Psychiatric resilience: longitudinal twin study†
Ananda B. Amstadter, John M. Myers and Kenneth S. Kendler

Background
The source of variability in people’s response to stressful life events is poorly understood.

Aims
We examine the genetic and environmental underpinning of resilience (i.e. the difference between the twins’ internalising symptoms and their predicted symptoms based on cumulative stressful life events).

Method
Stressful life event exposure and internalising symptoms were assessed at two time points in 7500 adult twins. Using the residual between actual and predicted internalising symptom total score, twin modelling was conducted for each wave separately and longitudinally.

Results
Resilience was found to have a moderate genetic heritability (~31%). Qualitative gender effects were found. Incorporating error of measurement into the model increased the estimated heritability for the latent construct of resilience (~50%). When measurement error and occasion-specific effects were removed, environmental influences contributed roughly equally to level of resilience.

Conclusions
Both genes and environment influence level of psychiatric resilience, and are largely stable over time. Environmental influences can have an enduring effect on resilience.

Declaration of interest
None.

Participants
Participants were derived from two interrelated Virginia Adult Twin Studies of Psychiatric and Substance Use Disorders (VATSPSUD) studies of White individuals,10 ascertained from the birth-certificate-based Virginia Twin Registry. Female–female (FF) twin pairs, born 1934–1974, were eligible if both members responded to a mailed questionnaire in 1987–1988. Data in this report utilised interviews completed at the first (FF1) and third interview wave (FF3), conducted in 1987–1989 and 1992–1995 respectively.10 Cooperation at the FF1 and FF3 interviews was 92% and 88% respectively (and 93% for the intervening second interview wave).10 Data on the male–male and male–female pairs (MMMF) came from a sample (birth years 1940–1974) initially ascertained, with a 72% cooperation rate, directly from registry records containing all twin births. The first interview (MMMF1) was completed largely by telephone in 1993–1996. The second wave of interviews (MMMF2), conducted in 1994–1998, had a retention rate of 83%. Zygosity was determined by discriminate function analyses using standard twin questions validated against DNA genotyping in 496 pairs.11 The mean age and years of education of the twins were 35.1 (s.d. = 7.5) years and 14.3 (s.d. = 2.2) at the FF3 interview, and 37.0 (s.d. = 9.1) years and 13.6 (s.d. = 2.6) at the MMMF2 interview.

These analyses utilised data from 7500 twins, including both members of 3084 pairs (503 monogygotic (MZ) female–female, 346 dizygotic (DZ) female–female, 703 MZ male–male, 485 DZ male–male, and 1047 opposite-gender DZ pairs) and 1325 twins without their co-twin. (These numbers do not sum because all possible pairings for triplet and quadruplet sets were included).
Internalising symptoms

Participants completed a shortened version of the Symptom Checklist-90 (SCL-90), which utilised a past-month time frame. The SCL scores was monotonic, with no evidence of a ceiling effect. Indeed, as the number of such events increases the average total internalising symptoms increase, as does the variability around the point estimate. We then utilised both waves of data simultaneously in a measurement model that assumes that there is a true latent liability to resilience. Each of the two estimates of resilience is considered to be a fallible index of the true latent liability. The paths $\lambda_1$ and $\lambda_2$ represent the degree to which the assessments of resilience obtained at the two waves reflect the true liability. The product of these paths reflects the reliability (or long-term stability) of the assessment of resilience. The latent liability, as well as wave-specific liability, for resilience was then modelled in a standard ACE twin design, as outlined above, including testing for gender effects. Two notable differences from the standard twin model and the measurement model are present. First, the model allows for separate estimates of occasion-specific influences (including measurement error) and true or enduring individual-specific environmental effects. Second, an estimate of the degree to which the time-specific measurements of resilience are reflective of the true latent liability for resilience is produced.

Twin modelling

Phenotypic variation in twin models is decomposed into additive genetic factors (A), which contribute twice as much to the correlations between MZ twins as they do for DZ twins; common environmental factors (C), which are the shared factors for example parental attitudes, economic disadvantage) that make twins reared together more similar and contribute equally to the correlation between MZ and DZ twins; and individual-specific environmental (E) sources, which reflect environmental experiences not shared by twins and therefore contribute to differences between the twins and errors of measurement.

We tested for qualitative gender differences (i.e. whether or not the same genetic factors influenced liability to resilience for males and females, quantified by $r_e$), and for quantitative gender differences (i.e. if there is equality in the estimates of the genetic contribution in males and females) by fitting three models that were compared with the saturated model. We constrained $r_e$ to 1.0 to test for qualitative gender effects. Second, we tested for quantitative gender effects by constraining the estimates of A, C and E to be equal in males and females. Finally, we computed a model to test for both qualitative and quantitative gender effects. To evaluate the fit of the twin models, full-information maximum likelihood approach to raw data was implemented in Classic Mx on a Windows-based PC. Akaike’s information criteria (AIC) is used as a guide to evaluating different models and produces an index of goodness of fit (i.e. balance of explanatory power and parsimony). Parsimony is important because in maximum likelihood estimation, log likelihoods will continue to decrease with additional model parameters potentially resulting in ‘over-fitting’. Akaike’s information criteria penalises models with increasing numbers of parameters thereby providing an appropriate balance between model complexity and explanatory power as manifest by the degree of misfit.

Univariate models

We conducted univariate models of the wave 1 and 2 data, fitting the saturated models and models with constraints for qualitative and quantitative gender effects. After choosing the best-fit ACE model from the gender-specific testing from the wave 1 and 2 models, we then fitted subsequent models in which parameters were dropped to determine the most parsimonious model (AE and CE models).

Measurement models

We then utilised both waves of data simultaneously in a measurement model that assumes that there is a true latent liability to resilience. Each of the two estimates of resilience is considered to be a fallible index of the true latent liability. The paths $\lambda_1$ and $\lambda_2$ represent the degree to which the assessments of resilience obtained at the two waves reflect the true liability. The product of these paths reflects the reliability (or long-term stability) of the assessment of resilience. The latent liability, as well as wave-specific liability, for resilience was then modelled in a standard ACE twin design, as outlined above, including testing for gender effects. Two notable differences from the standard twin model and the measurement model are present. First, the model allows for separate estimates of occasion-specific influences (including measurement error) and true or enduring individual-specific environmental effects. Second, an estimate of the degree to which the time-specific measurements of resilience are reflective of the true latent liability for resilience is produced.

Results

Our definition of resilience relies on two premises: first, stressful life events are associated with the level of internalising symptoms, and second, there is variability in responses to stressful life events. Indeed, as the number of such events increases the average total SCL score also increases, as does the variability around the point estimate (Fig. 1). The relationship between stressful life events and SCL scores was monotonic, with no evidence of a ceiling effect. Stressful life events were significantly and very similarly associated with SCL scores for wave 1 ($\beta = 0.21, R^2 = 0.03, P<0.001$) and wave 2 ($\beta = 0.20, R^2 = 0.03, P<0.001$). From these regressions we saved the standardised residuals, which represent resilience to the effects of stressful life events on internalising symptoms.

Univariate twin models

The between-twin Pearson correlation of resilience was roughly twice as large for MZ pairs as it was for DZ pairs for males ($rs = +0.30 and +0.16$ respectively) and females ($rs = +0.34 and +0.16$ respectively). Table 1 outlines the models that were fitted for wave 1 resilience. Models were compared with the saturated model (I). We began in model II by testing for qualitative gender effects. Model III tests for quantitative gender effects, and model IV tests for both gender effects. There was evidence for qualitative but not quantitative gender effects. Therefore, model III was chosen to fit the submodels (AE; CE). These results suggested that C could be dropped from the model without sacrificing fit (model VI), but A could not be dropped (model V). Consequently, the AE model with qualitative gender effects was the best-fit model (V;
The genetic correlation ($r_g$) is substantial (+0.76), but falls short of unity.

Models fit for wave 2 are presented in Table 1. In contrast to wave 1, for wave 2 the best-fit model was model X, without qualitative or quantitative gender effects. The AE and CE models were computed next, suggesting that C could be dropped from the model (XII) but A could not (XI).

**Longitudinal twin model**

We conservatively began by fitting the longitudinal ACE measurement model with qualitative but not quantitative gender effects estimated (model I). From this model we then fit the AE and CE submodels. As shown in Table 2, dropping C resulted in a superior model fit (model II), whereas dropping A resulted in a deterioration in fit (model III). From the AE model other constraints were tested to examine whether parsimony and fit could be improved. We constrained the $\lambda$ to equality to test whether the wave 1 and 2 assessments reflected with the same accuracy the underlying latent level of resilience. This constraint resulted in an improved fit (model IV; Fig. 3). We tested whether the time-specific A estimates could be set to 0 for waves 1 and 2, models V and VI, respectively. Model fit deteriorated for both of these constraints, making model IV our final model. In this model, $r_g$ was high, but statistically different from unity (+0.88).

There are five notable aspects of the final model. First, each wave provided a good index of the true level of resilience with the predicted stability of our measure across waves equal to +0.53. Second, for wave 1, the overall heritability was 31%, and of this, roughly 90% of this estimate is stable across time and driven by the latent measure, and 10% was occasion specific. Similarly, for wave 2, the overall heritability was 32%, with approximately 86% coming from the stable latent measure, and the remaining 14% was time specific. Third, when unconfounded by measurement error the heritability increased (~50%). Fourth, qualitative gender effects were found. Fifth, for both waves larger proportions of the variance in E was as a result of the occasion-specific effects (i.e. transient environmental effects and measurement error) than was as a result of environmental factors with an enduring effect (i.e. those accounted for by the latent E). Specifically, 61% of the variance in E was time specific for both waves, with the remaining variance attributed to stable E.

**Discussion**

This study marks the first estimation of the genetic and environmental contributions to psychiatric resilience to stressful life events, as well as the first estimation of the stability of the resilience phenotype and its aetiology assessed over time. The longitudinal nature of this study afforded the ability to incorporate unreliability of measurement of resilience (which is influenced by error of measurement from two sources: the stressful life events assessment and internalising symptom assessment) directly into

<table>
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<td></td>
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–2LL, –2 log likelihood; AIC, Akaike’s information criteria; A, additive genetic factor; C, common environmental factor; E, individual-specific environmental factor.
our twin model, allowing for the separation of error of measurement and occasion-specific effects from the enduring environmental differences. In doing so, a striking conclusion can be made: genetic constitution and enduring environmental influences contribute roughly equally to the latent construct of resilience. Here we discuss each component in turn, and then end with a comment on implications for measurement of this important construct.

Genetic contribution

We found a modest genetic influence for both waves of the study (heritability ~0.31), suggesting that genes have an impact on the way that individuals respond in the face of stressful life events. In a qualitative sense, we can interpret the wave-specific heritability estimates in the ‘mild to moderate’ range. Although we could not develop an aetiological model for resilience without incorporation of genetic liability, genes are only contributing a modest amount for each measurement occasion. These heritability estimates are similar to those of anxiety disorders, which are among the lower end of psychiatric disorders in terms of their genetic contribution. Another useful benchmark for comparison is with relevant personality factors (such as neuroticism, heritability ~0.41). Our results are consistent with the two prior genetic investigations of resilience, which have each adopted a different definition of the stressor as well as the outcomes. Nonetheless, both studies produced mild to moderate heritability estimates. In short, as human traits go, the wave-specific heritability estimates of resilience are on the lower end of the spectrum.

When unconfounded by measurement errors and occasion-specific influences, the heritability estimate for the latent factor increased substantially from the wave estimates; roughly 50% of the individual differences in resilience were attributed to the latent genetic factor. The construct of resilience is complex, and its measurement can introduce sources of error from how the stressor is measured to how the outcome is assessed. Thus, when these two sources of error are corrected for in the measurement model (which removes the transient environmental effects as well as true error), our heritability estimate was no longer in the ‘mild’ range (where the heritability estimates of most anxiety disorders are) but was well within the ‘moderate’ range, comparable with depression (see for example Kendler & Prescott). Our finding that the ‘stable component’ of resilience is approximately 50% heritable adds to our understanding of the genetic architecture of the construct and is of value in increasing understanding regarding individual differences in responses to stressful life events. Given that resilience is influenced by genetic factors, future studies aimed at identification of specific genetic variation that accounts for the heritability are warranted. The majority of the genetic influence was driven by the latent measure of resilience, yet there was a modest amount of genetic variance that was occasion specific. These ‘new genes’ that are related to resilience may be demonstrating a time-dependent gene-expression effect related to development (for example maturity, learning, treatment). However, we do not know what these processes may be.

In contrast to Broadman et al., we found that although the heritability of resilience is equivalent for males and females, the genes affecting resilience are not equivalent between males and females. This could suggest genetic differences between the genders in the contribution of sex hormones, fluctuations across the menstrual cycle in women, etc. This gene-environment correlation could also be accounted for by the ways in which men and women are socialised. Related phenotypes have also demonstrated qualitative gender effects (for example in major depression, alcohol dependence).10

### Table 2

<table>
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<th>Model</th>
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<tr>
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<tr>
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–2LL, –2 log likelihood; AIC, Akaike’s information criteria; A, additive genetic factor; C, common environmental factor; E, individual-specific environmental factor; L, lambda; A1, additive genetic factor for resilience at wave 1; A2, additive genetic factor for resilience at wave 2.

### Fig. 3

Percent of variance (with 95% CI) accounted for from the best-fitting model for genetic and environmental factors for resilience at waves 1 and 2.

- A1, additive genetic factor on the latent measure of resilience; A2, A3, additive to genetic factors for wave 1 and wave 2 respectively; E1, individual-specific environmental factor on the latent measure of resilience; E2, E3, individual-specific environmental factors for wave 1 and wave 2 resilience respectively.
How might we understand our demonstrated genetic contribution to resilience? Resilience research has its historical roots in developmental psychology, where temperamental factors (such as self-mastery, self-efficacy) have been associated with resilience in children exposed to adversity.\(^{18,19}\) Research on adults has also documented the role of temperamental factors in determining response to stressful life events. It is plausible that shared genetic variance between resilience and these personality factors may be in play. For example, neuroticism is moderately determining response to stressful life events. It is plausible that has also documented the role of temperament and personality in influencing by exposure to environmental events,\(^{26}\) particularly underscoring the plasticity of the brain structure and function.\(^{27}\) Alternatively, genetic influences may have an impact on resilience by directly shaping the way in which individuals react psychiatrically to stressful life events (for example affording psychological resources such as healthy coping techniques).

**Environmental contribution**

As important as our genetic findings are those regarding the environmental influences on resilience. This is the first study to quantify the degree to which environmental events affect resilience in an enduring fashion, as cross-sectional studies are confounded by errors of measurement. After partitioning out the measurement error and transient environmental effects, we found that roughly half of the contribution of variation in the latent level of resilience could be accounted for by individual-specific environmental events that had an enduring effect. This suggests that individual experiences are as important as genetic constitution for determining the latent level of resilience. Notably, the environmental estimates were influenced to a greater degree by occasion-specific effects than were the additive genetic effects, which were more stable and influenced to a higher degree by the latent genetic factor.

A large literature demonstrates the enduring negative health ramifications of exposure to types of extreme adverse events that may have been unshared by twins, such as childhood maltreatment (for example sexual abuse, physical abuse). Exposure can have lasting effects on acute stress reactivity,\(^{22,23}\) and psychiatric functioning.\(^{24,25}\) Evidence from neuroscience underscores the plasticity of the brain structure and function as influenced by exposure to environmental events,\(^{26}\) particularly with regard to circuit connectivity.\(^{27}\) Epigenetic modifications are also associated with extreme stressful life events in genes associated with the hypothalamic–pituitary–adrenal axis,\(^{28}\) which may be another mechanism through which the environmental influences exert their effect on resilience.

Alternatively, certain types of events unshared by twins may be ‘immunising’ (i.e. stress inoculation, steeling effects) and exposure may have buffered the individual from the effects of stressful life events. Findings from the preclinical literature have demonstrated that certain doses of stress can protect organisms from having high amounts of a stress reaction to adversity later in life.\(^{29}\) Notably, our quantification of resilience cannot recognise whether immunising or sensitising effects are at play, and this is an important direction for future research. Although the majority of the literature within traumatic stress suggests that trauma has a cumulative risk effect,\(^{30}\) there are studies that report that increased number of events can be related to an increased sense of locus of control.\(^{31}\) Supports could also have affected one twin (for example social support, higher education) that served to bolster resilience. However, whereas the majority of the additive genetic effects was stable and accounted for by the latent genetic factor, for the environmental estimates, larger proportions of the variance was accounted for by the time-specific effects (which also include measurement error) than was as a result of environmental factors with enduring effects. Nonetheless, identification of the sources of variation in the environmental events that have enduring effects on resilience is a critical step in this literature.

**Defining resilience**

Resilience has been defined in many different ways, with regard to the stressors that an individual would demonstrate resilience to when faced with and the phenotype that is characterised as a resilient response.\(^{30,32}\) Our statistical approach is similar to Kim-Cohen et al.’s,\(^{8}\) who examined the difference between children's scores on intelligence and behavioural scales and their predicted scores by level of socioeconomic deprivation; however, we focused on the types of stressors most likely to have an impact on internalising symptoms. Socioeconomic deprivation is a different type of stressor given its chronicity and diffuse impact. Our definition of resilience is appealing, as it does not rely on individuals to accurately report on a 'higher level' of observation. An alternative approach is to query individuals about their perceived ability to cope with adversities,\(^{36}\) which is more prone to reporting bias, errors in self-perception and variation with regard to the items included on the measure. Our objective approach provides a novel metric of resilience that statistically quantifies resilience.

**Limitations**

The present study should be interpreted in light of several weaknesses. First, resilience is a complex and dynamic process that unfolds over time and our assessment method cannot fully capture this process, only the outcome. Specifically, our quantification of resilience is inferred, in one domain (internalising symptoms), and is based off of the individual variation to the stressful life events experiences. Second, as noted above, exposure to stressful life events may be associated with immunising effects,\(^{32,33}\) as well as sensitising effects.\(^{22,23}\) Our quantification of resilience cannot differentiate between these effects. Second, although we assessed for a large number of stressful life events, our list was not exhaustive. To that end, we also did not assess for personal reaction to the event (such as fear, helplessness or horror) or the severity of the events. It is possible that individuals who may be resilient to some events may not show as much resilience to other events, thus a more thorough stressful-life-event assessment is warranted for future studies. Third, given demonstrated gene–environment correlation findings\(^{21}\) for exposure to stressful life events, it is possible that a modest proportion of the association between adversity and the SCL measure is non-causal (i.e. genetically mediated through personality factors that influence exposure to stressful life events and psychopathology symptoms). Fourth, our outcome of interest was the SCL, which is a broad measure of internalising symptoms; we did not examine broader psychiatric disorders (for example externalising disorders). Fifth, although our sample is likely representative of twins in the Mid-Atlantic region of the USA in the 1990s, the generalisability to other areas and time frames is limited. Finally, we conservatively fitted the measurement model with qualitative gender effects, which were found for wave 1 but not for wave 2.

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Psychiatric resilience

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