

WELFARE EFFECTS OF INCREASING  
TRANSFERS TO YOUNG ADULTS:  
THEORY AND EVIDENCE FROM FRANCE

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CREST

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# MOTIVATION - WHY SINGLE OUT YOUNG ADULTS?

## ▶ Financial fragility:

- Highest poverty rate (22% vs 17%) and deprivation rate
- Same pattern for other poverty indicators [Sav](#) [Pay](#)

## ▶ Inequality: [Evidence](#)

- Resources: Young adults income depends on parent's income
- Education choice: High social reproduction [Evidence](#)

## ▶ Little social assistance: Age category receiving the least social assistance, even if: [Level](#) [Ratio](#)

- Might help reducing financial fragility and inequalities
- Can shape education decision ([Fack and Grenet \(2015\)](#))

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## ▶ Financial fragility:

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## ▶ Inequality: (Evidence)

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## ▶ Little social assistance: Age category receiving the least social assistance, even if: (Level) (Ratio)

- Might help reducing financial fragility and inequalities
- Can shape education decision (Fack and Grenet (2015))

⇒ **Why so little assistance compare to other categories?  
Should it be increased?**

# MOTIVATION: CONCEPTUAL CHALLENGES

Shed light on the trade-off through **welfare analysis**: Two margins, **the benefits VS the costs**:

**(1) What is the social utility:**

- + Young adults are poor, so should value it a lot
- Interactions between parents and government transfers can decrease welfare effects
- ? weights that society put on young adults utility

**(2) What is the fiscal cost for the government:**

- Upfront cost of the policy
- ? Labor supply responses
- ? Education decision responses

# RESEARCH QUESTIONS

- ① What is the **welfare effect** of increasing transfers to young adults financed by older individuals?
- ② Should those transfers be tagged?

# OUTLINE

## 1. CONCEPTUAL FRAMEWORK

## 2. SOCIAL UTILITY OF TRANSFERS

2.1. MPC Estimation

2.2. Transfer derivatives estimation

2.3. Wrap-Up

2.4. Heterogeneity

## 3. FISCAL COST

## 4. CONCLUSION

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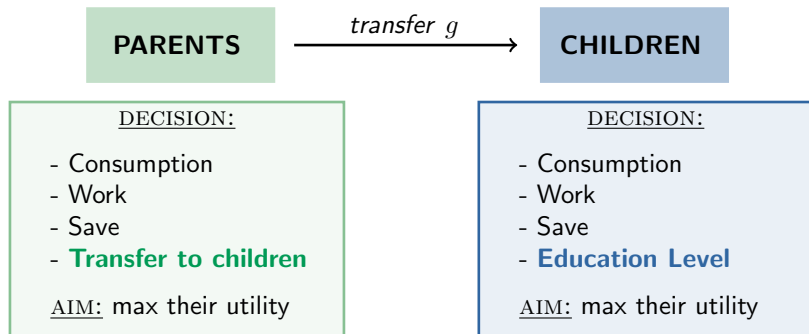
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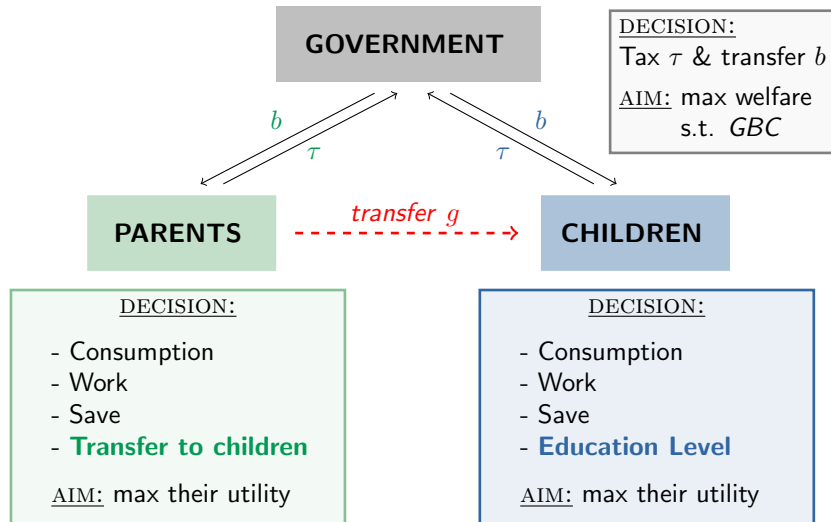
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# SETUP - PREVIEW





# SETUP - PREVIEW



## Should the government change the benefit age-profile?

- ▶ Trade-off when implementing a policy:
  - Social marginal utility (SMU)
  - Pareto weights ( $\omega$ )
  - Fiscal cost of the policy = 1 + Fiscal Externalities (FE)
- ▶ Comparing Policies: (*rely on MVPF literature*)
  - Compare **welfare effect** of small deviations from the actual policy for children vs parents via **cost-benefit ratios**:

$$\frac{\omega \times \text{SMU}^k}{1 + \text{FE}^k} \stackrel{?}{>} \frac{\omega^P \times \text{SMU}^P}{1 + \text{FE}^P}$$

- Compare benefit increase for children in **education** VS on **the labor market**.

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## ► Components of SMU:

- For transfer to parents:

- Utility gain of parents
- Utility gain of children from the amount **passed-through**

- For transfer to children:

- Utility gain of children
- Accounting for possible **crowding-out effect** on parent transfer

# SOCIAL MARGINAL UTILITY

## ► Components of SMU:

- For transfer to parents:

- Utility gain of parents
- Utility gain of children from the amount **passed-through**

- For transfer to children:

- Utility gain of children
- Accounting for possible **crowding-out effect** on parent transfer

## ► Need to estimate:

(1) Transfer derivatives: Crowding-out and pass-through

(2) Utility gain: Rely on Landais and Spinnewijn (2021)

- The higher the MPC, the larger the transfer value
- Ratio of parent-children mpc gives a lower bound of parent-children SMU ratio

## Challenges:

- (1) MPC: Need **comparable exogenous variation** in income for both **children** and **parents**.
- (2) Transfer derivatives: Identify parent-to-children transfer and its change

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- (1) MPC: Need **comparable exogenous variation** in income for both **children** and **parents**.
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## Data: Cr dit Mutuel Bank transaction data

- ▶ Granular information on **flows and household balance-sheet**:
  - Credit card expenditure at the transaction level
  - All incoming and outgoing transfers at a daily frequency
  - Balance of current and saving accounts, mutual funds and debt
- ▶ **Demographic characteristics** (Age, Sex, CSP, etc.)
- ▶ Period: 2019 - Now
- ▶ Random sample of **300,000 households** (> 500,000 individuals)

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### Exploit two one-shot transfers in 2020:

#### ▶ **Transfer to children (18-24):**

- Target: Entitled to housing benefits (APL) or students with grant
- Amount: 150 euros
- Number of treated in data:  $\simeq$  4,000 individuals

#### ▶ **Transfer to parents:** ARS (Allocation de Rentrée Scolaire)

- Target: Parents with a child between 6 and 18.
- Amount:  $\simeq$  500 euros per child
- Number of treated in data:  $\simeq$  20,000 individuals

► Compute MPC using an DID event study:

$$C_{it} = \alpha_i + \delta \text{Treated}_i \cdot \mathbb{1}_{[t > t^*]} + \gamma \mathbb{1}_{[t > t^*]} + \beta X_{it} + \varepsilon_{it}$$

- Weekly consumption ( $C_{it}$ ), treatment week  $t^*$ , incoming transfer ( $X_{it}$ )
- MPC retrieved from re-scaling  $\delta$  [Details](#)

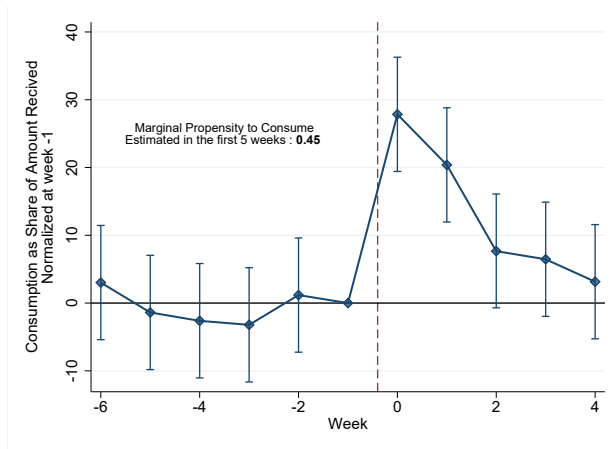
► Matching: Control group constructed using the one-to-one nearest-neighbor matching on pre-event characteristics:

- Exactly matched on demographics characteristics
- Mahalanobis distance on financial variables

# SMU - EMPIRICAL APPLICATION

## MPC - RESULTS

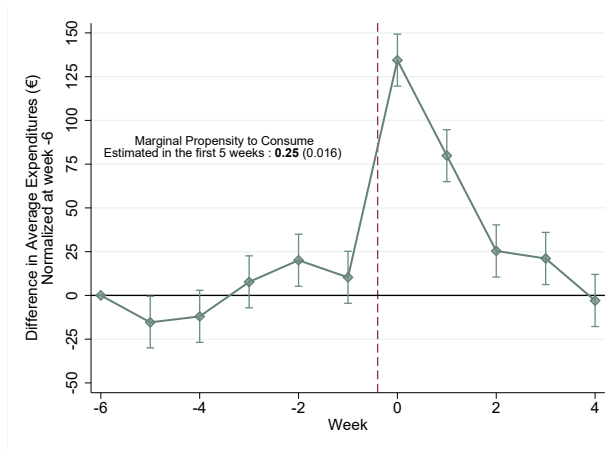
FIGURE: Children MPC (45%)



# SMU - EMPIRICAL APPLICATION

## MPC - RESULTS

FIGURE: Parents' MPC (25%)



Amount

Year

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# TRANSFER DERIVATIVE - METHODOLOGY

► Advantage:

- Data that pins down parent-child transfer
  - Panel follows individuals over several months/years
- ↪ Move away from standard survey cross-sectional analysis

► Regression: TWFE following individuals ( $i$ ) over months ( $t$ ):

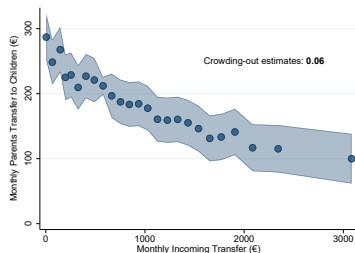
$$G_{it} = \alpha_i + \alpha_t + \beta_1 Y_{it} + \gamma_1 Y_{it}^2 + \beta_2 Y_{it}^P + \gamma_2 Y_{it}^{P^2} + \varepsilon_{it}$$

With  $G_{it}$  parents-to-child transfer,  $Y_{it}$  children and  $Y_{it}^P$  parent's total incoming transfers

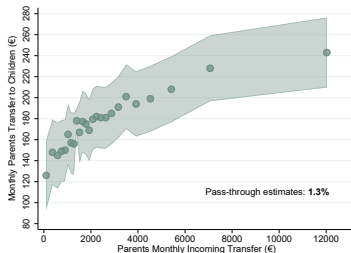
- Crowding-out recovered from  $\beta_1$  and  $\gamma_1$
- Pass-through recovered from  $\beta_2$  and  $\gamma_2$

# TRANSFER DERIVATIVES

(A) CROWDING-OUT (6%)



(B) PASS-THROUGH (1%)



- **Methodology 2:** Take advantage of heterogeneity in the scholarship amount (8.5%) OLS Non-Wire

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► Wrap-Up:

- MPC significantly higher for young adults
- Low crowding-out
- Low pass-through

► Implication for SMU: Putting pieces back in the model

$$SMU_t^{b^k} \geq 2.05 \times SMU_t^{b^P}$$

- Social benefit of transferring to children twice bigger

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**Idea:** Replicates the analysis for the most two fragile groups

## ① What about tagging low-income workers?

- MPC: 55% (0.1) [Plot](#)
- Crowding-out: 7% (0.008)

$$\Rightarrow SMU_t^{b^w} \geq 2.9 \times SMU_t^{b^P}$$

## ② What about tagging students with low-income parents?

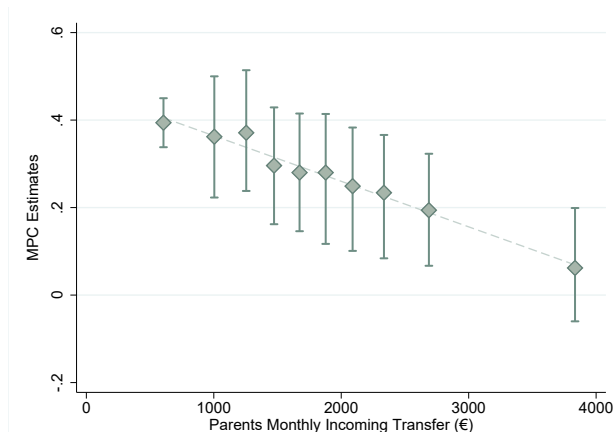
- MPC: 61% (0.15) [Plot](#)
- Crowding-out: 4.5% (0.017)

$$\Rightarrow SMU_t^{b^e} \geq 4.12 \times SMU_t^{b^P}$$

# SMU - TAGGING PARENTS

- So far the estimated effect is an **upper bound** (population of MPC with on average lower income)

FIGURE: Parents' MPC by Incoming Transfer



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## Benefit Cost Ratio Analysis:

$$\frac{SMU^k}{\underbrace{SMU^P}_{\text{Found to be large}}} \quad ? \quad \frac{1+FE^k}{\underbrace{1+FE^P}_{?}}$$

## Benefit Cost Ratio Analysis:

$$\frac{SMU^k}{SMU^P} \stackrel{?}{>} \frac{1+FE^k}{1+FE^P}$$

► Increasing transfer to parents:

- Labor supply

## Benefit Cost Ratio Analysis:

$$\frac{SMU^k}{SMU^P} \stackrel{?}{>} \frac{1+FE^k}{1.14}$$

► Increasing transfer to parents:

- Labor supply  $\Rightarrow$  Hendren (2016) FE = 0.14



## Benefit Cost Ratio Analysis:

$$\frac{SMU^k}{SMU^P} \stackrel{?}{>} \frac{1+FE^k}{1.14}$$

### ▶ Increasing transfer to parents:

- Labor supply  $\Rightarrow$  Hendren (2016) FE = 0.14

### ▶ Increasing transfer to children in education:

+ Education decision (✓)

? Return to education of pivotal children ( $\simeq$ )

- Wealth effects on labor supply (✓)

### ▶ Increasing transfer to children in labour market:

- Education decision (✓)

? Return to education of pivotal children ( $\simeq$ )

- Labor supply (✗)

► Children in education vs. parents:

- Fiscal cost = 0.9 Plot

$$\frac{SMU^e}{1 + FE^e} \geq 6 \times \frac{SMU^P}{1 + FE^P}$$

- Redistribution is highly welfare enhancing

► Children in education vs. parents:

- Fiscal cost = 0.9 Plot

$$\frac{SMU^e}{1 + FE^e} \geq 6 \times \frac{SMU^P}{1 + FE^P}$$

- Redistribution is highly welfare enhancing

► “Poor” children in the labor market vs. parents:

- If labor supply elasticity is the same for parents and children:

$$\frac{SMU^w}{1 + FE^w} \geq 2 \times \frac{SMU^P}{1 + FE^P}$$

Plot

# CONCLUSION

- ▶ The social utility of increasing transfer to young adult is at least **two times higher** than targeting older individuals
- ▶ Fiscal cost depends on whether targeting students or workers, but **never justify such low assistance**
- ▶ **Higher welfare effect when targeting:**
  - Students with low income parents
  - Young workers with low income

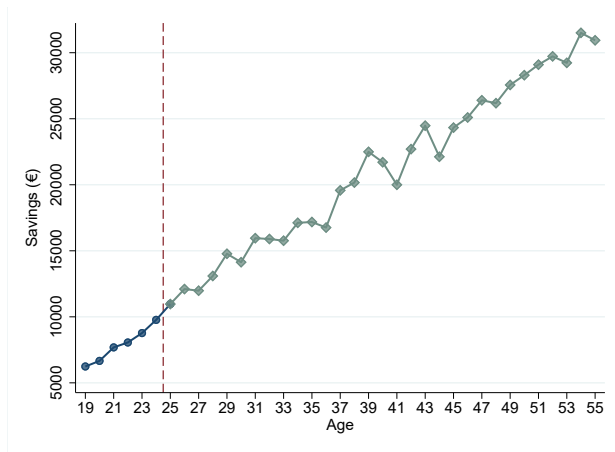
⇒ **Redistribution from older to younger individual  
would be highly welfare enhancing**

Thank you!

# APPENDIX

# MOTIVATION - FINANCIAL FRAGILITY

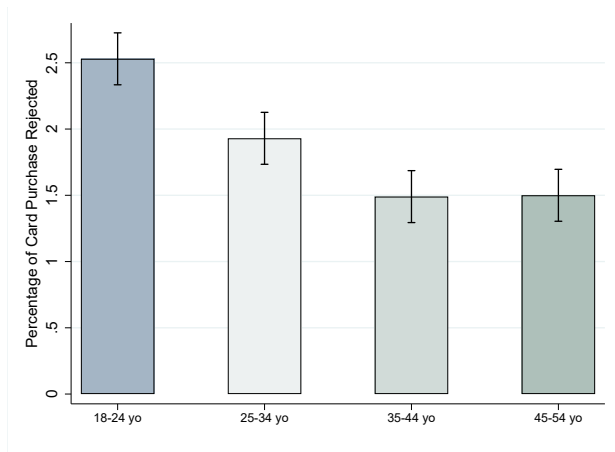
FIGURE: Savings by Age



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# MOTIVATION - FINANCIAL FRAGILITY

FIGURE: Share Card Purchase Rejected

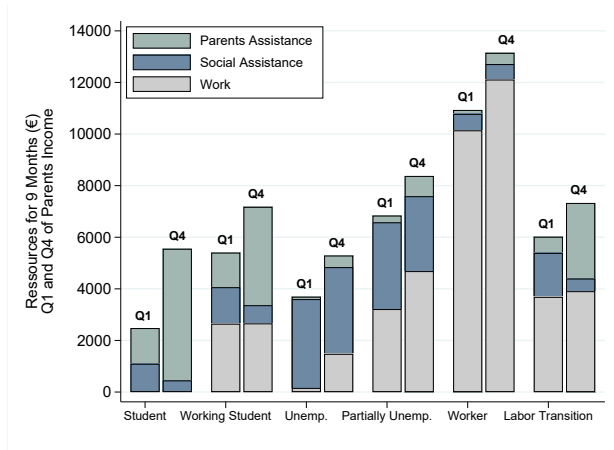


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# MOTIVATION - INEQUALITY

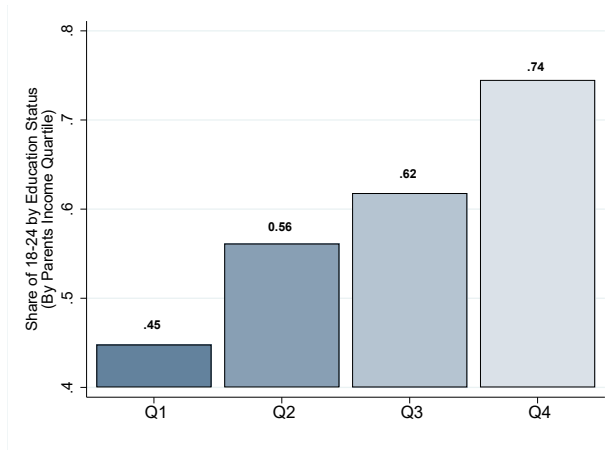
FIGURE: Income of Young Adults



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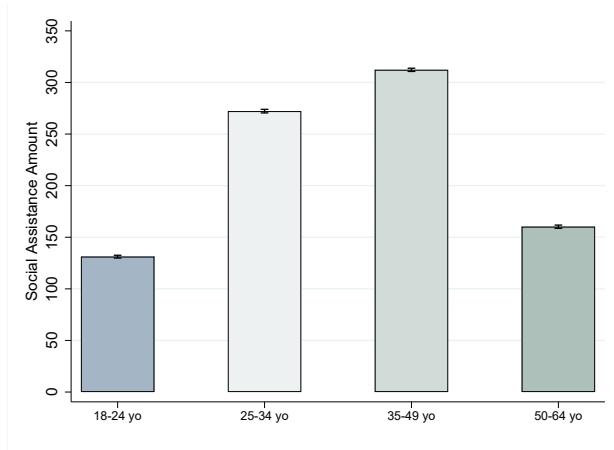
# MOTIVATION - INEQUALITY

FIGURE: Share of 18-24 in Education



# MOTIVATION - SOCIAL ASSISTANCE

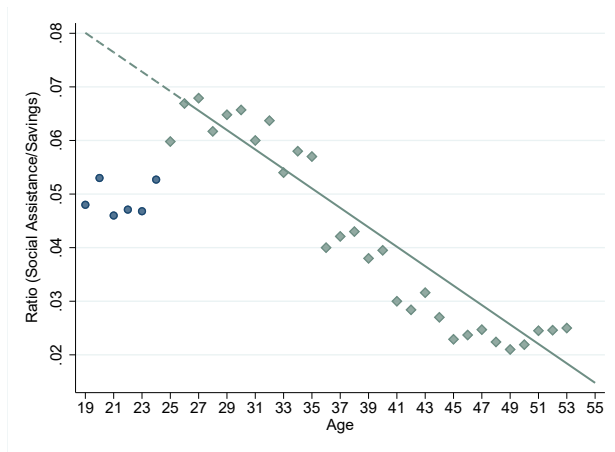
FIGURE: Social Assistance by Age



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# MOTIVATION - SOCIAL ASSISTANCE

FIGURE: Ratio Social Assistance by Savings by Age



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## ► MPC Computation:

$$C_{it} = \alpha_0 + \alpha_i + \gamma \mathbb{1}_{[t > t^*]} + \beta \text{Treated}_i \cdot \mathbb{1}_{[t > t^*]} + X_{it} + \varepsilon_{it}$$

$$MPC = \frac{\beta \cdot 5}{\text{Transfer Amount}}$$

## ► DID runs to plot graphs

$$C_{it} = \alpha_0 + \alpha_i + \gamma \sum_{t \neq -1} \text{Week}_t + \beta \sum_{t \neq -1} \text{Week}_t \cdot \text{Treated}_i + X_{it} + \varepsilon_{it}$$

$C_{it}$  = Weekly consumption,  $t^*$  = Treatment week,  $X_{it}$  = Incoming transfer [Back](#)

FIGURE: MPC by Amount of Transfer

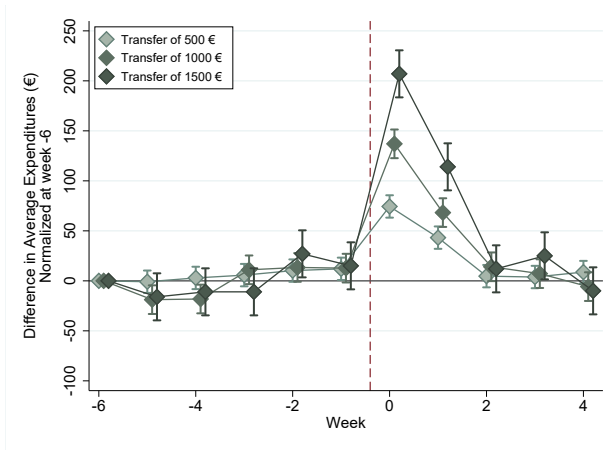
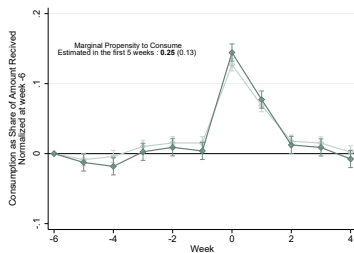
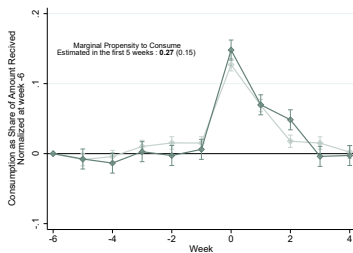


FIGURE: MPC Parents Additional Year

(A) MPC 2019 (0.25)



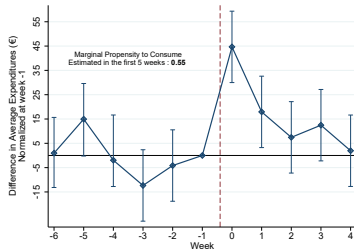
(B) MPC 2021 (0.27)



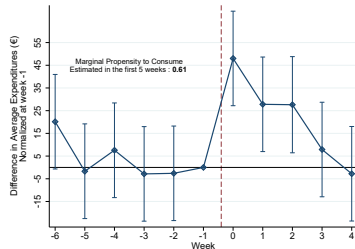
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# MPC HETEROGENEITY

(A) WORKERS (0.55)



(B) STUDENTS GRANT (0.61)



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# CROWDING-OUT - SCHOLARSHIP

TABLE: OLS Estimates, Crowding-out of Scholarship

	(1)	(2)
<b>1. Scholarship</b>		
Scholarship Amount	-0.095*** (0.027)	<b>-0.085***</b> (0.028)
Parent Ref Wage	0.0002*** (7e-5)	0.2e-4 (0.7e-4)
<b>2. Covariates</b>		
Age		3.3 (4.05)
Nb Siblings $\geq 18$ yo		-11.01** (4.3)
Gender		-19** (8.18)
Incoming Transfers Parents		0.0005*** (0.8e-5)
<b>Number of Observations</b>	2,833	2,833

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FIGURE: Decomposition Parent's Assistance

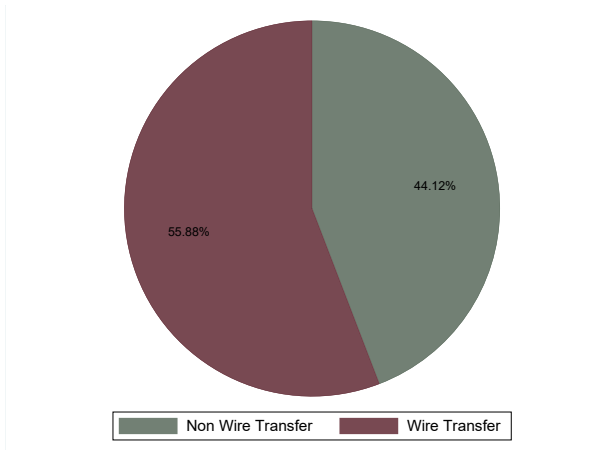
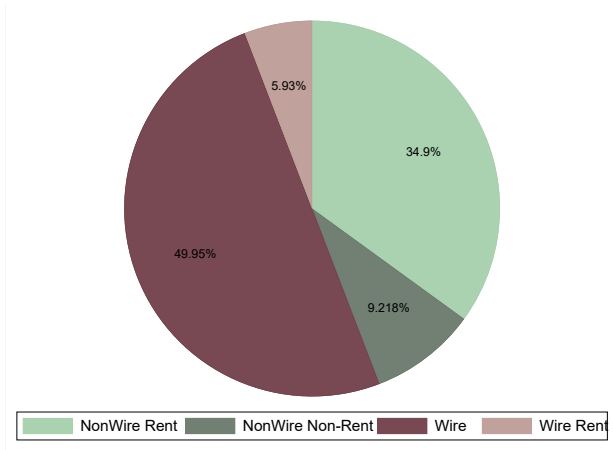


FIGURE: Decomposition Parent's Assistance



# BEHAVIOURAL RESPONSES - ESTIMATES

TABLE: Estimates used

	Parameters	Estimates
<b><u>1. Labor Supply</u></b>		
Cesarini et al (2017)	Wealth elasticity	0.01
Hendren (2016)	Labor Supply FC	0.14
<b><u>2. Education distortion</u></b>		
Fack et Huillery (2021)	Return to education	0.10
Fack and Grenet	Education responses wrt $b^e$	0.07/1500
Blundell et al (2016)	Education responses wrt $b^w$	0.01/270

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