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- Other moments may be of interest, like dynamic choice transitions

Wandering astray: Teenagers' choices of schooling and crime

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We build and estimate a dynamic model of teenagers' choices of schooling and crime, incorporating four factors that may contribute to the different paths taken by different teenagers: beterogeneous endowments, unequal opportunities, uncertainties about one's own ability, and contemporaneous shocks. We estimate the model using administrative panel data from Chile that link school records with juvenile criminal records. Counterfactual policy experiments suggest that, for teenagers with disadvantaged backgrounds, interventions that combine mild improvement in their schooling opportunities with free tuition (by adding 157 USD per teenager-year to the existing high school youcher) would lead to an 11% decrease in the fraction of those ever arrested by age 18 and a 13% increase in the fraction of those consistently enrolled throughout primary and secondary educa-

KEYWORDS. Teenage crime, education, information friction, institutional friction. dynamic model, structural estimation.

TEL CLASSIFICATION 12 VA2

1. INTRODUCTION

Teenage years are a critical period in life, featuring major physical, psychological, and attitudinal transitions. Faced with all these complications, some teenagers may experience a particularly difficult transition to adulthood and wander astray, dropping out of school and/or engaging in criminal activities. Invenile delinquency is a serious problem worldwide. For example, in the U.S., over 725,000 teenagers were in detention centers

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We thank Tatiana Reves and Damián Vergara for excellent research assistance. We thank seminar and conference participants at Duke, PUC. Rio. Universidad de los Andes (Colombia). Universidad de Chile, and Econometric Society Meetings for helpful comments. Nicolás Grau thanks the Centre for Social Conflict and Cohesion Studies (ANID/ FONDAP/15130009) for financial support. Jorge Rivera thanks the support from the ANID PIA/APOYO AFB180003. Powered@NLHPC: This research was partially supported by the supercomputing infrastructure of the NLHPC (ECM-02). All errors are ours.

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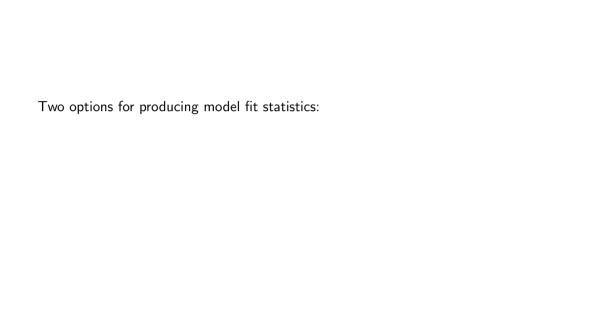
 $\label{lem:model} \mbox{Model fit results for youth with parents who are not well educated:}$

Variable	Data	Model
Ever arrested %	5.6	5.2
Always enrolled, 0 arrest %	72.2	72.0
GPA (standardized)	-0.39	-0.43
Retention %	7.3	7.8
Grade Completed by T	11.1	11.3

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Model fit for other groups is similar



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$ullet$ Use model assumptions and estimates to simulate a new dataset, $ ilde{Y}$ vs. Y

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$ullet$ Use data and estimates to compute \hat{Y} and compare with Y

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- Repeat process multiple times and take average to limit simulation error

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