# **Parental Investment and Child Development**

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### **Objectives**

- Formulate a dynamic production function model for the various dimensions of human capital with endogenous parental investments to examine dynamic complementarities and interactions among different inputs in forming child human capital.
- Use a maximum likelihood approach by the minorization-maximization algorithm to estimate the distribution of the latent factors and dynamic CES production functions of human capital.

#### Contributions

- This research is the first attempt to analyze the determinants and interactions of three important dimensions of human capital: cognitive skills, noncognitive skills and health.
- It provides rare evidence in developing country settings on parental investments and child development
- It uses a latent factor approach to identify the latent, unobserved factors instead of (noisy) proxy variables to correct measurement error problems and explore the endogeneity of investments.

## Model: Dynamics of Skill Formation

 The framework builds on the dynamic factor models of Cunha et al. (2010), Cunha and Heckman (2007), Attanasio et al. (2017), Agostinelli and Wiswall (2016) and Aucejo and James (2021):

$$\Theta_{i,t}^{k} = f(\Theta_{i,t-1}^{C}, \Theta_{i,t-1}^{NC}, \Theta_{i,t-1}^{H}, I_{i,t}, P_{i}^{C}, P_{i}^{NC}, X_{i,t}, A_{t}^{k}, v_{i,t}, \varepsilon_{i,t}^{k})$$
(1)

$$k \in \{C, NC, H\}$$

I use a Constant Elasticity of Substitution (CES) production function

$$\Theta_{i,t}^{k} = \left[ \gamma_{1t}^{k} (\Theta_{i,t-1}^{C})^{\rho^{tk}} + \gamma_{2t}^{k} (\Theta_{i,t-1}^{NC})^{\rho^{tk}} + \gamma_{3t}^{k} (\Theta_{i,t-1}^{H})^{\rho^{tk}} + \gamma_{4t}^{k} (I_{i,t})^{\rho^{tk}} \right]$$

$$\gamma_{5t}^{k} (P_{i}^{C})^{\rho^{tk}} + \gamma_{6t}^{k} (P_{i}^{NC})^{\rho^{tk}} \right]^{1/\rho^{tk}} e^{X'_{i,t} \delta_{t}^{k} + A_{t}^{k} + \mu^{k} v_{i,t} + \varepsilon_{i,t}^{k}}$$
(2)

Where  $\gamma_{1t}^k + \gamma_{2t}^k + \gamma_{3t}^k + \gamma_{4t}^k + \gamma_{5t}^k + \gamma_{6t}^k = 1$ .

## Measurement System

- The latent factors  $(\Theta_{i,t}^C, \Theta_{i,t}^{NC}, \Theta_{i,t}^H, \Theta_{i,t-1}^C, \Theta_{i,t-1}^{NC}, \Theta_{i,t-1}^H, P_i^C, P_i^{NC}, \text{ and } I_{i,t})$  are not directly measured, the factor model approach is used to extract these unobserved variables from a large set of observed data.
- Define the natural log of the factors as  $\theta_{i,t}^k = ln(\Theta_{i,t}^k)$ ,  $\theta_{i,t-1}^k = ln(\Theta_{i,t-1}^k)$ ,  $\mathcal{P}_i^C = ln(\mathcal{P}_i^C)$ ,  $\mathcal{P}_i^{NC} = ln(\mathcal{P}_i^{NC})$  and  $\mathcal{I}_{i,t} = ln(\mathcal{I}_{i,t})$
- The observed measures proxy the natural log of the factors.

## Measurement System

- There are two types of the observed measures, continuous and binary measures.
- The continuous measures are described by:

$$T_{i,j,\tau}^k = \alpha_{j,\tau}^k + \beta_{j,\tau}^k \theta_{i,\tau}^k + u_{i,j,\tau}^k \tag{3}$$

The binary measures are described by:

$$T_{i,j,\tau}^{k} = \mathbb{1}[\alpha_{j,\tau}^{k} + \beta_{j,\tau}^{k} \theta_{i,\tau}^{k} + u_{i,j,\tau}^{k} > 0]$$
 (4)

where:

- $T_{i,j,\tau}^k$ : the *j*th measure relating to latent factor *k* for individual *i* at time  $\tau$ ,  $\tau \in \{t-1,t\}$ .
- o  $\theta_{i,\tau}^k$ : unobserved/latent factors:  $\theta_{i,t}^k$ ,  $\theta_{i,t-1}^k$   $\mathcal{P}_i^{\mathcal{C}}$ ,  $\mathcal{P}_i^{\mathcal{NC}}$  and  $\mathcal{I}_{i,t}$
- $\circ$   $u_{i,h,t}$ : error terms.

#### Parental Investments

$$InI_{i,t} = \alpha_{1,t} + \alpha_{2,t} In\Theta_{i,t-1}^{C} + \alpha_{3,t} In\Theta_{i,t-1}^{NC} + C\alpha_{4,t} In\Theta_{i,t-1}^{H} + \alpha_{5,t} InP_{i}^{C} + \alpha_{6,t} InP_{i}^{NC} + \alpha_{7,t} X_{i,t} + \alpha_{8,t} Z_{i,t} + v_{i,t}$$
(5)

#### where:

- $\circ$   $X_{i,t}$ : child gender, urban/rural residence and the number of siblings.
- $\circ$   $Z_{i,t}$ : instrumental variables including the log of wealth index reflecting parental resources, household economic shocks and the log of regional prices.
- $\circ$   $v_{i,t}$ : an error term.

Measurement System:

$$T_{i,j,\tau}^{k} = \alpha_{j,\tau}^{k} + \beta_{j,\tau}^{k} \theta_{i,t-1}^{k} + u_{i,j,\tau}^{k} T_{i,j,\tau}^{k} = \mathbb{1}[\alpha_{j,\tau}^{k} + \beta_{j,\tau}^{k} \theta_{i,\tau}^{k} + u_{i,j,\tau}^{k} > 0]$$
(6)

Parental Investments:

$$InI_{i,t} = \alpha_{1,t} + \alpha_{2,t} In\Theta_{i,t-1}^{C} + \alpha_{3,t} In\Theta_{i,t-1}^{NC} + C\alpha_{4,t} In\Theta_{i,t-1}^{H} + \alpha_{5,t} InP_{i}^{C} + \alpha_{6,t} InP_{i}^{NC} + \alpha_{7,t} X_{i,t} + \alpha_{8,t} Z_{i,t} + v_{i,t}$$

$$(7)$$

Production functions:

$$\Theta_{i,t}^{k} = \left[ \gamma_{1t}^{k} (\Theta_{i,t-1}^{C})^{\rho^{tk}} + \gamma_{2t}^{k} (\Theta_{i,t-1}^{NC})^{\rho^{tk}} + \gamma_{3t}^{k} (\Theta_{i,t-1}^{H})^{\rho^{tk}} + \gamma_{4t}^{k} (I_{i,t})^{\rho^{tk}} \right]$$

$$\gamma_{5t}^{k} (P_{i}^{C})^{\rho^{tk}} + \gamma_{6t}^{k} (P_{i}^{NC})^{\rho^{tk}} \right]^{1/\rho^{tk}} e^{X'_{i,t}} \delta_{t}^{k} + A_{t}^{k} + \mu^{k} v_{i,t} + \varepsilon_{i,t}^{k}$$
(8)

- The distributions of the log factors,  $f(\theta)$ ,  $f(\mathcal{P})$ ,  $f(\mathcal{I})$ , are assumed to be distributed as a mixture of two normals.
- Model estimation consists of two steps.
  - First step: estimate the measurement system to recover the parameters  $\beta_{j,\tau}^{C}$ ,  $\beta_{j,\tau}^{NC}$ ,  $\beta_{j,\tau}^{H}$ ,  $\beta_{j}^{PC}$ ,  $\beta_{j}^{PNC}$ ,  $\beta_{j,t}^{PNC}$ ,  $\alpha_{j}^{C}$ ,  $\alpha_{k}^{NC}$ ,  $\alpha_{j}^{PC}$ ,  $\alpha_{j}^{NC}$ ,  $\alpha_{j,t}^{NC}$  and the latent factor distributions
  - Second step: use the estimated parameters from the first step to take individual-specific draws and use these draws as observable data to estimate investment and production functions.

• The likelihood of all the observed measures conditional on  $\theta_i$ :

$$L(T_i|\theta_i) = \prod_{k=1}^K \prod_{i=1}^J \prod_{\tau=1}^T f(T_{i,j,\tau}^k|\theta_i)$$
(9)

Where  $\theta_i = [\theta_{i,\tau}^k]' = [\theta_{i,\tau}^C, \theta_{i,\tau}^{NC}, \theta_{i,\tau}^H, \mathcal{P}_i^C, \mathcal{P}_i^{NC}, \mathcal{I}_{i,\tau}]'$ 

The log-likelihood function:

$$\mathcal{L}(\Psi) = \sum_{i=1}^{N} ln \mathcal{L}(T_i) = \sum_{i=1}^{N} ln \left( \int \mathcal{L}(T_i | \theta_i) dF(\theta) \right) = \sum_{i=1}^{N} ln \left( \int \mathcal{L}(T_i | \theta_i) f(\theta) d\theta \right)$$
(10)

Where:

- o  $f(\theta) = \sum_{c=1}^{\infty} \tau_c f(\theta | \mu_c, \Omega_c)$ , where  $\mu_c$ ,  $\Omega_c$  and  $\tau_c$  are the mean, covariance and the mixture probability of the two normals.
- $\quad \ \ \, \Psi \,\, \text{is all the parameters of the model,} \,\, \Psi = \{\alpha,\beta,\sigma,\tau_c,\mu_c,\Omega_c\}.$



- The log-likelihood function is estimated using the minorization-maximization algorithm.
- Given the unobservable nature of the factors, the log-likelihood function is constructed by integrating over the distributions of the unobservable factors.

#### Data and variables

• Data is from Round 2 (at age 12) and Round 3 (at age 15) of the Young Lives survey for the Older cohort in Vietnam.

Latent factors		Observed variables
Child's cognitive skills - Round 2	$ heta_2^C$	<ol> <li>PPVT test</li> <li>Math Test</li> <li>Reading level</li> <li>Writing level</li> </ol>
Child's cognitive skills - Round 3	$\theta_3^C$	<ol> <li>PPVT test</li> <li>Math Test</li> <li>Cloze test</li> </ol>
Child's noncognitive skills - Round 2 and Round 3	$ heta_2^{NC},\  heta_3^{NC}$	<ol> <li>Self-esteem score</li> <li>Self-efficacy score</li> <li>Self-respect and inclusion score</li> </ol>
Child's health - Round 2 and Round 3	$\theta_2^H$ , $\theta_3^H$	<ol> <li>Child height for age z-score</li> <li>Child weight</li> <li>How is child health?</li> </ol>

### Data and variables

Latent factors		Observed variables
Parental cognitive skills	$\mathcal{P}^{C}$	<ol> <li>Mother's years of education</li> <li>Father's years of education</li> </ol>
Parental noncognitive skills	$\mathcal{P}^{NC}$	<ol> <li>Self-esteem score</li> <li>Self-efficacy score</li> <li>self-respect and inclusion score</li> </ol>
Parental Investments	$\mathcal{I}_3$	<ol> <li>Expenditure on the Young Lives child</li> <li>Number of hours studying outside school as a proxy for the time that parents dedicate to the child</li> <li>Quality of relationship between child and parents</li> </ol>

#### Estimates: Determinants of Parental Investments

Variables	Parental investments	
Child's cognitive skills, age 12	-0.024	
	(0.030)	
Child's noncognitive skills, age 12	-0.102	
	(0.076)	
Child's health, age 12	0.218***	
	(0.032)	
Parental cognitive skills	0.402***	
	(0.057)	
Parental noncognitive skills	0.172***	
	(0.066)	
Wealth index	0.177***	
	(0.045)	
Observations	961	

*Note:* Standard errors in parentheses based on 100 bootstrap replications of the entire estimation process; \*\*\* p<0.01, \*\* p<0.05, \* p<0.1.

<sup>-</sup> Selected covariates are reported.

#### Estimates of Production Functions

	Cognitive skills,	Noncognitive	Health, age 15
	age 15	skills, age 15	
Child's cognitive skills at age 12	0.613***	0.001	0.064**
	(0.044)	(0.039)	(0.029)
Child's noncognitive skills at age 12	0.145***	0.339***	0.130**
	(0.055)	(0.077)	(0.053)
Child's health at age 12	0.042*	0.065	0.670***
	(0.025)	(0.053)	(0.073)
Parental Investments	0.261***	0.505***	0.346**
	(0.075)	(0.167)	(0.170)
Parental cognitive skills	0.017	-0.005	-0.184
	(0.041)	(0.122)	(0.193)
Parental noncognitive skills	-0.079	0.096	-0.026
	(0.050)	(0.081)	(0.065)
$A_t$	-0.010	0.211***	0.205***
	(0.024)	(0.031)	(0.024)
Control Function (investment	-0.126*	-0.132	-0.420**
residuals)	(0.068)	(0.262)	(0.179)
Complementarity( $ ho$ )	-0.168	-1.851**	0.034
	(0.117)	(0.757)	(0.327)
Elasticity of substitution	0.856***	0.351*	1.036***
	(0.141)	(0.211)	(0.195)
Observations		961	

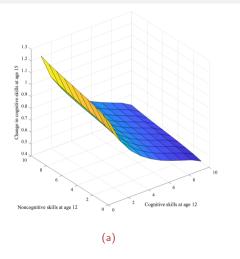
Note: Standard errors in parentheses based on 100 bootstrap replications of the entire estimation process; \*\*\* p<0.01, \*\* p<0.05, \* p<0.1.

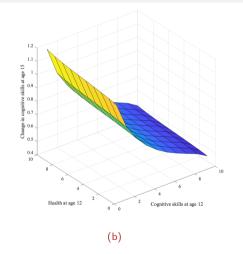
# Estimates of Production Functions - Marginal Effects

	Cognitive skills, age 15	Noncognitive skills, age 15	Health, age 15
Child's cognitive skills at age 12	0.614***	0.002	0.064**
	(0.043)	(0.039)	(0.029)
Child's noncognitive skills at age 12	0.145***	0.316***	0.130**
	(0.055)	(0.076)	(0.052)
Child's health at age 12	0.042*	0.085	0.670***
	(0.025)	(0.062)	(0.065)
Parental investments	0.261***	0.513***	0.346**
	(0.075)	(0.162)	(0.169)
Parental cognitive skills	0.017	-0.008	-0.184
	(0.042)	(0.123)	(0.194)
Parental noncognitive skills	-0.079	0.093	-0.026
	(0.050)	(0.078)	(0.067)
Observations	. ,	961	. ,

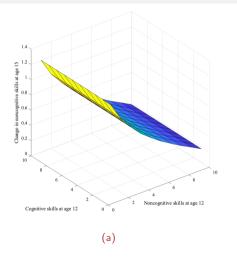
Note: Standard errors in parentheses based on 100 bootstrap replications of the entire estimation process; \*\*\* p < 0.01, \*\* p < 0.05, \* p < 0.1.

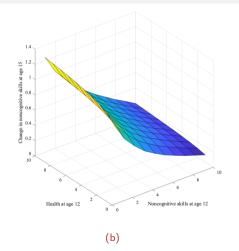
# Cognitive Skills: Self-productivity $(\partial \Theta_t^{\mathcal{C}}/\partial \Theta_{t-1}^{\mathcal{C}})$



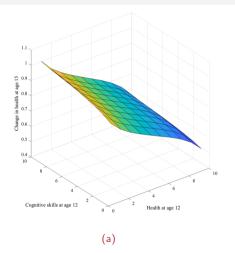


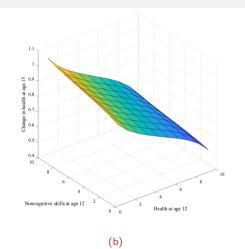
# Noncognitive Skills: Self-productivity $(\partial \Theta_t^{NC}/\partial \Theta_{t-1}^{NC})$





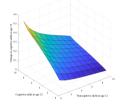
# Health: Self-productivity $(\partial \Theta_t^H/\partial \Theta_{t-1}^H)$

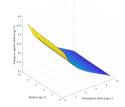




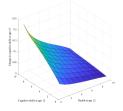
# Cognitive skill: Cross-productivity

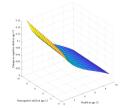
• Cross-productivity from Noncognitive Skills to Cognitive Skills  $(\partial \Theta_t^C/\partial \Theta_{t-1}^{NC})$ 





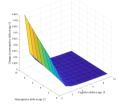
• Cognitive Skills: Cross-productivity from Health to Cognitive Skills  $(\partial \Theta_t^C/\partial \Theta_{t-1}^H)$ 

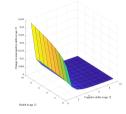




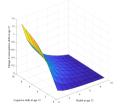
# Noncognitive skill: Cross-productivity

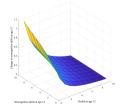
• Cross-productivity from Cognitive Skills to Noncognitive Skills  $(\partial \Theta_t^{NC}/\partial \Theta_{t-1}^C)$ 





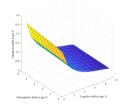
• Cross-productivity from Health to Noncognitive Skills  $(\partial \Theta^{NC}_t/\partial \Theta^H_{t-1})$ 

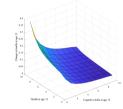




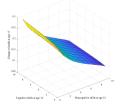
# Health: Cross-productivity

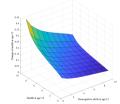
• Cross-productivity from Cognitive Skills to Health  $(\partial \Theta_t^H/\partial \Theta_{t-1}^C)$ 



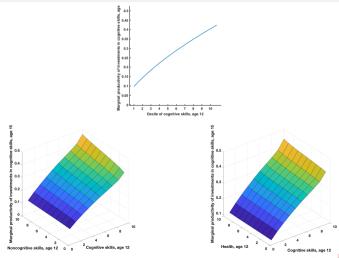


• Cross-productivity from Noncognitive Skills to Health  $(\partial \Theta_t^H/\partial \Theta_{t-1}^{NC})$ 

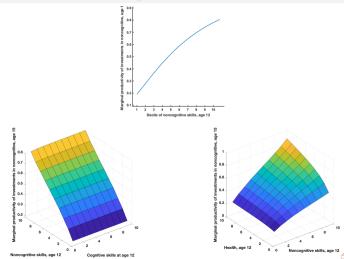




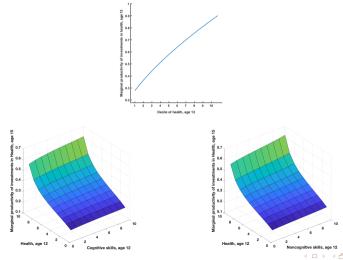
# Dynamic Complementarity between Investments and Cognitive Skills



# Dynamic Complementarity between Investments and Noncognitive Skills



# Dynamic Complementarity between Investments and Health



#### **Conclusions**

- Self-effects are present and strong in the production of all human capital dimensions.
- There is the existence of cross-productivity from noncognitive skills and health to cognitive skills and from cognitive and noncognitive skills to health.
- Parental investments strongly and directly affect the accumulation of skills and health
- There are strong dynamic complementarities between parental investments and three dimensions of human capital.

#### **Conclusions**

- The self-effects, cross-effects and dynamic complementarities together become a dynamic multiplier effect mechanism of human capital accumulation.
- Parental skills and resources are positively associated with parental investments.
- These effects could lead to substantially different growth rate of human captial and substantial increases in inequality.
- These results indicate the importance of interventions by boosting investments at early ages that can alter child development path, especially for disadvantaged children.
- Lack of parental investments can seriously hinder the development of a child.

Thank you!