Intergenerational educational trajectories and early mortality from chronic diseases among Swiss middle-aged adults: A registry population-based study

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Consistent evidence of inequalities in all-cause and cause-specific mortality due to individual attained education

Experimental and quasi-experimental evidence: Galama et al, 2018; Lager and Torssander, 2013

Observational evidence: Mackenbach et al, 2016 and 2019; Huisman et al, 2015; Welsh et al., 2021

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<u>Trends</u> (Mackenbach et al., 2019)

Life course origin of chronic diseases (CDs): how about parental education ?



from Olsen et al., 2016

<u>Putative mechanisms underlying the effect of parental education on mortality</u>:

Intergenerational educational inequalities in CD-related mortality: what drives them?

1. Only individual education (Theory of resource distribution; Ross and Mirowski 2011)

2. Both parental and individual edu (Theory of resource multiplication; Willson et al. 2007)

3. Direction of change between parental and individual edu (Theory of social mobility; Erikson and Goldthorpe 1992)

Intergenerational educational inequalities in CD-related mortality: our study

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Previous studies - Only 2 with cause-specific CD mortality, mainly from Finland and Belgium.

- Cancer, CVD, Substance use in Finnish born <1950: Elo et al., 2014;
- Cancer in Belgian born 1970-1980: Da Grande et al., 2014

Summary of findings: Individual education as main driver for Cancer / Substance use Parental and individual edu for CVD

Data: Swiss National Cohort

- Sociodemographic information for individuals resident in CH at 1990 and 2000 censuses •
- Mortality follow-up until end of 2018 •

Study population: Adolescents 10-19y at 1990 census, N = 544 743. • Individuals born 1971 to 1980. Mortality follow-up up to 38-47y of age.



Swiss National Cohort (SNC)

Intergenerational education : exposure





Individual Edu

- High Low
- High
- Low

Characteristics of study population

Characteristics	N (%)
Sex	
Men	283 761 (52%)
Women	260 982 (48%)
Household type	
Couple	488 919 (90%)
Single parent	55 824 (10%)
Country of birth of parents	
CH	429 589 (79%)
EU	78 584 (14%)
Other	36 570 (7%)
Birth cohort	
1971-1975	242 761 (45%)
1976-1980	301 982 (55%)

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- Sex as effect modifier







Low-Low vs High-High



20 25 30 35 40 45 Age [years]







Results - Women



Limitations and strengths

Limitations

- Information bias from education/mortality missclassification ullet
- Residual confounding ullet

<u>Strengths</u>

- Large sample size allowing to study cause-specific premature mortality ullet
- We estimated inequalities via marginal risk differences and not conditional hazard ratios > Avoid non-collapsibility and implicit selection bias of hazard ratios (Young et al., 2020)

Summary and public health consequences

- Sex- and cause of death- specific inequalities in premature mortality
- Inequalities in CVD premature mortality driven by individual education
- Inequalities in Substance use premature mortality driven by change between parental and individual edu

Public health consequences

• Intergenerational prevention strategies may mitigate premature mortality

Thank you for your interest

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