Abstract  The aim of this multi-informant twin study was to determine the relative role of genetic and environmental factors in explaining variation in trait resilience in adolescents. Participants were consenting families (N = 2,638 twins in 1,394 families), from seven national cohorts (age 12–18 years, both sexes) of monozygotic and dizygotic twins reared together. Questionnaire data on the adolescents’ Ego-resilience (ER89) was collected from mothers, fathers and twins, and analysed by means of multivariate genetic modelling. Variance in trait resilience was best represented in an ADE common pathways model with sex limitation. Variance in the latent psychometric resilience factor was largely explained by additive genetic factors (77% in boys, 70% in girls), with the remaining variance (23 and 30%) attributable to non-shared environmental factors. Additive genetic sources explained more than 50% of the informant specific variation in mothers and fathers scores. In twins, additive and non-additive genetic factors together explained 40% and non-shared environmental factor the remaining 60% of variation. In the mothers’ scores, the additive genetic effect was larger for boys than for girls. The non-additive genetic factor found in the twins’ self ratings was larger in boys than in girls. The remaining sex differences in the specific factors were small. Trait resilience is largely genetically determined. Estimates based on several informants rather than single informants approaches are recommended.

Keywords  Resilience · Adolescents · Twin study · Heritability · Multiple informants

Introduction

Many studies on human resilience have focused on how contextual factors may act as buffers against stressful life events and adversity (Cicchetti et al. 1993; Cowen et al. 1994; Hurd and Zimmerman 2010; Luthar 1991). However, it has also long been assumed that there must be a genetic component in human resilience against stress and adversity (Rutter 2003). Moderate to strong genetic effects have been reported in twin studies for several normal personality traits and competencies (Ganiban et al. 2008; Kandler et al. 2009; Koenig et al. 2008; Nes et al. 2006; Raevuori et al. 2007; Vernon et al. 2009) as well as in exceptional talents (Haworth et al. 2010; Vinkhuyzen et al. 2009) However, there are very few genetically informative studies on positive psychological personality traits related specifically to the reactivity to events.

Stein et al. (2009) recently reported a higher frequency of the homozygous ‘LL’ allele of the 5HTTLPR in a group of undergraduate students with high scores on trait resilience as measured by the Connor-Davidson Resilience Scale (CD-RISC) (Connor and Davidson 2003), but the authors duly emphasize the necessity of replication of this result. A Swedish twin study (Hansson et al. 2008) found moderate heritability estimates (h² ranging between .24 and .49), and zero order shared environmental effects, on measures of well-being, negative and positive mental health. The authors related this to individual salutogenic
factors and indirectly to trait resilience. Boardman et al. (2008) found $h^2 = 0.52$ for men and $h^2 = 0.38$ for women in a study of US adult twins, where resilience was indirectly defined as the residual variance in positive mood after adjusting for life stressors. Apart from these studies, we are not aware of other studies that have approached the study of resilience from a behaviour genetic perspective. Thus, the present state of knowledge in this field does not seem to differ much from that described by Shiner and Caspi (2003) almost 10 years ago: ‘Resilience researchers have called for increasing focus on the processes underlying resilience (ref.), and personality research should be an important part of future work in this area. It is curious, also, that genetic studies, which are so integral to research on vulnerability associations, have played such a minor role in research on resilience. From an evolutionary perspective genes are equally likely to protect against environmental insult as they are to create vulnerability to disease…’ (pp. 20).

The aim of the present study was to study the aetiology of stress resilience by investigating the relative contribution of genetic and environmental causes to the variability of the personality trait of resilience in adolescents. Adolescence was chosen because it is characterized by developmental change in physical as well as social domains. It is also a phase when several mental disorders, among these anxiety, depression, substance abuse and antisocial behaviours, have their first onset (Moffitt et al. 2007; Nolen-Hoeksema et al. 2007; Wittchen 2008), for a review, see (Kessler et al. 2007). Studying resilience as an individual personality buffer against developing mental difficulties through the normal developmental challenges as well as stressful life events of adolescence can lead to a better understanding resilience in humans in general. Knowledge of the genetic and environmental causes behind individual differences in trait resilience may also have important implications for intervention and prevention.

In the study of personal resilience, two different conceptualizations have traditionally been employed. One is the empirically based personality type approach, where a certain personality profile defines the resilient typology. Studies are often based on measures of the Five Factor Model, where a combination of high scores on Agreeableness, Conscientiousness, Extraversion and Openness and low score on neuroticism, has been found to characterize the resilient type (Asendorpf et al. 2001; Asendorpf and van Aken 1999; Barbanelli 2002; Boehm et al. 2002; Costa et al. 2002; DeFruyt et al. 2002; Rammstedt et al. 2004; Robins et al. 1996; Schnabel et al. 2002), see (Torgersen and Vollrath 2006) for a review. However, resilience has also been approached as a theoretically derived personality construct that can be measured dimensionally. Block & Block (Block 1950; Block 1951) were among the first to explore resilience from this angle, introducing the concept of Ego-resilience as derived from psychoanalytic theory and explored empirically through inverse factor analysis of Q-sort (Block 1978) personality descriptions. A short form of their original scale, the Ego Resilience Scale 89 (ER89) (Block and Kremen 1996) has later been used on various samples, settings and age groups (Bonanno et al. 2011; Cohn et al. 2009; Ong et al. 2006, 2010; Sahdra et al. 2011; Schaubroeck et al. 2011; Sherman et al. 2010). Several other dimensional resilience scales are also available (Connor and Davidson 2003; Friborg et al. 2003; Jew et al. 1999; Oshio and Koneko 2003; Wagnild and Young 1993). Comparisons of the two traditions (Huey and Weisz 1997; Waakaar and Torgersen 2010) concerning their ability to predict adaptive behaviors and mental difficulties in children and adolescents have concluded in favor of the FFM typology approach. However, as shown by Waakaar and Torgersen (2010), this advantage was limited to the prediction of mental difficulties. The ER89 resilience scale performed equally well compared to the FFM in explaining variance in adolescents’ adaptive functioning. This, and the fact that the resilience scales are often shorter and thus more easily applied in epidemiological studies and community samples than the comprehensive FFM based measurement instruments, constitute sound reasons for employing the dimensional approach when the scope of the study is narrowed to the resilience phenotype and not personality more generally.

A multiple informant survey of trait resilience in seven Norwegian cohorts of monozygotic and dizygotic adolescent twins of both sexes reared together constituted the data that was entered into the multivariate genetic modelling analyses. The primary hypothesis was that trait resilience would be moderately to strongly heritable. In the case where variation in scores across different informants can be attributed to the same common etiological sources, the multi informant design offers an error free estimate of the relative impact of these on the trait in focus. Earlier studies have shown that not all variation between informants will be due to measurement error or rater bias, and that experience based on unique interactions may allow mothers, fathers and adolescents separately to provide substantive additional information about an adolescent’s behavior (Bartels et al. 2007; Hoekstra et al. 2008). The inclusion of scores from mothers and fathers in addition to the twins’ self ratings in the present study allows for the expression of such valid observer based differences. In cases where the significant rater specific additive genetic component can be ruled out, exploration of alternative systematic rater bias models may be relevant (Hartman et al. 2007). There was not sufficient previous empirical basis for formulating specific hypotheses about differences in heritability between different informants nor between the sexes.
Methods

Participants and procedure

Sample

The population consisted of all twin pairs born in Norway between 1988 and 1994. A total of 5,374 twin pairs (10,748 individuals) were registered in the Norwegian Medical Birth Registry in the relevant birth cohorts. Of these, 574 pairs (11% of the twin relevant population) had to be excluded from the study due to being deceased or having missing addresses. All remaining twins born in 1988–1994 (aged 12–18 years at the time of entering the study) were invited to participate. Written informed consent was provided from 2,486 families, which constituted 53% of the invited families. A total of 1,394 families (56% of the consenting families, and 26% of the total twin birth cohorts) returned the survey forms. A flowchart of the study attrition is presented in Fig. 1.

Significant differences in educational level between participating mothers and same age women in the general population (sample mothers mean level = 2.6; population women mean = 2.2, χ² = 3,131.9, df = 12, P < 0.000) as well as between participating fathers and same age men in the general population (sample fathers, mean level = 2.6; population men, mean level = 2.1, χ² = 505.3, df = 12, P < 0.000) indicated that some selective attrition based on socioeconomic status may have occurred. Educational levels was measured on a scale from 1 to 4, with 1 signifying basic education only and 4 higher university grade. Data on the general population were based on available tables from Statistics Norway (2007).

Procedure

The Norwegian Medical Birth Registry provided the names and addresses of all twins and their parents born in Norway in the seven relevant birth cohorts. The eligible families received a letter by post with information about the study and a letter of consent to be returned from parents as well as from the twins. All consenting families then received posted paper-and-pencil inventories to be filled in by the twins and both parents. Approval was received from The Norwegian Data Inspectorate and the Regional Committee for Ethics in Medicine.

Measures

Trait resilience

Trait resilience was measured by five items from the original 14 items Ego-Resilience scale (ER89) by Block and Kremen (1996). The items were selected on the basis of having demonstrated superior item-to-scale associations through a series of pilot studies on community high school and college students. The following items were included: (1) I enjoy dealing with new and unusual situations; (2) I usually succeed in making a favorable impression on people; (3) I like to do new and different things; (4) My daily life is full of things that keep me interested; (5) I would be willing to describe myself as a pretty strong personality. Twins’ self-rating forms and parental forms were equal except from the substitution of ‘I’ and ‘me’ in the twins’ forms with ‘The twin’ and ‘his/her’ in the parental forms. Items were scored on a five-point scale from ‘Not typical’ to ‘Very typical’. Final scale inter-item reliability Cronbach’s alpha across informants ranged from 0.76 to 0.80.

Zygosity

The zygosity was determined by means of a questionnaire for the total sample of twins. A part of the questionnaire has earlier been used in one Norwegian twin study (Torgersen 1979), another part in another twin study (Harris...
et al. 1995; Tambs et al. 2009). The questionnaire consisted of items about similarity in appearance, how often the twins had been mixed-up with each other, and whether they believed that they were monozygotic or dizygotic. DNA was drawn from a subsample consisting of 204 twin pairs. Eighteen genetic markers were tested, with a probability of misclassification of less than 0.0001. The non-tested twins were allocated to zygosity group by means of discriminant analysis of the questionnaire data from each twin, mother and father. The misclassification was less than 2%. As twin pairs with equivocal questionnaires were oversampled, the misclassification in the whole sample is expected to be much less.

**Data analyses**

**General issues**

**Data preparation**

Missing variables on item level were imputed with Expectation Maximization (Shafer 1997) (Missing Variables Analysis in SPSS Statistical Package, version 15.01). This imputation method uses other non-missing items from the same scale to impute a value (Expectation) which is then checked as to whether that is the most likely value (Maximization). If not, a more likely value is re-imputed, and this two-step iterative procedure is repeated until the most likely value has been found (SPSS 2007).

To avoid the possibility of spurious inflation of the twin correlations and between informant correlations, imputation was carried out on a within person basis. Thus, mothers’ scores on Twin B were not included in estimation of missing items of the mothers’ Twin A scores, nor were any other informant’s scores. Scales with more than two missing items were set as missing on the whole scale in further analyses. The total proportion of items imputed for any scale was <2%. Due to significantly negatively skewed distributions (the ratio of skewness/standard error of skewness > −2) on several of the ER89 scale scores, square root transformations were performed for these variables before entered into the genetic analyses.

Separate linear regression analyses performed on the ER scores for each informant and twin group as dependent showed no significant effects of age. However, there were significant sex effects on the twins’ scores. In order to avoid spurious inflation of the common environmental influences, all genetic analyses were performed with the effect of sex on the means structure controlled through the use of residualized scores. Preliminary univariate analyses (within informant) using a correlated factors model (S*R*S’) (Gaussian decomposition) showed no deviation from the basic assumption of equality of means and variances across twins and across zygosity groups in any of the informants’ scores (results of data preparation analyses are available upon request).

Means and standard deviations were calculated using the SPSS Statistical Package, version 15.01. All further data analyses were performed using the open source statistical software package R, version 2.12.0 (R Development Core Team 2008). Specifically, genetic modelling analyses were performed using the OpenMx version 1.0.3 (Boker et al. 2011). OpenMx is an open source software for fitting Structural Equation Models (SEM) to observed data, integrated within the statistical open source statistical program R (R Development Core Team 2008).

**Genetic modelling**

The twins reared together design is based on the model assumption that monozygotic (MZ) twins share all (or practically all) their genetic characteristics, while dizygotic (DZ) pairs on the average have 50% of their segregating genes in common. Thus, any differences in phenotype between MZ twins have to be caused by unique environmental factors and/or measurement error. Phenotypic MZ correlations above the size of DZ correlations are indications of genetic effects. DZ correlations more than half the size of the MZ correlations must be due to shared environmental factors, while DZ correlations smaller than 50% of the MZ correlations are indicative of non-additive genetic effects, stemming from either dominance effects (alleles interacting within a particular locus) or epistasis (alleles interacting across different loci). Furthermore, the non-additive genetic and common environmental parameters are fully negatively confounded in this design. This means that any model within this design can contain only one of these two parameters. Non-additive effect models are fit when the MZ correlations exceed the DZ correlations by more than 50%. There is generally little empirical evidence for violations of the two remaining assumptions that must be met for the twin model to be valid: (1) that there is no assortative mating for the phenotype measured; and (2) that MZ and DZ twins are equally exposed to the relevant environmental stimuli for the trait being studied (Bulik et al. 2000; Neale et al. 1998).

Phenotypic correlations do not provide tests of the fit of a specific etiological model, nor of the comparative fit of competing models. Thus, employing a biometrical modelling approach (Neale and Cardon 1992), covariances based on raw data between the twins were fitted to a structural equation model by means of maximum likelihood estimation (FIML). Models were compared using the likelihood-ratio $\chi^2$ (LRC) statistic and the Akaike information criterion (Akaike 1987), and the sample size adjusted Bayesian
Information Criterion (BICadj) (Markon and Krueger 2004) were considered during model selection. While the AIC takes into account the overall fit and the parsimony of the model, the BICadj also allows for the comparison of nonnested models. Lower (negative) numbers signify better model fit.

Model fitting

The genetic models employed in the present study assume that, besides a common view, each informant assesses a rater-specific aspect of the adolescent’s behavior (Hewitt et al. 1992). Thus, in addition to contributing to a set of common genetic, common shared environmental and a common-nonshared environmental variance estimates, each informant will constitute a unique source of variation that may meaningfully be partitioned into a unique genetic, a unique shared environmental and a unique non-unshared part (Baker et al. 2007; Bartels et al. 2007). Based on this, three models were tested in the present study: (1) the Cholesky decomposition (this is a descriptive model where all parameters are allowed to vary freely); (2) the independent pathways model and (3) the common pathways model.

In the trivariate Cholesky model, a first set of additive genetic, common environmental (or dominant genetic) and non-shared environmental factors are allowed to influence the scores of all three raters. Then, a second set is allowed to influence the scores of the second and third rater, but not the rater’s scores entered first in the model. A last set is allowed only to affect the scores of the informant entered last. Since in the case of modeling multiple informants the order of entry of the variables into the model is insubstantial, a correlated factors solution was employed to secure that the order of the variables would not affect the estimates (Neale et al. 2006; Rijsdijk 2011). The IP model assumes that there is only one common set of additive, shared environmental (or dominant genetic) and nonshared environmental factors influencing variance in all three informants’ scores. In addition to the common influences, this model includes informant-specific genetic, shared environmental, and nonshared environmental influences. This tests whether the informants’ scores can be expressed by a common set of genetic and environmental causes affecting trait resilience to a different degree, and whether there are etiological aspects of resilience that are uniquely represented by each specific informant.

The common pathways model (CP) is more stringent than the IP model. Within the CP model, genetic and environmental influences upon the different informants scores on trait resilience may be estimated via their effects on a shared underlying latent resilience factor. A particularly interesting aspect of the common pathways model is that the common latent psychometric factor will be unaffected by informant-specific bias and measurement error. The nonshared environmental pathway on the common resilience factor only include environmental influences that in their effect make the twins different from each other. The nonshared environmental pathways on the informant-specific resilience factors will contain measurement error as well as such environmental effects (Neale and Cardon 1992). Thus, given that the common pathways model fits the data, the common factors effects on the latent psychometric resilience factor provide a highly reliable estimate of the effect of the different etiological effects on trait resilience in adolescents.

Results

Descriptives

Table 1 shows sample means and standard deviations by gender, zygosity groups and informants.

Correlation structure

Within person correlations across informant’s scores, as well as cross person within informant and cross informant correlations for each zygosity groups (MZ, DZ) and each sex are shown in Table 2. The correlations were generated through ML estimation, and are thus built on the same logic that constitute the basis for the later the twin models.

Table 1  Descriptive data of informants (mothers, fathers and twins), individual ratings by zygosity group

| Zygosity group | Informant | N    | Mean | SD   |
|---------------|-----------|------|------|------|
| MZ Males      | Mother    | 423  | 13.75| 3.05 |
|               | Father    | 314  | 13.57| 2.75 |
|               | Twins     | 413  | 12.99| 3.00 |
| MZ Females    | Mother    | 576  | 13.53| 3.16 |
|               | Father    | 415  | 13.44| 3.14 |
|               | Twins     | 595  | 12.62| 3.47 |
| DZ Males      | Mother    | 381  | 14.08| 2.96 |
|               | Father    | 288  | 13.58| 3.16 |
|               | Twins     | 373  | 13.13| 3.40 |
| DZ Females    | Mother    | 456  | 13.60| 3.20 |
|               | Father    | 328  | 13.55| 2.99 |
|               | Twins     | 477  | 12.51| 3.27 |
| DOS           | Mother    | 775  | 13.85| 3.12 |
|               | Father    | 581  | 13.38| 2.96 |
|               | Twins     | 752  | 13.07| 3.29 |

Note: MZ monozygotic, DZ dizygotic
Inspection of the correlation matrix may give a cursory impression of the associations present in the data. Table 2 shows that in all groups MZ scores were higher than DZ scores. Thus, genetic variance was indicated for all informant scores. Most of the MZ twin correlations were more than double the size of the DZ correlations, indicating that the ADE model would be a more adequate model than the ACE model in the subsequent genetic modelling analyses. There seem to be some differences in the size of the MZ/DZ correlations between same-sex boys and same-sex girls. However, apart from the markedly lower correlations for the twins scores in the opposite sex group, the correlation between same sex and opposite sex DZ twins were generally of equal scale. Thus, any sex-specific effects were not likely to be of large scale. As expected, the cross informant-within person correlations (left side of Table 2) were generally in the range of medium to low, indicating informant specific as well as common effects.

| Table 2 Estimated twin correlations across sex for the different informant scores and the two zygosity groups |
|--------------------------------------------------|--------------------------------------------------|--------------------------------------------------|--------------------------------------------------|--------------------------------------------------|--------------------------------------------------|
| Within person correlations | Cross person correlations: within informant (diagonal) and cross informant (off diagonals) |
| Mo | Fa | Tw | Mo | Fa | Tw | Mo | Fa | Tw | Mo | Fa | Tw |
|--------------------------------|--------------------------------------------------|--------------------------------------------------|--------------------------------------------------|--------------------------------------------------|--------------------------------------------------|
| Mother Males 1 | Mother MZ 0.75 | 0.55 | – | – |
| Female 1 | DZ 0.29 | 0.15 | 0.37 | – | – |
| Father Males 0.47 | Father MZ 0.40 | 0.73 | 0.35 | 0.66 | – | – |
| Female 0.47 | DZ 0.07 | 0.17 | 0.14 | 0.36 | 0.11 | 0.38 |
| Twins Males 0.32 | Twins MZ 0.21 | 0.23 | 0.48 | 0.26 | 0.19 | 0.40 | – | – | – |
| Female 0.38 | DZ 0.10 | 0.15 | 0.12 | 0.09 | 0.12 | 0.21 | 0.13 | 0.04 | 0.10 |

Multivariate model testing

Based on the results from the correlation matrix and preliminary univariate analyses (available from the first author on request), the multivariate full heterogeneity ADE model was chosen as the starting point for a series of subsequent genetic model testing. The results of model are shown in Table 3.

As we can see from Table 3 models I–II, the ADE model gave an acceptable fit compared to the fully saturated model that allowed all variables covariate with all other variables without any constraint to any parameter. Testing multivariate quantitative and qualitative heterogeneity based on a correlated factors approach (Neale et al. 2006) showed no qualitative sex differences in the variance estimates (ref. models II–IV in Table 8). However, there were significant quantitative sex differences in the variance estimates, implying that the same genetic and environmental factors were influencing the informants’ scores, but to a different amount in their evaluation of girls and boys (comparison of models IV–V in Table 3).

Table 3 also shows a comparison between the multivariate Independent pathways model and the Common pathways model, both allowing all path estimates to vary freely between the sexes (models VI–VII). Using the multivariate Correlated factors heterogeneity model as the baseline model, the results show the best fit for the Common pathways model. The CP model was chosen as the preferred model for later analyses.

Further model testing was undertaken to simplify the structure of the CP factor model (models VIII–XVI in Table 3). First, we tested a full homogeneity model to see whether all parameters could be equated across sex (model VIII). This model fit the data significantly more poorly compared to the model with parameters estimated freely for males and females. Then, each common factors and specific factors were dropped individually (but simultaneously for both sexes) from the full ADE common
pathways heterogeneity model. This strategy resulted in that Common D and specific D could separately be removed without significant drop in fit compared to the full CP model (models X and XIII). However, dropping both (model XIV) was rejected. A model dropping the common D as well as the two non-significant Ds paths (i.e. the specific d in the mothers’ and fathers’ scores) did not cause any change fit compared model X. Thus, being the most parsimonious model (as judged by the lowest AIC value), Model XV ended up as the preferred model.

Figure 2 a (boys) and b (girls) show the standardized path estimates for all factors in the final model, separate for each sex. Even though a model setting the common factor path estimates equal between sexes was rejected, a model dropping the common D as well as the two non-significant Ds paths (i.e. the specific d in the mothers’ and fathers’ scores) did not cause any change fit compared model X. Thus, being the most parsimonious model (as judged by the lowest AIC value), Model XV ended up as the preferred model.

Table 3 Selected results of multivariate behavior genetic model fitting of the mothers’, fathers’ and twins’ ratings on the ER89 scale

| No. | Name | \(-2LL\) | df | AIC | BICadj | \(\Delta LL\) | \(\Delta df\) | \(P\) | Comparison model |
|-----|------|---------|----|-----|--------|-------------|-----------|-----|----------------|
| I. | Saturated (5 zygosity groups) | 40283.05 | 7012 | 26259.05 | 40287.11 |
| II. | ADE full heterogeneity | 40407.59 | 7102 | 26203.59 | 40407.12 | 124.55 | 90 | 0.01 | I |
| III | ADE nonscalar sex limitation | 40403.54 | 7105 | 26193.54 | 40407.6 | 0.49 | 3 | 0.92 | II |
| IV. | ADE, quantitative, but no qualitative | 40417.43 | 7114 | 26189.43 | 40421.49 | 13.89 | 9 | 0.13 | III |
| V. | ADE full homogeneity | 40460.76 | 7123 | 26214.76 | 40464.82 | 43.33 | 9 | 0 | IV |
| VI. | ADE heterogeneity IndepPathw | 40412.04 | 7105 | 26202.04 | 40416.1 | 8.5 | 0 | 0 | II |
| VII. | ADEade heterogeneity ComPathw | 40417.59 | 7113 | 26191.59 | 40421.65 | 14.05 | 8 | 0.08 | II |
| VIII | ADEade homogeneity ComPathw | 40463.44 | 7128 | 26207.44 | 40467.5 | 45.85 | 15 | 0 | VII |
| IX. | ComPathw DEade (drop Ac) | 40470.84 | 7115 | 26240.84 | 40474.91 | 53.26 | 2 | 0 | VII |
| X. | ComPathw AEade (drop Dc) | 40420.35 | 7115 | 26190.35 | 40424.41 | 2.76 | 2 | 0.25 | VII |
| XI. | ComPathw ADEade (drop Ec) | 40490.88 | 7115 | 26260.88 | 40494.94 | 73.29 | 2 | 0 | VII |
| XII. | ComPathw ADEade (drop As) | 40446.1 | 7119 | 26208.1 | 40450.17 | 28.52 | 6 | 0 | VII |
| XIII. | ComPathw ADEae (drop Ds) | 40432.14 | 7119 | 26194.14 | 40436.2 | 14.55 | 6 | 0.02 | VII |
| XIV. | ComPathw ADEae (drop Dc and Ds) | 40512.99 | 7121 | 26270.99 | 40517.05 | 95.4 | 8 | 0 | VII |
| XV. | ComPathw AEadnse (drop Dc, And n.s. Ds.) | 40420.35 | 7119 | 26182.35 | 40424.41 | 2.76 | 6 | 0.84 | VII |

Note: Best fitting model marked in bold

Table 4 shows the standardized variance components in the final model separate for each sex, with the proportion of the total variance that was explained by genetic and environmental factors in each informant’s scores.

Looking at the total variance estimates for each informant’s scores, we see a somewhat different pattern in the twins’ scores compared to that of the mothers’ and fathers’. While the parental scores were mainly explained by additive genetic factors (between three-thirds and three quarters of the mothers’ and fathers scores) and some remaining non-shared environmental factors, the genetic source of variance in the twins scores were distributed between additive and dominant genetic factors. The strongest dominant genetic effect was found in the twin boys’ scores, while in the twin girls’ scores, the proportion of dominant genetic effect was small. The twins’ self ratings yielded higher non-shared environmental estimates than the estimates built upon mothers and fathers scores. This was most pronounced in girls, where more than half of the variance in the female twins’ scores came from non-shared environmental sources.

Furthermore, a very high proportion of the additive genetic variance in the mothers’ and twin boys’ scores was due to additive variance in the common factor (between 57 and 99%), indicative of a rather high degree of agreement between the mother and the twins in the additive genetic aspects of resilience. The fathers’ additive genetic variance in both sexes was about half due to variance in the common factor, half due to additive factors specific for the fathers’ scores (Table 4). Thus, the additive genetic variance in the fathers’ scores were to a larger extent specific to the fathers’ perspective. Dominant genetic factors were only represented through factors specific for the twins’ scores. The main proportion of the total non-shared environmental factors were due to factors specific for each informant (52–91%). In interpreting this result one must keep in mind that any measurement error will be allocated to, and cannot be separated from, the effect of specific non-shared environment factor in this model.
The aim of this multi-informant twin study was to provide reliable estimates of the relative impact of genetic and environmental causes of variation in trait resilience in adolescents.

The results showed that the covariation between the different informants’ scores for both sexes could best be modelled by means of a latent psychometric factor, where additive genetic and unique environmental influenced the different informants’ scores through the same mechanisms. There were significant sex differences in the size of the estimates in the model. Additive genetic factors explained 78% of the variation in the latent psychometric resilience factor in the boys’ scores, and 70% in the girls’ scores. The remaining variation in the latent resilience factor was explained by unique environmental factors (23% in boys, 30% in girls), while the impact of non-additive genetic factors was negligible.

The heritability estimates of trait resilience found in this study exceeds that from an earlier study by Boardman et al. (2008). One possible explanation is that it may be due to the multivariate design, which is able to produce more reliable estimates than those based on any one single rater. Table 4 shows how the heritability estimates in the present study varied between raters (from 75% in mothers’ ratings of boys...
to 15% in boy twins’ self ratings). Although genetic factors were also the responsible for the majority of the variance in the fathers’ scores, the highest heritability was clearly found in the scores of the mothers. The broad-sense heritability (including the dominant genetic factor) in the twins’ scores was around 50% for boys and 40% for girls.

All informants’ scores loaded substantially on the latent common genetic factor for both sexes, thus there was support for an additive genetic factor that was represented in all informants’ ratings of the adolescents’ trait resilience. In addition, there was evidence of modest to small genetic effects contributing to differences between each informant’s ratings. These effects were most pronounced in the fathers scores, where more than half of the genetic variance was explained by informant specific genetic effects. Rater specific genetic effects contributed to between 1 and 57% of the total genetic variance in the mothers’ and twins’ scores.

The main part of the genetic variation in the latent resilience factor came from additive genetic sources, that is, contributions of genes that are independent of each other. Small but significant non-additive genetic effects were specific for the twins’ scores only. Non-additive genetic effects have been reported in other personality traits in adolescents (Rebollo et al. 2006; Rettew et al. 2008). It is, however, possible that the significant specific D found in the twins’ self ratings may contain some aspect of negative contrast or competition effects, indicating that the scores of one twin influence the scores of the other twin in the opposite direction (Simonoff et al. 1998). Sibling competition effects act to deflate both MZ and DZ variances, but MZ variances to a higher extent. This also affects covariances, resulting in MZ phenotypic correlations much larger than MZ correlations, thus imitating a dominance effect. Such systematic within informant cross twin rater bias effects are generally hard to explore in the classic twin design due to statistical power issues (Rietveld et al. 2003).

The results of the present study showed that just about one quarter of the total variation in trait resilience was attributable to environmental factors. With no indication of any shared environmental effects in the present data (based on inspection of the correlation structure \((r_{DZ} < \frac{1}{2} r_{MZ})\)), all environmental variation in the latent trait resilience must have originated from sources that were not shared between the twins within the same family (the error-free common \(E\) in the model). This is in accordance with what is generally found for most complex human traits (Plomin et al. 2001). Earlier longitudinal studies have pointed to factors within the community, such as friends, school, leisure time and neighbourhood as predictors of better developmental outcomes in the face of stress and adversity (Werner 2000). Several resilience enhancing factors within the family environment have also been suggested, such as intra-familial social support and low marital conflict (Pinkerton and Dolan 2007; Rutter 1999). However, among such potential influences within the family, only those that in effect would make twins within the same family different from one another on trait resilience could be of significance.

There were significant sex differences in the size of the standardized estimates of the common factor model of trait resilience in this study. The estimates of the boys additive genetic effects on the maternal scores were larger, while the nonshared environmental estimates were smaller than the same estimates for the girls. The dominant genetic effect found in the twins’ scores was larger for boys than for girls. The differences between the sexes for the remaining informant specific estimates were very small.

Limitations

The following limitations should be kept in mind when evaluating the results of this study.

### Table 4 Variance estimates in final model for each sex

|               | Broad-sense total heritability | Proportion (%) due to Ac | Proportion (%) due to As | Proportion (%) due to Ds | Total environment | Proportion (%) due to Ec | Proportion (%) due to Es |
|---------------|-------------------------------|--------------------------|--------------------------|--------------------------|-------------------|--------------------------|--------------------------|
| Mothers       |                               |                          |                          |                          |                   |                          |                          |
| Boys          | 0.75                          | 56                       | 44                       | 0.24                     | 50                | 50                       |                          |
| Girls         | 0.55                          | 78                       | 22                       | 0.45                     | 40                | 60                       |                          |
| Fathers       |                               |                          |                          |                          |                   |                          |                          |
| Boys          | 0.69                          | 46                       | 54                       | 0.31                     | 29                | 71                       |                          |
| Girls         | 0.66                          | 39                       | 61                       | 0.34                     | 32                | 68                       |                          |
| Twins         |                               |                          |                          |                          |                   |                          |                          |
| Boys          | 0.50                          | 30                       | 00                       | 0.70                     | 0.49              | 8                        | 92                       |
| Girls         | 0.41                          | 41                       | 54                       | 0.05                     | 0.60              | 12                       | 88                       |

*Note: Ac common additive genetic factor, As informant specific additive genetic factor; Ds informant specific dominant genetic factor, Ec common nonshared environmental factor, Es informant specific nonshared environmental factor*
The twins reared together design generally yields common environmental estimates that are lower than those reported in adoption studies (Buchanan et al. 2009).

A comparison between level of education in the participating families compared to age equivalent levels for women and men within the total population give reason to believe that there may be selection bias in the participating families based on socioeconomic indicators. Earlier genetically informative studies have yielded differential heritability estimates across socioeconomic groups (South and Krueger 2011; Tuvblad et al. 2006) as well as in groups with different exposure to various stressful life events (Distel et al. 2011). Unfortunately, limited statistical power impeded the testing of additional moderators beyond this sex limitation model.

Measurement invariance may constitute a challenge in studies where different groups are being compared on measures of complex traits. In fact, the significant differences in latent factor loadings between informants and across sex in the present study may be seen as forms of measurement non-invariance in how trait resilience is being perceived in different groups. As noted by Neale et al. (2005), measurement non-invariance on item level betweenzygosity groups may impact the estimates in studies based on sum scores. This type of bias would be most pronounced in the case of binary items, and the direction of the bias on the variance estimates will depend on which of the MZ or DZ scores are the best indicators of the latent trait in question. Although preliminary exploratory factor analyses (PCA) in the present study showed preference for a one factor solution for all zygosity groups and factor loadings of each item typically ranging between 0.65 and 0.80, only a full multivariate simultaneous analysis of measurement model and variance decomposition analytic approach would provide an effective handling of the measurement invariance issue. This kind of analyses have recently contributed informatively in the search for the causal structure between diagnostic criteria within and between DSM diagnostic categories (Torgersen et al. 2008). Further, a comparison between the sum score approach and a combined variance decomposition/item response theory (IRT) measurement model approach (van den Berg et al. 2007) on attention problems resulted in markedly higher heritability estimates with the latter. However, the sum scores approach employed in the present study yielded very high heritability estimates, and analyses at an even more increased level of precision would be premature in this early phase of the study of the resilience phenotype.

Conclusion

Resilience in adolescents as measured by three informants (mothers, fathers and the twins’ self ratings) was best conceptualized as a common latent factor, with additional genetic, shared environmental factors as well as non-shared environmental factors that were specific for each informants’ scores. There were significant sex differences in the sizes of the standardized estimates within the model. In the mothers’ scores, the genetic factor was larger and the non-shared environmental smaller in the boys, whereas the opposite pattern was representative for the girls. Resilience as a latent personality trait was highly genetically determined for both sexes, with additive genetics effects explaining nearly three quarters of the variance. There was no indication of any common environmental factor in trait resilience, and non-shared environmental factors had a moderate to low effect. There were significant informant specific contributions to the additive genetic effect in all three informants’ scores, indicating that a multiple rater approach is necessary in order to reach a more complete understanding of the etiological basis of the phenomenon.

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