

# **Heterogeneity in Childhood Residential Mobility Trajectories: Implications for Adult Preventative Healthcare Use**

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# Introduction

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- **Preventative healthcare** is key to well-being, reducing costs, and improving lifespan.
- **Life-course factors** (e.g., poverty, parental separation, **residential instability**) shape preventative health behaviours (Abel & Frohlich, 2012; Kuh & Ben-Shlomo, 2004).
- Moves **disrupt relationships with healthcare providers**, reducing access to care (Busacker & Kasehagen, 2012; Hutchings et al., 2016; Nathan et al., 2022).

# Introduction

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- Limitations in research:
  - Mobility is treated as a uniform experience
  - **Long-term effects** on preventative healthcare utilization remain underexplored.
- Research questions:
  - RQ1: How do residential mobility trajectories in childhood vary among individuals?
  - RQ2: Do residential mobility trajectories in childhood influence patterns of preventative healthcare use in adulthood?

# Theoretical background

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## Frequency

- Frequent moves disrupt continuity of care and relationships with providers (Hutchings et al., 2016).
- Harder to maintain regular check-ups or consistent health monitoring (Busacker & Kasehagen, 2012).
- Repeated disruptions create long-term instability, shaping how individuals engage with healthcare in adulthood (Bures, 2003; Mollborn et al., 2018; Vogel et al., 2017).

## Timing of moves

- Moves during adolescence may be especially disruptive.
  - Adolescence = stage of identity formation and growing autonomy (Li et al., 2019; Steinberg & Morris, 2001).
- Moving can break social networks and peer support → higher risks of mental health challenges and school dropout (Li et al., 2019; Tønnessen et al., 2016).

# Theoretical background

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## **Distance of moves**

- Long-distance moves often cause major disruptions: families must re-register with new providers and adapt to unfamiliar systems.
- These transitions can delay or reduce engagement with preventive care (Fowler et al., 1993; Jatrana et al., 2013).

## **Socioeconomic context of moves**

- Moving into disadvantaged areas limits access to quality care and health-promoting resources (Kirby & Kaneda, 2005).
- Lower local access → lower preventive care use.
- Early exposure to disadvantage can shape health literacy and ability to navigate healthcare systems in adulthood.

# Swedish context

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## Primary care organisation (pre-2010)

- Locations planned by counties based on population health needs
- No provider choice → patients assigned to nearest centre

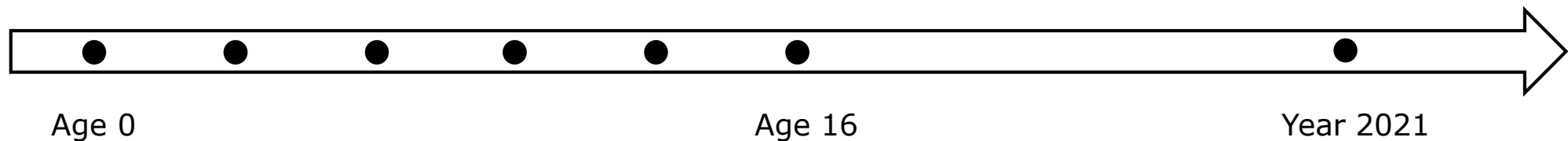
## Implications of moving

- Switching primary care centre
- Disrupted continuity of care, especially for children

# Data

## Swedish register data

- 1990-1993 cohorts
  - $N = 417,850$



- Mobility → change in DeSO

# Variables

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Usual preventive care measures:

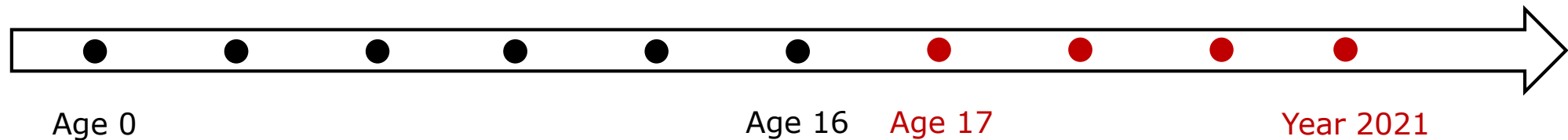
- Primary prevention (e.g. vaccinations)
  - Secondary prevention (e.g. cancer screenings)
  - Tertiary prevention
- 
- Another important aspect is whether healthcare is accessed in a *timely manner* to prevent serious health issues.



# Variables

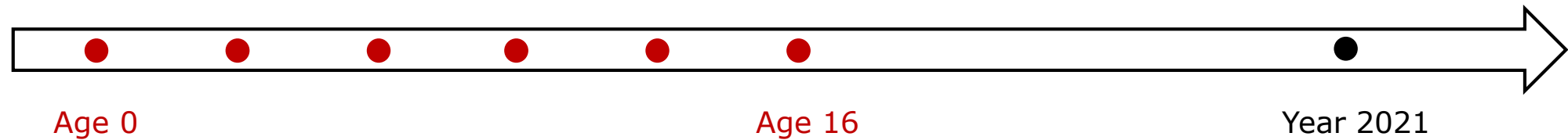
## Preventative healthcare:

- potentially avoidable hospitalization
  - Measures both chronic and acute conditions
  - Binary outcome



# Variables

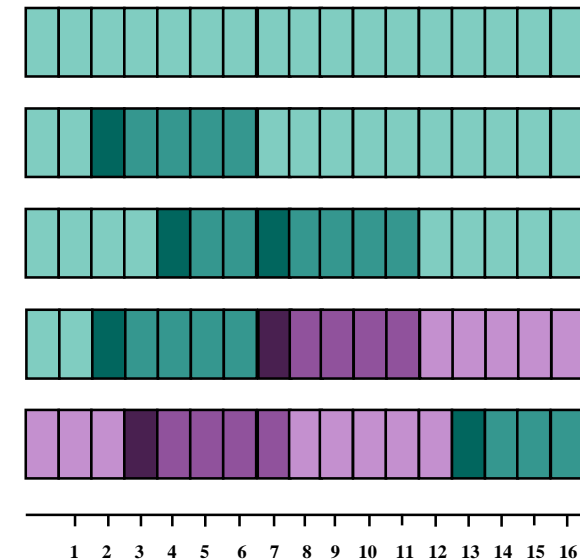
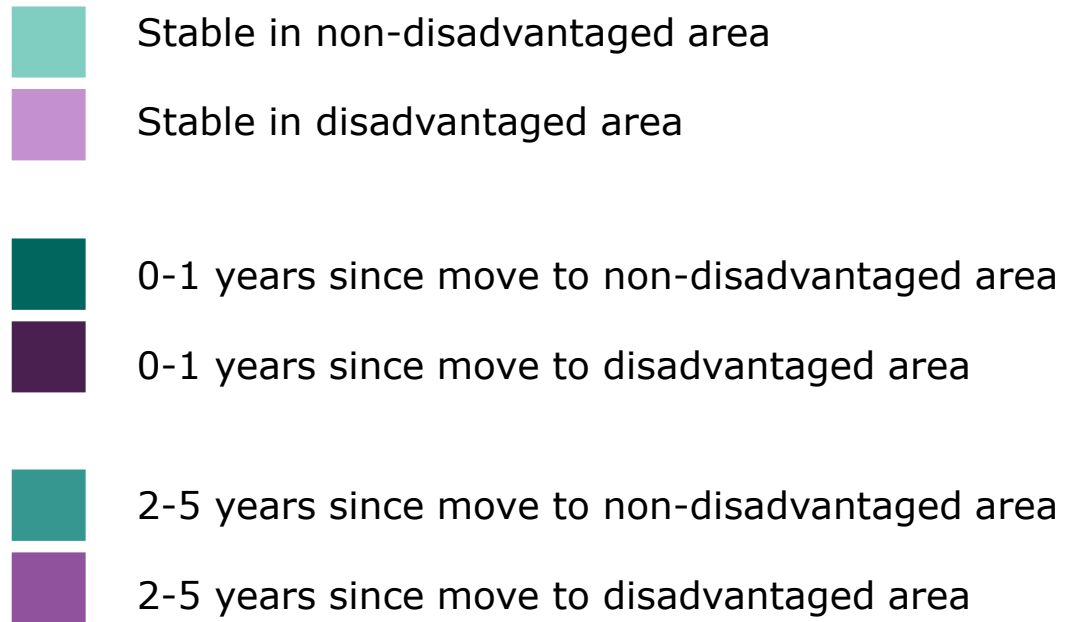
## Residential mobility:



- Stability/move → time since move
- Socio-spatial context of the area → deprivation index
  - individuals aged 25–64 years:
    - low educational status;
    - low income, defined as <50% of the median individual income;
    - unemployment;
    - receipt of social welfare.
  - Each indicator is standardised (converted to z-scores) and summed.
  - Top 20% → disadvantaged.

# Variables

## Residential mobility:



# Variables

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## **Controls:**

- gender
- parental migration background
- cohort
- parental education at the age of 5

# Method

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## **Step 1:**

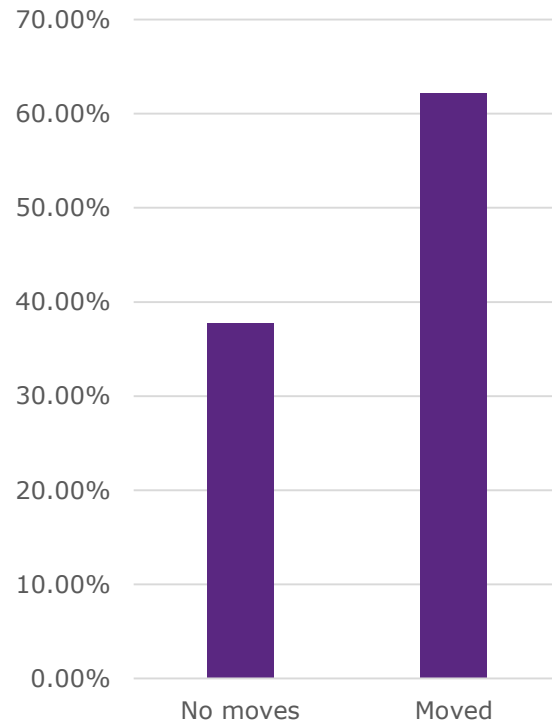
- Sequence analysis
  - Dynamic Hamming Distance (DHD) algorithm,
- Clustering
  - CLARA (clustering in large applications).

## **Step 2:**

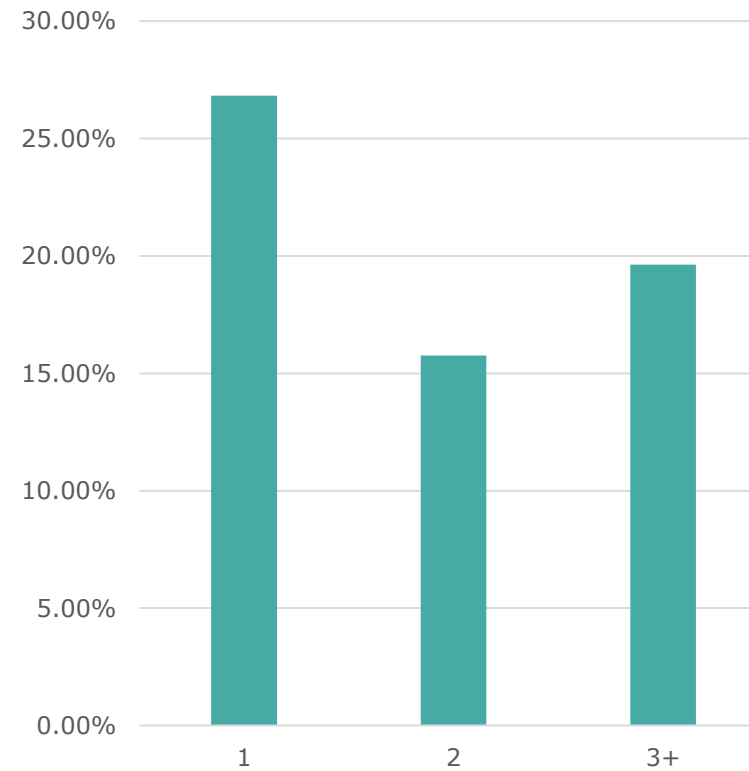
- Regression on key indicators—ever moved, frequency, age of move.
- Logistic regression to predict preventative healthcare use with typology.

# Results

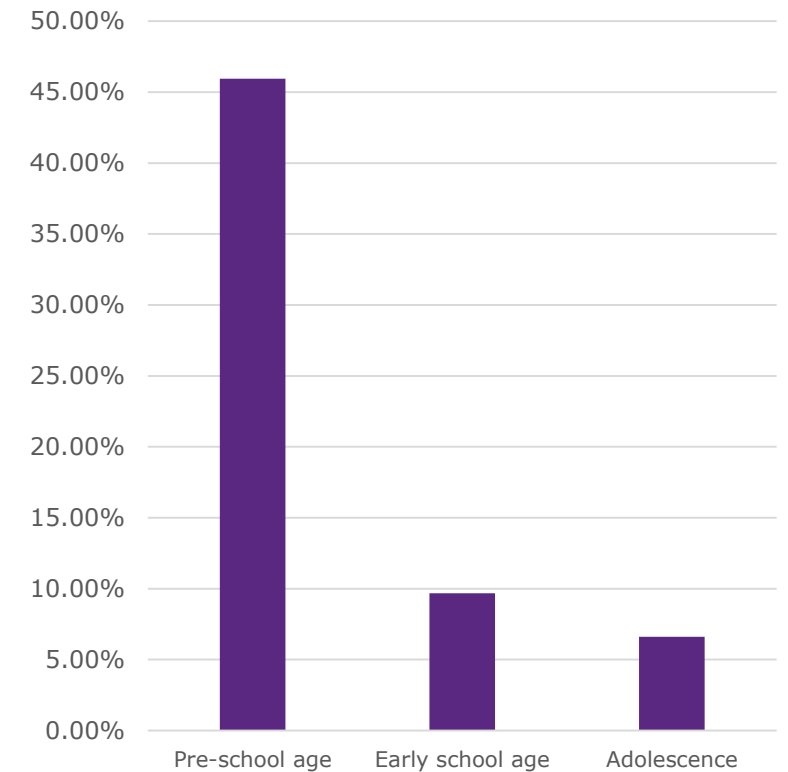
Ever moved in  
childhood



Number of moves in childhood

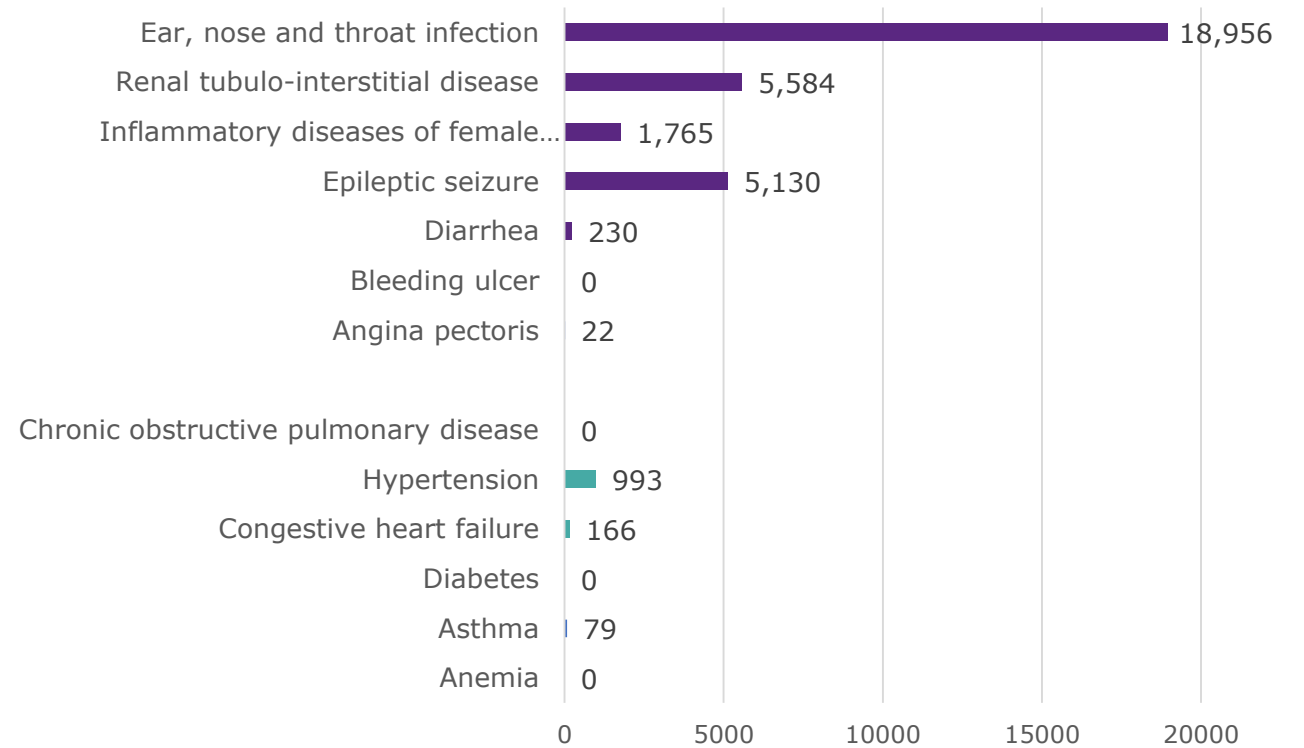
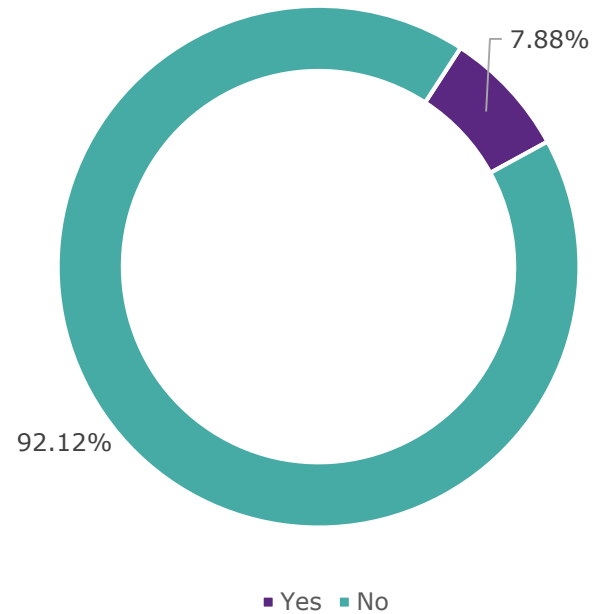


Age at first move in childhood



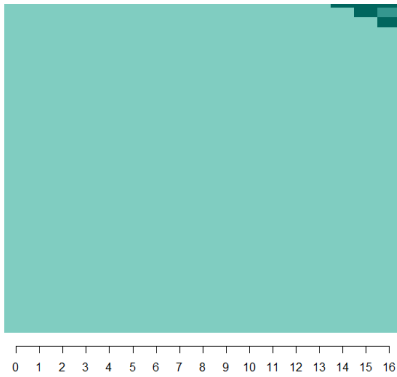
# Results

Potentially avoidable hospitalisations during adulthood

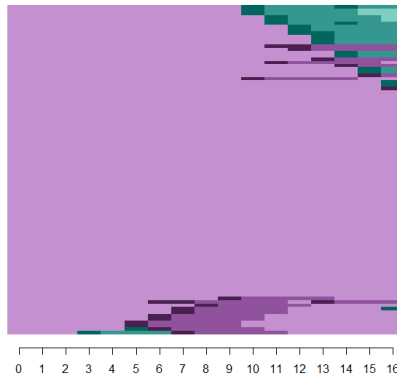


# Results

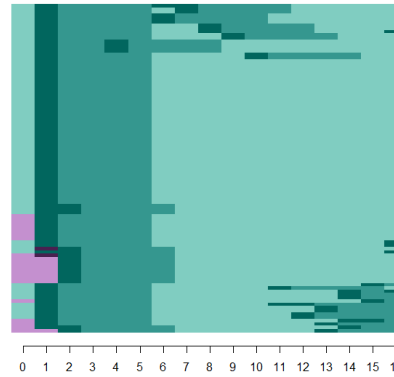
**1. Stable non-DI  
(N = 153,818)**



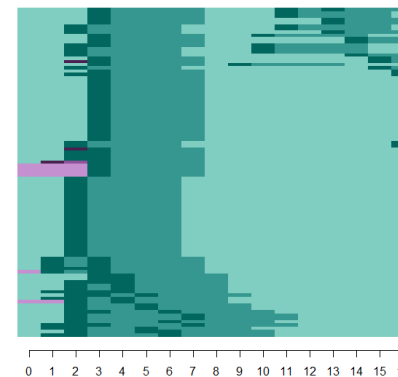
**2. Stable DI (N = 25,151)**



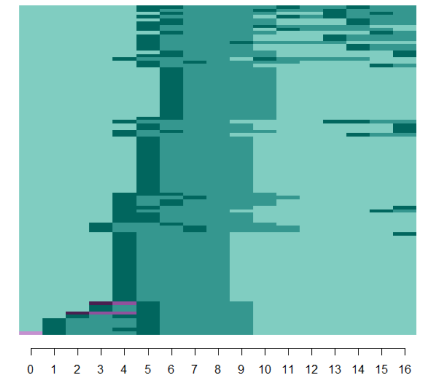
**3. Pre-school mobility (age 1)  
non-DI (N = 34,954)**



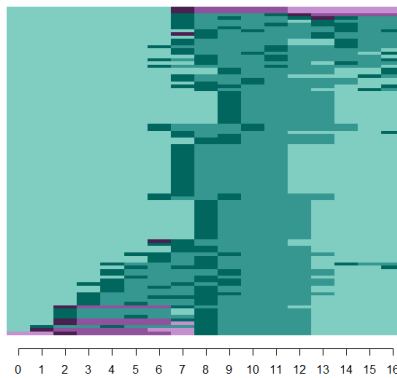
**4. Pre-school mobility (age 2-3)  
non-DI (N = 40,539)**



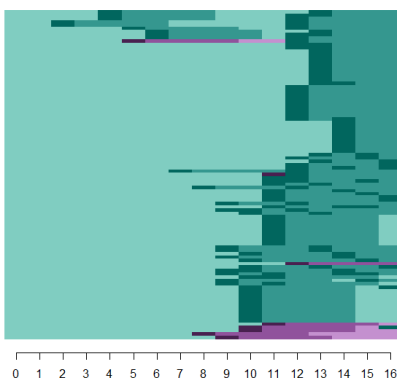
**5. Pre-school mobility (age 5-6)  
non-DI (N = 34,891)**



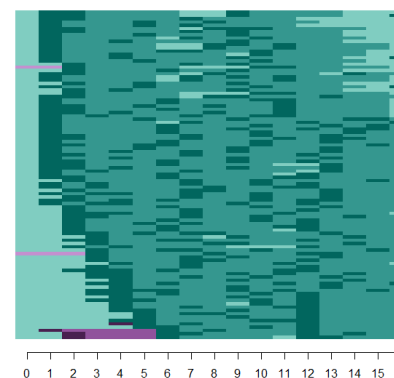
**6. Early-school mobility non-DI  
(N = 29,873)**



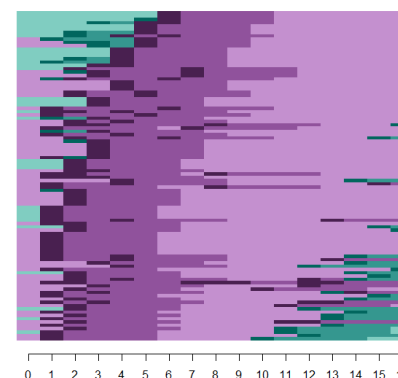
**7. Adolescence mobility non-DI  
(N = 32,384)**



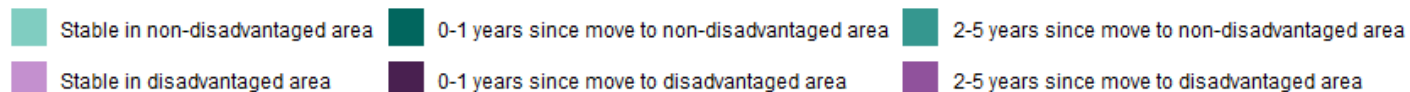
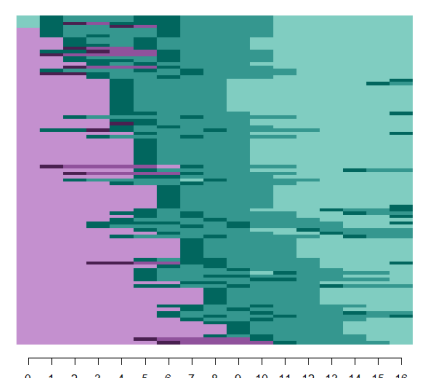
**8. Frequent mobility non-DI  
(N = 31,402)**



**9. Mobility DI (N = 19,002)**

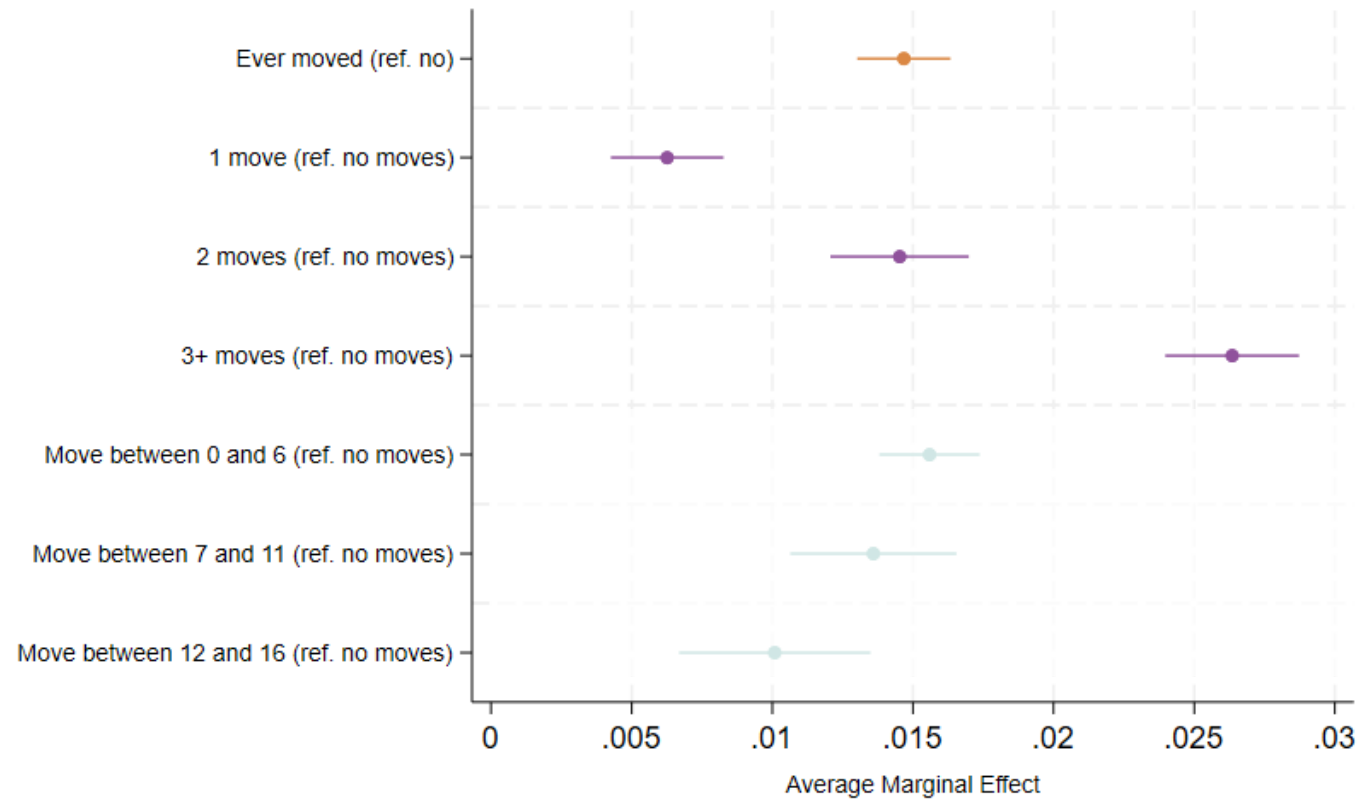


**10. Upward mobility  
(N = 15,836)**





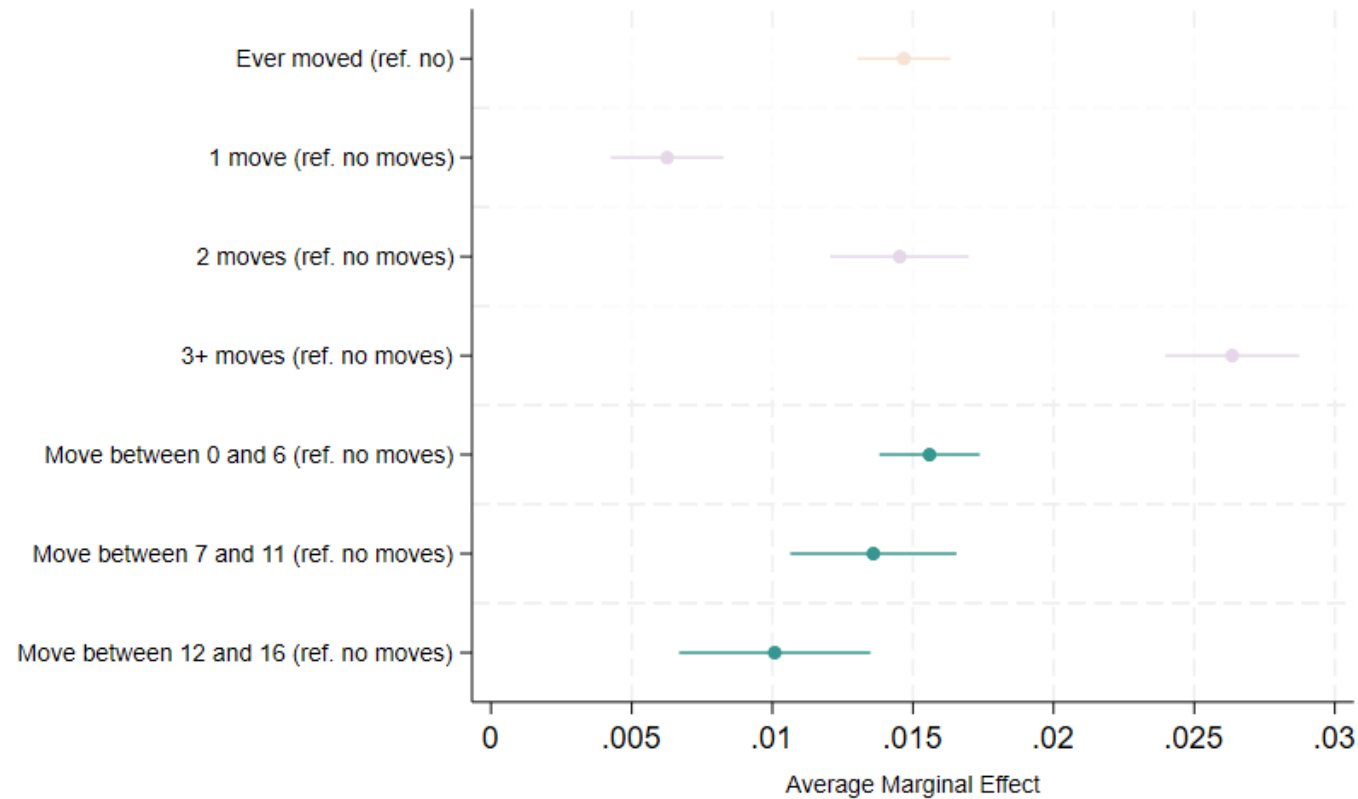
# Results



**Figure 1. Average marginal effects (AMEs) coefficients for basic indicators of mobility trajectories across logistic models predicting PAH**

*Notes: gender, parental migration background, parental education, cohorts are added as controls.*

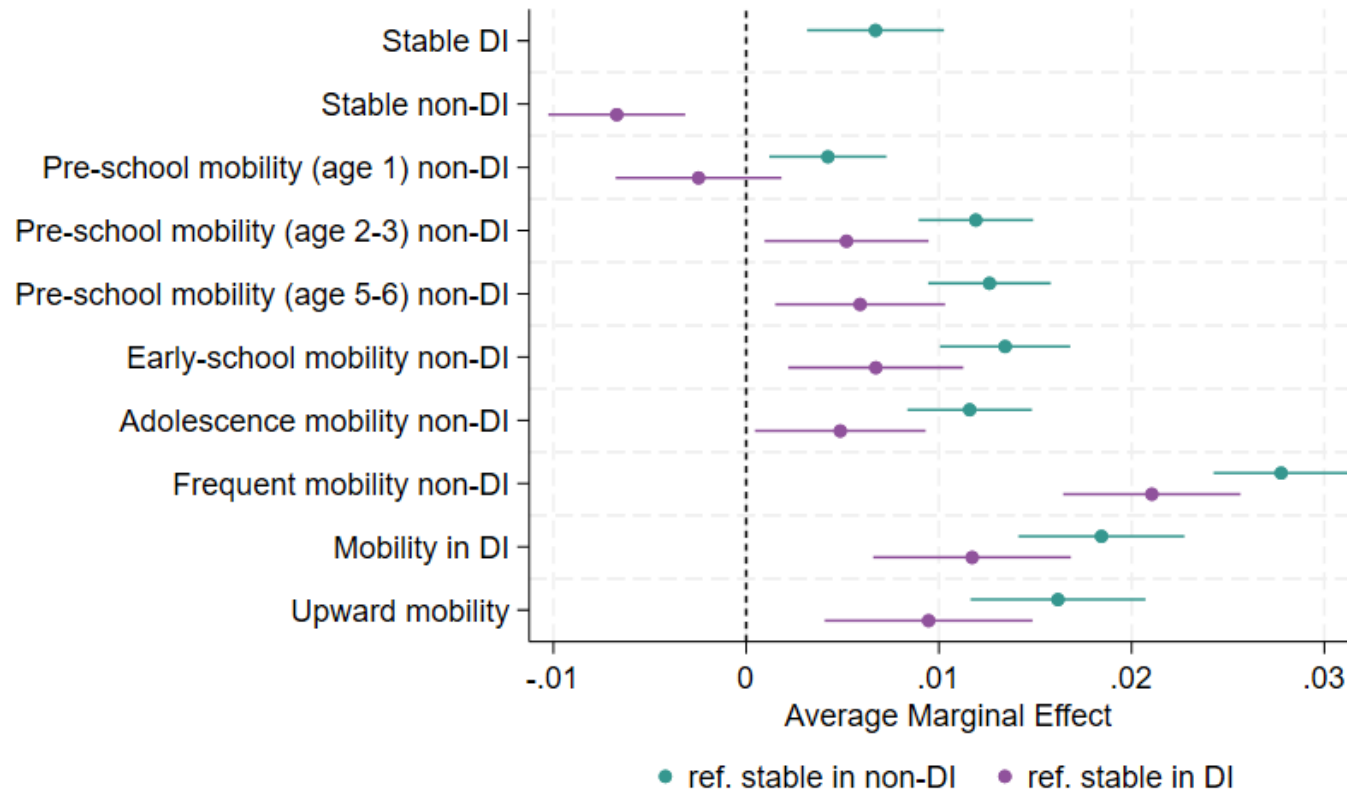
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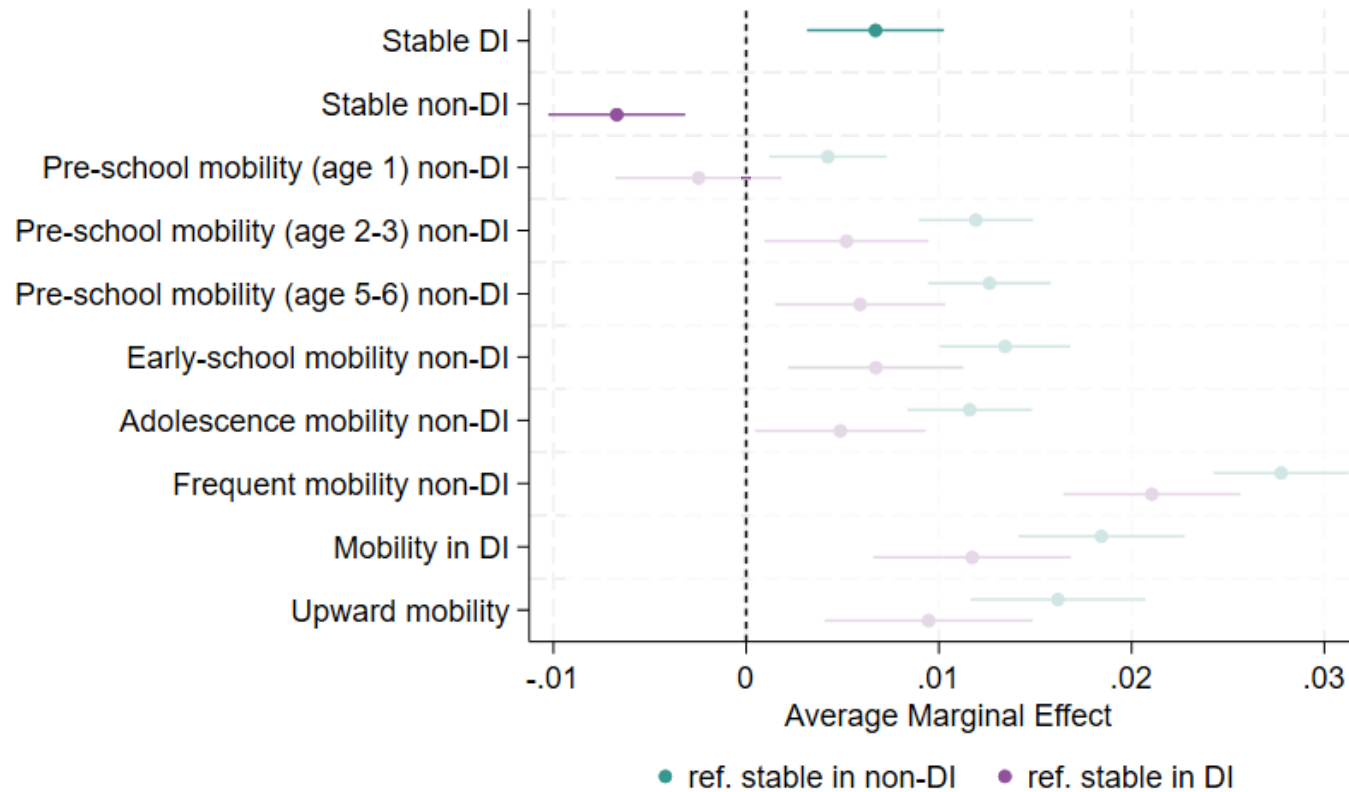
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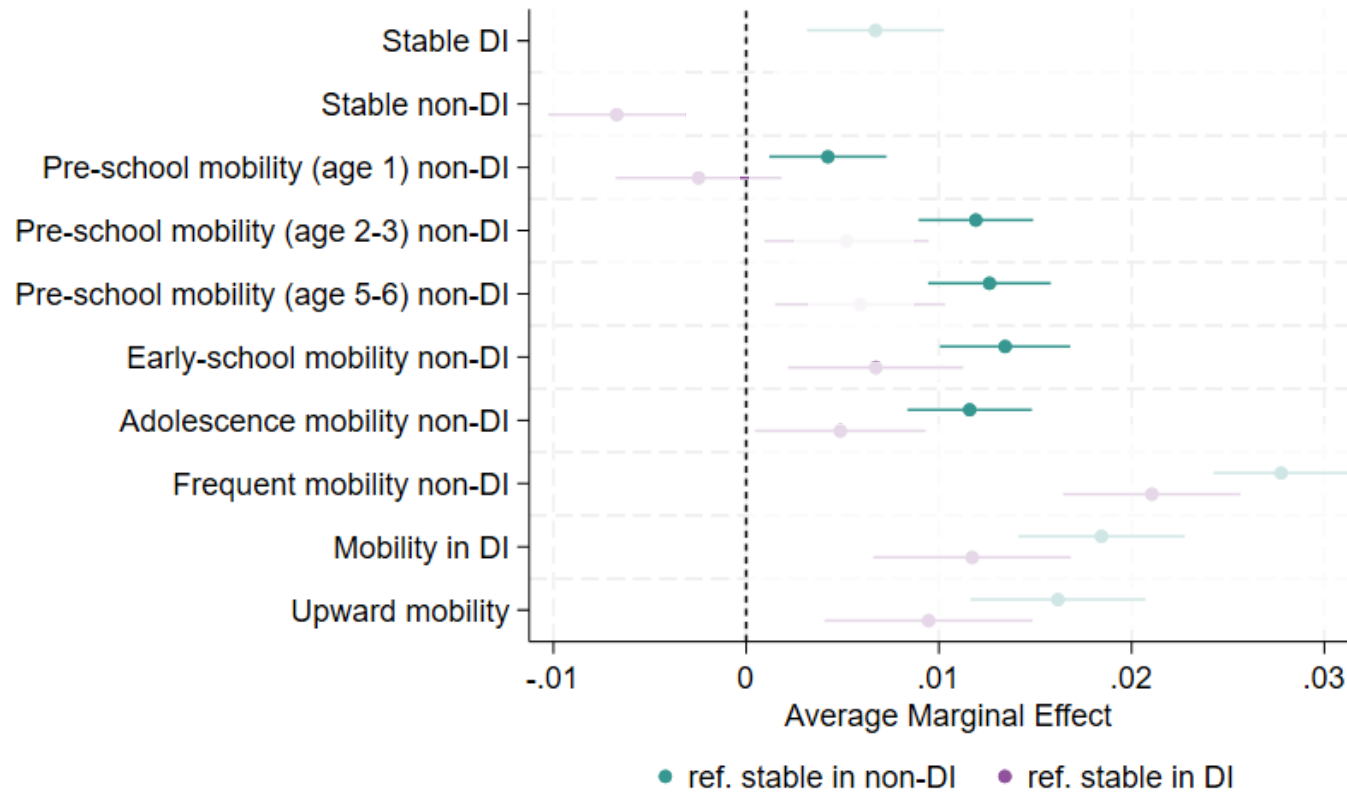
**Figure 2. Average marginal effects (AMEs) coefficients for residential mobility in childhood typologies across nested logistic models predicting PAH**  
*Notes: gender, parental migration background, parental education, cohorts are added as controls.*

# Results



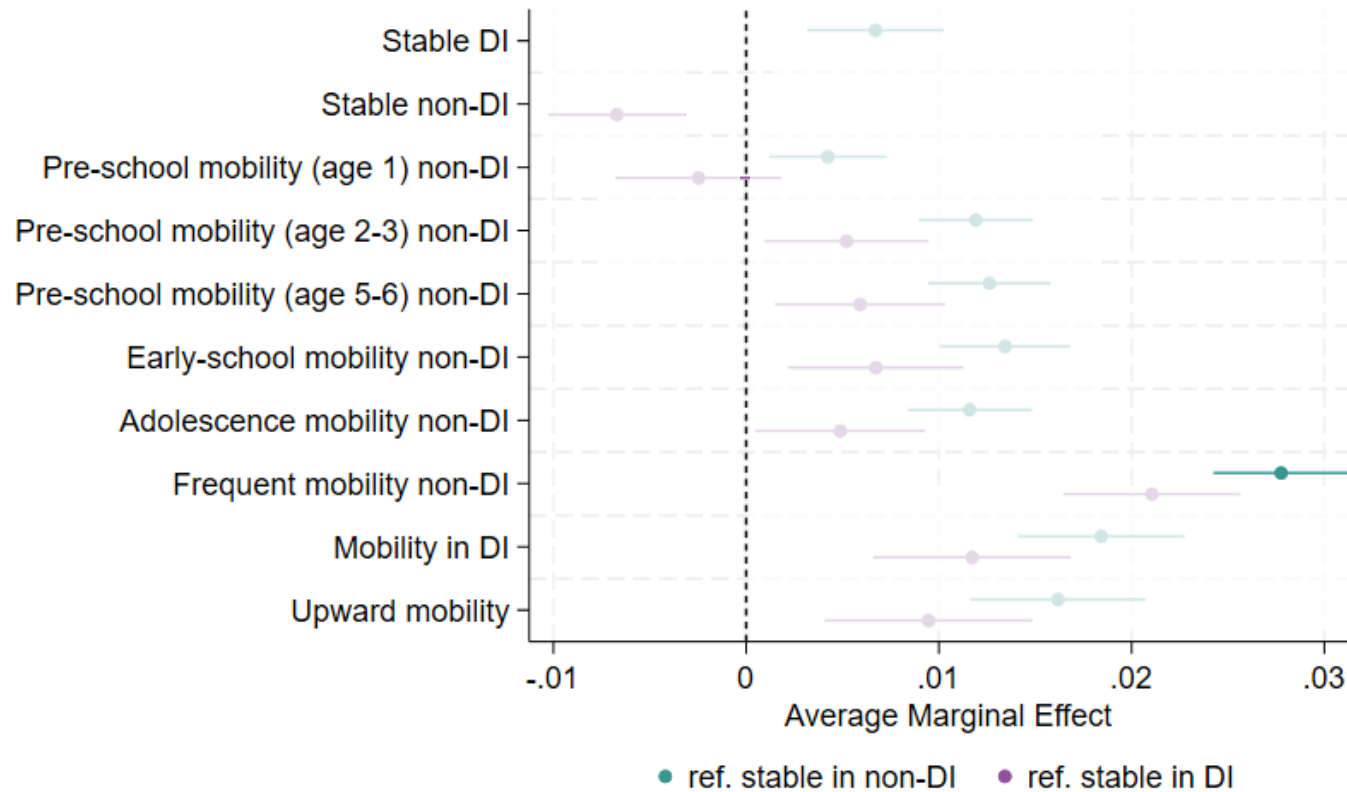
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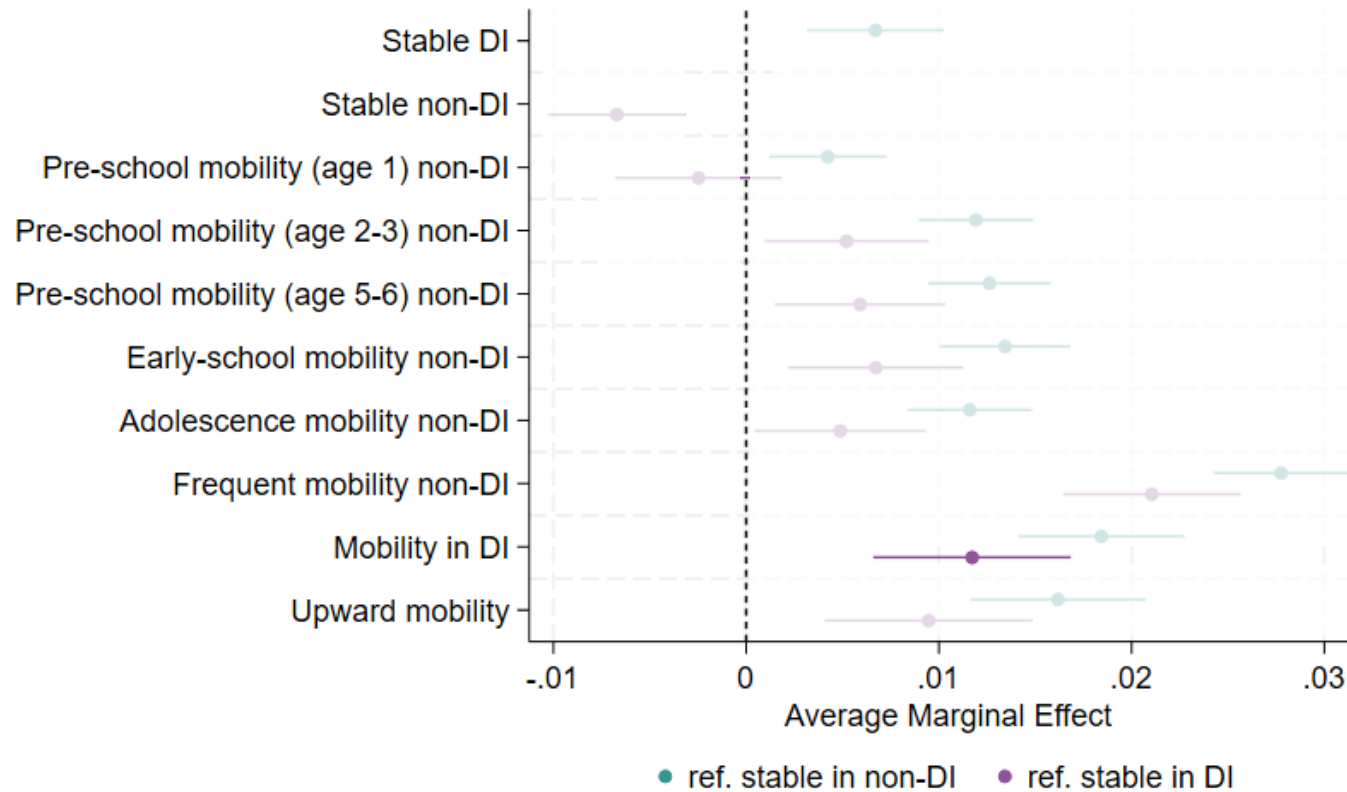
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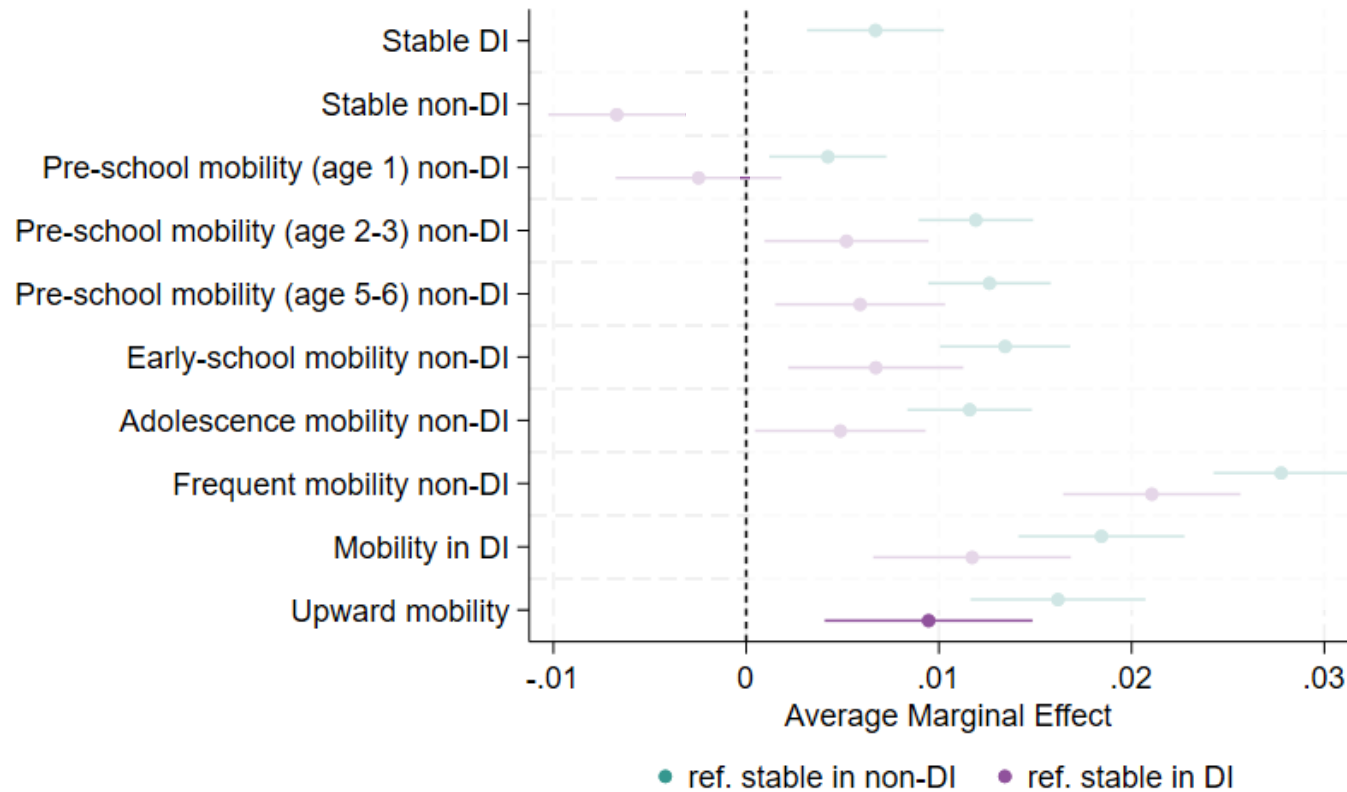
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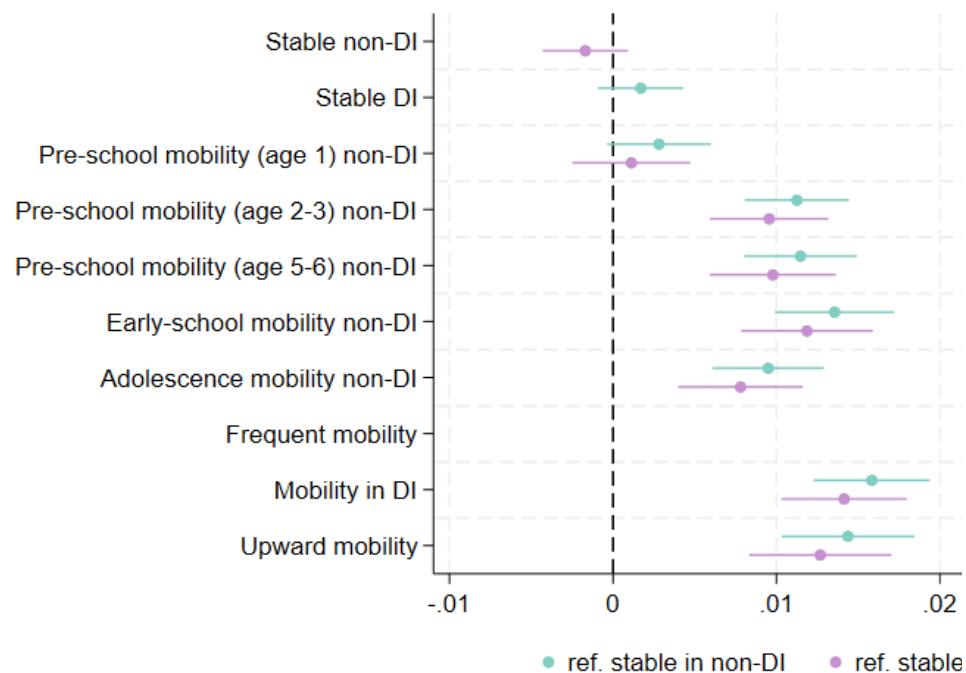
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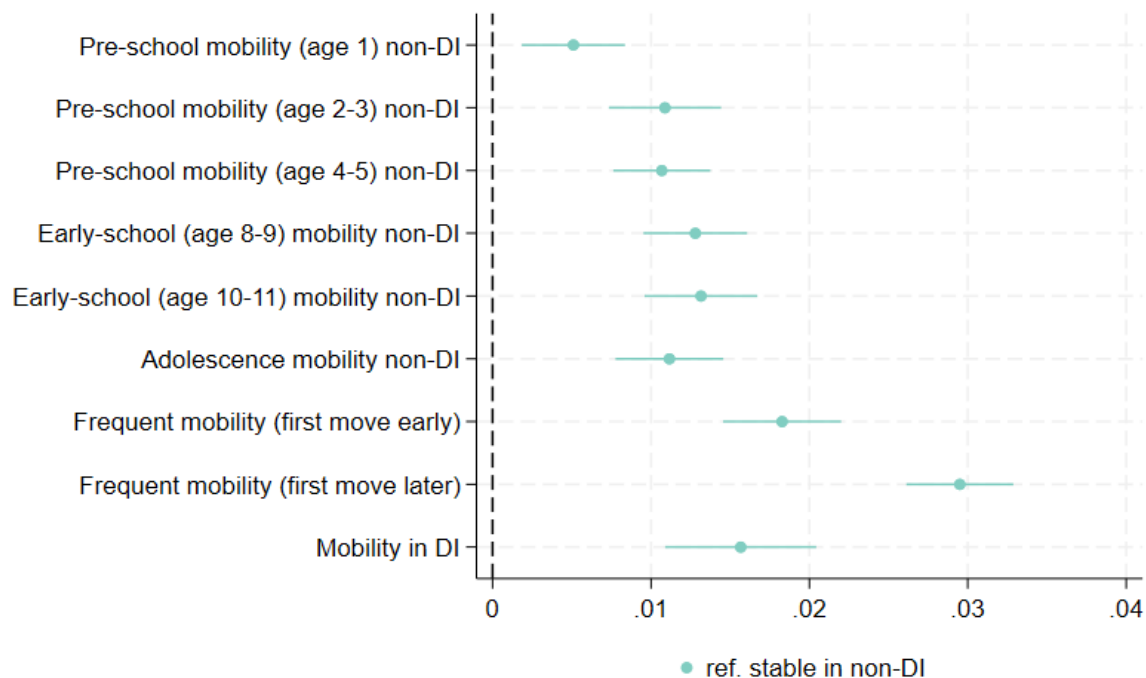


# Different thresholds



**Figure 3. Average marginal effects (AMEs) coefficients for residential mobility in childhood typologies across nested logistic models predicting PAH with 30% threshold**

*Notes: gender, parental migration background, parental education, cohorts are added as controls.*



**Figure 4. Average marginal effects (AMEs) coefficients for residential mobility in childhood typologies across nested logistic models predicting PAH with 10% threshold**

*Notes: gender, parental migration background, parental education, cohorts are added as controls.*

# Limitations

- PAH may capture hospitalisations that would have occurred regardless of primary care access or prevention efforts.
  - Future research should examine broader preventive measures (e.g., screenings, dental care, GP visits).
- Cohort still relatively young (early 30s) → chronic PAHs (e.g., diabetes, COPD) not yet prevalent.
  - Current results reflect acute care responsiveness more than chronic prevention.
  - Residential mobility might have more effect on chronic conditions as individuals age.

# Future steps

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- Childhood health
- Mobility in adulthood
- Basic mobility indicators vs trajectories

# Main takeaways

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- Childhood residential mobility is linked to lower engagement with preventative healthcare in adulthood.
  - Nature of moves is key:
    - Frequent movers
    - Moves in disadvantaged context
- Basic indicators vs. sequence analysis
- Magnitude of effects
  - Comparable to or greater than parental education

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**Thank you for your attention!**

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# LIFELONGMOVE

Understanding spatial mobility  
from early life into adulthood

**European Research Council**  
**Consolidator Grant (CoG)**

Ref: 101043981

Period: Jan 2023 – Dec 2027

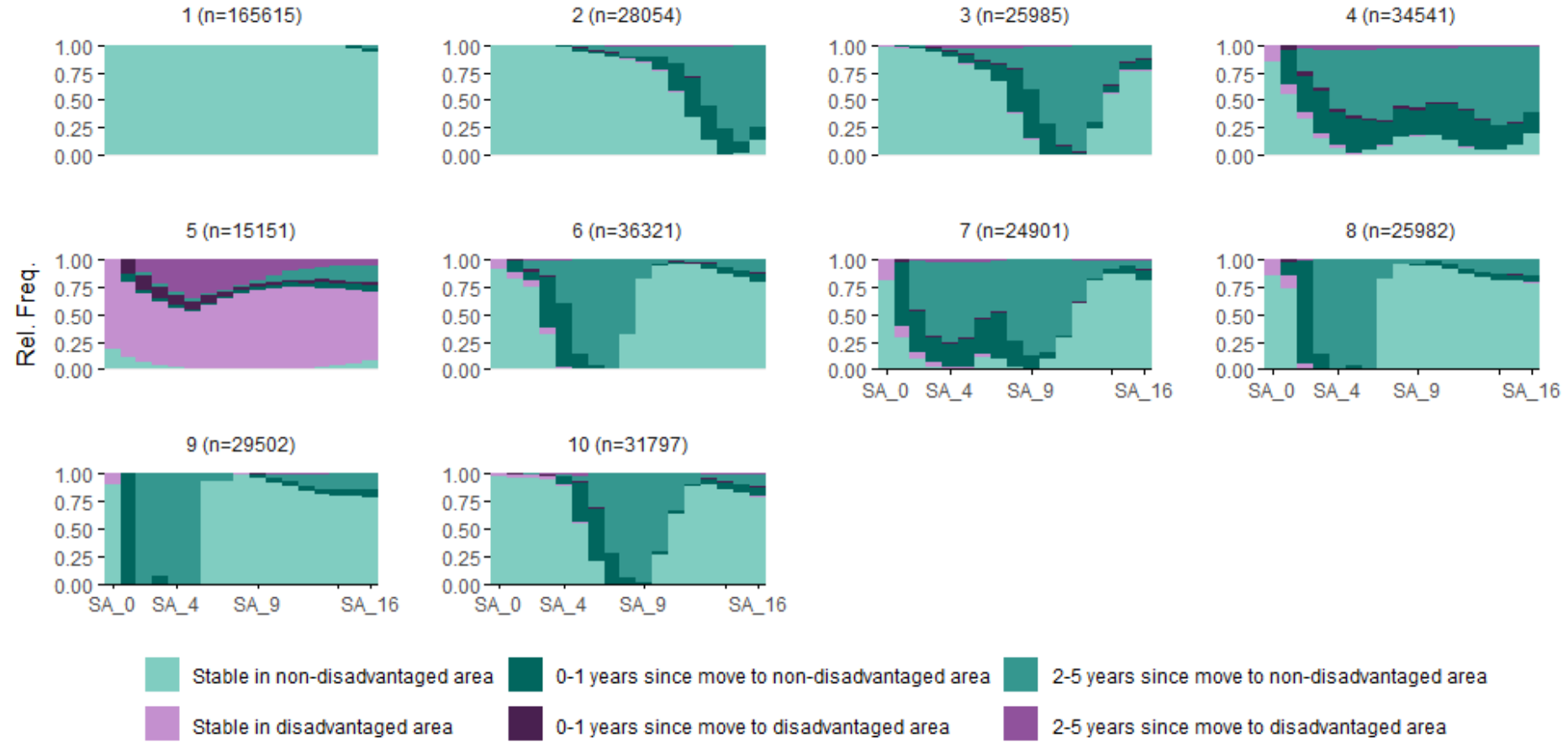


**European Research Council**  
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# Appendix

Condition	ICD-10 coding	N
<b>Chronic conditions</b>		
<b>Anemia</b>	D501, D508, D509	0
<b>Asthma</b>	J45, J46	79
<b>Diabetes</b>	E101-E108, E110-E118, E130-E138, E140-E148	0
<b>Congestive heart failure</b>	I50, I110, J81	166
<b>Hypertension</b>	I10, I119	993
<b>Chronic obstructive pulmonary disease</b>	J41, J42, J43, J44, J47 J20 if secondary diagnosis J41, J42, J43, J44 or J47	0
<b>Angina pectoris</b>	I20, I240, I248, I249	22
<b>Acute conditions</b>		
<b>Bleeding ulcer</b>	K250, K251, K252, K254, K255, K256, K260, K261, K262, K264, K265, K266, K270, K271, K272, K274, K275, K276, K280, K281, K282, K284, K285, K286	0
<b>Diarrhea</b>	E86, K522, K528, K529	230
<b>Epileptic seizure</b>	O15, G40, G41, R56	5,130
<b>Inflammatory diseases of female pelvic organs</b>	N70, N73, N74	1,765
<b>Renal tubulo-interstitial disease</b>	N390, N10, N11, N12, N136	5,584
<b>Ear, nose and throat infection</b>	H66, H67, J02, J03, J06, J312	18,956

# 10%





30%

