

The causal effect of early retirement on medication use across sex and occupation: Evidence from Danish administrative data

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Why?

Challenges in the literature:

- ▶ Directionality?
- ▶ Adequate measures of health

Common patterns:

- ▶ self-reported health based studies: positive effect¹
- ▶ administrative data: decline in doctors' appointments and hospital utilization²
- ▶ mental health & medication use: mixed results³

¹e.g. Coe & Zamarro (2011), Eibich (2015), Shai (2018), Gorry, Gorry & Slavov (2018)

²Bíró & Elek (2018), Nielsen (2019), Frimmel & Pruckner (2020), Perdrix (2020), Hallberg, Johansson & Josephson (2015), Kuusi, Martikainen & Valkonen (2020)

³Coe & Zamarro (2011), Mazzonna & Peracci (2012), Kolodziej & García-Gómez (2019), Bíró & Elek (2018), Hagen (2018), Perdrix (2020), Gorry et.al. (2018), Frimmel & Pruckner (2020)

Why?

Medication data:

- ▶ large administrative data source
- ▶ specific health conditions

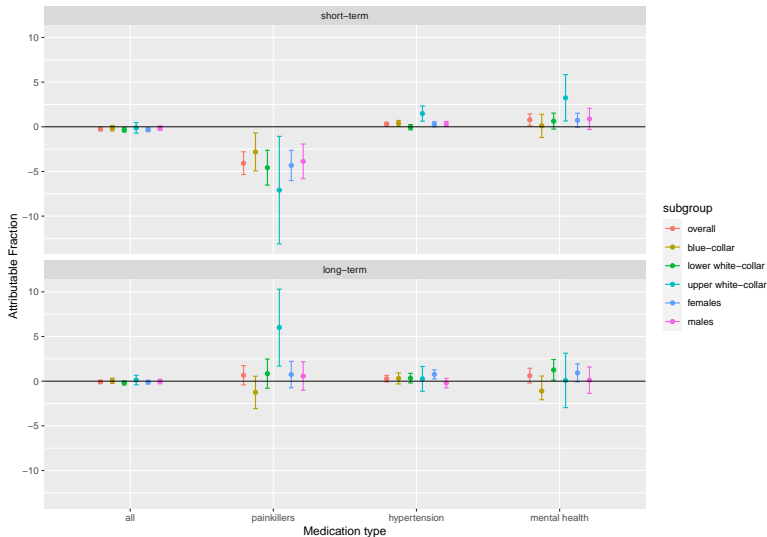
Approach:

- ▶ Recent data → recent reform → improved identifiability
- ▶ Monthly data
- ▶ Heterogeneity by occupation at age 45-55 & sex
- ▶ Improved IV technique for binary outcomes⁴

⁴Vansteelandt, Bowden, Babnezhad & Ghoetghebeur (2011)

Results

Percentage point change in medication use if all would have retired

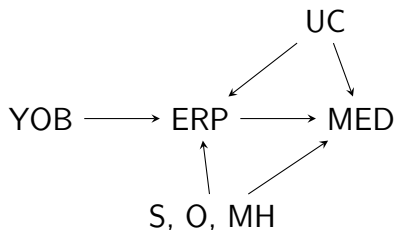


Design: reform

Table 1: Ages at which Early Retirement Pension (ERP) is available according to date of birth.

Date of Birth	ERP age
< 1954	60
≥ 01-January-1954	60.5
≥ 01-July-1954	61
≥ 01-January-1955	61.5
≥ 01-July-1955	62
≥ 01-January-1956	62.5
≥ 01-July-1956	63
≥ 01-January-1959	63.5
≥ 01-July-1959	64
> 1963	computed in relation to life expectancy

Design: IV model



$$P(\text{MED}_i = 1) = \text{logit}^{-1}(\beta_0 + \beta_1 \text{ERP}_i + \beta_2 \text{S}_i + \beta_3 \text{O}_i + \beta_4 \text{MH}_i + u_i)$$

$$P(\text{ERP}_i = 1) = \text{logit}^{-1}(\gamma_0 + \gamma_1 \text{YOB}_i + \gamma_2 \text{S}_i + \gamma_3 \text{O}_i + \gamma_4 \text{MH}_i + \nu_i)$$

Design: variables

- ▶ ERP: Early retirement pension, indicator of retirement between age 60-60.5
- ▶ YOB: year of birth, either pre-1954 (unaffected cohorts) or ≥ 1954 (affected cohorts)
- ▶ MED: medication use, indicator of medication use between age 60-60.5 (short-term) or 60-63 (long-term)
- ▶ MH: medication history, categorical medication use between age 59.5-60, 0, ≤ 7 , ≤ 31 or >31 daily defined doses a month
- ▶ S: sex (male/female)
- ▶ O: longest held occupation type between age 45-55.

Design: medication types

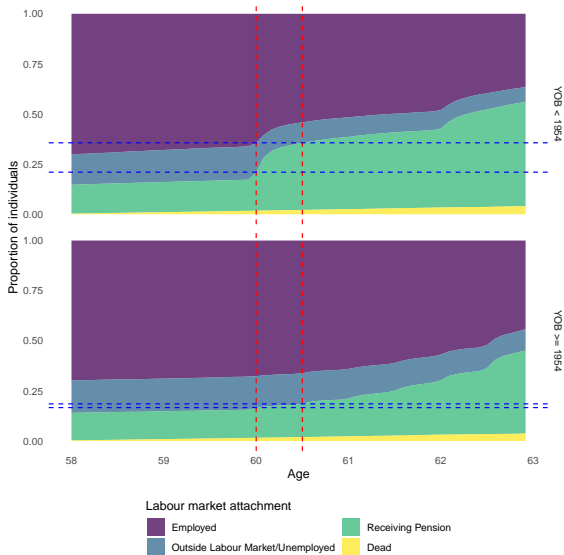
Painkillers: M02AA, N02A, N02B (antiinflammatory preparations, non-steroids for topical use, opioids, other analgesics and antipyretics)⁵

Medication to reduce blood pressure: C02, C03, C07, C08, C09 (antihypertensives, diuretics, beta blocking agents, calcium channel blockers, agents acting on the renin-angiotensin system)

Antidepressants and medications for anxiety and sleep disorders: N05B N05C, N06A, N06C (anxiolytics, hypnotics and sedatives, antidepressants, psycholeptics and psychoanaleptics in combination)

⁵N02BE01, N02BE05, N02BA01, N02BA51, N02BB51 and 200 mg M01AE01 are not included

Design: labour market attachment

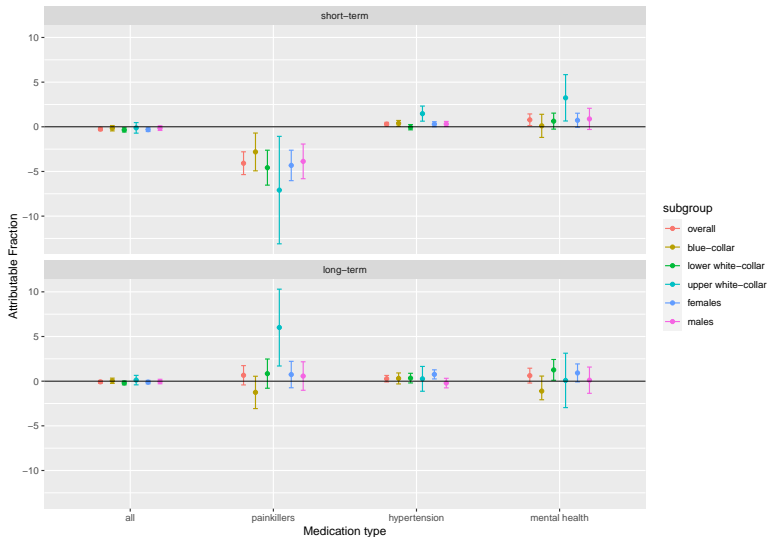


IV assumptions

- ▶ IV associated with exposure
- ▶ Exclusion restriction
- ▶ Randomization assumption

Results

Percentage point change in medication use if all would have retired



Discussion

- ▶ Upper white-collar, income/wealth effects
- ▶ No long-term effects?
- ▶ Effects on amount of medication use (IV models for count/zero-inflated outcomes)
- ▶ Different types of medication, split up mental health medication?

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Descriptives

		Year of birth					
		<1954 (N = 222,269)			≥1954 (N = 231,722)		
		All	ERP = yes	ERP = no	All	ERP = yes	ERP = no
S	male	0.51	0.39	0.54	0.51	0.45	0.51
	female	0.49	0.61	0.46	0.49	0.55	0.49
O	blue-collar	0.27	0.35	0.25	0.25	0.13	0.25
	lower white-collar	0.41	0.46	0.39	0.42	0.43	0.41
	upper white-collar	0.22	0.12	0.25	0.20	0.29	0.20
	no occupation	0.10	0.07	0.10	0.14	0.15	0.14
MH	0 DDD/month	0.30	0.26	0.31	0.30	0.22	0.30
	≤ 7 DDD/month	0.14	0.12	0.14	0.14	0.12	0.14
	≤ 31 DDD/month	0.19	0.19	0.19	0.19	0.19	0.19
	> 31 DDD/month	0.37	0.43	0.36	0.37	0.47	0.37
MH type	none	0.30	0.25	0.31	0.30	0.22	0.30
	hypertension	0.32	0.36	0.31	0.31	0.36	0.31
	painkillers	0.02	0.03	0.02	0.03	0.05	0.03
	mental	0.06	0.07	0.05	0.05	0.09	0.05
	other	0.30	0.29	0.30	0.31	0.29	0.31
ERP	no	0.80	0	1	0.98	1	0
	yes	0.21	1	0	0.02	0	1

Medication frequencies

Table 2: Most frequently used types of prescription medication ($\leq 15\%$) in the period 2008-2017 among individuals aged 58-63 that have used prescription medication at least once.

medication type (ATC)	percentage	subtypes ($\leq 10\%$)
antiinfectives for systemic use (J01)	68%	penicillins
antiinflammatory and antirheumatic products (M01)	46%	ibuprofen
analgesics (N02)	45%	paracetamol, tramadol
agents acting on the renin-angiotensin system (C09)	32%	ACE inhibitors (plain), Angiotensin II receptor blockers (plain)
lipid modifying agents (C10)	30%	HMG CoA reductase inhibitors
drugs for acid related disorders (A02)	29%	proton pump inhibitors
ophthalmologicals (S01)	28%	antibiotics
corticosteroids, dermatological preparations (D07)	28%	corticosteroids, (potent, group III)
psycholeptics (N05)	21%	benzodiazepine derivatives + related drugs
antithrombotic agents (B01)	21%	platelet aggregation inhibitors
diuretics (C03)	20%	thiazides and potassium in combination
calcium channel blockers (C08)	19%	dihydropyridine derivatives
antifungals for dermatological use (D01)	17%	imidazole and triazole derivatives
beta blocking agents (C07)	17%	beta blocking agents (selective)
psychoanaleptics (N06)	17%	selective serotonin reuptake inhibitors
drugs for obstructive airway diseases (R03)	16%	selective beta-2-adrenoreceptor agonists
cough and cold preparations (R05)	16%	-
sex hormones and modulators of the genital system (G03)	15%	estradiol

