in aggression (also known in the literature as $r^2$) ($V_{TBV} / V_{TOT} = 0.016, SE = 0.030$; where $V_{TBV} = V_{DGE} + V_{IGE} + 2COV_{DGE,IGE} = 0.051 + 0.026 - 2 \times 0.030 = 0.016$; $V_{TOT} = 0.99$) was considerably smaller (namely, 3.19 times) than what ‘traditional’ estimates of heritability based on DGEs would (inappropriately) conclude ($h^2 = 0.051, SE = 0.024$).

Table 1. Results of the univariate mixed 'animal model' fitted to partition variation in aggressive behaviour with random intercepts for focal and opponent identity. Estimates of variance components and their correlations are given with associated standard errors. Random effects are expressed as the proportion of total phenotypic variation not attributable to fixed effects explained by each effect. Focal and opponent variances, as well as their covariance, are partitioned into environmental (PE) and genetic (G) components. For each model, variance terms are provided with a likelihood ratio test (LRT) between the given model and the previous model, with associated degrees of freedom (df) and values of $P$. The most parsimonious model (model 7) is denoted in bold face. *tested in addition over an equal mix of df = 1 and df = 2 (representing a test of variance and covariance together, against model 5), $\chi^2 = 7.88$, $p < 0.05$.

| Model | Variance $\sigma^2$ (SE) | Correlations r Foc-Opp (SE) | Test |
|-------|-------------------------|-----------------------------|------|
|       | Focal                   | Opponent                    |      |
|       | PE(f)                   | DGE                         | G    | LogL | $\chi^2$ | df | $P$ |
| 1     | —                       | —                           | —    | —    | —        | 1168.01 | —  | —   |
| 2     | 0.17 (0.02)             | —                           | —    | —    | —        | 1131.75 | 72.52 | 0/1 < 0.01 |
| 3     | 0.17 (0.02)             | 0.11 (0.02)                 | —    | —    | —        | 1116.89 | 29.27 | 0/1 < 0.01 |
| 4     | 0.17 (0.02)             | 0.11 (0.02)                 | —    | —    | —        | 1115.34 | 3.1  | 1.08 |
| 5     | —                       | 0.12 (0.03)                 | 0.05 (0.02) | 0.11 (0.02) | —    | —    | —        | —    | —    |
| 6     | —                       | 0.12 (0.03)                 | 0.05 (0.02) | 0.08 (0.03) | 0.03 (0.02) | —    | —    | —        | 1109.15 | 3.68 | 0/1 < 0.05 |
| 7     | 0.12 (0.03)             | 0.05 (0.02)                 | 0.08 (0.03) | 0.03 (0.02) | 0.71 (0.03) | 0.01 (0.18) | —    | —    | —        | 1107.05 | 4.2  | 1* < 0.05 |

very rapidly. This is because selection for increased aggression would drive the evolution of a social environment in which aggression is more readily elicited by interacting conspecifics. Therefore, IGEs arising from social interactions can both provide a source of additional genetic variation that either facilitates rapid selection responses or serves as a source of evolutionary constraint on phenotypes. However, to date IGEs have largely been ignored as a potential mechanism explaining evolutionary stasis in individual behaviour research.

IGEs are expected to exist on traits such as aggression and dominance, i.e., traits that are expressed explicitly as part of social interactions. Interestingly, IGEs can also affect the evolution of other aspects of phenotype, including behavioural traits not expressed within a social context, provided these covary genetically with traits that do harbour IGEs. For example, the literature on ‘behavioural syndromes’ often reports that traits expressed in social interactions (e.g., aggressiveness, sociability) are phenotypically correlated with other risky behaviours expressed in non-social contexts, such as exploratory tendency, or anti-predator boldness. Of course, these correlations are important for evolutionary dynamics only if they are underpinned by genetic processes. Thus, if IGEs are present for a social behaviour such as aggression, the evolution of any trait genetically correlated either with the social behaviour or its IGEs may be affected.

Here, we investigated whether IGEs contribute to the genetic architecture of behavioural variation expressed in social and non-social contexts. We repeatedly measured two behavioural traits (exploration, non-social, and aggression, social) in a pedigreed laboratory population of Mediterranean field crickets (Gryllus bimaculatus) descended from wild-caught grandparents. For the data presented in this paper, we show elsewhere that exploratory behaviour and aggressiveness are both repeatable and heritable (subject to DGEs) but not genetically correlated (Santostefano et al. under review). Here we expand upon these analyses by quantifying (i) whether IGEs also contributed to genetic variance in aggressiveness, (ii) whether, for aggressiveness, DGEs (tendency to act aggressively) and IGEs (tendency to elicit aggressiveness) were correlated, and (iii) whether IGEs on aggression were also correlated with DGEs for exploration, a trait not directly involved in social interactions. Our approach thus implies that drawing evolutionary predictions while ignoring IGEs not only on the focal trait, but also on other seemingly independent traits, can be greatly misleading.
Among-trait correlations. Multivariate models corroborated the strong negative genetic correlation between DGEs and IGEs on aggression ($r_G = -1.02$, SE 0.40, $P < 0.05$) (Table 2). We note this estimate is slightly greater than that presented above (though based on SE the confidence intervals will be strongly overlapping) and very slightly outside the permissible parameter space for a (true) correlation (we also note that not constraining the parameter space in the model fit allows better convergence and an estimate of the uncertainty associated with $r_G$). However, the genetic correlation between DGEs on exploration and DGEs on aggression was close to zero and non-significant ($r_G = -0.04$, SE 0.24, $P > 0.05$) (Table 2), contrary to predictions from the behavioural syndrome literature. Multivariate models also provided some evidence for a positive genetic correlation between IGEs on aggression and DGEs on exploration, although the estimated was marginally non-significant ($r_G = 0.59$, SE 0.28, $P = 0.056$) (Table 2).

Using the estimated $G$ matrix, we compared the fit of five model structures (considered a priori) using AIC (Table 3, Fig. 1). This approach is warranted because a multivariate rather than a pair-wise bivariate approach greatly increases statistical power. A model where both the correlation between DGEs on exploration and IGEs on aggression, as well as the correlation between DGEs and IGEs on aggression were included (Model 3) fitted the data best, consistent with our inferences from likelihood ratio tests derived from the multivariate model detailed in the main text. The direct genetic correlation between aggression and exploration was not included in this model, consistent with this correlation being close to zero in the full model estimated above. This full pattern is somewhat difficult to interpret since, given the magnitude of estimated correlations between IGEs for aggression and DGEs on both behaviours, we might have expected a stronger (direct) genetic correlation between aggression

### Table 2. Estimated additive genetic ($G$) covariances and correlations (with SE) between two behaviours (aggression and exploration), and IGEs on aggression. We present covariances (lower-off diagonals) and correlations (upper-off diagonals) for each set of traits. Correlations printed in bold-face are significant ($P < 0.05$) based on likelihood ratio tests derived from the multivariate model detailed in the main text.

|                | Aggressiveness (DGE) | Exploration (DGE) | Aggressiveness elicited (IGE) |
|----------------|----------------------|-------------------|-------------------------------|
| Aggressiveness (DGE) | —                    | $-0.04$ (0.24)    | $-1.02$ (0.40)               |
| Exploration (DGE)   | $-0.01$ (0.03)       | —                 | $0.59$ (0.28)                |
| Aggressiveness elicited (IGE) | $-0.04$ (0.02) | 0.05 (0.03)       | —                            |

### Table 3. Relative fit of five multivariate models differing in architecture of genetic correlations between direct genetic (DGE) and indirect genetic (IGE) effects based on the Akaike's information criterion (AIC). We present each model's AIC-value relative to the model with the lowest AIC-value ($\Delta$AIC), its weight, and relative likelihood. Model denominations refer to Fig. 1: A is the correlation between DGEs and IGEs on aggressiveness; B is the correlation between DGEs on exploration and DGEs on aggressiveness; C is the correlation between DGEs on exploration and IGEs on aggressiveness. Model 5 (the complete model) is presented in Table 2.

| Model          | $\Delta$AIC | Akaike Weight | Relative LL |
|----------------|-------------|---------------|-------------|
| 3. $B = 0$    | 0           | 0.78          | 1.00        |
| 4. $C = 0$    | 3.62        | 0.13          | 0.16        |
| 5. $A, B, C$ estimated | 5.49       | 0.05          | 0.06        |
| 1. $A, B, C = 0$ | 6.06      | 0.04          | 0.05        |
| 2. $A = 0$    | 8.64        | 0.01          | 0.01        |

### Figure 1. Correlation structure of the five hypothesized multivariate model structures presented in Table 3 (detailed in the Methods). A is the correlation between DGEs and IGEs on aggressiveness; B is the correlation between DGEs on exploration and DGEs on aggressiveness; C is the correlation between DGEs on exploration and IGEs on aggressiveness. Estimated correlations with corresponding SEs derived from the full model (Model 5, presented in Table 2) are shown with each arrow; bolded arrows represent paths with statistical support from the LRT and AIC.
and exploration. As this was not the case, it is possible that the IGEs and DGEs for aggression are not as tightly correlated as implied by the point estimate (see also our discussion above). With this caveat noted, we find by AIC comparison that individuals with a high genetic merit for explorative tendency in novel environments tended to elicit more aggression (Table 3, Fig. 1). Taken together with the strong (and significant) genetic correlation between DGEs and IGEs on aggression (Table 1; Table 2), we view this as evidence that the social environment can indeed influence the evolution of behaviours including those expressed outside the social context.

Discussion
This study investigated a largely overlooked mechanism, indirect genetic effects, which may contribute to the observed behavioural variation in social traits under selection and impact their evolutionary dynamics. Our study on male Mediterranean field crickets confirmed that the phenotypic expression of aggression and exploration was repeatable, and showed that the former depended on opponent, as well as focal identity. Both behaviours harboured additive genetic variance, but—importantly—heritable variation in focal aggressiveness arose jointly from the genotypes of the focal (DGEs) and opponents (IGEs) (Table 1). As aggressiveness represents an important component of an often-documented "aggression-boldness syndrome"26, the evolutionary consequences of these IGEs may extend to other associated traits. Indeed, we found evidence for a genetic architecture suggesting that the evolution of a non-social trait such as exploration may not be independent from the evolution of a social trait, and vice versa, given that its DGEs were correlated with the IGEs acting on aggression. Our study therefore identifies IGEs as an important overlooked component of the (multivariate) genetic architecture of behaviour that should be considered when making predictions on the evolution of traits studied in 'personality' research. Our results generally imply that IGEs can have consequences for the evolutionary trajectories of a wide range of traits, including those not expressed as part of social interactions (e.g., exploratory tendency, body size, etc.).

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The sign of this correlation can also be interpreted in terms of behavioural feedback processes and the functional role of aggression. For example, in species (or contexts) where individuals escalate agonistic behaviour through positive feedbacks (i.e. aggression elicits aggression14) direct-indirect (genetic) covariance will be positive. Conversely, negative correlations arise when aggression is asymmetric, being directed by more competitive (or dominant) individuals towards subordinate social partners. This is because, in a dyadic contest, a genotype predisposing to contest winning by the focal will necessarily predispose to losing when encountered in an opponent20,28,31,32. Thus, the negative genetic correlation found here actually suggests that, at least within the context of the behavioural trials conducted, aggression is being used to assert social dominance in this species. The importance of such correlations applies to any species displaying aggressive interactions, regardless of whether aggression is part of stereotyped escalated context or linked to dominance.

A question not previously considered is whether IGE on aggression (or indeed other social traits) will also have evolutionary implications for non-social aspects of 'animal personality'23. For example, traits such as boldness and exploratory tendency are often correlated with aggression (e.g. mediated by proximate mechanisms such as variation in metabolism12,13), leading to the suggestion of an integrated 'aggression-boldness syndrome' (meta-analysis, ref. 26). When we thus extended our analysis to include a non-social behaviour, we found evidence of a genetic covariance structure that would preclude independent evolution of exploration and aggressiveness. Interestingly, this was manifest as a correlation between IGEs on aggression and DGEs for exploration, rather than the conventional (i.e. direct additive) genetic covariance structure that is normally estimated in studies seeking to understand multivariate selection responses (e.g. using the Lande equation, refs 15 and 33).

Specifically, a high genetic merit for exploration is associated with a tendency to elicit more aggressive behaviour from conspecific partners (Table 2, Fig. 1). The correlation between DGEs in exploration and IGEs in aggression mirrors, at the genetic level, conclusions of a phenotypic study on the closely related cricket species G. campestris24. In this species we found a positive correlation between individual (phenotypic) merits for exploration and aggression elicited in conspecifics (r = 0.45, SE 0.17) (Note the corresponding among-individual phenotypic correlation estimated in the present experiment is also significantly positive and similar in magnitude: r = 0.37, SE 0.09; Table S2). Thus, had we not considered IGEs, we would incorrectly have concluded that exploratory behaviour and aggressiveness were evolutionarily independent10,34. Instead we expect that selection on exploratory behaviour will cause correlated evolution of the social environment with consequences for mean aggression (and vice versa). However, it does not follow that the IGEs constraining evolution of mean aggression will necessarily constrain the evolution of exploration behaviour too. In general, IGEs arising from competition related processes are expected to impose constraints on traits that are consequent, rather than causal to, contest outcomes (and thus resource acquisition19), a scenario that is not clearly the case here. We fully acknowledge that our study is not directly informative for the causal pathways linking aggression to exploration, but several possibilities can be hypothesised. For example, the positive association could arise if exploration in a novel environment increases the likelihood of encountering rivals (and thereby provoking more attacks from conspecifics). Exploration could also be favoured in individuals eliciting aggression as a result of competition for territories in...
the population. Alternatively, exploratory tendency may be (genetically) correlated with other traits that directly mediate agonistic behaviour in competitive interactions (e.g. size, weapon morphology).

We also note that the variance partitioning approach used to model IGEs in this paper is mathematically equivalent to the alternative (but complementary) ‘trait based’ approach advocated by others15, 16, 36. In this latter framework, an interaction effect coefficient \( \psi \) (‘psi’), captures the effect of a measured conspecific trait (or traits) on focal phenotype. \( \psi \) represents a standardized reaction norm slope, hence the level of phenotypic plasticity to a social environmental gradient21. In the context of our study, \( \psi \) is captured by the correlation between DGEs and IGEs: individuals responded to the aggressiveness and explorative tendency expressed by social partners (because IGEs on aggression are correlated to DGEs of both behaviours), implying that \( \psi \) is multivariate in nature. A hot question in quantitative genetics revolves around the issue of whether genotypes differ in their responsiveness to phenotypes of conspecífics, which would imply heritable variation in \( \psi \). An interesting follow-up question is thus whether responsiveness to other individuals (\( \psi \)) varies according to behavioural ‘types’, as has recently been suggested in the personality literature23, 39. Importantly, a genetic architecture that includes genetic variation in \( \psi \) and its covariance with other DGEs and IGEs would likely reveal further interesting repercussions for evolutionary processes of behavioural traits.

In conclusion, a crucial consequence of social interactions is that they generate IGEs that not only contribute to the observed variance but also impact evolutionary dynamics of traits under selection. In this case, constraints on the phenotypic evolution of mean aggression arise from the negative correlation of direct and indirect genetic effects. More generally, we note that the role of IGEs has received little attention in ‘animal personality’ research, on the phenotypic evolution of mean aggression arise from the negative correlation of direct and indirect genetic effects. More generally, we note that the role of IGEs has received little attention in ‘animal personality’ research, on the selection of all individuals42, 43. All individuals within a group were tested on the same day (8 individuals simultaneously), randomized for time of the day and test location. Dyads of males paired to selection can be profoundly misleading if effects of social interactions mediated by IGEs are not considered.

Methods

Cricket collection, breeding, and housing. The parental generation of crickets was collected from a tomato field of approximately 2500 m² near Capalbio, Italy (42°42’46.7” N 11°33’99.3’ E) in July 2013. We collected a total of 100 individuals: 34 adult males, 33 adult females, 12 near-final instar males, and 21 near-final instar females. Following capture, crickets were transported to a climate controlled chamber at the Ludwig Maximilians University of Munich (Planegg-Martinsried, Germany), where they were housed at 26 °C (± 0.5) and 65% (± 0.5) humidity, under a 14:10 light dark photoperiod (h) that wild crickets experienced at the time of capture.

Sexually mature wild-caught individuals from the parental generation were randomly paired 4 days after arrival in the laboratory. A total of 35 males and 35 females produced a total of 34 clutches from which offspring hatched. We raised 40 offspring (F1) per parental pair (1360 offspring in total), from which we randomly selected 455 males were selected and screened for behavioural phenotypes (335 from the F2 and 120 from the F3). 15 individuals of the F2 were only tested twice, because they were subsequently used for other purposes. The number of available adult offspring (of both sexes) per female was \( n = 622 \) for the F2 and \( n = 281 \) for the F3 (per female mean ± SD: 8.64 ± 2.46 for the F2 and 5.51 ± 2.44 for the F3). Of these, a total of 455 males were selected and screened for behavioural phenotypes (335 from the F2 and 120 from the F3).

Experimental protocol. Behavioural trials were conducted between January and June 2014. Each individual was repeatedly assayed for each of 2 behaviours on the same day (exploration and aggression, described in detail below) following46; the same individual was assayed for each behaviour 6 times, with measurements taken approximately one week apart (range 7–9 days). Because individual identification is required for the aggression test (detailed below), subjects were marked with coloured tape on the pronotum (red or blue, randomly assigned each time) the day before a focal trial (see also47). The two tests were always done sequentially and in the same order; carry-over effects could therefore not be modelled. We chose this set-up because it ensured that all individuals were given the exact same treatment since this greatly facilitates comparison between individuals46, 47. The 455 males were divided into 7 groups of 40 individuals (F2), one group of 55 individuals (F2), and 3 groups of 40 individuals (F3). 15 individuals of the F2 were only tested twice, because they were subsequently used for other purposes. Individuals were divided into groups according to their estimated age (days post-moult) to avoid any possible age-related effects on aggression (see also48). All individuals within a group were tested on the same day (8 individuals simultaneously), randomized for time of the day and test location. Dyads of males paired for the aggression tests were randomly assigned amongst the non-related individuals within the same group to produce social environments that were homogenous with respect to relatedness.

All trials were performed on a rack fitted with two shelves, each equipped with a camera, in the same climate room where the individuals were housed (detailed in ref. 24. All trials were recorded using high-resolution digital video cameras (Basler GenICam, Germany) fitted 43 cm above each testing arena. The cameras were connected to a computer outside of the climate room and managed using the software MediaRecorder (Noldus, Netherlands). Videos were recorded at 27.81 frames per second and 1600 × 1200 pixels resolution.
A small number of trials were excluded from the final dataset: 31 of 1888 (F2) and 3 of 608 (F3) for exploration trials (respectively 1.64% and 0.49%), and 27 of 944 (F2) and 5 of 304 (F3) for aggression trials (respectively 2.86% and 1.64%) due to technical problems with data recording or video-tracking. Note that the total number of aggression trials is approximately half of that of other trials since two individuals are involved in each aggression test. The final sample size (behavioural tests) was therefore 2462 for exploration (mean number per individual: 5.27, SD 1.23) and 1195 for aggression (mean number per individual: 5.16, SD 1.28) tests.

**Behavioural trials and scoring.** Exploration and aggression behaviour were assayed following the protocol in\(^4\) (for an illustration of the setup, see Fig. 2 in\(^4\)). Briefly, at the onset of the exploration test, each individual was moved (inside its own shelter) from its home container to the exploration arena. Exploration activity was then recorded automatically for 30 minutes. Following the exploration test, the shelters were removed and the individuals given a further 10 minutes to acclimatize. The divider between two arenas was then lifted, after which we filmed each dyad engaging in social interactions for a period of 10 minutes. We then returned the crickets to their home containers in the allotted housing slots within the climate room.

Exploration and aggression videos were analysed using Ethovision version 11.0 (Noldus, the Netherlands). This software package enables tracking of isolated individuals and extracts the spatial coordinates for each video frame. We summed up all distances to calculate the total distance moved in the novel environment (exploration test), viewed as proxy for ‘exploration behaviour’ (following\(^45\)). For the aggression test, we calculated the total time each individual spent moving towards the opponent (‘relative movement’ for simplicity), by summing up only the consecutive samples (frames) where the relative distance between subjects decreased (see User manual of Ethovision v11.0, Noldus Information Technology 2014, for details). We set a maximum interaction distance between the two subjects of 8 cm based on pilot trials to define a range in which the directional movement towards the other cricket would be meaningful. We validated the choice of the variable ‘relative movement’ for aggression both for a related cricket species and for a subset of the current dataset. Relative movement was highly correlated with the variable ‘approach’ towards the opponent that we scored manually, and is commonly used in aggression tests\(^46\). The choice and validation of relative movement as a measure for aggression is detailed in the Supplementary Material.

**Quantitative genetics analysis.** Univariate models. We conducted two sets of statistical analyses. First, using univariate mixed-effects models we partitioned the total phenotypic variance (\(V_P\)) for each measured trait (aggression, exploration) into its underlying components: residual within-individual variance (\(V_R\)) and among-individual variance (\(V_I(f)\)) for the focal individual. The latter component represents the statistical signature of ‘personality’ variation\(^45\) and so was tested in its own right before we further partitioned it in another comparison of variance components across traits. Throughout, we assumed a Gaussian error distribution, which restricted maximum likelihood; dependent variables were mean-centred and variance standardized to facilitate 

\[
\tau^2 = \frac{V_{TBV}}{V_{TOT}}
\]

**Multivariate models.** As a next step, we used a multivariate extension of the framework described above to estimate patterns of between-trait (aggression, exploration) covariance at the among-individual (I) level, further
partitioned into the permanent environmental and genetic levels by respectively estimating the PE and G matrices. This allowed us to estimate the correlation between the opponent identity effect on aggressiveness and the focal identity effect on exploration (I matrix), and to partition it into its genetic and environmental components. We fitted exploration and aggression as response variables and included only fixed effects that explained significant variation in univariate analyses (detailed above).

Significance testing in mixed-effects models. We tested model fixed effects using conditional F-tests with denominator degrees of freedom (df) estimated from the algebraic algorithm in ASReml 4.1. We used a hierarchical stepwise forward approach to evaluate the statistical significance of random effects by likelihood ratio tests (LRTs). We started with a phenotypic model that contained only fixed effects and residual variation (Model 1). We then tested for differences among individuals in the focal (Model 2) and the opponents (Model 3) by sequentially fitting individual and opponent identities respectively. Model 4 tested for the phenotypic correlation between the two. We repeated the same structure when testing for genetic variation and added DGEs (Model 5), IGEs (model 6), and their correlation (model 7). We assumed a χ²-distribution for the test statistic which is calculated as twice the difference in log-likelihood between a model where a target random effect was fitted versus not fitted. Variances are bound to be positive, therefore in testing them we applied the LRT assuming (for testing a single variance component) an equal mixture of χ² and χ².

For multivariate models, we compared the fit of the model where all covariances at a specific level were estimated with one where those covariances were instead all constrained to zero (with degrees of freedom equal to the number of covariance terms). This provides an overall (i.e. matrix level) test for nonzero covariance structure. We further tested the significance of each covariance separately by applying a LRT (assuming χ²) as described above. This led to 5 alternative multivariate models, differing in the correlation structure (See Table 3 for details). We also compared the fit of the alternative models (both for univariate and multivariate analyses separately) using the Akaike information criterion (AIC), calculating ΔAIC relative to the model with the lowest AIC. We calculated the Akaike weight and model likelihood for each model using the package ‘qpcR’ in R 3.1.0.
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Author Contributions
F.S. and N.J.D. conceived the study. F.S. collected the data. F.S. analysed the data with input from N.J.D., A.J.W., and P.T.N. F.S. wrote the first draft of the manuscript. All authors contributed to revisions and approved the most recent version of the manuscript.

Additional Information
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