



# **Does more education reduce the wage gap?**

## **An analysis of labour market outcomes for native and foreign PhD (*preliminary*)**

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# Background

- Mobility of tertiary educated >> mobility others over past 30 years;
- Widespread evidence of skill wastage: occupational downgrade, over-education.... regardless of selective immigration policies;
- Wastage attributed to L-demand (discrimination) and supply (host country language skills, no network...), and more recently to institutional features (e.g. occupational licensing) and
- imperfect transferability of human capital...

# Transferability of human capital

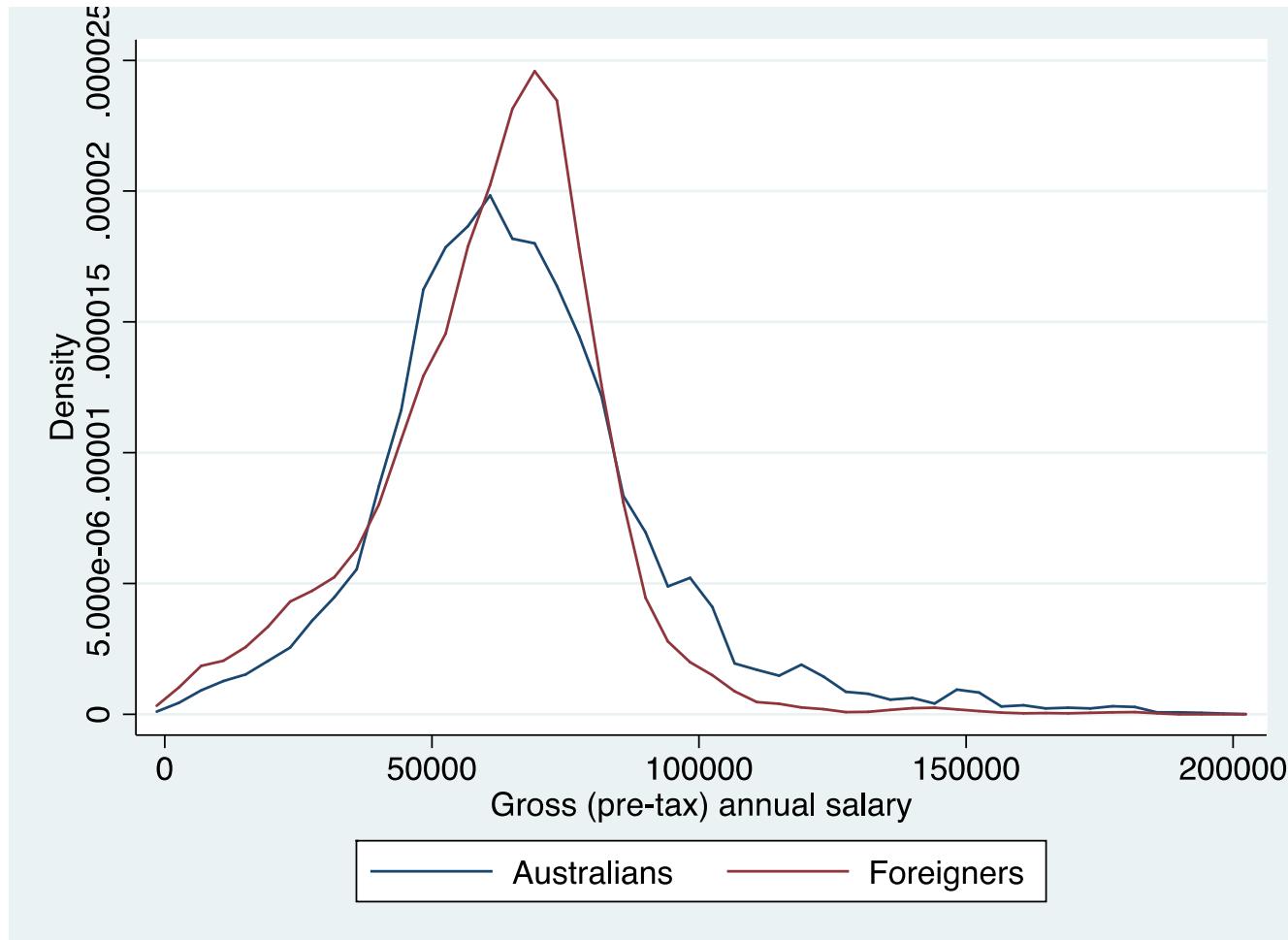
- Deemed imperfect but it conflicts with:
  - Migrants self-selection - they tend to be better able and motivated than stayers (Borjas, 1987, 1991)...
  - Migrants education – it is often higher than among stayers and natives (Carrington and Degatriache, 1998, Docquier, Ozden and Peri, 2014).....
  - Prominence of education and host country language skills in point awarded in selective immigration countries like Australia.
- Also difficult to test - hard to identify perceived quality of foreign schooling from migrants' information about host country labour market

→ Is transferability imperfect in extreme case of PhD education? (small contribution but also little research)

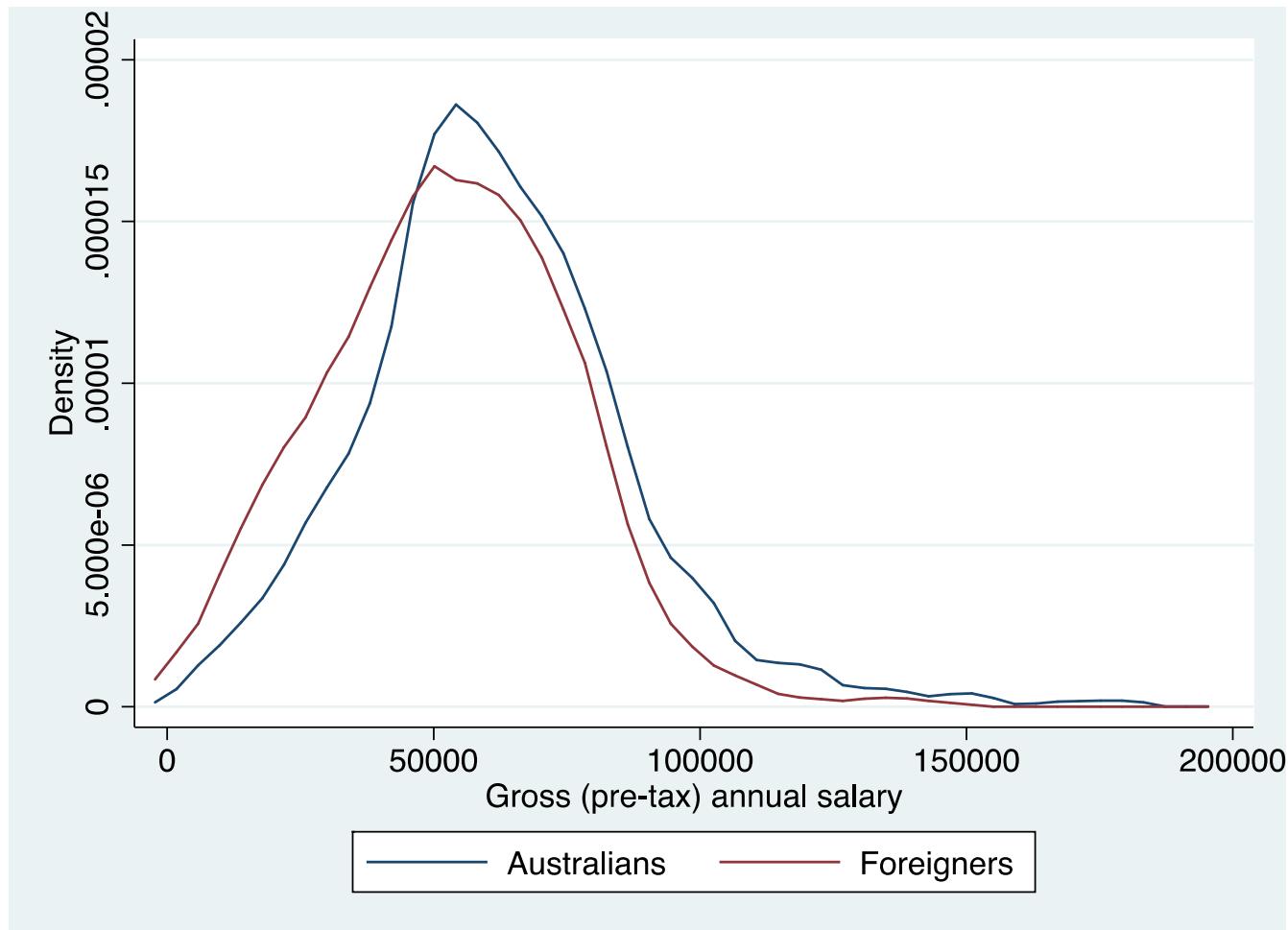
# This paper

- Focuses on PhD labour market in Australia
- Data from Graduate Destination Survey 1999-2015
- Focus on decomposition of native-foreign wage difference

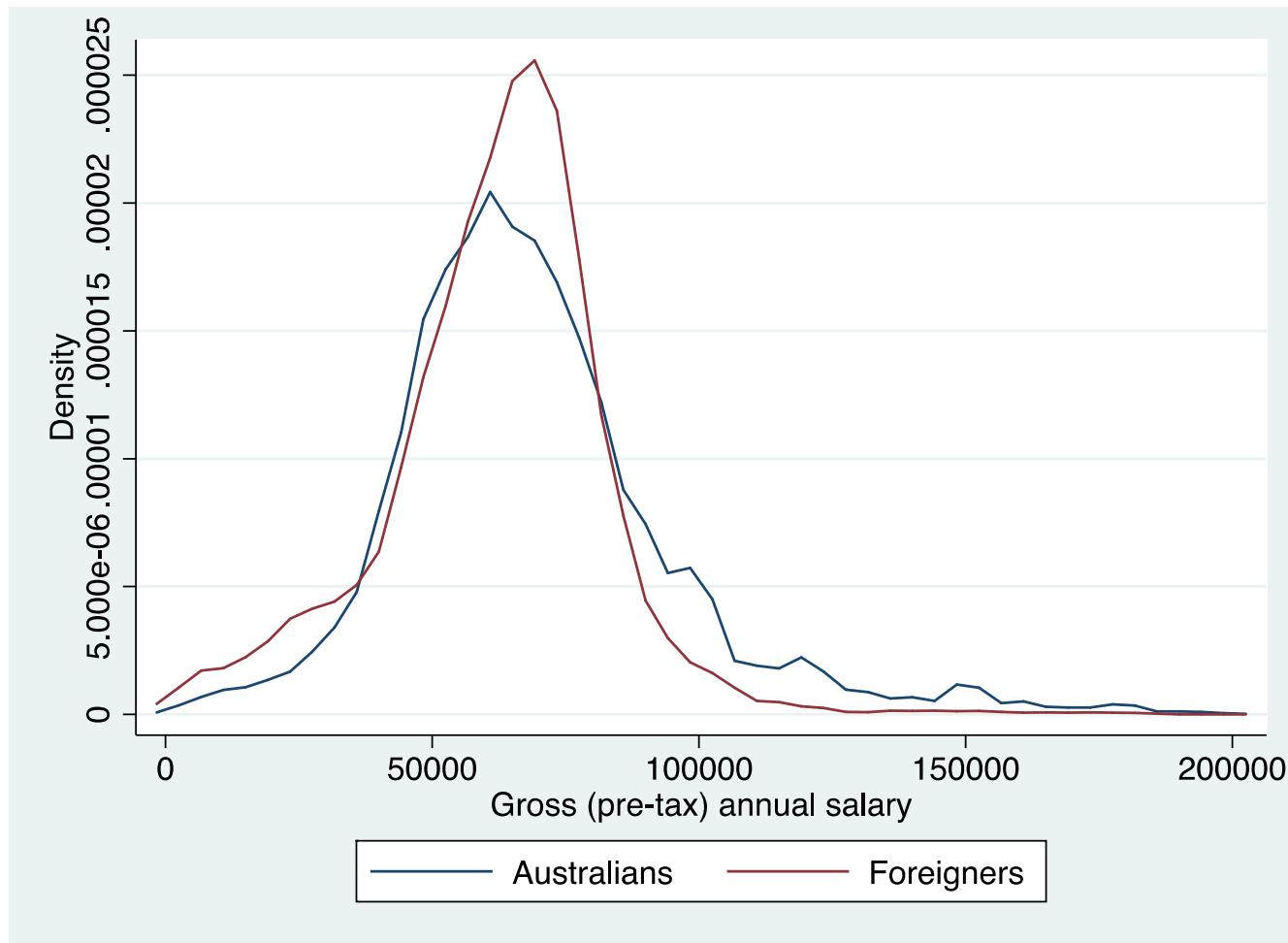
# Salary by residence: working in Australia



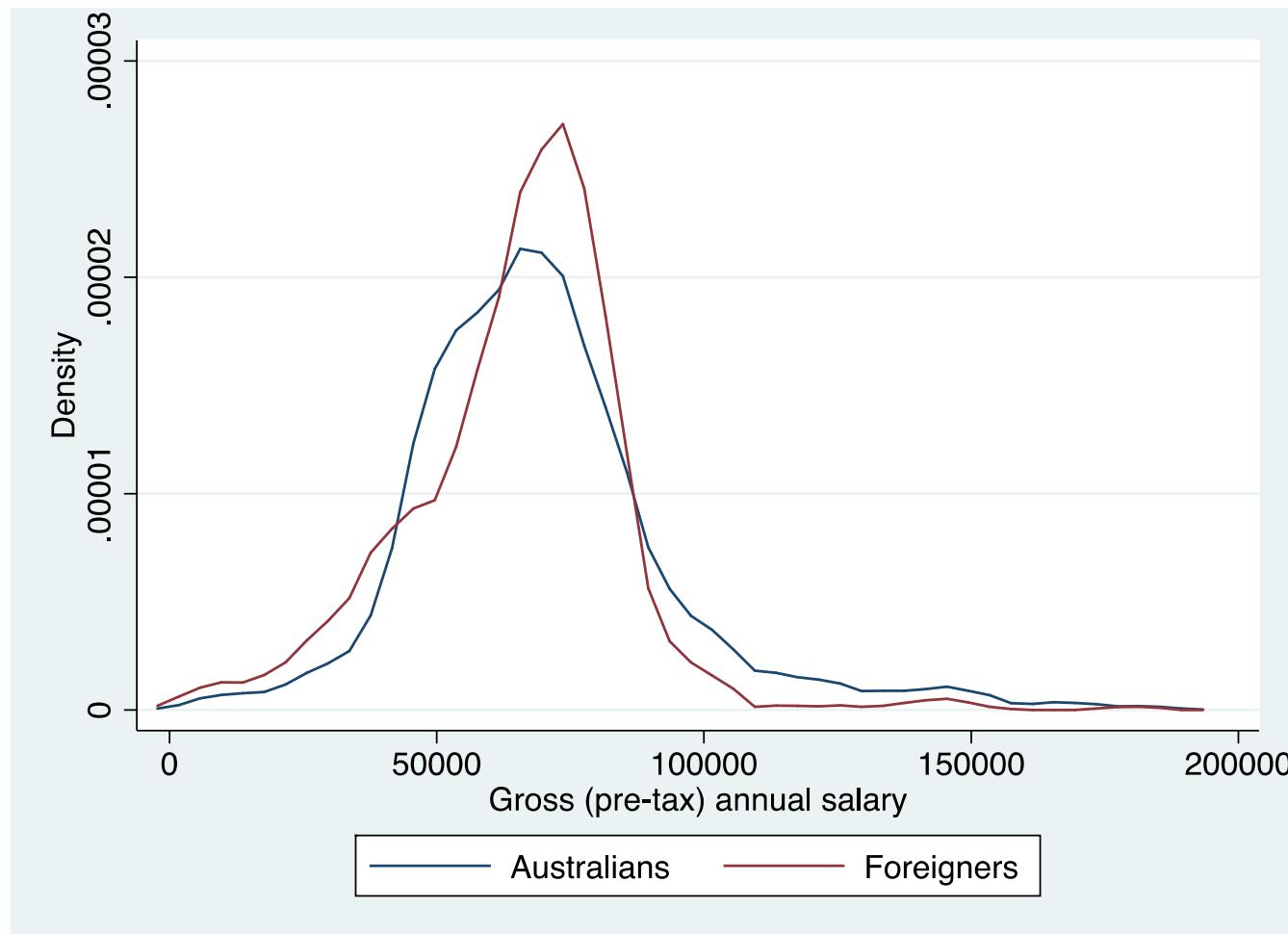
# STEM



# Humanities



# Medicine



# Literature

- Features of PhD programme, occupation, and trends: on-the-job training for research (Mangematin and Mangran, 1998), international labour demand dominated by universities and research centres (Auriol, 2010; Levin, 1996). Rising volumes of graduates (Cyranoski et al, 2011), temporary work (Stephan and Ma, 2005), private sector employment, and mismatch (Di Paolo, 2014) now a feature;
- Determinants of graduates' mobility: work in prestigious institution more important than pay or skill use (De Grip et al, 2010). Return home related to family not work (Franzoni et al, 2012);

# Literature

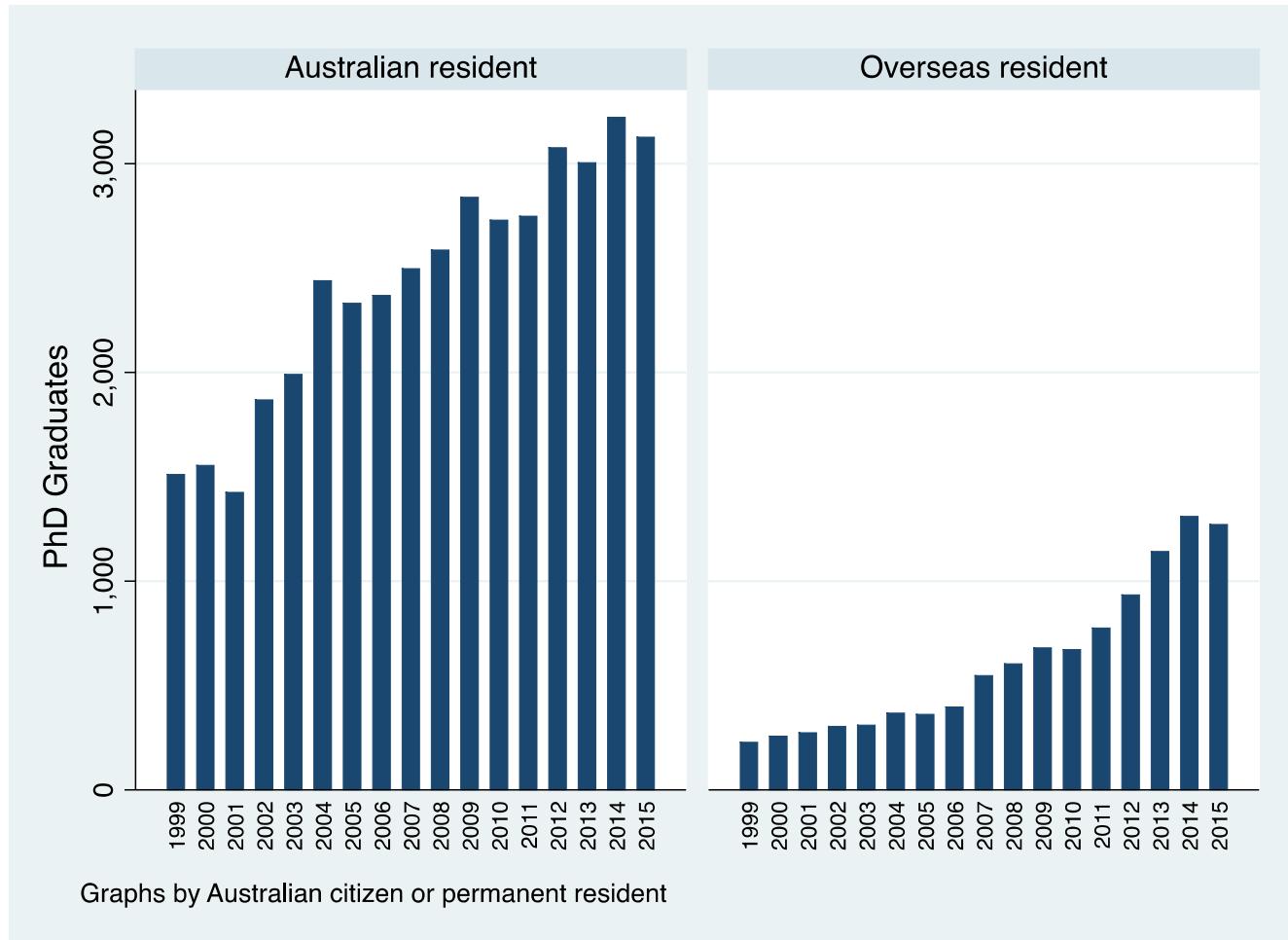
- Labour market outcomes: raise innovation activity via patents and start-ups (Gauthier-Loiselle, 2010), raise international collaboration and productivity at individual (Jonkers et al, 2013), institutional (Carillo et al, 2013) and geographical level (Freeman, 2014).
- Language is barrier to managerial positions, especially in the Sciences (Hunt, 2013). Evidence of over-education also in Sciences (Engineering - Hunt, 2013).
- No research on Australia: some descriptives of graduates' preference for university-industry partnerships (Hartman, 2002)

# Data

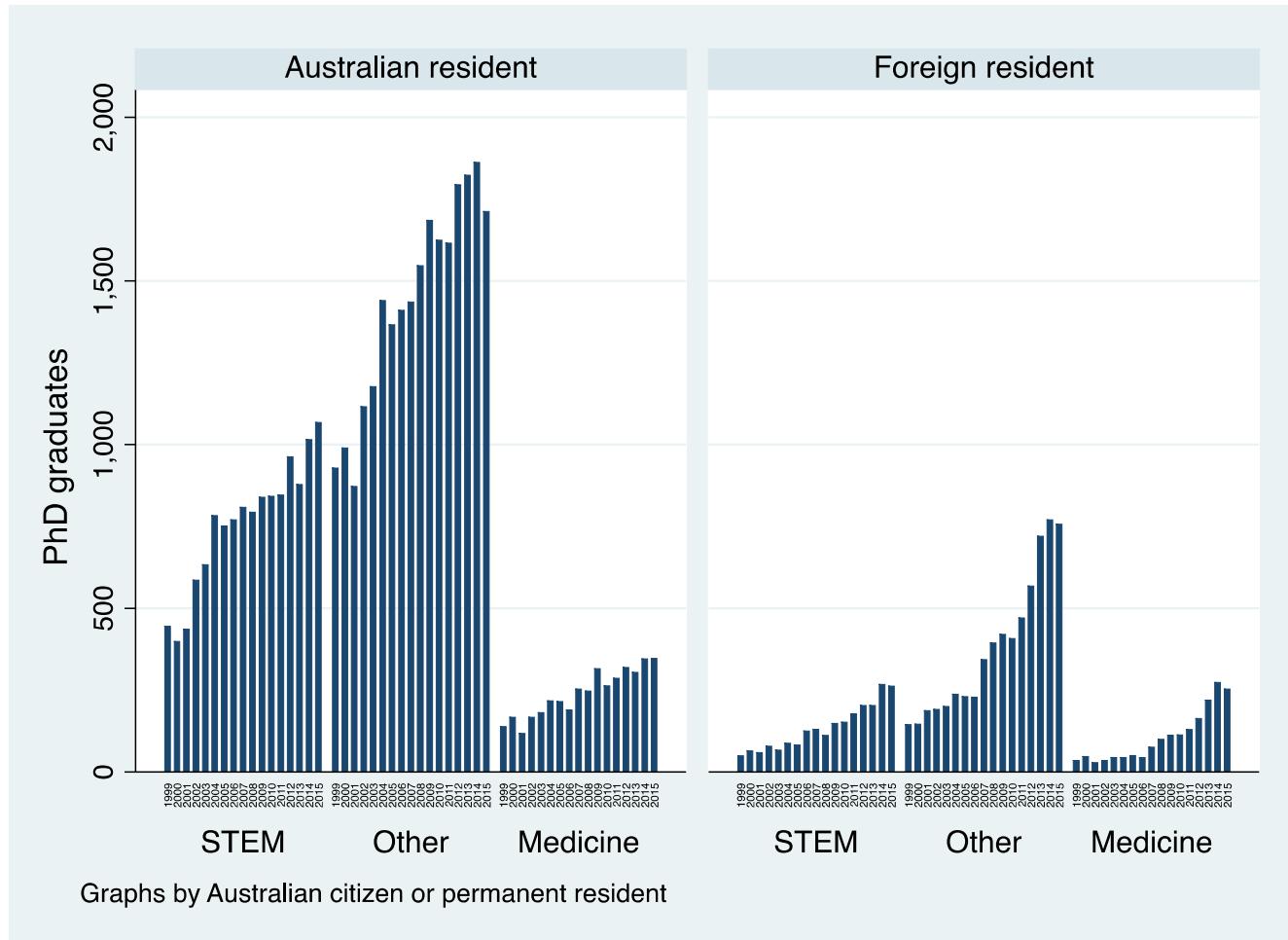
## Graduate Destination Survey

- annual survey asking questions on supervision, labour market outcomes, employer location, and basic demographics (e.g. no marital status, number kids...);
- Pooled cross section across time with good coverage ~ 50-60% graduates;
- Limitations: limited demographic information, no data on dropouts, no information on GPA
- Sample: 16 rounds (1999-2015): 51,761 observations → work in Australia (35,602) with no missing data on salary/hours (29,304): 91.2% Australians (26,678) and 8.8 foreign resident (2,586)

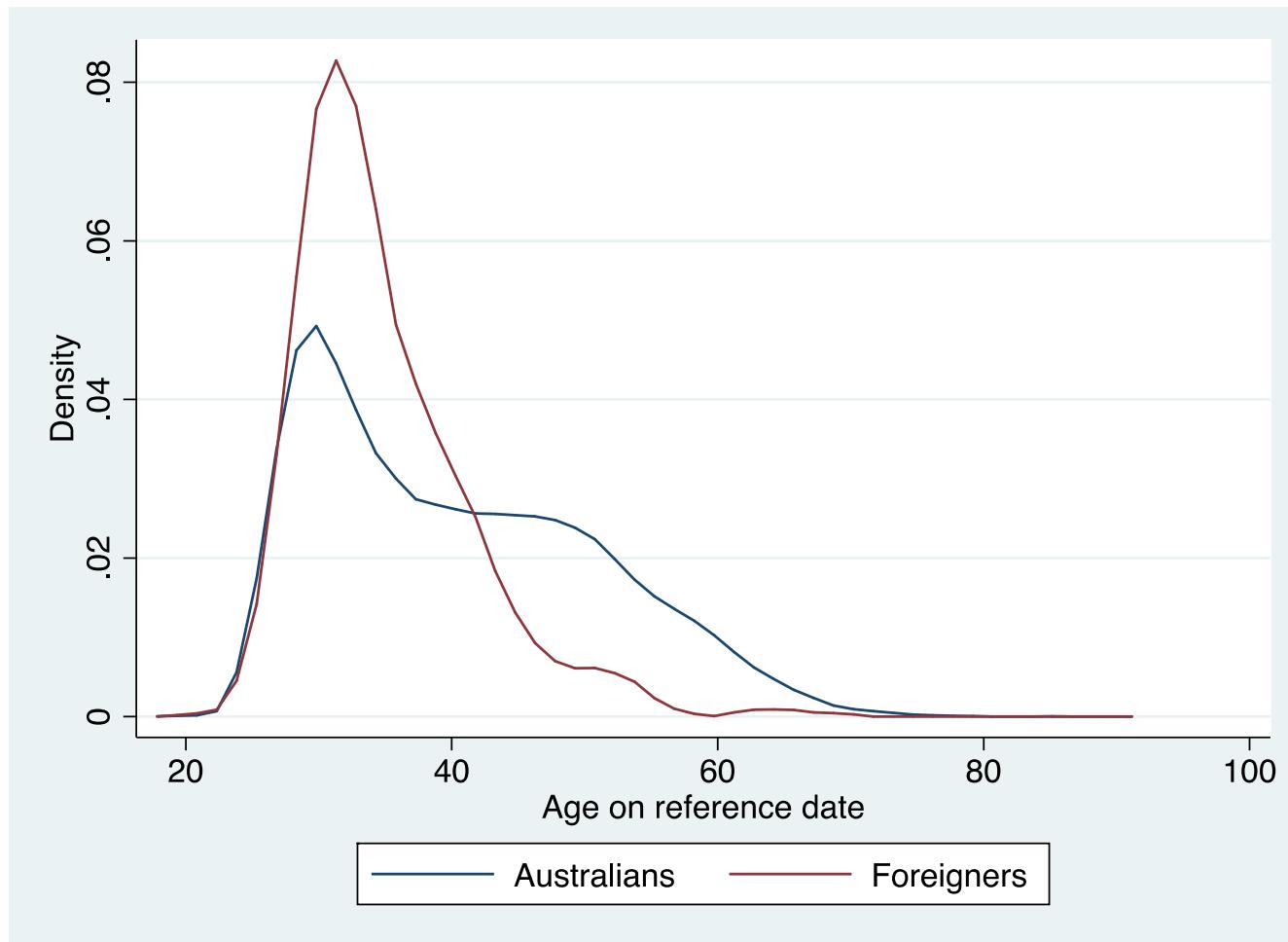
# Rising numbers of PhD graduates



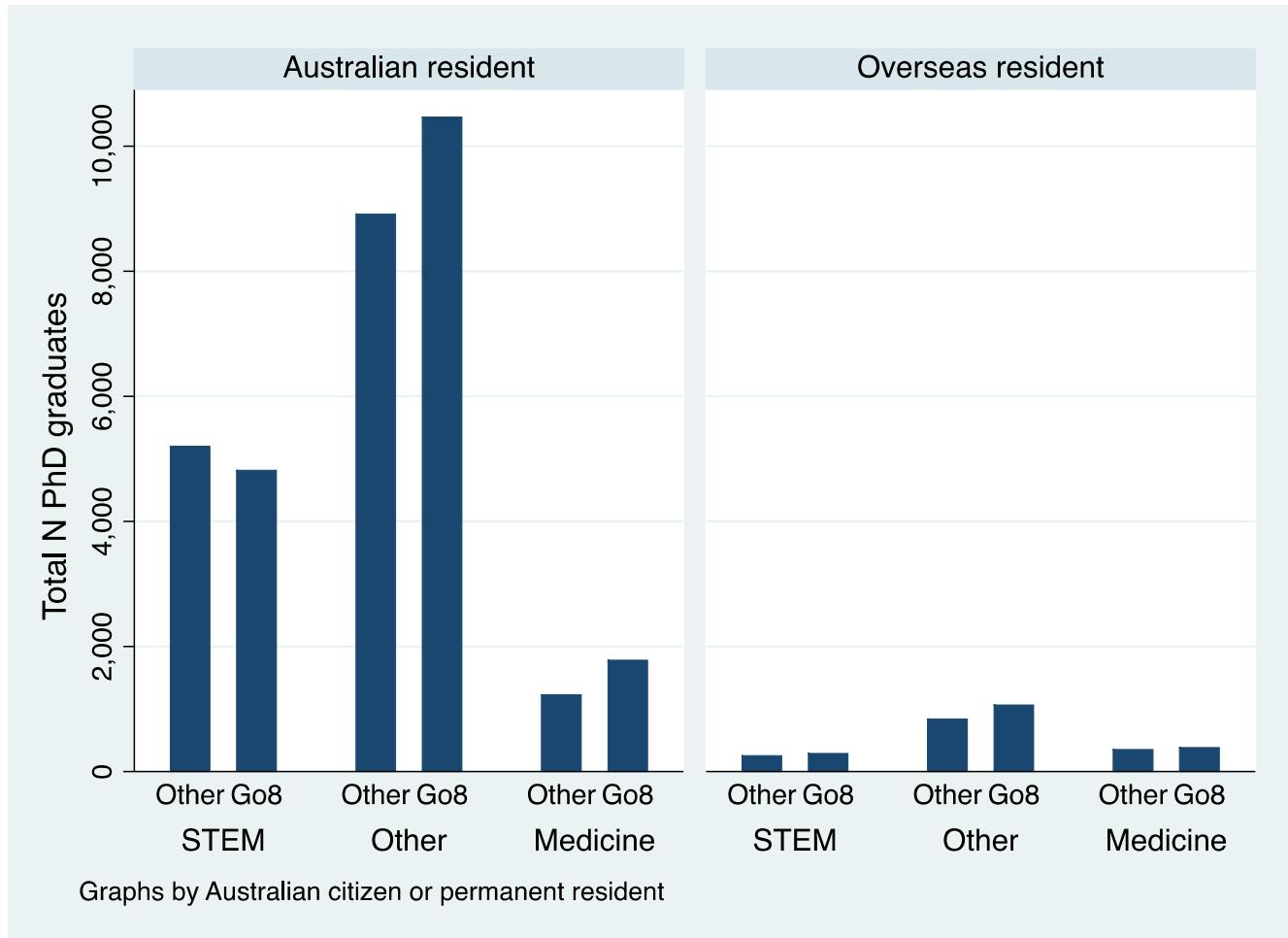
# Rise across all disciplines



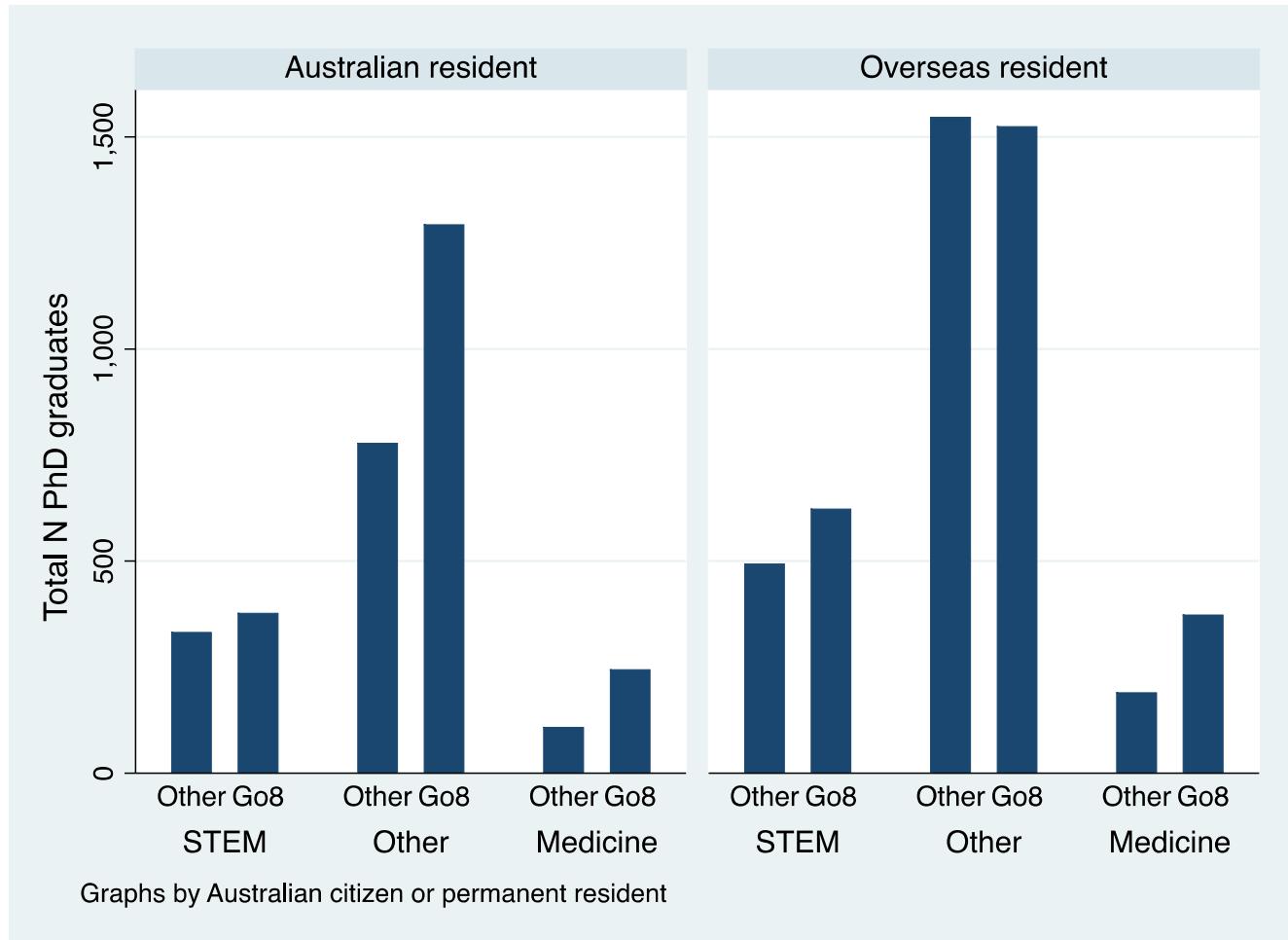
# An Australian anomaly: age at completion



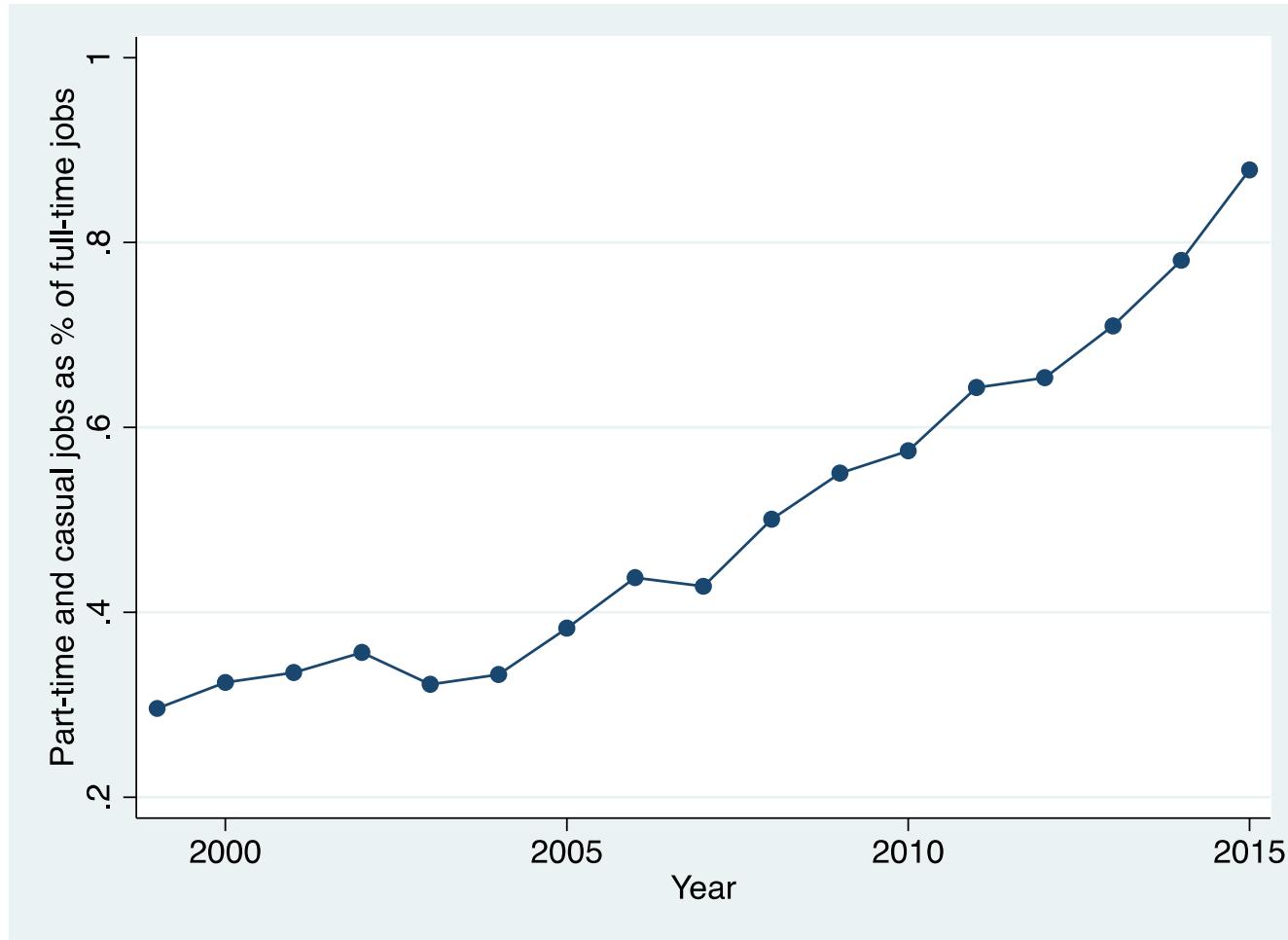
# Who stays and who goes: work in Australia



# Who stays and who goes: work abroad



# Temporary jobs on the rise



# Unconditional means

Variable	Natives	Foreigners	Difference
Unemployed	.061 (.239)	.103 (.305)	-.042***
Works in Australia	.912 (.283)	.402 (.490)	.510***
<b><i>Working in Australia</i></b>			
Wage (annual)	\$65,925 (\$47,072)	\$56,135 (\$53,119)	\$9,790***
Ln hourly wage	3.437 (.652)	3.317 (.910)	.120***
Age	38.98 (10.49)	33.60 (5.98)	5.38***
Females (share)	.530 (.499)	.372 (.483)	.158***
Years of enrolment	6.16 (2.33)	4.82 (1.32)	1.34***
NESB – English at home	.818 (.386)	.254 (.436)	.564***
Go8 university	.526 (.499)	.545 (.498)	-.019**
Field of study: STEM	.309 (.462)	.171 (.376)	.138***
Humanities	.598 (.490)	.598 (.491)	-.000
Medicine	.093 (.290)	.231 (.423)	-.138***
Worked in last year	.860 (.347)	.672 (.470)	.188***
Employer public sector	.678 (.467)	.660 (.473)	.018**

# Some notable features

- Little difference in employment rate
- Large difference in emigration/return rate
- Females are majority among natives
- Almost 20% natives do not speak English at home
- Humanities most common overall (60%). Foreigners prefer Medicine to STEM (23% vs 17%). Opposite among natives
- Temporary job for reaching almost 50%. Rise exponential since GFC (2007-8)

# Methodology

- Oaxaca-Blinder decomposition at the mean (Stata oaxaca), and
- As second stage on quantiles (25<sup>th</sup>, 50<sup>th</sup>, 75<sup>th</sup>) using Recentred Influence Function at quantile of interest (Firpo et al, 2009; Fortin et al, 2011): use 25<sup>th</sup>, 50<sup>th</sup>, 75<sup>th</sup> quantile – Stata rifreg then oaxaca

# Some choices

- Ln hourly wage as dependent variable
- Covariates:
  - Demographics: gender, age and age squared, speaks English at home, if disabled or from an aboriginal background;
  - Education: Go8, % foreign students in the same field of study and university, mode of attendance;
  - Labour market: lagged average wage and lagged unemployment rate by year and field of education.
  - Fixed effects: year, state of employment.
- Adjustment to control for selection on emigration decision (IMR): work in last year of study and years since enrollment as exclusion restrictions

# Results at mean – all ages and jobs

**Table 2 – Baseline results Oaxaca-Blinder decomposition at the mean: all observations**

	OLS	Adjusted Pooled OLS	STEM	Humanities	Medicine
Difference in ln hourly wage: $\Delta_t$	.091*** (.028)	.091*** (.026)	.0001 (.074)	.192*** (.040)	-.025 (.057)
Nr observations	29,304	29,304	8,724	17,568	3,012
<i>Decomposition:</i>					
Explained (E)	-.024 (.044)	-.0005 (.009)	-.006 (.021)	.009 (.010)	.008 (.016)
Unexplained (U)	.096*** (.028)	.096*** (.028)	.0006 (.077)	.184*** (.039)	-.033 (.061)
Interaction	.019 (.045)	-	-	-	-
<i>E contributors:</i>					
Gender	-.011* (.006)	-.014*** (.002)	-.004** (.002)	-.014*** (.001)	-.001 (.002)
Age	-.105 (.135)	.212*** (.021)	.190*** (.033)	.232*** (.029)	.002 (.004)
NESB	.068*** (.019)	.032*** (.008)	.020 (.016)	.037*** (.007)	.022** (.010)
Go8	-.003 (.002)	-.002** (.0007)	.0007 (.0008)	-.002** (.0008)	.002 (.001)
Years of enrolment	-.004 (.017)	-.005 (.003)	.0002 (.003)	-.009*** (.003)	.017*** (.007)

# Results at mean: age<=45 and full time job

**Table 3 – Results Oaxaca-Blinder decomposition at the mean (age <=45 and FT only)**

	OLS	Adjusted Pooled OLS	STEM	Humanities	Medicine
Difference in ln hourly wage: $\Delta_t$	-.031 (.047)	-.031 (.045)	-.136 (.170)	.054 (.049)	-.069 (.043)
Nr observations	13,813	13,813	3,584	8,419	1,810
<i>Decomposition:</i>					
Explained (E)	-.083* (.048)	.033*** (.011)	-.019 (.036)	-.017 (.013)	-.036 (.028)
Unexplained (U)	.0007 (.049)	.002 (.045)	-.117 (.166)	.071 (.045)	-.032 (.053)
Interaction	.051 (.049)	-	-	-	-
<i>E contributors:</i>					
Gender	-.009 (.008)	-.009*** (.002)	.00004 (.002)	-.010*** (.002)	-.002 (.003)
Age	.058 (.097)	.079*** (.019)	.014 (.021)	.101*** (.025)	.029 (.029)
NESB	.042 (.027)	.020*** (.008)	.013 (.019)	.029*** (.010)	.011 (.015)
Go8	-.001 (.001)	-.0009 (.0009)	.0004 (.002)	-.0002 (.002)	.001 (.002)
Years of enrolment	-.023 (.023)	-.007** (.003)	.002 (.006)	-.013*** (.003)	.018 (.011)
..	..	..	..	..	..

# Results at mean: age<=45 and part time job

**Table 4 – Results Oaxaca-Blinder decomposition at the mean (age <=45 and PT only)**

	OLS	Adjusted Pooled OLS	STEM	Humanities	Medicine
Difference in ln hourly wage: $\Delta_t$	.109*** (.037)	.109*** (.040)	.118 (.111)	.165*** (.048)	-.014 (.093)
Nr observations	8,163	8,163	2,394	4,840	929
<b>Decomposition:</b>					
Explained (E)	-.027 (.045)	-.033*** (.013)	-.064** (.031)	-.025 (.016)	-.011 (.029)
Unexplained (U)	.146*** (.038)	.143*** (.043)	.182 (.118)	.191*** (.050)	-.003 (.096)
Interaction	-.009 (.046)	-	-	-	-
<b>E contributors:</b>					
Gender	-.009 (.010)	-.011*** (.003)	-.003 (.003)	-.015*** (.002)	-.0004 (.002)
Age	.028 (.025)	-.003 (.006)	-.018 (.031)	-.001 (.016)	-.021 (.044)
NESB	.058 (.036)	.041*** (.008)	.005 (.023)	.049*** (.013)	.017 (.014)
Go8	.004 (.004)	.0007 (.0010)	-.001 (.002)	.003* (.001)	-.0004 (.003)
Years of enrolment	.008 (.017)	-.007 (.004)	-.014** (.007)	-.006 (.006)	.023* (.010)

# Results along distribution – all ages and jobs

Table 5 – Main results Oaxaca-Blinder decomposition along the wage distribution

	OLS - Pooled			STEM			Humanities			Medicine		
	25q	50q	75q	25q	50q	75q	25q	50q	75q	25q	50q	75q
$\Delta_t, \Delta_t(\tau)$ adjusted	-.009 (.024)	-.004 (.015)	.041*** (.012)	.045 (.045)	.024 (.041)	.059 (.042)	.052* (.027)	.019 (.019)	.048*** (.017)	-.059* (.033)	-.048** (.023)	.002 (.020)
<i>Decomposition:</i>												
Explained (E)	-.052*** (.005)	-.051*** (.005)	.003 (.006)	-.028* (.015)	-.028* (.015)	-.003 (.012)	-.041*** (.007)	-.038*** (.007)	.019* (.010)	-.071*** (.013)	-.055*** (.011)	-.019 (.013)
Unexplained (U)	.043* (.024)	.046*** (.016)	.038** (.015)	.073* (.045)	.052 (.042)	.062 (.043)	.094*** (.028)	.058*** (.020)	.030 (.019)	.012 (.035)	.007 (.022)	.021 (.020)
<i>Contribution to E</i>												
Gender	-.008*** (.001)	-.008*** (.001)	-.009*** (.001)	-.001 (.001)	-.0008 (.0008)	-.002 (.001)	-.008*** (.001)	-.009*** (.001)	-.009*** (.002)	-.001 (.001)	-.002* (.001)	-.003** (.002)
Age	.162*** (.014)	.195*** (.012)	.205*** (.014)	.168*** (.027)	.173*** (.027)	.127*** (.019)	.177*** (.016)	.206*** (.016)	.229*** (.018)	.047** (.020)	.029** (.015)	.059*** (.016)
NESB	.015*** (.003)	.012*** (.003)	.021*** (.004)	.011 (.011)	.013 (.009)	.006 (.009)	.022*** (.005)	.016*** (.004)	.027*** (.005)	.016*** (.006)	.008* (.005)	.014** (.006)
Go8	-.0005** (.0002)	-.0005** (.0002)	-.0008** (.0003)	.0009 (.0009)	.001 (.0009)	-.0002 (.0008)	-.0008* (.0004)	-.0007** (.0003)	-.0009** (.0004)	.001 (.0009)	.0008 (.0006)	.0008 (.0009)
Years of enrolment	-.0005 (.002)	-.002 (.001)	-.003 (.003)	.003 (.003)	.005** (.002)	.0007 (.003)	-.004* (.003)	-.005** (.002)	-.006** (.002)	.003 (.006)	.006 (.006)	.008 (.005)

# Results along distribution: age <=45 and FT

**Table 6 – Main results Oaxaca-Blinder decomposition along the wage distribution (age <=45 and FT only)**

	OLS - Pooled			STEM			Humanities			Medicine		
	25q	50q	75q	25q	50q	75q	25q	50q	75q	25q	50q	75q
$\Delta_t, \Delta_t(\tau)$ adjusted	-.012 (.067)	.013 (.060)	.078 (.066)	.212 (.237)	.168 (.245)	.301 (.311)	-.037 (.093)	-.016 (.097)	.025 (.110)	-.046 (.121)	-.016 (.169)	.052 (.112)
<i>Decomposition:</i>												
Explained (E)	-.064*** (.014)	-.067*** (.020)	-.016 (.017)	.006 (.044)	-.014 (.040)	.010 (.043)	-.058** (.022)	-.054** (.022)	.012 (.022)	-.103*** (.033)	-.087** (.040)	-.058* (.035)
Unexplained (U)	.051 (.068)	.080 (.063)	.095 (.069)	.206 (.242)	.182 (.261)	.290 (.304)	.020 (.095)	.038 (.100)	.013 (.114)	.056 (.112)	.071 (.166)	.110 (.113)
<i>Contribution to E</i>												
Gender	-.010*** (.002)	-.010*** (.003)	-.010*** (.003)	-.003 (.004)	-.003 (.004)	-.007 (.005)	-.009*** (.002)	-.009*** (.002)	-.008*** (.003)	-.008 (.005)	-.008 (.005)	-.009* (.005)
Age	.054* (.031)	.082** (.035)	.093*** (.027)	.063 (.051)	.061 (.058)	.071 (.064)	.024 (.033)	.056 (.038)	.075** (.036)	.063 (.052)	.068 (.047)	.058 (.046)
NESB	.024** (.012)	.022* (.013)	.039*** (.013)	.027 (.030)	.022 (.030)	.030 (.034)	.036** (.016)	.037** (.015)	.063*** (.018)	-.011 (.021)	-.011 (.021)	-.005 (.020)
Go8	-.0004 (.0007)	-.0005 (.0006)	-.0007 (.0008)	-.002 (.002)	-.001 (.002)	-.003 (.003)	-.00007 (.0007)	-.00008 (.0007)	-.0001 (.001)	.004 (.004)	.004 (.004)	.004 (.004)
Years of enrolment	.002 (.004)	.002 (.005)	.002 (.005)	.003 (.009)	.008 (.011)	.008 (.010)	.005 (.005)	.003 (.006)	.003 (.006)	-.024* (.010)	-.014 (.015)	-.011 (.013)

# Summary of results at mean

- Natives enjoy higher hourly wages than foreigners, but this is driven by Humanities:
  - Effect mostly due to unexplained.
  - Among explained component effect of age and NESB (up) compensated by gender (down)
  - Similar implications to results at mean: maybe difficult to find objective productivity measures as signal from university (Go8) matters
  - Effect for STEM and Medicine arise only at 10 and 90 quantile.

# Summary of results along distribution

- Natives have similar hourly wages to foreigners, aside from Humanities:
  - Gap wider at higher hourly wage level
  - Unexplained component predominant
  - Among explained component effect of NESB (up) always present (age too but only if include the 45+), but reduced by gender (down)
  - Maybe difficult to find objective productivity measures as signal from university (Go8) matters
  - Result characterises both full-time and part-time job for under-45
  - Implies top quality candidates face no discrimination but average quality candidates are subjected to it (controversial?).

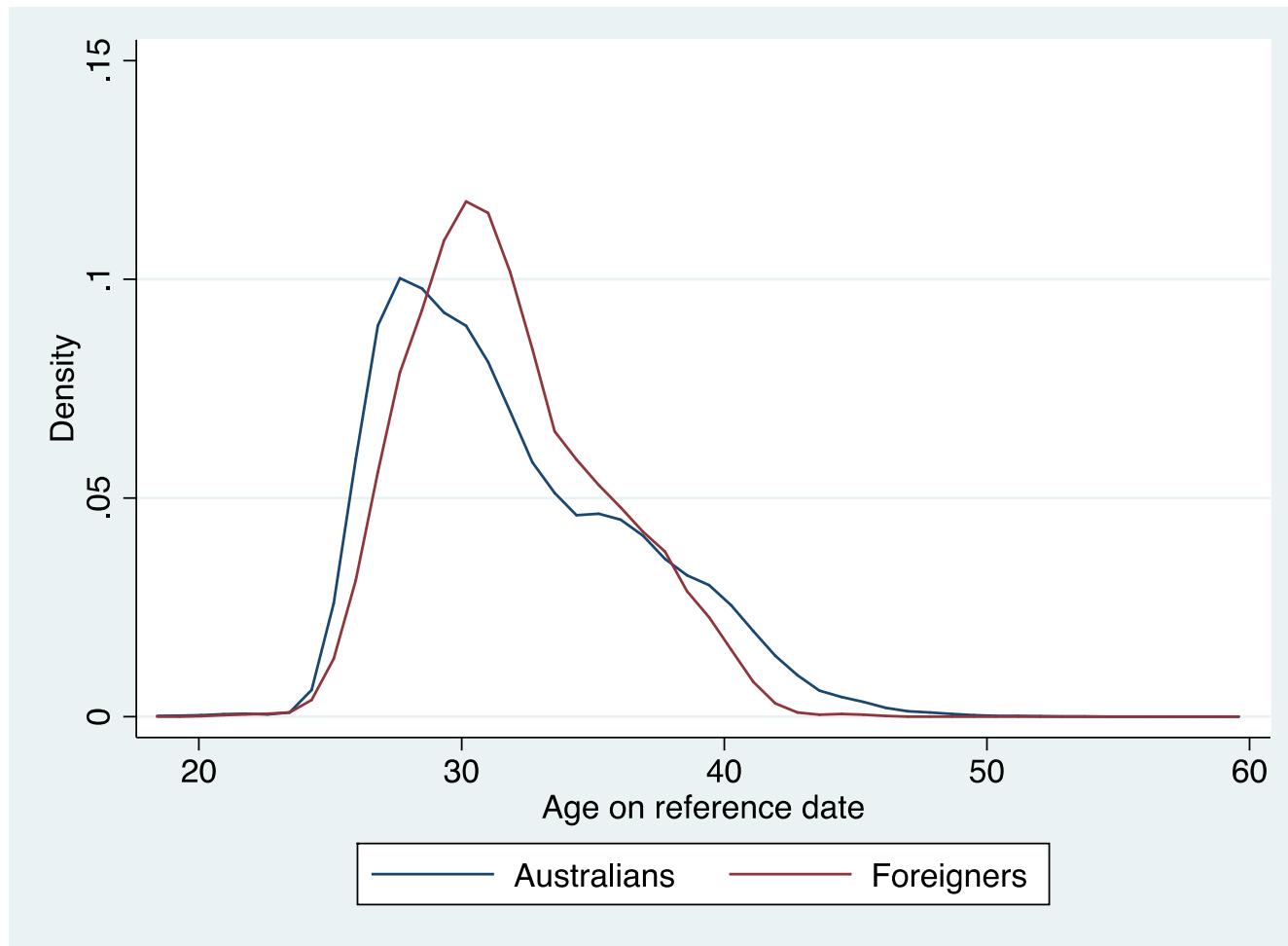
# Implications (preliminary)

- No evidence of wage gap in PhD trained and working in Australia aside from Humanities;
- Gender effect intriguing: why males get penalised?
- NESB: language matters but probably also includes culture, habits, beliefs → “if speaks with accent then thinks and does with accent”
- Lack of difference reassuring but growing lack of permanent jobs since GFC maybe is not → competition on observables where they exist (STEM, Medicine) but less in Humanities
- Other evidence point to natives moving to occupations with comparative advantage (language/culture). Results from graduate labour market reflect this.

# Q&A

Thanks for your attention and feedback

# Age distribution for age start < 36



# Hourly salary of foreigners staying or returning

