Developmental Trajectories of Physical Aggression from School Entry to Late Adolescence

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The developmental perspective applied to psychopathology has led to the concept of early- and late-onset disorders. This study explores the application of the early- and late-onset concepts of antisocial behavior to physical aggression. Are there two categories of chronically physically violent adolescents: those who are physically aggressive throughout childhood and those who start being physically aggressive during adolescence? The estimation of developmental trajectories for repeated measures of two different response variables—physical aggression in childhood as measured by teacher reports and physical aggression in adolescence as measured by self-reported violent delinquency—is achieved with a semi-parametric, group-based method. This new method is applied to a large sample of males from Montreal who have been assessed repeatedly since kindergarten. Several salient findings emerge from the analysis. First, we find considerable change in the levels of childhood and adolescent physical aggression. Thus, there is little evidence of stability of behavior in an absolute sense. A second key finding concerns the connection of childhood aggression to adolescent aggression. Boys with higher childhood physical aggression trajectories are far more likely to transition to a higher-level adolescent aggression trajectory than boys from lower childhood physical aggression trajectories. However, for all childhood physical aggression trajectory levels the modal transition is to a relatively low-level adolescent aggression trajectory. Third, we find little evidence of "late onset" of high-level physical aggression. Specifically, the joint trajectory analysis finds no evidence of transition from a low physical aggression trajectory in childhood to a high trajectory in adolescence.

Keywords: Adolescence, aggression, child development, violence.

Abbreviations: BIC: Bayesian Information Criterion.

Introduction

Over the past decade the use of a developmental psychopathology perspective has led investigators to look for the age of "onset" of specific disorders. A rapid consensus was reached that there were two categories of cases: those who initiated that kind of behavior during childhood and those who initiated it during adolescence (e.g., American Psychiatric Association, 1994; Moffitt, Caspi, Dickson, Silva, & Stanton, 1996; Patterson, DeBaryshe, & Ramsey, 1989). The consensus is surprising because the two categories solution aggregates a large variety of antisocial phenomena that could have different etiologies and developmental trajectories, and thus would probably need different prevention and treatment strategies.

Within 14 years the DSM classification of conduct disorder went from five categories (DSM-III: under-socialized aggressive, undersocialized nonaggressive, socialized aggressive, socialized nonaggressive, atypical), to three (DSM-III-R: group, solitary aggressive, undifferentiated), to two (DSM-IV: childhood-onset, adolescent-onset). The major loss in this simplification process has been the differentiation between antisocial behavior with and without physical aggression. This is an especially surprising orientation, since physical aggression is probably the main reason the general public and funding agencies support research on antisocial behavior. Most courts of law will treat differentially a juvenile who physically aggressed and juveniles who shoplifted, ran away from home, or used marijuana. There is good evidence that the mean frequency of physical aggression increases during the first year after birth, and then decreases steadily from the third year after birth (Goodenough, 1931; Nagin & Tremblay, 1999; Restoin et al., 1985; Tremblay et al., 1996, 1999). One would expect that running away from home, shoplifting, and marijuana use would not follow the same trajectory as physical aggression.

Most tests of the early- and late-onset antisocial behavior classification aggregated different forms of antisocial behavior (e.g., Hodgins, 1994; Moffitt et al., 1996; Nagin & Farrington, 1992; Tibbetts & Piquero,
follow-up study involving five additional prospective evidence of the onset of physical aggression after age 6. A physical aggression from age 6 onward. There was no identified; all had either stable or declining trajectories of the population. Four such distinct trajectory groups were distinctive groupings of developmental trajectories within Nagin (1999), which is specifically designed to identify parametric, group-based methodology, described by ever, Nagin and Tremblay (1999) find no evidence of such evidence of physical aggression among subpopulations. How-

little evidence that is available suggests that humans begin physically aggressing as young children become physically violent in life circumstances has been implicated in the onset of delinquency.

This study builds from the findings of Nagin and Tremblay (1999) and Broidy et al. (1999), by exploring the linkage between trajectories of physical aggression in childhood and adolescence. The key question we address is whether there is any substantial evidence of onset of physical aggression in adolescents among individuals with no history of physical aggression during the elementary school years, and if there is, whether that physical violence is chronic and sustained rather than episodic.

We also have a methodological objective. Specifically, we develop an important extension to the semiparametric, group-based method for identifying and analyzing developmental trajectories described in a number of recent articles (Land, McCall, & Nagin, 1996; Nagin, 1999; Nagin & Land, 1993; Nagin & Tremblay, 1999; Roeder, Lynch, & Nagin, 1999). Prior applications of the method analyzed repeated measures of a single response variable (e.g., the number of convictions from ages 10 to 32). Here we extend the method to permit the estimation of trajectories for repeated measures of two different response variables—physical aggression in childhood as measured by teacher reports and physical aggression in adolescence as measured by self-reported violent delinquency.

We offer this method to complement the conventional approach for measuring stability of behavior that uses measures of association—such as correlation coefficients—to index the strength of the relationship between the measured behavior of the same individuals at two points in time (see, e.g., Backteman & Magnusson, 1981; Farrington, 1991; Huesmann, Eron, Lefkowitz, & Walder, 1984; Loebel & LeBlanc, 1990; Olweus, 1979). In our judgement, stability coefficients suffer from several important limitations. First, the magnitude of a stability coefficient is not easily interpreted. What, for example, does a stability coefficient of +0.6 imply about the proportion of individuals who exhibit behavioral continuity in comparison to the proportion of individuals who exhibit behavioral change? How should we interpret a stability coefficient of .3 in comparison to a stability coefficient of .6? Clearly, one is larger but how much larger? Second, the size of stability coefficients are quite sensitive to inclusion of outlier data from highly skewed distributions such as those often encountered in research on crime and antisocial behavior (Moffitt, 1993). Third, stability coefficients are based on measurements at two points in time and, thus, make inefficient use of longitudinal data with more than two assessment periods. If
one wishes to investigate stability for more than two periods, it is necessary to calculate multiple stability coefficients. Even more fundamentally, stability coefficients may be sensitive to the choice of assessment periods used in their calculation. Fourth, the usual and customary interpretation of a correlation coefficient assumes that it applies equally to all individuals within the population under study. However, research showing substantial heterogeneity in the development of antisocial behavior over the life span calls into question the plausibility of such an assumption (e.g. the theories of Moffitt, 1993; and Patterson et al., 1989).

The method we use here has several noteworthy features that avoid these problems. First, it allows us to identify—in concrete, visual terms—the proportion of individuals who exhibit behavioral continuity and the proportion of individuals who exhibit behavioral change. Second, because the approach is based on mixtures of simple, but realistic, probability distributions, it provides a basis for identifying population heterogeneity in developmental trajectories. Relatedly, this feature also allows us to distinguish between random noise and systematic population variation in patterns of childhood physical aggression and adolescent violence. This avoids an important pitfall associated with constructing groups based on subjective or adhoc classification rules—overfitting the data with the creation of groups that reflect only random variation. Third, it makes use of all the available data, not just assessments at two periods in time.

**Methods**

**Sample and Instruments**

The data used in this analysis were obtained from a long-term longitudinal study of 1161 white males who were enrolled in kindergarten classes in 53 low socioeconomic status neighborhood schools in Montreal. Only those boys whose biological parents were born in Canada and whose parents’ primary language was French were included in the study. After applying these criteria and eliminating individuals who refused to participate or who could not be located, a sample of 1037 subjects remained.

The first wave of data collection occurred in 1984, when all of the boys were approximately 6 years old. At ages 6, 10, 11, 12, and 13 the boys were rated by their teachers on the Social Behavior Questionnaire (Tremblay et al., 1991). For purposes of our analysis, we are interested in the distribution of the physical aggression measures in this instrument. At each wave of data collection, a physical aggression index was created by combining data on three items: (1) kicks, bites, hits other children; (2) fights with other children; and (3) bullies or intimidates other children. The final index ranges from 0 to 6 at each wave of data collection. The internal consistency index (Cronbach’s alpha) ranges from .78 to .87.

Our analysis also relies on a measure of self-reported physical aggression that was collected from each boy at ages 13, 14, 15, 16, and 17. The internal consistency index (Cronbach’s alpha) ranges from .62 to .75. To obtain this information, each boy was asked to report the frequency with which he had committed each of the following acts within the previous 12 months: (1) gang fighting; (2) using a deadly weapon; (3) attacking someone who had done nothing to you; and (4) throwing rocks, bottles, or other objects at someone. Response options included: (0) never, (1) once or twice, (2) sometimes, and (3) often. Thus, the combined measure at each wave is an integer index of the frequency of physical aggression over the preceding year. This index does not control for the seriousness of the incident. However, we have conducted our analysis with different subsets of the items and our findings do not materially differ.

A small number of the 1037 boys had missing data on either all of the childhood or adolescent aggression outcomes. After deleting these boys from the data set, we were left with a final analysis sample of 926 boys. While some of the 926 boys had missing data at some waves, we allowed each of them to contribute information to the analysis for each wave in which they were observed. For additional details about these data, see Tremblay, Pihl, Vitaro, and Dobkin (1994) and Tremblay et al. (1991).

**Statistical Analyses**

In this section, we provide an overview of our analysis methods. A more detailed discussion is presented in the Appendix. The statistical model advances prior work of Nagin and colleagues in the use of finite mixture models to identify distinctive developmental trajectories within the population. As developed in Nagin (1999), this modeling strategy is particularly well suited for analyzing individual-level developmental trends in which development does not vary regularly and smoothly across population members. Rather, the population is viewed as being comprised of distinct clusters of developmental trajectories.

Mixture models assume that the population is comprised of a finite number of unobserved groups of individuals. The groups are defined by an expected developmental trajectory that relates the expected level of the behavior of interest with age. Technically, we model this linkage between expected behavior and age by up to a second-order polynomial equation. The parameters of this relationship are allowed to vary freely across groups. This provides for great flexibility in describing the developmental trajectory of each group. In addition to estimating a parametric representation of each group’s trajectory, the proportion of the population comprising each trajectory group is also estimated.

The mixture model framework applied here is designed to combine trajectories of distinct but related behaviors—in this case trajectories of physical aggression from ages 6 to 13 years based on teacher reports are combined with trajectories of physical aggression from ages 13 to 17 years based on self-reports. Each group is characterized by two separate trajectories—one for childhood aggression and the other for adolescent aggression. The trajectories are the product of a joint estimation procedure that estimates each group’s trajectories in combination. In addition to the two trajectories, which in combination specify the developmental course of physical aggression from ages 6 to 17 years, an estimate of the proportion of the population following this developmental course is also produced.

Model estimation is achieved by maximization of the likelihood that is derived in the Appendix. This likelihood brings together earlier mixture model work that was designed to accommodate special features of the data to be analyzed. Nagin and Tremblay (1999) develop a mixture model based on the censored normal distribution. The censored normal is well suited to accommodate a common feature of psychometric scale data such as the childhood physical aggression data analyzed here. Typically a sizable contingent of the sample exhibits none of the behaviors measured by the scale. The result is a clustering of data at the scale minimum. Also, there is usually a smaller contingent that exhibits all of the behaviors measured by the scale. The result is another cluster of data at the scale maximum. Nagin and Land (1993), Land et al. (1996), and Roeder et al. (1999) develop a mixture model based on the Poisson distribution and its more general relative, the zero-inflated Poisson distribution. This Poisson-based model is designed to estimate trajectory models in which the response variable is an integer-valued index or count and is used to model self-reported physical aggression during adolescence.
In addition to a depiction of the shapes of a group’s joint behavioral trajectories and estimation of the proportion of the population following such a joint trajectory, another important output is an estimate of the probability that each individual belongs to each of the groups. This probability is calculated on the basis of the data and the maximum likelihood parameter estimates associated with the mixture; we call it the posterior probability of group membership. Using this probability, individuals can be assigned to the group to which they have the highest probability of belonging.

Final model selection requires a determination of the number of groups that best describes the data. As described in Nagin (1999), a conventional chi-squared-based test of goodness-of-fit cannot be used to test whether the addition of a group results in a significant improvement in explanatory power. Instead we follow the recommendation of Kass and Raftery (1995) and Raftery (1995) to use the Bayesian Information Criterion (BIC) as a basis for selecting the optimal model. For any given model BIC is calculated as: \( \text{BIC} = -2 \log(L) + \log(n)k \), where \( L \) is the model’s maximized likelihood, \( n \) is the sample size, and \( k \) is the number of parameters in the model.

**Results**

To introduce the joint trajectory analysis we begin by reporting results for separate trajectory models for childhood and adolescent aggression, respectively. Based on the BIC criterion a three-group model was found to best fit the childhood aggression data. For the adolescent aggression data a six-group model was found to best fit the data. However, here we describe the four-group model because the results from this more parsimonious solution are qualitatively similar. Figure 1 presents the results of this analysis.

The left panel of Fig. 1 shows the forms of the three childhood physical aggression trajectories. The largest group, an estimated 50\% of the population, is defined by a low and declining aggression trajectory; the next largest group, about 30\% of the population, is described by a medium and generally declining trajectory; and the smallest group, about 20\% of the population, by a trajectory of high aggression. These findings generally conform with Nagin and Tremblay (1999). However, in addition to the three groups identified here, they also found a small group of boys who showed no sign of desistance from their high level of physical aggression. This chronic group emerges with the use of physical aggression measurements made at ages 14 and 15. To avoid overlap with the adolescent violent data these measurements were not used in this analysis.

The right panel of Fig. 1 reports the trajectories for adolescent physical aggression. The largest group (64\%) self-reports very little physical aggression over the observation period. The smallest group (5\%) reports high levels of aggression from ages 13 to 17 years old. In between are two groups of about equal size, an estimated 15\% of the population each, that display opposite trajectories. One is a desistance trajectory in which group members start off with a high level of physical aggression at age 13 but by age 17 have desisted from such behavior. The other is a rising trajectory group that starts off reporting no physical aggression at age 13 but by age 17 is reporting modest levels.

We turn now to the results of the joint analysis in which the trajectories of physical aggression in childhood are linked with those in adolescence. Based on the BIC criterion, a seven-group model was found to best fit the data. The posterior probabilities of group membership also suggest that the fit is good. These probabilities provide a quantitative and objective basis for assigning individuals to the trajectory group that best matches their behavior. For example, an individual who displayed a high level of physical aggression in childhood but who ceased being violent in adolescence would probably be assigned to a group with a comparable predicted trajectory. Ideally, for each individual this probability should be one for the group to which they are assigned and thereby zero for all others. To the degree that there are large numbers of individuals whose behavior is not well matched by a trajectory group, assignment proba-

![Figure 1. Marginal trajectories of childhood physical aggression and adolescent physical aggression (n = 926).](image-url)
Aggression during adolescence.

Reports a relatively high but declining level of physical aggression. The larger of the two groups, estimated to account for about 10% of the population, includes individuals the low childhood aggression group. A second group has a trajectory of medium but declining levels of childhood physical aggression but divergent trajectories of adolescent aggression. Despite the similarity of their childhood aggression but divergent trajectories of adolescent aggression, however, these two groups display markedly different trajectories of aggression in adolescence. One group, which we call the high childhood aggression/high adolescent aggression group, is estimated to account for about 3% of the population. At age 13, the self-reported physical aggression of this group is higher than that of any other group at any age. Their violence rises steadily to a peak at age 15 and thereafter declines. This single-peaked pattern conforms to conventional findings based on aggregate data on the relation of age and crime (Farrington, 1986; Hirschi & Gottfredson, 1983). Criminal involvement as measured by arrest rises steadily through adolescence and then begins a steady decline thereafter. It is noteworthy, however, that this is the only single-peaked individual level trajectory identified in this analysis. The second group is comprised of individuals who, despite their high aggression in childhood, report a negligible level of violence in their youth. This group is estimated to comprise 10% of the population.

Consider the implication of these trajectory plots for the continuity of physical aggression. One definition of continuity is that behavior remains stable over time. By this definition, the results provide little evidence of stability. All seven of the childhood physical aggression trajectories are generally declining. For the adolescent trajectories two are declining, one is rising, and one is hump shaped. Another definition of continuity is that behavior remains stable over time. In this group-based modeling approach, relative stability is supported by trajectories not intersecting; intersection implies change in the relative standing of groups vis-à-vis the measured behavior. By this definition there is very substantial evidence of continuity. Inspection of Fig. 2 reveals that there are only two instances of intersection in the childhood trajectories. Among the adolescent trajectories, there is only one trajectory that intersects others. Still, the other six trajectories maintain their relative positions throughout the observation period. Overall, the results provide strong evidence of...
We now consider continuity across the two life course stages—childhood and adolescence. A striking feature of the adolescent aggression plots in Figs. 3a–c is the variability of adolescent outcomes for groups with similar trajectories of childhood physical aggression. Consider the two high childhood aggression groups (Fig. 3c). One displays a clear form of continuity—their high childhood physical aggression is followed by high adolescent physical aggression. However, the other group with comparably high childhood physical aggression goes on to report virtually no physical aggression in adolescence. This group is three times larger than the high adolescent aggression group. Thus, even among those individuals who were highly physically aggressive in their childhood, the probability of their engaging in little or no adolescent aggression is still very high—about .77 (.77 = .10/(.03+.10)).

Next, consider the two middle childhood aggression trajectories (Fig. 3b). One group displays a clear form of behavioral continuity and goes on to exhibit considerable adolescent aggression. However, this group is half the size of the counterpart group that goes on to self-report virtually no physical aggression during adolescence.

Finally, consider the low childhood aggression cluster (Fig. 3a). The group with the majority of cases reports virtually no physical aggression in adolescence. A second group proceeds to display a pattern of adolescent physical aggression that rises to a moderate level, while the third group gives a mirror image; physical aggression is moderate at age 13 and decreases with age. These three
groups are respectively estimated to account for 30%, 13%, and 11% of the population. Thus, for those who display low levels of physical aggression during childhood, the probability of their going on to follow a rising trajectory of physical aggression in adolescence, even only to a moderate level, is a low .23 (.23 = .13/.13 + .11 + .33). The counterpart probabilities for the declining and never adolescent aggression trajectories are .19 and .58, respectively. Again, a wide variety of adolescent aggression trajectories attend to comparable trajectories of childhood aggression.

Another salient aspect of the change captured by the joint trajectories is that change is commonly toward less physical aggression. About 13% of the population are estimated to belong to the two high childhood aggression trajectory groups. About 75% of these individuals go on to belong to the trajectory group that reports virtually no physical aggression in adolescence. Similarly, for the two middle childhood aggression trajectories and for the three low childhood aggression trajectories the modal transition is to the nonviolent adolescent trajectory group.

The joint trajectory model combines trajectories of childhood and adolescent physical aggression. An alternative approach allows for a probabilistic transition from the three childhood physical aggression trajectories identified depicted in Fig. 1 to the four adolescent trajectories also depicted in that figure. Specifically, we examine the overlap of group membership between the childhood and adolescence trajectory groups. For example, we measure the proportion of individuals in the high childhood aggression trajectory group who go on to follow a high-level trajectory of physical aggression in adolescence. To compute these transition probabilities we first assign individuals to the childhood and adolescence physical aggression trajectory groups based on the posterior probability of group membership. The transition probabilities are computed by cross-tabulating the childhood and adolescent trajectory group assignments.

Table 1 reports these transition probabilities and in addition reports a conventional measure of association, $\gamma = .462$, for ordinal contingency tables. Consistent with expectations, the trajectories for childhood and adolescent aggression are not independent. For all childhood aggression groups—including the high aggression group—the modal transition probability is to the no-aggression trajectory in adolescence. This finding matches the pattern observed in the joint analysis—a general tendency to transition to less physical aggression. That said, the probability of transitioning to the no adolescent aggression group declines from .75 for the low childhood aggression trajectory to .40 for the high childhood aggression trajectory. Also, the probability of transitioning into the high adolescent aggression group rises from .02 for the low childhood aggression group to .13 for the high childhood aggression group. This finding is also mirrored in the joint analysis—children with a history of aggression are much more likely to be aggressive as adolescents. Conversely, the analysis suggests that adolescent initiation of high levels of sustained physical aggression among those without a history of childhood physical aggression is a rare event. Only 1.7% of the individuals in the low childhood aggression group make the transition to the high physical aggression group in adolescence. Those individuals from the low childhood aggression group who go on to engage in some physical aggression during adolescence most commonly follow the two moderate level adolescent aggression groups.

The distribution of responses to the individual items that comprise the adolescent physical aggression scale provides further perspective on the prevalence of sustained physical aggression in adolescence among the childhood physical aggression groups. For any given item (e.g., attacked someone with a weapon), the individual had four response options—never in the past year, one to two times, occasionally, and frequently. We counted the number of individuals by childhood aggression group who in any year from ages 13 to 17 responded that they frequently engaged in any of the violent acts in our index. Only 3% of the individuals in the low childhood physical aggression group responded that they frequently engaged in any of these violent behaviors at any time from ages 13 to 17 whereas 8% and 17% of the middle and high aggression groups gave at least one affirmative reply.

**Discussion**

The objective of this analysis was to examine the linkage between childhood and adolescent physical aggression. Our study advances research on this issue in two important ways. First, we have focused exclusively on physical aggression. We have maintained this focus because physical violence is arguably the most socially destructive form of antisocial behavior. Also, the developmental course of physical aggression may be different than for other forms of antisocial behavior.
including nonphysical aggression (Crick, Casas, & Ku, 1999; Nagin & Tremblay, 1999; Tremblay, 2000; Tremblay et al., 1999). A second key contribution of this study is that it links trajectories of distinct but analogous behaviors—teacher-rated childhood physical aggression and self-reported adolescent physical aggression. Previous research has used a static analytic approach in which behavior at one point in time is associated with behavior at a later point in time. This analytical strategy does not adequately capture the developmental course of the behavior of interest.

Several salient findings have emerged from the analysis. First, we find considerable change in the levels of childhood and adolescent physical aggression within different groups. Thus, there is little evidence of stability of behavior in an absolute sense. Instead we find strong evidence of stability only in a more narrow sense of the term—the relative ranking of the groups. Our findings on the stability of relative standing closely conform with much prior research. Understanding the mechanisms that underlie the changes in absolute levels of physical aggression for the different groups should help identify more effective interventions to prevent physical violence.

A second key finding concerns the connection of childhood aggression to adolescent aggression. Boys with higher childhood physical aggression trajectories are far more likely to transition to a higher-level adolescent aggression trajectory than boys from lower childhood physical aggression trajectories. However, for all childhood physical aggression trajectory levels the modal transition is to a relatively low-level adolescent aggression trajectory. This finding is reminiscent of Robins' (1978, p. 611) observation that “adult antisocial behavior virtually requires childhood antisocial behavior [yet] most antisocial children do not become antisocial adults”. Since our results apply only to physical violence, future research should try to identify the specific antisocial behaviors for which this statement is not valid. This knowledge would be extremely useful for targeting the appropriate type of preventive intervention to the appropriate age group.

A noteworthy result from a theoretical and practical perspective is that we find little evidence of “late onset” of high-level physical aggression. The joint trajectory analysis finds no evidence of transition from a low physical aggression trajectory in childhood to a high trajectory in adolescence. Within the three-group cluster of low childhood physical aggression individuals (Fig. 3a), a majority of the boys report no physical aggression in adolescence. For the minority who do indicate some physical aggression, self-reported levels are moderate, not high. The results of the “classify-analyze” procedure reported in Table 1 also reveals little evidence of late onset of physical aggression. The boys assigned to the low aggression trajectory constitute only about 15% of the population of the high adolescent physical aggression group.

Thus, by any standard our analysis suggests that individuals without any substantial history of physical aggression during the elementary school years are a small minority of adolescents who engage in high levels of physical aggression. This conclusion calls into question the applicability of theories that implicate late onset in their explanation of physical violence.

That said, we are not suggesting that adolescents without a history of physical aggression in childhood never commit acts of serious violence. We are, however, not comfortable with the use of the term “onset” to describe such incidents. Onset is meant to imply the emergence of an enduring pattern of behavior. We would not use the word onset to describe events that are episodic and without real continuity. For example, we do not use the term “onset of flu” to describe a person’s first bout with the flu. Periodic bouts with the flu are part of the human condition, not an enduring syndrome or ongoing pattern of related events. Adolescence is commonly a tumultuous period in human development. For the first time, physically but not socially mature individuals have ample time to socialize without supervision. It should not be surprising that the physical aggression that virtually all individuals display in the first years of life (Tremblay et al., 1999) might emerge sporadically in the turbulent and largely idiosyncratic circumstances of adolescence. Thus, we view these unusual instances of aggression among the individuals with no history of childhood physical aggression (from age 6 onward) as incidents that do not reflect an enduring behavioral pattern. Whether we can predict and prevent these incidents will need to be specifically investigated.

Finally, contrary to widespread belief, our results clearly indicate that frequent use of physical aggression is relatively uncommon during adolescence, even in a sample of males who started school in low socioeconomic areas of a large North American city. The chronic cases of physical aggression during adolescence had chronic problems with physical aggression at least since kindergarten, and most probably since infancy (Tremblay et al., 1999). The identification of the different developmental trajectories should help identify the reasons why with age these boys fail to reduce their use of physical aggression, while most of their peers do.

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References


In this Appendix, we provide the details of our joint trajectory modeling procedure. For details on the univariate trajectory model see Nagin (1999). Let $c_i$ denote the childhood aggression score associated with individual i at time t and let $a_i$ denote the adolescent aggression score associated with individual i at time t. In the special case where $c_i$ and $a_i$ are independent and conditional upon membership in group j, we can write their joint probability distribution as the product of two independent probability distributions:

\[ p(c_i, a_i | \beta_0, \beta_1^j) = f(c_i | \beta_0^j) h(a_i | \beta_1^j) \]

where $\beta_0^j$ and $\beta_1^j$ denote the parameters that govern the probability distribution for $c_i$ and $a_i$, $f(\cdot)$ and $h(\cdot)$, respectively.

For this application $c_i$ is assumed to follow the censored normal distribution and $a_i$ is assumed to follow the Poisson distribution.

For each class $j$, we assume the parameter governing the childhood aggression score for individual i at time t is given by

\[ z_{it} = \theta_{it} + \xi_t, \]

where $\xi_t$ can range from ages 6 to 13. Similarly, the parameter governing the probability mass function for the adolescent aggression distribution is given by

\[ \beta_i = \exp (\beta_0 + \beta_1^j t + \beta_1^j t^2) \]

where $t$ can range from ages 13 to 17.

Let $C_i$ and $A_i$ denote each individual's longitudinal sequence.
of measurement of $c_{it}$ and $a_{it}$. Again conditional on group membership we assume $c_{it}$ and $a_{it}$ are independently distributed over time. Thus,

$$P(C_i, A_i | J) = \prod_j f(c_{it} | \alpha_{it}) h(a_{it} | \beta_{it}).$$

Because group membership is not observed, construction of the unconditional likelihood of $A_i$ and $C_i$ requires that we sum over the $J$ conditional likelihood functions and weight each by $\pi_j$, the probability of membership in class $j$:

$$P(A_i, C_i) = \sum_j \pi_j P(C_i, A_i).$$

Finally, the likelihood function that is used for estimation of the model parameters, $\alpha$ and $\beta$, is constructed as the product of $P(A_i, C_i)$ over the $I = 1, 2, \ldots, N$ individuals comprising the estimation sample.