

# **Spread out your gown and fly? Early career effects of graduate migration in the Danish labour market**

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# Mobility and Early Career Effects

## Introduction

- We contribute to the empirical literature on **causes and consequences of human capital migration**
- Specifically, we use detailed **registry data for graduates** from Danish higher educational institutions (HEIs)...
- ... estimate **returns to migration** while treating mobility decision as endogenous with regard to observed labour market outcomes (employment probability and entry wage)
- Reduced form and endogenous treatment model estimates
  - Migration decision **endogenous** to labour market outcomes
  - **IV strategy:** (Aggregate) local labour market signals in origin region of graduates at time period of graduation (exogenous supply push)

# Mobility and Early Career Effects

## Related Literature / Contribution

- Growing number of studies on **labour market returns** to education, migration, urbanization (e.g. urban wage premium)
- *What is missing?*
  - Identification of **causal effects for graduate migrants**
  - Accounting for heterogeneous HEI systems (**academic/professional**)
  - Decomposing **spatial patterns** of migrants and associated returns
- *Closest references to...*
  - ...our research focus: Wage differences among U.S. graduates as studied in Card and Lemieux (2001), Yankow (2003) and Wozniak (2010)
  - ...our identification strategy: Probability model to control for selection into mobility (Böheim and Taylor, 2007; Fitzenberger et al., 2015)

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## Empirical Approach

- We estimate a linear POLS model for **outcome**  $y_{it}$  as

$$y_{it} = \beta_0 + \beta_1 \mathbf{x2}_{it} + \delta d_i + \lambda_t + \mu_{n(i)} + u_{it} \quad (1)$$

where:  $y_{it} = [\text{Pr}(Emp_{it}), \log(wage_{it})]$

$\mathbf{x2}_{it}$  = individual factors,  $\lambda_t$  = time-fixed effects

$\mu_{n(i)}$  = municipality-fixed effects,  $\varepsilon_{it}$  = error term

- Combes (2008):  $\mu_{n(i)}$  to account for **spatial sorting effects**
- Binary **treatment variable**  $d_i$  defined as

$$d_i = \begin{cases} 1 & \text{if individual } i \text{ is a } \mathbf{Graduate Mover} \\ 0 & \text{otherwise} \end{cases}$$

- Coefficient  $\delta$  as parameter of interest (return to migration)

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## Empirical Approach

- Problem of reduced-form estimation of eq.(1) is that a "causal" interpretation of parameter  $\delta$  is not possible due to **endogenous selection into treatment**
  - Endogeneity arises from underlying **correlation structure** of unobservables affecting treatment ( $d_i$ ) and outcome ( $y_{ijt}$ )
- Endogenous **treatment effects model** specifies  $d_i$  as

$$d_i = \begin{cases} 1 & \text{if } \theta_0 + \theta_1 \mathbf{x} \mathbf{1}_i + \theta_2 \mathbf{z}_{(i)n} + \epsilon_i \\ 0 & \text{otherwise} \end{cases} \quad (2)$$

- Estimation of eq.(2) as first-step **(linear) probability model** for becoming a Grad Mover as
  - Use  $\hat{d}_i$  in structural form IV regression of eq.(2)

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## Empirical Approach

- To account for endogeneity problem, we aim at formulating **exclusion restrictions** on basis of individual ( $x1_i$ ) and municipality ( $z_{(i)n}$ ) characteristics
- **Key criterion:** Exclusion restrictions shall be directly correlated with **mobility decision of graduates**, but only indirectly correlated with outcome variable(s)

$z_{(i)n}$  = [current and lagged labour market conditions at place of residence for time period of graduation]

$x1_i$  = [age, sex, family status,  $\Delta$ family status, education status, employment status(t-n), commuting(t-n), migration(t-n)]

$x2_i$  = [age, sex, education status, employment status, commuting status, industry, job classification]

[Data Description](#)

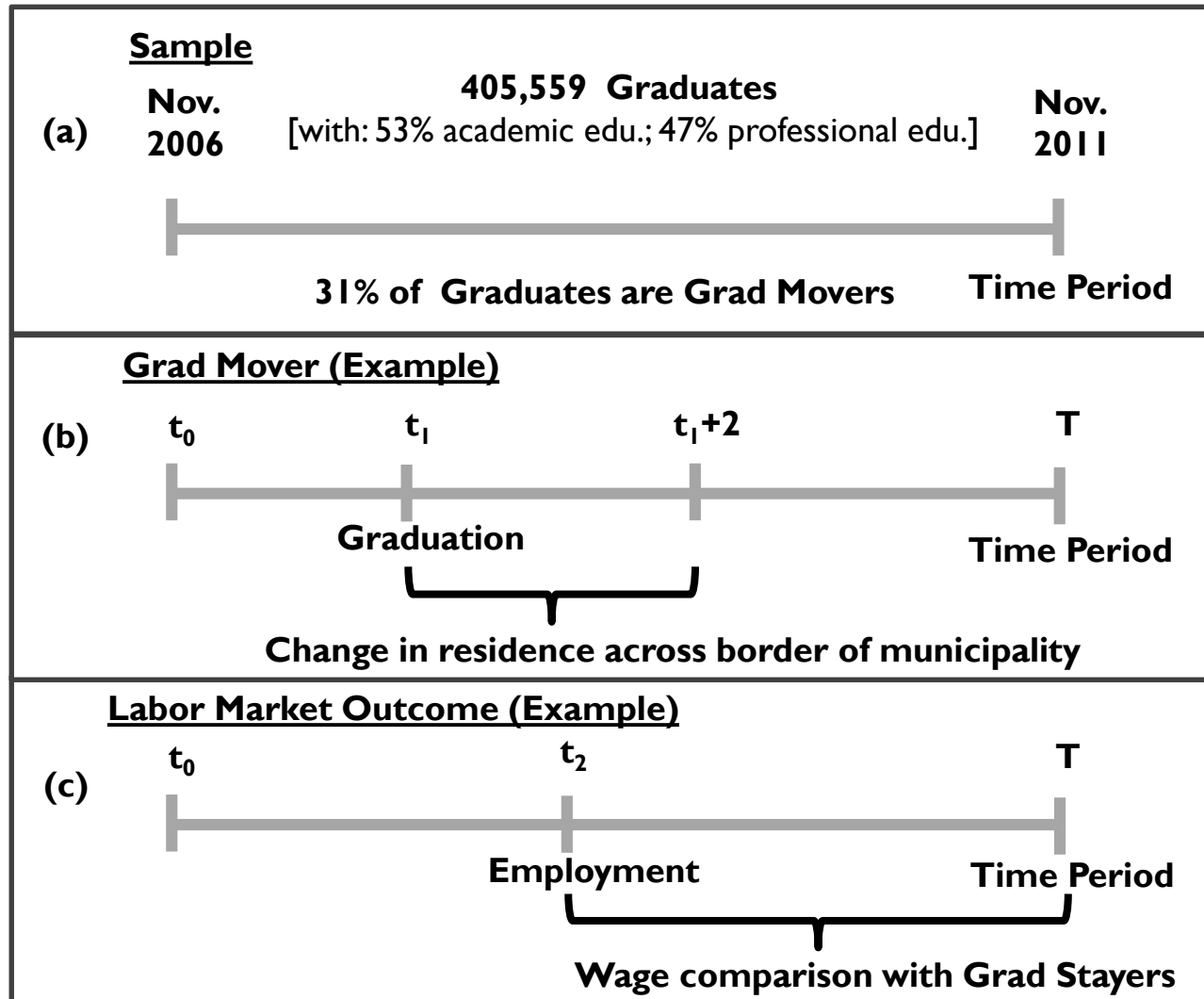
# Mobility and Early Career Effects

## Data

- Longitudinal micro-database from the residents' registry in Denmark from Statistics Denmark 2006-2011
- *Sample:* All residents aged between 16-70, **who graduated** between 2006-2011 (final graduation record)
  - 405,559 registered graduates
  - 31% are identified as graduate movers
- We associate migration activities through registered **movements across Danish municipalities** with graduation event...
  - ...if they occur in the period between  $t_{1(i)}$  and  $t_{1(i)} + 2$
  - ...where  $t_{1(i)}$  time period of graduation for each individual

# Mobility and Early Career Effects

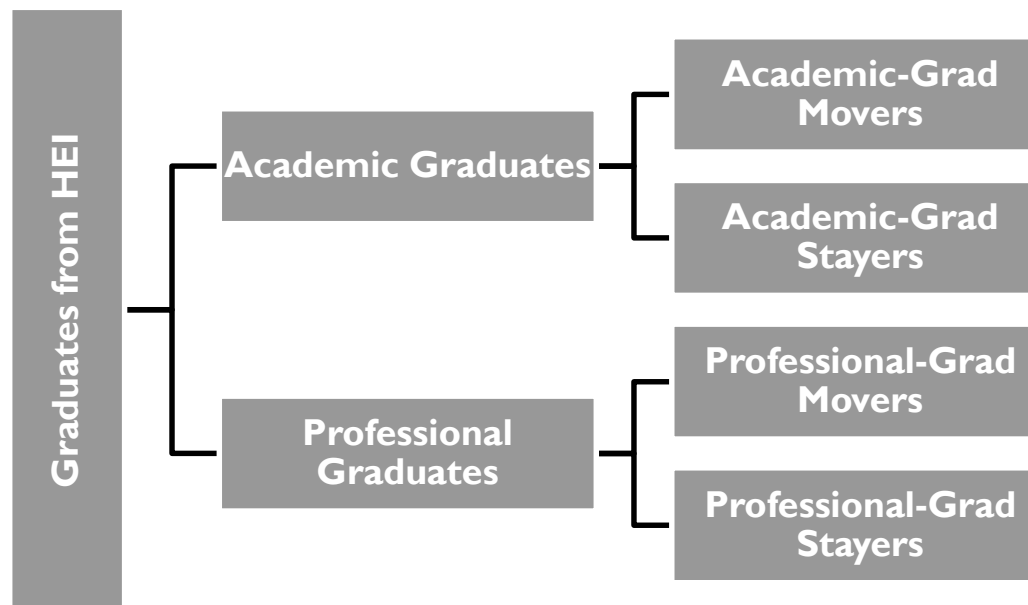
## Data



# Mobility and Early Career Effects

**Data**

**Decomposition I**



**Decomposition II**

**Academic-Grad**  
**Professional-Grad**

- stayers in rural municipality
- stayers in urban municipality
- movers from rural to urban
- movers from urban to rural
- movers from rural to rural
- movers from urban to urban

# Mobility and Early Career Effects

## Empirical Results

Table 1: Reduced form estimates

<i>Dep. Var.</i>	<i>Pr(Emp)</i>			<i>log(Wage)</i> §		
	Overall	Academic	Professional	Overall	Academic	Professional
Grad Mover	<b>0.004***</b>	<b>0.006***</b>	<b>0.003***</b>	<b>0.071***</b>	<b>0.085***</b>	<b>0.057***</b>
(S.E.)	(0.0004)	(0.0005)	(0.0006)	(0.0016)	(0.0023)	(0.0024)
Obs.	1,279,899	662,784	617,115	871,453	466,281	405,172
Individual Factors	✓	✓	✓	✓	✓	✓
Time Dummies	✓	✓	✓	✓	✓	✓
Municipality Fixed-Effects	✓	✓	✓	✓	✓	✓

Note: § = only full-time employment. \*\*\*, \*\*, \* denote statistical significance at the 1%, 5% and 10% significance level, respectively. Results reported for pooled OLS specification with two-way clustered standard errors at the individual and municipality level. Individual-specific controls include age, sex, dummies for sector of employment, occupational types, educational level and a dummy whether the person commutes to work or not

# Mobility and Early Career Effects

## Empirical Results

Table 2: Endogenous treatment model

<i>Dep. Var.</i>	<i>Pr(Emp)</i>			<i>log(Wage)</i> §		
	Overall	Academic	Professional	Overall	Academic	Professional
Grad Mover	<b>-0.003*</b>	<b>0.007***</b>	<b>-0.015***</b>	<b>0.038***</b>	<b>0.093***</b>	<b>-0.030**</b>
(S.E.)	(0.0018)	(0.0009)	(0.0020)	(0.0096)	(0.0153)	(0.0111)
Obs.	1,069,861	554,246	515,615	705,775	382,060	323,715
Weak IV Test	$F=9.82 > 15\%$ crit. val. of Stock-Yogo			$F=11.14 > 15\%$ crit. val. of Stock-Yogo		
Individ. Factors	✓	✓	✓	✓	✓	✓
Time Dummies	✓	✓	✓	✓	✓	✓
Municipality Fixed-Effects	✓	✓	✓	✓	✓	✓

Note: § = only full-time employment. \*\*\*, \*\*, \* denote statistical significance at the 1%, 5% and 10% significance level. Results for linearized 2SLS with two-way clustered standard errors at individual and municipality level. Instruments: graduates' family status at time of graduation, change in family status at time of graduation, graduate's commuting and mobility history as well as for current and one-period lagged values for local labour market variables in HEI region at time period of graduation.

# Mobility and Early Career Effects

<i>Dep. Var.:</i>	<i>Pr(Emp)</i>			<i>log(Wage)</i> §		
	Overall	Academic	Professional	Overall	Academic	Professional
Mover Rural-Urban	-0.003*	<b>0.011***</b>	<b>-0.018***</b>	<b>0.143***</b>	<b>0.206***</b>	<b>0.101***</b>
(S.E.)	(0.0019)	(0.0023)	(0.0019)	(0.0137)	(0.0197)	(0.0094)
Mover Urban-Rural	0.003*	<b>0.004**</b>	0.003	<b>0.071***</b>	<b>0.077***</b>	<b>0.071***</b>
(S.E.)	(0.0015)	(0.0021)	(0.0019)	(0.0072)	(0.0115)	(0.0078)
Mover Urban-Urban	<b>-0.014***</b>	-0.002*	<b>-0.022***</b>	<b>0.126***</b>	<b>0.136***</b>	<b>0.144***</b>
(S.E.)	(0.0019)	(0.0011)	(0.0014)	(0.0098)	(0.0141)	(0.0046)
Mover Rural-Rural	<b>0.006***</b>	<b>0.008***</b>	<b>0.005***</b>	<b>0.061***</b>	<b>0.087***</b>	<b>0.058***</b>
(S.E.)	(0.0015)	(0.0019)	(0.0016)	(0.0044)	(0.0078)	(0.0047)
Stayer Urban-Urban	<b>-0.016***</b>	<b>-0.006***</b>	<b>-0.023***</b>	<b>0.055***</b>	<b>0.058***</b>	<b>0.084***</b>
(S.E.)	(0.0025)	(0.0012)	(0.0023)	(0.0061)	(0.0076)	(0.0042)

# Mobility and Early Career Effects

## Conclusions

- Being mobile after graduation offers positive labour market returns
- If we control for endogeneity of mobility decision (unobservable individual productivity) effects complex:
  - i.e. if there are **two academic graduates** with similar productivity and individual characteristics, but differences in their graduation environment and core personal traits, then the one who decides to move, will have higher employability and higher wage premium
  - Results for **professional graduates** show mixed returns due to different labour market embeddedness (e.g. apprenticeships); mobility decisions that does not reflect productivity have a negative impact on their labour market outcomes

# Mobility and Early Career Effects

## Conclusions

- Results for spatial decomposition of migration patterns show
  - **Moving to urban area** has a largest positive effect both on employment and wages of the academic graduates
  - For professional graduates, **employment probability is highest in rural areas**, both for stayers and movers (however, professional graduates moving to the urban areas get higher wage premium)
- Results underline importance of spatial mobility for optimising the labour market allocation and individual outcomes
- Results may guide **regional policy** with regard to impact channels of graduate migration and mobile human capital
  - Academic graduates favour the urban areas and “thick” labour markets
  - Professionals can contribute for building the local labour market in mainly rural areas

# Mobility and Early Career Effects

## Data

Table 1: Variables ( $y_{it}$ ,  $x1_{it}$ ,  $x2_{it}$ )

Category	Variable	Graduate mover			Graduate stayer		
		Obs.	Mean	S.D.	Obs.	Mean	S.D.
<b>Outcome</b>	Annual wage (in DKK)	575,057	228897	150398	585,749	217544	158128
<b>Outcome</b>	Employment (0=no, 1=yes)	634,321	0.861	0.346	645,578	0.843	0.363
<b>Control (x1, x2)</b>	Age (in years)	634,321	28.74	5.94	645,578	32.79	9.30
<b>Control (x1, x2)</b>	Male (0=female, 1=male)	634,321	0.473	0.499	645,578	0.419	0.493
<b>Control (x1)</b>	Family status: Married partnership	634,321	0.228	0.419	645,578	0.399	0.490
<b>Control (x1)</b>	Family status: Unmarried partnership	634,321	0.102	0.343	645,578	0.089	0.333
<b>Control (x1)</b>	Family status: Alone with no partner	634,321	0.670	0.937	645,578	0.512	0.855
<b>Control (x2)</b>	Sector: Agriculture, fishing, mining	634,321	0.038	0.192	645,578	0.033	0.180
<b>Control (x2)</b>	Sector: Manufacturing	634,321	0.043	0.204	645,578	0.037	0.190
<b>Control (x2)</b>	Sector: Energy and water supply	634,321	0.013	0.112	645,578	0.011	0.104
<b>Control (x2)</b>	Sector: Building and construction	634,321	0.166	0.372	645,578	0.126	0.332
<b>Control (x2)</b>	Sector: Trade, hotel, restauration	634,321	0.057	0.232	645,578	0.049	0.216
<b>Control (x2)</b>	Sector: Transport, post and telecom	634,321	0.078	0.268	645,578	0.068	0.252
<b>Control (x2)</b>	Sector: Finance and business	634,321	0.077	0.266	645,578	0.076	0.265
<b>Control (x2)</b>	Sector: Public and personal service	634,321	0.406	0.491	645,578	0.473	0.499
<b>Control (x2)</b>	Occupation: High-skilled	634,321	0.689	0.462	645,578	0.706	0.455
<b>Control (x2)</b>	Occupation: Medium-skilled	634,321	0.473	0.499	645,578	0.484	0.499
<b>Control (x2)</b>	Occupation: Service-skilled	634,321	0.317	0.465	645,578	0.324	0.468
<b>Control (x2)</b>	Occupation: Manual workers	634,321	0.224	0.417	645,578	0.241	0.428
<b>Control (x2)</b>	Occupation: Military workers	634,321	0.193	0.394	645,578	0.210	0.407
<b>Control (x2)</b>	Occupation: Unknown	634,321	0.193	0.394	645,578	0.210	0.407
<b>Control (x2)</b>	Education: Vocational education	634,321	0.423	0.494	645,578	0.386	0.487
<b>Control (x2)</b>	Education: Short higher education	634,321	0.077	0.267	645,578	0.078	0.268
<b>Control (x2)</b>	Education: Medium level education (BA)	634,321	0.281	0.449	645,578	0.309	0.462
<b>Control (x2)</b>	Education: Higher education (MA/PhD)	634,321	0.218	0.413	645,578	0.227	0.419
<b>Control (x2)</b>	Commuter (0=no, 1=yes)	634,321	0.405	0.491	645,578	0.380	0.485

# Mobility and Early Career Effects

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## Data

Table 2: Variables ( $z_{(i)n}$ )

Variables (z)	Description	NUTS4 regions	Mean	S.D.
<b>Unemployment Rate</b>	Unemployment rate (in %)	98	5.209	1.703
<b>Disposable Income</b>	Disposable income per capita (in DKK)	98	187223	22785
<b>% Human Capital</b>	Share of persons with Bachelor, Master and PhD university degrees in total population aged 15-69 (in %)	98	10.04	6.930
<b>Net Migration</b>	Net in-migration balance as share of total population	98	1.001	0.005
<b>Net Commuting</b>	Number of gross in-commuters relative to gross out-commuters	98	1.000	0.214
<b>Δ House Prices</b>	Annual growth rate of average property prices per square meter (in %)	98	15.60	6.928
<b>% Home Ownership</b>	Share of home owners in total population (in %)	98	0.212	0.074
<b>% Manufacturing</b>	Number of workplaces in the manufacturing sector relative to total number of manufacturing workplaces in Denmark	98	91.96	32.16
<b>% Business Services</b>	Number of workplaces in business-related services relative to total number of workplaces in business-related services in Denmark	98	105.9	19.35
<b>% Other Services</b>	Number of workplaces in non-business-related services relative to total number of workplaces in other services in Denmark	98	100.0	8.529
<b>Crime Rate</b>	Total number of crime offenses per population	98	11.11	4.324
<b>Population Density</b>	Total population per area ( in square km)	98	1893.7	2928.3
<b>City Type1</b>	City of residence in municipality: Capital city/surrounding (binary dummy)	98	0.308	0.461
<b>City Type2</b>	City of residence in municipality: Pop. ≥100,000 (binary dummy)	98	0.146	0.354
<b>City Type3</b>	City of residence in municipality: Pop. 30,000 – 99,999 (binary dummy)	98	0.139	0.345
<b>City Type4</b>	City of residence in municipality: Pop. 2,000 – 29,999 (binary dummy)	98	0.401	0.490

Source: Authors' calculations based on data from STATBANK (Statistics Denmark). NUTS4 region = Number of municipalities per year, S.D. = Standard deviation. Variables take values for the individual's place of residence at the time period of graduation.