Motivation	Data	Model	Results	Discussion

# Education, cognitive ability and Cause-Specific Mortality: A structural approach

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Motivation				

- Education is negatively associated with mortality for most causes of death
- Standard Cox method: Interpretation of coefficients difficult in competing risks Ignores Cause-specific hazard rates are interdependent Does not provide importance of cause Education and mortality both depend on cognitive ability
- Proposed solution:

Focus on months lost due to specific cause of death Structural model that derives cognitive ability from IQ-scores

• Using Swedish conscription data 18-63 year

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# Causal impact of education on mortality

- Recent results deriving from natural experiments and from twin studies in education suggest that causal effect of education on health is small or even absent
- Suggest an important role for confounding factors, such as cognitive ability
- Educational attainment and cognitive ability strongly correlated. Difficult to disentangle.
- Using structural models: models interdependence
  Half of mortality disparity across education levels due to selection of the healthier into higher education (Bijwaard et al. 2015a,b).
- Studies on educational differences in cause-specific mortality ignore endogeneity

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Estimate the education-mortality gradient for major causes of death.

Contribution is threefold:

Q Causal effect of education on months-lost due to specific cause

- Account for endogeneity of education attainment
- Oerive selection effect both on observed and unobserved characteristics (cognitive ability)

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Swedish N	Ailitarv exami	mation Data		

Examinations for military service men born 1951-1960: 446,545 individuals.

- Linked to parental info: Detailed info on individual demographic and socioeconomic characteristics, including parental SES and education
- Intelligence test: IQ in 9 categories
- Education classified in 4 levels: primary education, Secondary education (2 years), Full Secondary education (3 years) and, Higher education
- Linked to death registers: Mortality by cause of death, till end 2012. neoplasms, CVD, external causes ,and other causes.

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# Descriptive statistics: distribution cause of death

	primary	Sec edu (2yr)	Sec edu (3yr)	Higher
# of deaths	8,770	9,451	2,506	3,829
deaths per 1000	90.8	59.1	45.3	28.4
		causes c	of death	
neoplasm	18.2	14.0	13.1	10.0
Cardiovascular diseases	18.4	13.9	10.4	6.3
External causes	31.5	16.5	11.7	6.8
Other causes	22.6	14.7	10.1	5.3

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# Cumulative incidence curves by cause of death and education level



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# Cox hazard ratios, adjacent education

	Sec edu (2 yr)	Sec edu (3 yrs)	Higher
neoplasm	0.77**	0.88**	0.79**
CVD	0.72**	0.73**	0.63**
external causes	0.51**	0.69**	0.59**
other causes	0.59**	0.64**	0.54**

\*\*p < 0.01



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# Inference in competing risks model

- Cause-specific Cox hazard models, λ<sub>k</sub>(t) Difficult interpretation if one covariate appears in several competing hazards and assumes independence of causes of death.
- Cumulative incidence: probability dying from cause k before t

$$F_k(t) = \int_0^t \lambda_k(s) S(s) \, ds$$

Fine-Gray model sub-distribution hazard also difficult to interpret

• Months lost due a specific cause, (from age 18 till age 63)

$$L_{k}(18,63) = \int_{18}^{63} F_{k}(s) \, ds$$

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# Structural model of education and cause-specific mortality

Extension of structural model of Bijwaard et al. (2015a,b)

Model the interdependence between education and cause-specific mortality, because both are affected by cognitive ability.

#### Educational attainment D

Ordered probit model depending on observed characteristics and latent cognitive ability,  $\boldsymbol{\theta}$ 

#### 2 Potential cause-specific hazard $\lambda$

Depending on education attained and latent cognitive ability: only observe hazards for observed education. Proportional Gompertz with shape and scale depending on education and cause of death

#### Measurement, M

Measuring (a proxy) of cognitive ability, IQ, depending on observed characteristics and latent cognitive ability\_\_\_\_\_

## Graphical representation



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## • Educational gain $G_c(\tau_0, \tau_1)$ ;

Average educational difference in months due cause cEducational gain (difference) implied by structural model

#### Selection effect;

Effect of selecting education: difference with non-parametric estimate  $G_{NP,c}(\tau_0, \tau_1)$ 

 selection on observables G<sub>NP,c</sub>(τ<sub>0</sub>, τ<sub>1</sub>) - G<sub>sep,c</sub>(τ<sub>0</sub>, τ<sub>1</sub>) with G<sub>sep,c</sub>(τ<sub>0</sub>, τ<sub>1</sub>) is the educational gain based on a stratified model (ignoring cognitive ability)

• selection on cognitive ability  $G_{sep,c}(\tau_0, \tau_1) - G_c(\tau_0, \tau_1)$ difference structural model and stratified model

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# Model estimates of months lost due to specific cause 18-63

#### Educational gains

Non-parametric estimates

Based on the non-parametric estimate of survival, Kaplan–Meier and cumulative incidence, Aalen–Johansen

#### Structural model

Gompertz hazard models by education level and cause of death, including observed individual characteristics Model accounting for (latent) cognitive ability influencing both education and cause-specific hazards.

## Non-parametric: Months lost and gain 18-63





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## Structural model: Months lost 18-63



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# Structural model: educational gain 18-63



# Total selection effect 18-63



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## Selection: observed and cognitive ability





## Structural model versus Cox models



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# Summary: Educational gains on cause specific mortality

Developed structural model which accounts for interdependence of education and cause-specific mortality rates.

- educational gain in months lost due to specific cause (accounting for cognitive ability)
- Selection effects: observed and (latent) cognitive ability

Main empirical results: accounting for selection

- Highest educational gain for primary education: 9 mo
- Largest gain due reduction in external causes: 1–7 mo small gains for CVD: < 1 month and neoplasms

- Largest selection effect lowest 2 groups: 2 mo
- Largest selection effect for external causes

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- Months lost better measure than hazard ratios Accounts for interdependence between causes and easy to interpret and additive measure
- Structural model accounts for interdependence of education and mortality due to cognitive ability Ignoring this leads to underestimate educational gains for low educated and overestimate for higher education

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#### Limitations

- Other personal traits might affect education non-cognitive skills
   Educational gain is likely to be upper-bound
- Only men
- short follow up: max age 63