

# Comprehensive Early Childhood Development Support Systems and Academic Achievement

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NBER SI Children 2022  
July 29th, 2022



# Comprehensive Support Systems

- Early childhood interventions aims to reduce early inequalities.
- [Shonkoff and Philips \(2000\)](#): fragmented + multiple entry points  $\Rightarrow$  inconsistent early **detection** systems and lack of **complementarity** of interventions to solve complex problems.
- **Comprehensive support systems** Array of early detection systems and array of services for children and their families.
- 2007: **Chile Crece Contigo** (*Chile Grows with You*)

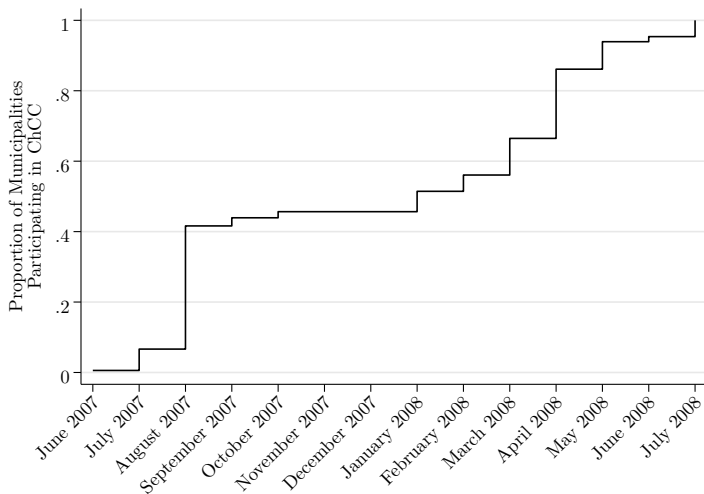
# Chile Crece Contigo Institutional Background

- Objective: *promote the development of children who participate in the public health network*
- Target Population: all children and their families from gestation until the first years of schooling.
- Changes introduced:
  - ▶ Introduce **coordination** of all services for children and their mothers from age zero to five.
  - ▶ Supply of services **expanded** with the focus changed from biomedical to bio-psychosocial approach.
- ChCC offers multiple services to all children that attends the public health system, but some services are offered only based on need.
- Gradual Implementation: June 2007 - July 2008

▶ Services Provided

▶ Annual Budget

# Chile Crece Contigo Rollout



# This paper

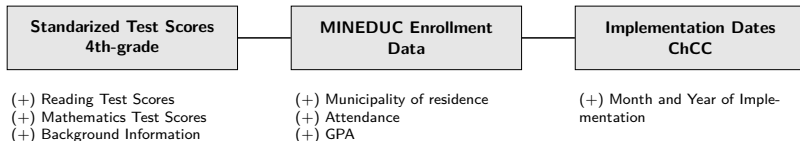
## Research Question

What are the long-term effects of exposure to the comprehensive support system in human capital accumulation?

- Main results:
  - ▶ Exposed since gestation vs Non-eligible: 0.25SD in mathematics and 0.30SD language standardized test scores.
  - ▶ Heterogeneous returns and heterogeneous exposure
    - ★ Male recipients have higher returns to exposure than female recipients.
    - ★ Referral mechanism partially explains these differences.
- Main contributions:
  - ▶ Contributes to the existing literature on the effects of early childhood interventions later in school outcomes.
  - ▶ Provides evidence of the returns of comprehensive support systems
  - ▶ Contributes to the limited evidence of Chile Crece Contigo.

# Data

- Combines three sources of data: Ministry of Education, Quality of Education Agency and Ministry of Social Development.
- Seven cohorts of first-time 4th graders (2012-2018) in public or private subsidized schools
- Born between July 2001 and June 2009.



► Descriptive Statistics

# Empirical Strategy: Measuring Exposure to ChCC

- Leverage institutional implementation: (i) gradual roll-out of the policy and (ii) only children younger than 60 months eligible.
- Child  $i$  who lived in municipality  $m$  born in period  $t$  had  $A_{imt}$  when ChCC was implemented

$$A_{imt} = t - E_{mt}$$

- Level of exposure  $D_{imt}$

$$D_{imt} = \begin{cases} 1 & \text{if } A_{imt} < -8 \\ (67 - A_{imt})/67 & \text{if } -8 < A_{imt} < 60 \\ 0 & \text{if } A_{imt} \geq 60 \end{cases} \quad (1)$$

► Exposure Distribution



# Empirical Model: Cumulative Exposure Event-study design

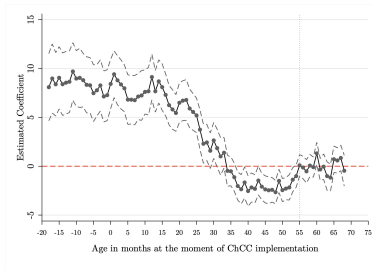
- Event-study design

$$Y_{imt} = \alpha + \sum_{a=-18[a \neq 60]}^{70} \pi_a \mathbb{1}[A_{imt} = a] + X_{imt}\beta + \gamma_m + \tau_r t + \delta_s + \epsilon_{imt} \quad (2)$$

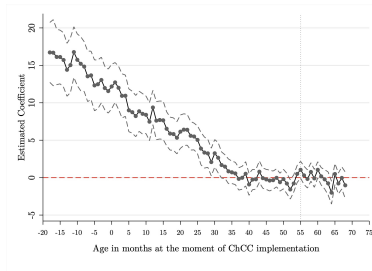
Where:

- ▶  $Y_{icst}$ : outcome (annual attendance, language and mathematics test scores)
  - ▶  $X_{ics}$ : vector of control variables: maternal schooling, and gender and age of the student (in months).
  - ▶  $\gamma_m$ : municipality time-invariant effects.
  - ▶  $\delta_s$ : school time-invariant effects.
  - ▶  $\theta_r \times t$ : regional month-of-birth trends
  - ▶  $\epsilon_{imst}$ : error term clustered at municipality level.
- This models allow me to estimate difference only “from above” (Hoyne et al. (2016), Bailey et al. (2021), Bailey et al. (2020))

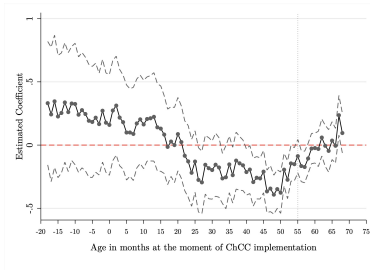
# Event Study Parameters, full sample



(a) Mathematics

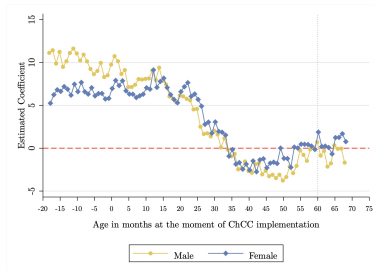


(b) Language

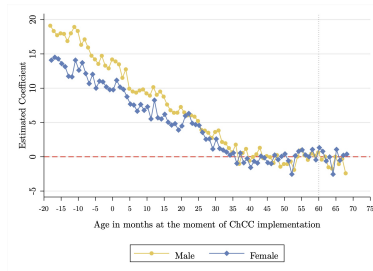


(c) Annual Attendance

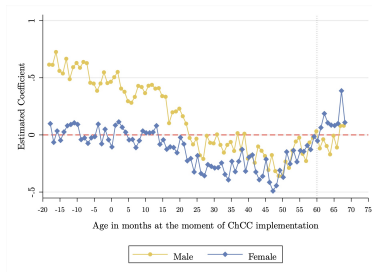
# Event Study Parameters, by student gender



(a) Mathematics



(b) Language



(c) Annual Attendance

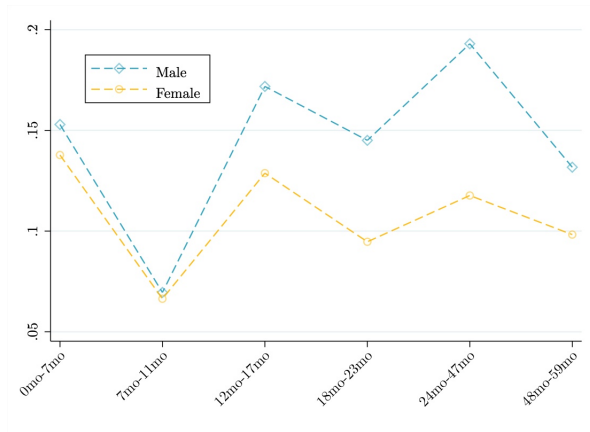
## Gender Differences (cont)

- Previous research have shown higher returns for girls compared to boys (Conti et al., 2019; García et al., 2018; Magnuson et al., 2016; Heckman et al., 2013).
- If  $\hat{\pi}_a^j = Pr(N = 1|j) \times \Delta^j$  for  $j = M, F$
- Then, from the previous results,

$$\begin{aligned}\hat{\pi}_a^M - \hat{\pi}_a^F &= Pr(N = 1|M)\Delta^M - Pr(N = 1|F)\Delta^F > 0 \\ \underbrace{\frac{Pr(N = 1|M)}{Pr(N = 1|F)}}_{>1} \frac{\Delta^M}{\Delta^F} &> 1\end{aligned}$$

- If  $Pr(N = 1|M) > Pr(N = 1|F)$  it is possible have higher returns for female recipients but exposure returns higher for male recipients.

# Ratio of Tested Children Referred to Stimulation Services, by gender and age



Note: Author's calculation based on the 2018 Monthly Statistical Reports from the Chilean Department of Health Statistics and Indicators (DEIS). Each point represents the proportion of children that a psycho-motor screening test was applied and receive a referral to assist to stimulation services. This can be calculated based on the sum of service code 06902601, 06902602, and 06902603 divided service code 02021740 for each age and gender group.

# Conclusion

- Children exposed since the prenatal period have returns of  $0.25\sigma$  in mathematics and  $0.3\sigma$  in language.
- Returns concentrate in the first 36 months of age.
- Male recipients have 50% higher returns in mathematics and 30% in language when exposed before 12 months of age.
- Intensity of treatment is higher for male recipients (need-based services).

Thank you

## Bio-psychosocial evaluation applied during prenatal care stage

[illegible]



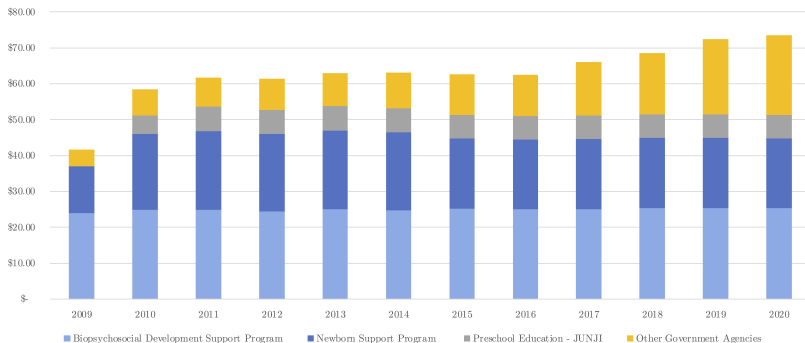
# Services Provided by ChCC

Three main components:

- Newborn Support Program (PARN)
  - ▶ Materials (trousseau) and educational workshops about newborn care and basics of early parenting and respectful parenting
  - ▶ Annual Budget (2019): \$19.61M USD (107 USD per beneficiary)
- Biopsychosocial Development Support Program (PADB)
  - ▶ Main program in ChCC (“Programa Eje”)
  - ▶ Detection Instruments + Interventions + Nutrition
  - ▶ Annual Budget (2019): 25.36M USD (46 USD per beneficiary)
- Mental Health Support Program (PASMI)
  - ▶ Target population: 5-9 years old
  - ▶ Screening services and interventions to detect and support the alterations in the socio-emotional development

▶ Background

# Chile Crece Contigo Annual Budget



Note: Figure based on the annual budget report by the Chilean Budget Office. The amount are in US dollars of October 2021.

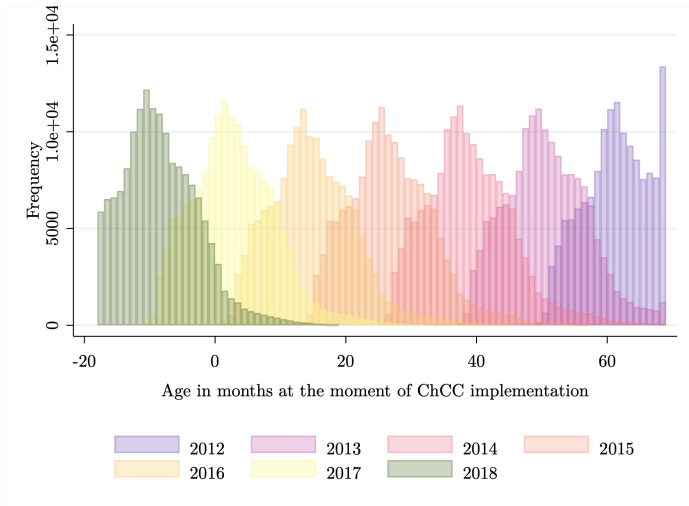
► Background

## Descriptive Statistics

|                            | Full-Sample |          | Low-SES Sample |          | Fully Exposed |          | Not Exposed |          |
|----------------------------|-------------|----------|----------------|----------|---------------|----------|-------------|----------|
|                            | Mean        | SD       | Mean           | SD       | Mean          | SD       | Mean        | SD       |
| Outcomes                   |             |          |                |          |               |          |             |          |
| Attendance                 | 93.20       | ( 5.89)  | 92.91          | ( 6.21)  | 93.53         | ( 5.70)  | 93.03       | ( 6.19)  |
| Language                   | 266.04      | ( 51.09) | 258.98         | ( 50.78) | 270.50        | ( 52.44) | 263.45      | ( 51.89) |
| Mathematics                | 260.76      | ( 48.10) | 253.71         | ( 48.02) | 261.99        | ( 46.83) | 259.07      | ( 50.07) |
| Background Characteristics |             |          |                |          |               |          |             |          |
| Gender (Male=1)            | 0.50        | ( 0.50)  | 0.49           | ( 0.50)  | 0.48          | ( 0.50)  | 0.52        | ( 0.50)  |
| Attended PreK              | 0.88        | ( 0.33)  | 0.85           | ( 0.35)  | 0.90          | ( 0.30)  | 0.79        | ( 0.41)  |
| Attended Kindergarten      | 0.95        | ( 0.22)  | 0.93           | ( 0.25)  | 0.94          | ( 0.24)  | 0.89        | ( 0.31)  |
| Mother MS Dropout          | 0.09        | ( 0.28)  | 0.13           | ( 0.33)  | 0.06          | ( 0.23)  | 0.13        | ( 0.34)  |
| Mother HS Dropout          | 0.23        | ( 0.42)  | 0.30           | ( 0.46)  | 0.19          | ( 0.39)  | 0.28        | ( 0.45)  |
| Mother HS Graduate         | 0.38        | ( 0.49)  | 0.38           | ( 0.49)  | 0.42          | ( 0.49)  | 0.36        | ( 0.48)  |
| Mother College Dropout     | 0.11        | ( 0.31)  | 0.08           | ( 0.28)  | 0.09          | ( 0.29)  | 0.07        | ( 0.25)  |
| Mother Two-Year College    | 0.11        | ( 0.32)  | 0.07           | ( 0.26)  | 0.14          | ( 0.35)  | 0.10        | ( 0.30)  |
| Mother Four-Year College   | 0.08        | ( 0.28)  | 0.03           | ( 0.17)  | 0.10          | ( 0.30)  | 0.06        | ( 0.23)  |
| Low SES                    | 0.51        | ( 0.50)  | 1.00           | ( 0.00)  | 0.51          | ( 0.50)  | 0.50        | ( 0.50)  |
| Age                        | 9.58        | ( 0.58)  | 9.60           | ( 0.60)  | 9.51          | ( 0.51)  | 9.82        | ( 0.66)  |
| N                          | 1,148,296   |          | 584,716        |          | 215,796       |          | 124,089     |          |

▶ Data

# Cumulative Distribution of Exposure to ChCC IU-5, by year of birth



# Empirical Model: Linear Cumulative Exposure

- Based on Equation 1 it is possible to estimate the linear cumulative exposure effect to ChCC.
- Linear Cumulative exposure model

$$Y_{imt} = \alpha + \varphi D_{imt} + X_{imt}\beta + \gamma_m + \tau_r t + \delta_s + \epsilon_{imt} \quad (3)$$

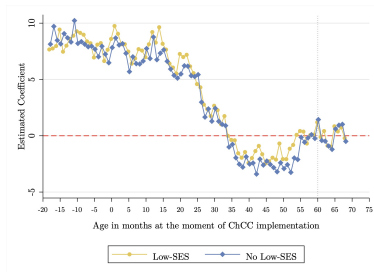
Where:

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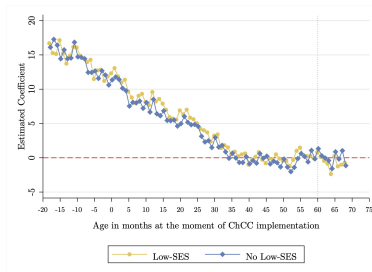
# Estimation Results of Cumulative Exposure to ChCC on Educational Achievement

|                              | Attendance           | Language             | Mathematics          |
|------------------------------|----------------------|----------------------|----------------------|
| Exposure to ChCC IU-5        | 1.069<br>(0.179)***  | 8.457<br>(1.557)***  | 12.708<br>(1.105)*** |
| Low SES                      | -0.729<br>(0.030)*** | -4.527<br>(0.116)*** | -3.335<br>(0.107)*** |
| Gender (Male=1)              | -0.071<br>(0.011)*** | -8.846<br>(0.163)*** | 3.758<br>(0.151)***  |
| Mother HS Dropout            | -0.238<br>(0.028)*** | 3.754<br>(0.208)***  | 3.946<br>(0.186)***  |
| Mother HS Graduate           | 0.294<br>(0.032)***  | 10.780<br>(0.231)*** | 10.914<br>(0.248)*** |
| Mother College Dropout       | -0.312<br>(0.041)*** | 14.234<br>(0.292)*** | 13.173<br>(0.280)*** |
| Mother Two-Year College      | 0.264<br>(0.037)***  | 14.972<br>(0.313)*** | 14.123<br>(0.322)*** |
| Mother Four-Year College     | 0.140<br>(0.040)***  | 22.192<br>(0.407)*** | 20.359<br>(0.403)*** |
| R2                           | 0.15                 | 0.16                 | 0.23                 |
| N                            | 1,148,296            | 1,148,296            | 1,148,296            |
| Municipality Dummy           | Yes                  | Yes                  | Yes                  |
| Age Polynomial               | Yes                  | Yes                  | Yes                  |
| School Dummy                 | Yes                  | Yes                  | Yes                  |
| Region Month of Birth Trends | Yes                  | Yes                  | Yes                  |

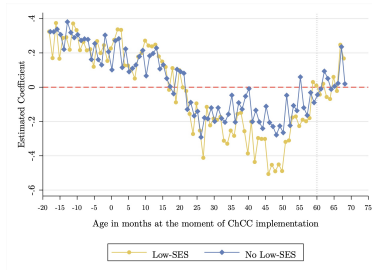
# Event Study Parameters, by low-SES classification



(a) Mathematics



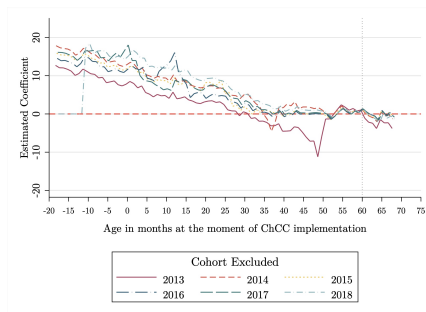
(b) Language



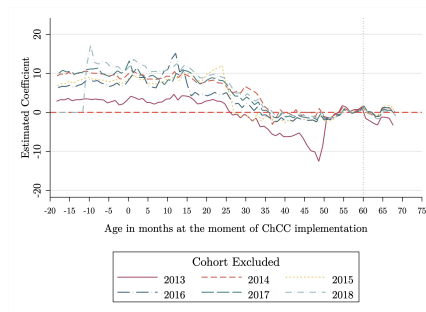
(c) Annual Attendance

# Sensitivity Analysis: Cohort Exclusion

Figure: Event-study estimations excluding cohorts



(a) Language

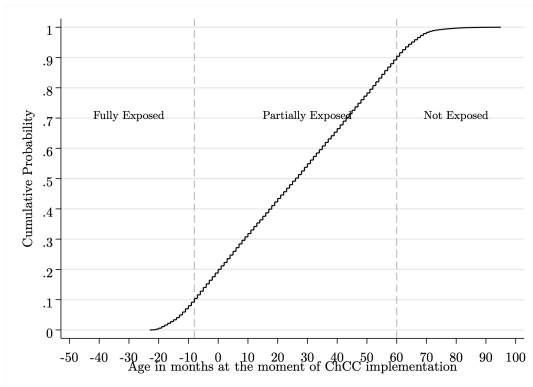


(b) Mathematics

► Event-Study Results



# Distribution of Age at Implementation of ChCC



Note: The figure shows the distribution of the age of the child at the moment when ChCC was implemented  $A_{mt}$ . The implementation dates used to calculate this are the ones reported by from the Ministry of Social Development.

- J Shonkoff and D Philips. *From neurons to neighborhoods: The science of early childhood development*. 2000. ISBN 0309069882. doi: 10.1097/00004583-200205000-00022.  
w28268.pdf.
- Hilary Hoynes, Diane Whitmore Schanzenbach, and Douglas Almond. Long-Run impacts of childhood access to the safety net, apr 2016. ISSN 00028282.
- Martha Bailey, Hilary Hoynes, Maya Rossin-Slater, and Reed Walker. Is the Social Safety Net a Long-Term Investment? Large-Scale Evidence from the Food Stamps Program. *Review of Economic Studies*, 2021.
- Martha J Bailey, Hilary W Hoynes, Maya Rossin-Slater, and Reed Walker. Is the Social Safety Net a Long-Term Investment? Large-Scale Evidence from the Food Stamps Program. 2020. URL <https://www.acf.hhs.gov/ofa/resource/tanf-financial-data-fy-2016>.
- Gabriella Conti, Stavros Poupakis, Malte Sandner, and Sören Kliem. The Effects of Home Visiting on Mother-Child Interactions : Evidence from Dynamic Micro-Level Data. *Hceo Working Paper Series*, 2019.
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- Katherine A. Magnuson, Robert Kelchen, Greg J. Duncan, Holly S. Schindler, Hilary Shager, and Hirokazu Yoshikawa. Do the effects of early childhood education programs differ by gender? A meta-analysis. *Early Childhood Research Quarterly*, 36: 521–536, 2016. ISSN 08852006. doi: 10.1016/j.ecresq.2015.12.021. URL <http://dx.doi.org/10.1016/j.ecresq.2015.12.021>.
- James Heckman, Rodrigo Pinto, and Peter Savelyev. Understanding the mechanisms through which an influential early childhood program boosted adult outcomes. *American Economic Review*, 103(6):2052–2086, 2013. ISSN 00028282. doi: 10.1257/aer.103.6.2052.
- Damian Clarke, Gustavo Cortés Méndez, and Diego Vergara Sepúlveda. Growing together: assessing equity and efficiency in a prenatal health program. *Journal of Population Economics*, 2020. ISSN 14321475. doi: 10.1007/s00148-019-00761-6.