

# Working on Wall Street or Relaxing on the Riviera? Age-related impacts of income and wellbeing on regional migration

Arthur Grimes & Judd Ormsby

Pathways Conference, Wellington, July 2015

*Research funded by Royal Society of NZ Marsden Fund, &  
Resilient Urban Futures programme (MBIE)*

Understand relationship between internal migration, age, subjective wellbeing & income.

1. Does SWB of a place matter for migration?
2. Do incomes in a place matter for migration?
3. How do patterns change with age?
4. What impacts of education?
5. What impacts of the rate of time preference?



# ROAD MAP

Definition of subjective wellbeing

Theory and hypotheses

Data: HILDA

*Ex-ante* analysis

(Future extension: *ex-post* analysis)

Conclusions

# SUBJECTIVE WELLBEING

All things considered, how satisfied are you with your life? ... pick a number between 0 and 10.

Previous lit shows SWB correlates with:

- Income
- Blood pressure
- Friends' & colleagues' perceptions
- Objective measures of a place's well-being
- Sunshine

But does SWB matter for behaviour, especially migration?

# RESULTS PREVIEW:

- **People migrate to happy places**
  - A revealed preference validation of SWB
  - Younger people are most responsive to well-being (with an upturn again when old)
  - SWB more important for highly educated people (especially for young & old adults)
- **Income not (separately) significant**
  - But we do see declining coefficients over time
- **Distance matters (negatively)**
- **People less mobile as they age**
  - And more mobile with higher education

# (INTERNAL) MIGRATION MODEL

- Most migration models concentrate on impacts of income ( $Y$ ) rather than SWB [or utility]
- Glaeser et al. (2014) find some people move to 'unhappy cities'
  - they conclude that  $\text{SWB} \neq \text{Utility}$
  - but interpretation is not based on a life-cycle model
- Our theoretical life-cycle model shows:
  - location choice (& SWB) is a  $f^n$  of income & amenities across multiple periods
  - people make different migration choices faced with same external parameters (because of different rates of time preference and different income profiles)

# LIFE-CYCLE MODEL

(LOCATION A IS HIGH AMENITY/LOW WAGE; B IS HIGH WAGE/LOW AMENITY)

Location attributes:

Income  $y_A < y_B$

Amenities  $n_A > n_B$

Payoffs:

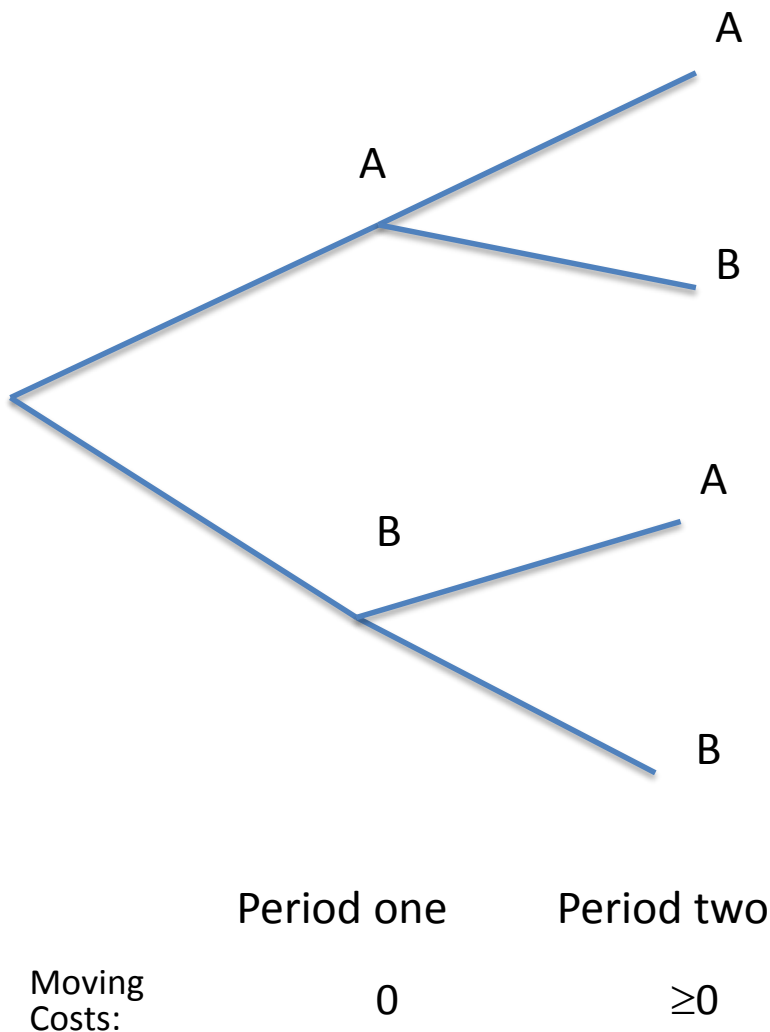
$$u_t = \log(c_t) + \log(n_t)$$

Problem:

$$\max U = u_1 + \frac{1}{1+\rho} u_2$$

s.t.

$$c_2 = y_2 + (1+r)(y_1 - c_1)$$



# KEY THEORETICAL RESULTS: STABLE INCOME PATH

(LOCATION A IS HIGH AMENITY/LOW WAGE; B IS HIGH WAGE/LOW AMENITY)

1. Interest rates matter: high  $r$  may enable  $(B, A)$  rather than  $(B, B)$
2. Moving costs may keep one in an otherwise less preferred location [e.g.  $(B, B)$  instead of  $(B, A)$ ]
3. Someone who lives for the present (surfer) may choose  $(A, B)$ 
  - and will have low  $U_2$  relative to  $U_1$
4. So Gold Coast may comprise young surfers & old savers





# THEORETICAL IMPLICATIONS

(LOCATION A IS HIGH AMENTY/LOW WAGE; B IS HIGH WAGE/LOW AMENITY)

1. Different individuals choose different location combinations
2. May move to or from high wage areas
3. May increase or decrease utility over time
4. Moving costs → greater immobility
  - may keep people either in  $(A, A)$  or  $(B, B)$
5. People with high rates of time preference move from high-amenity area in youth to high-wage area when older
6. Rising income path adds further complexity
  - Person may use higher future income to purchase more amenities when young, so educated may choose high amenities initially

# DATA: OVERVIEW

HILDA – unit record panel dataset.

$t = 12$  years (drop first year) [2002-2013]

Supplement with ABS Labour force stats.

## Sample:

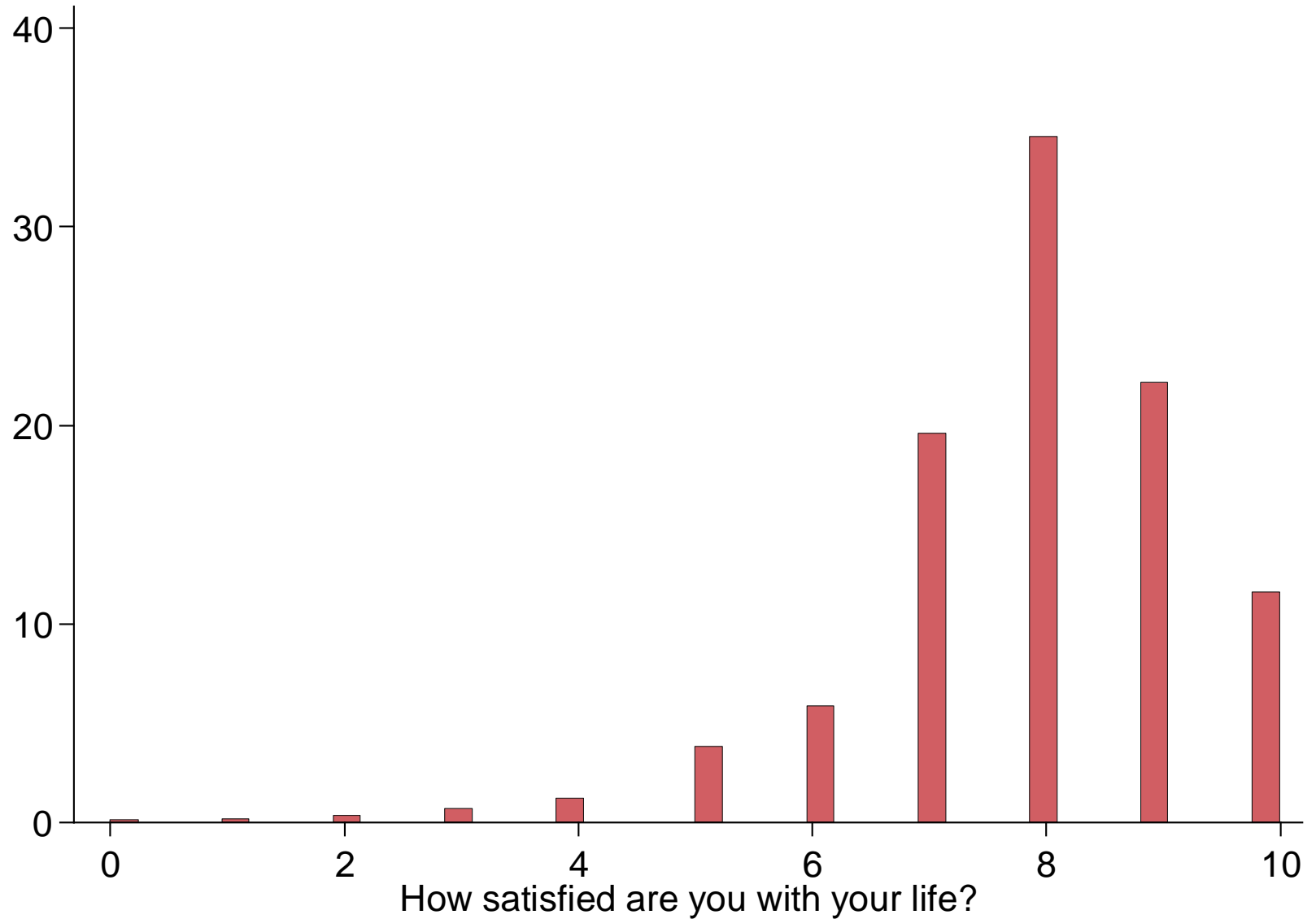
All people in HILDA (originally ~37,000) less those:

- Not in two consecutive waves (~7,000)
- Younger than 25 (another ~10,000)
- Temporary sample members (another ~2,000)
- In the defence force, missing attributes (<60 more)

~17,700 people & ~126,000 person-time observations.

Robustness check: Look only at (smaller) balanced panel

# SWB – INDIVIDUAL DISTRIBUTION



Wave 13

N = 17,493

Median = 8

Mean = 7.9

Sd = 1.45

## Estimate McFadden's choice model:

Probability of individual  $i$  choosing to live in region  $j$  depends on:

- individuals' characteristics,  $\mathbf{w}_i$ , and on:
- all regions' attributes in relation to individual  $i$ ,  $\mathbf{x}_{i,k}$ .

$$\text{Prob}(\text{Location}_{i,t} = \text{Region}_j) = f\left(\sum_k \mathbf{x}_{i,k,t-1}, \mathbf{w}_{i,t-1}\right)$$

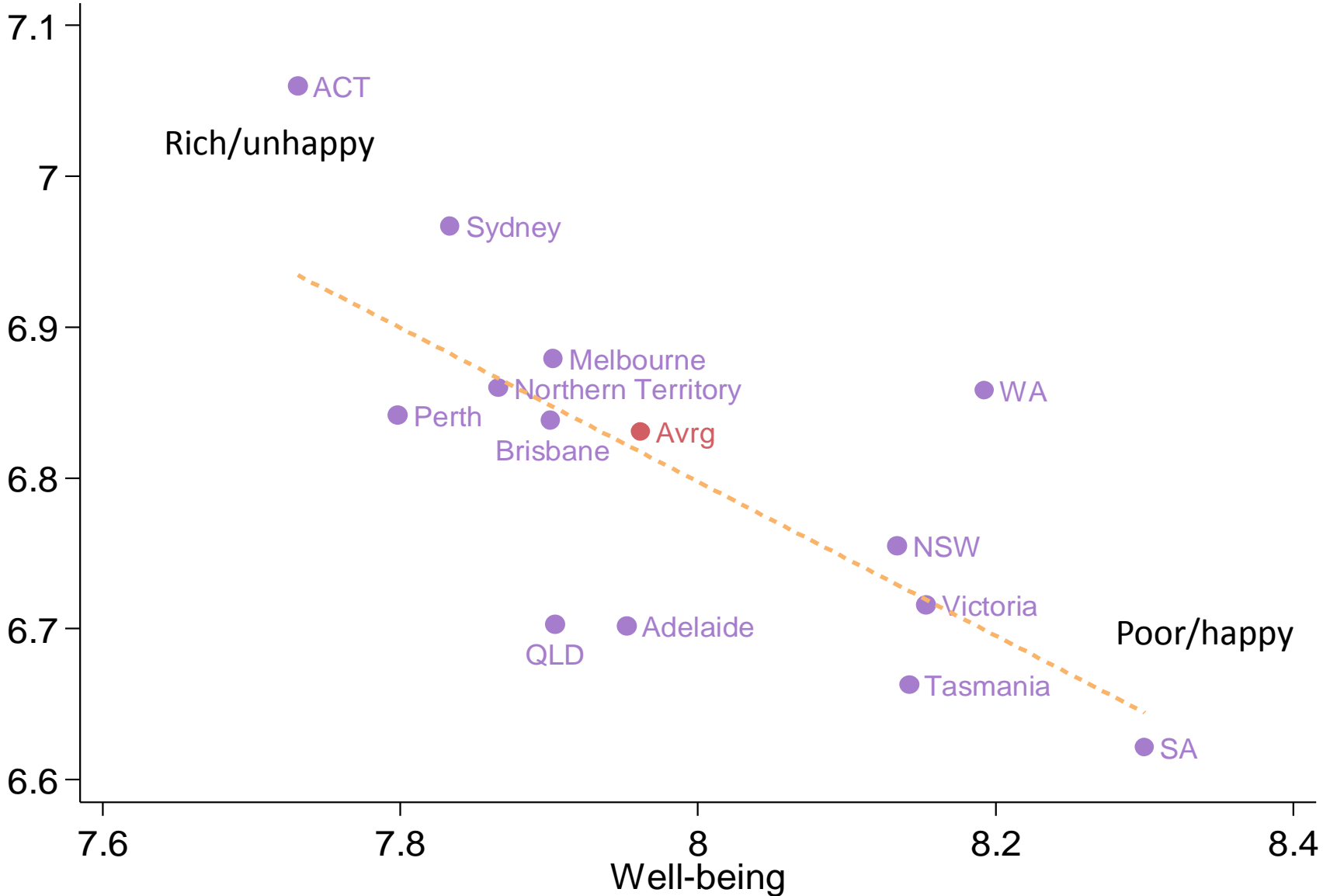
## Region attributes, $x_{i,k,t-1}$

Mean SWB, log mean income, log mean rent, currently live in region, current distance to region, log population, unemployment rate, employment rate, regional fixed effects

## Individual characteristics, $w_{i,t-1}$

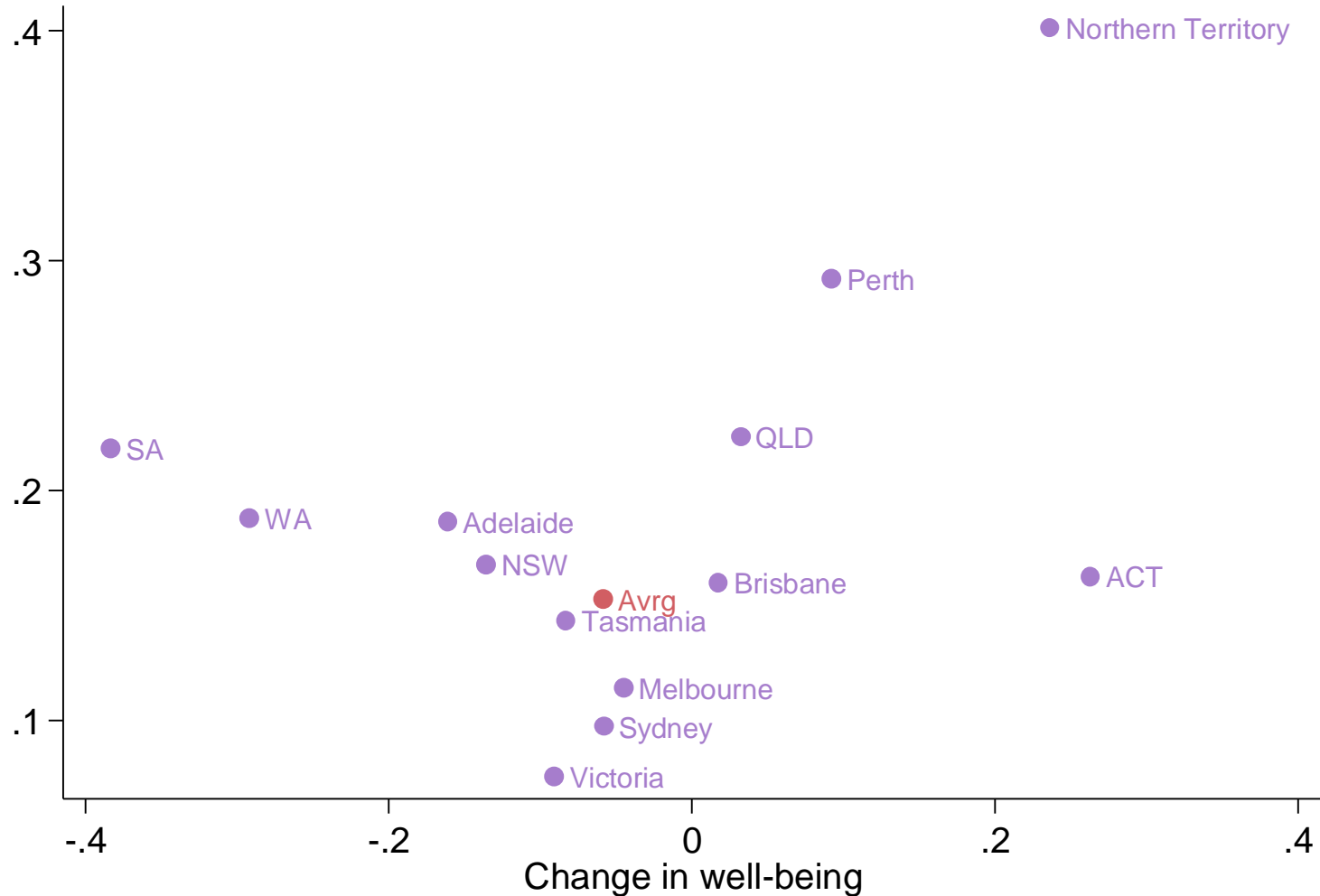
Different regional effects for different types of people according to: gender, ethnicity, education, age, family type, smoker

# MEAN LOG(Y) VS MEAN SWB (2001)



# LOG(Y) & SWB CHANGES (2001-2013)

Changes in average log income and well-being from 2001 to 2013



Coefficients are odds ratios (i.e. coeff=1  $\Rightarrow$  zero effect).

Dependent variable is location choice.

# BASE RESULTS

|   | income           | SWB              | comparison       | with controls    |
|---|------------------|------------------|------------------|------------------|
| <b>Region attributes:</b>                 |                  |                  |                  |                  |
| log (mean income)                         | 1.00             |                  | 0.90             | 0.80             |
| mean SWB                                  |                  | <b>1.63***</b>   | <b>1.63***</b>   | <b>1.50*</b>     |
| current region                            | <b>463.06***</b> | <b>463.36***</b> | <b>463.37***</b> | <b>201.18***</b> |
| distance                                  |                  |                  |                  | <b>.93***</b>    |
| log (population)                          |                  |                  |                  | <b>3.77**</b>    |
| employment rate                           |                  |                  |                  | 0.10             |
| unemployment rate                         |                  |                  |                  | 0.04             |
| log (mean rent)                           |                  |                  |                  | 0.85             |
| Regional intercepts                       | Yes              | Yes              | Yes              | Yes              |
| Personal Characteristics                  | No               | No               | No               | Yes              |
| <i>N</i>                                  | 1,615,445        | 1,615,445        | 1,615,445        | 1,615,445        |
| <i>Number of person-time observations</i> | 124,265          | 124,265          | 124,265          | 124,265          |
| <i>Number of people</i>                   | 17,347           | 17,347           | 17,347           | 17,347           |
| <i>Number of regions</i>                  | 13               | 13               | 13               | 13               |

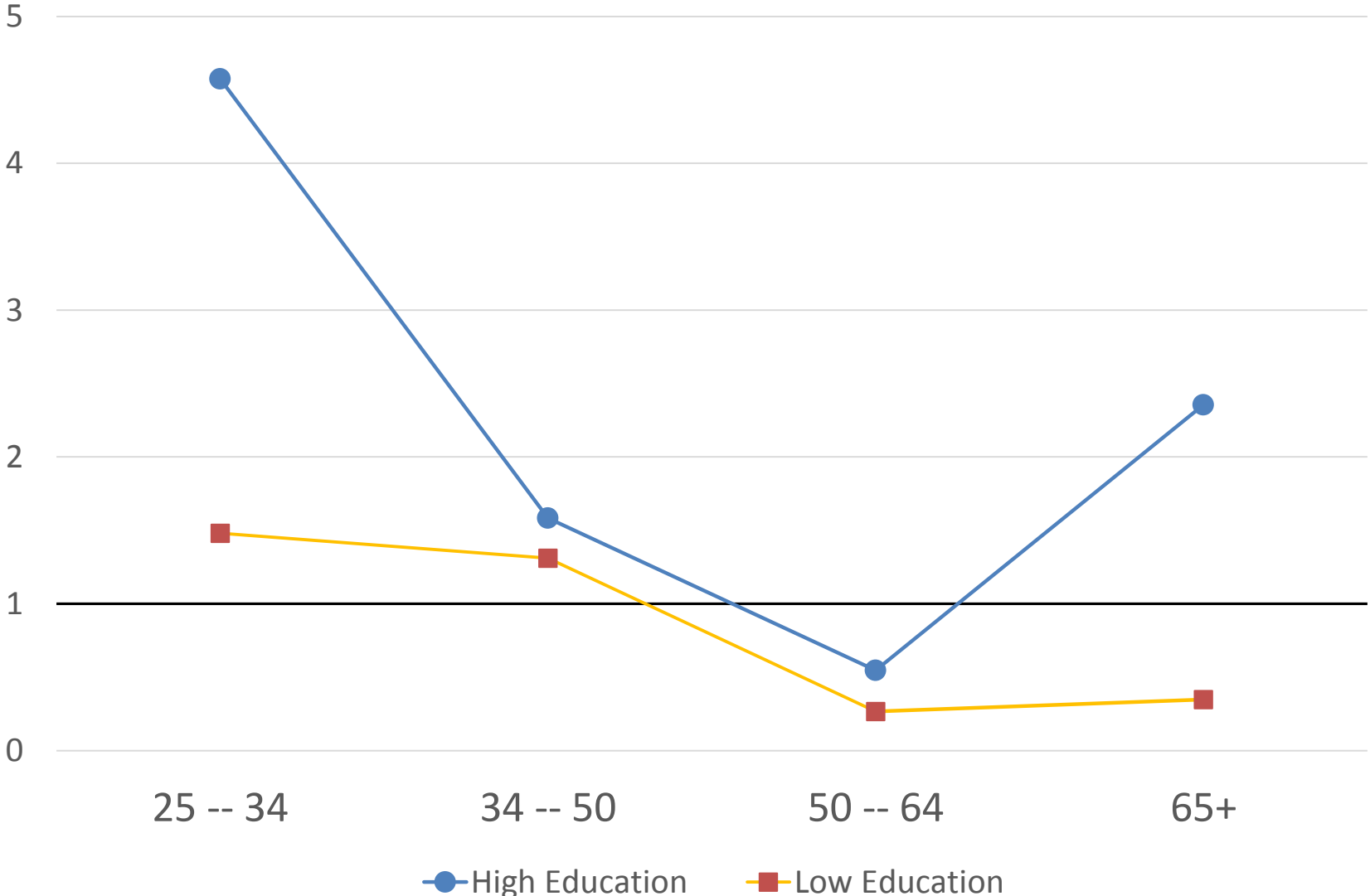
\* p<0.10, \*\* p<0.05, \*\*\* p<0.01. Standard errors clustered at person level.



# RESULTS BY AGE

Odds ratios

SBE coefficients by age for high and low education



# KEY CONCLUSIONS

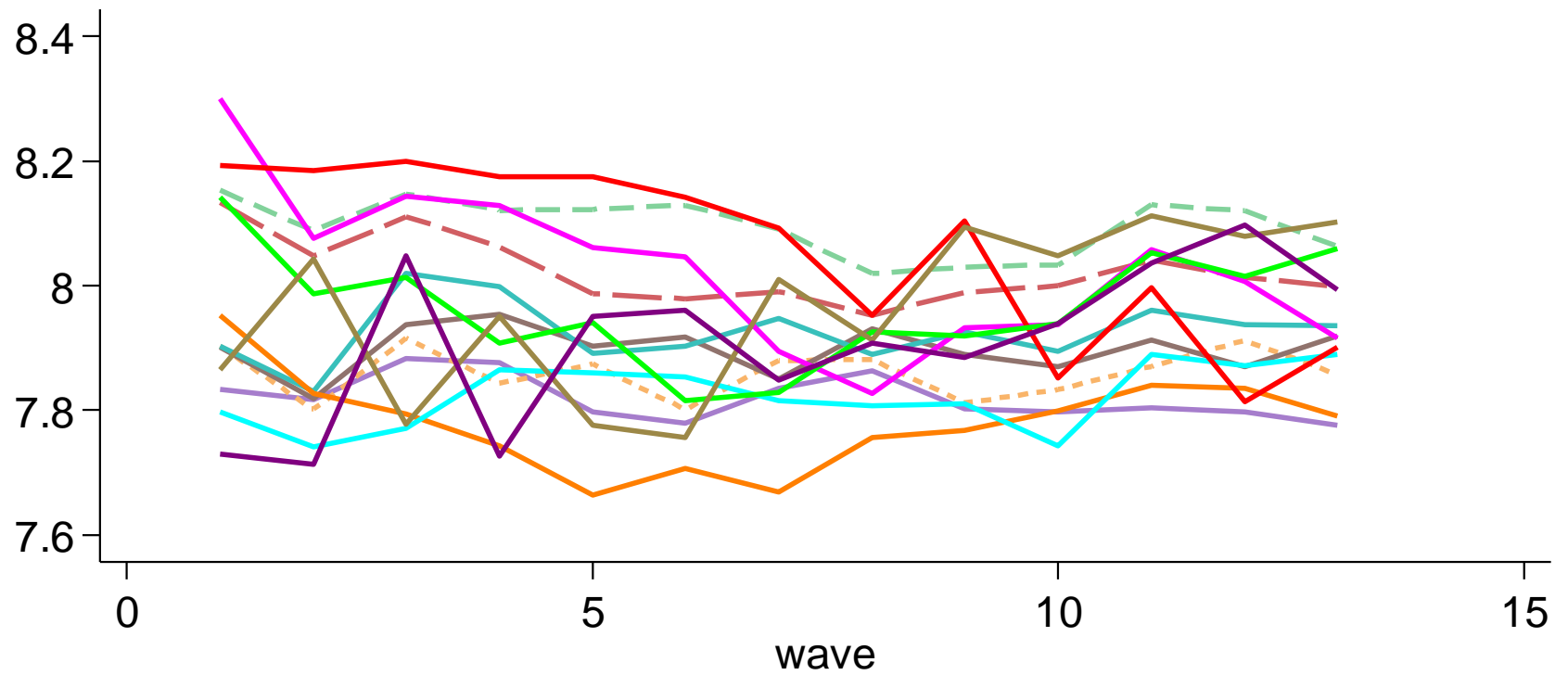
- SWB is a measure that matters to people
  - verified by revealed preference migration choices
- Income is not separately important
  - But distance is important
- Younger people more responsive to SWB than older people
- Tertiary educated people more responsive to SWB than less educated people
- Educated older people also tend to respond to SWB

# NEXT STEPS

- Include SWB & income for 'like people' in model
- Breakdowns by rate of time preference (& age)
  - Include SWB & income of older people as influences (interacted with time preference)
- Run *ex-post* analysis
  - Estimate what is the actual effect of migration on SWB



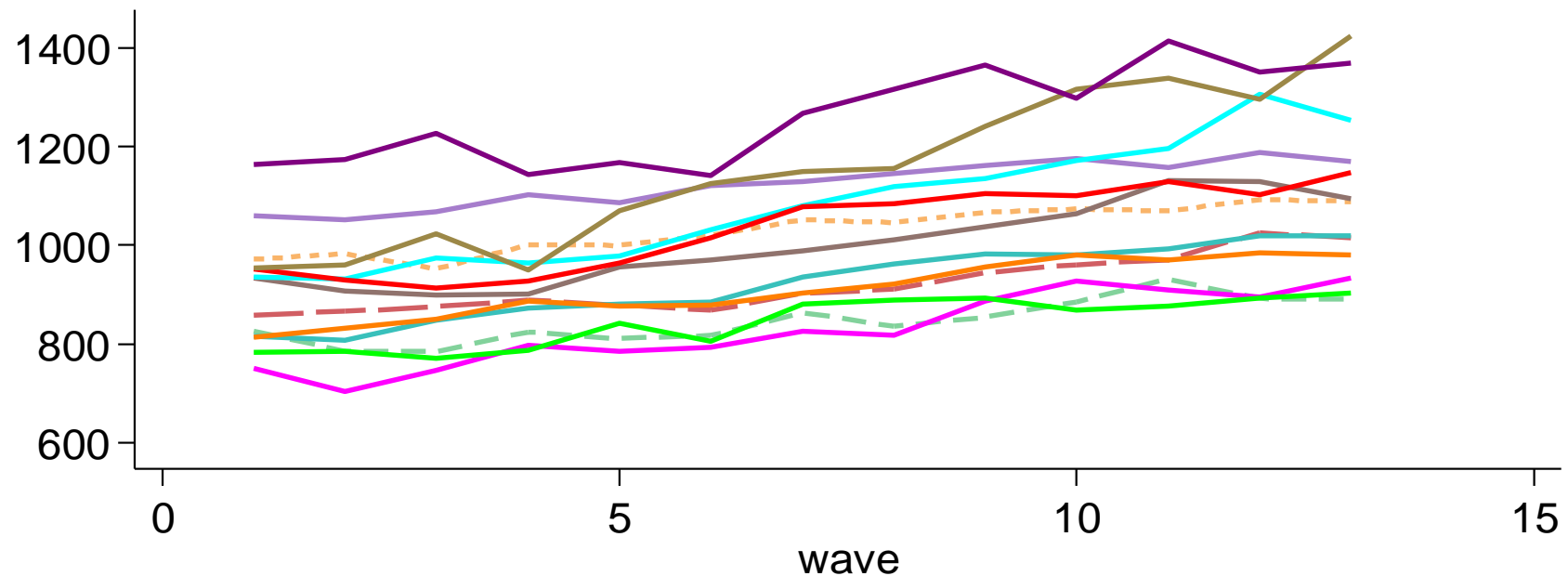
# APPENDIX: SWB



- [11] Sydney
- [21] Melbourne
- [31] Brisbane
- [41] Adelaide
- [51] Perth
- [61] Tasmania
- [81] ACT
- [19] Balance of NSW
- [29] Balance of Victoria
- [39] Balance of QLD
- [49] Balance of SA
- [59] Balance of WA
- [71] Northern Territory



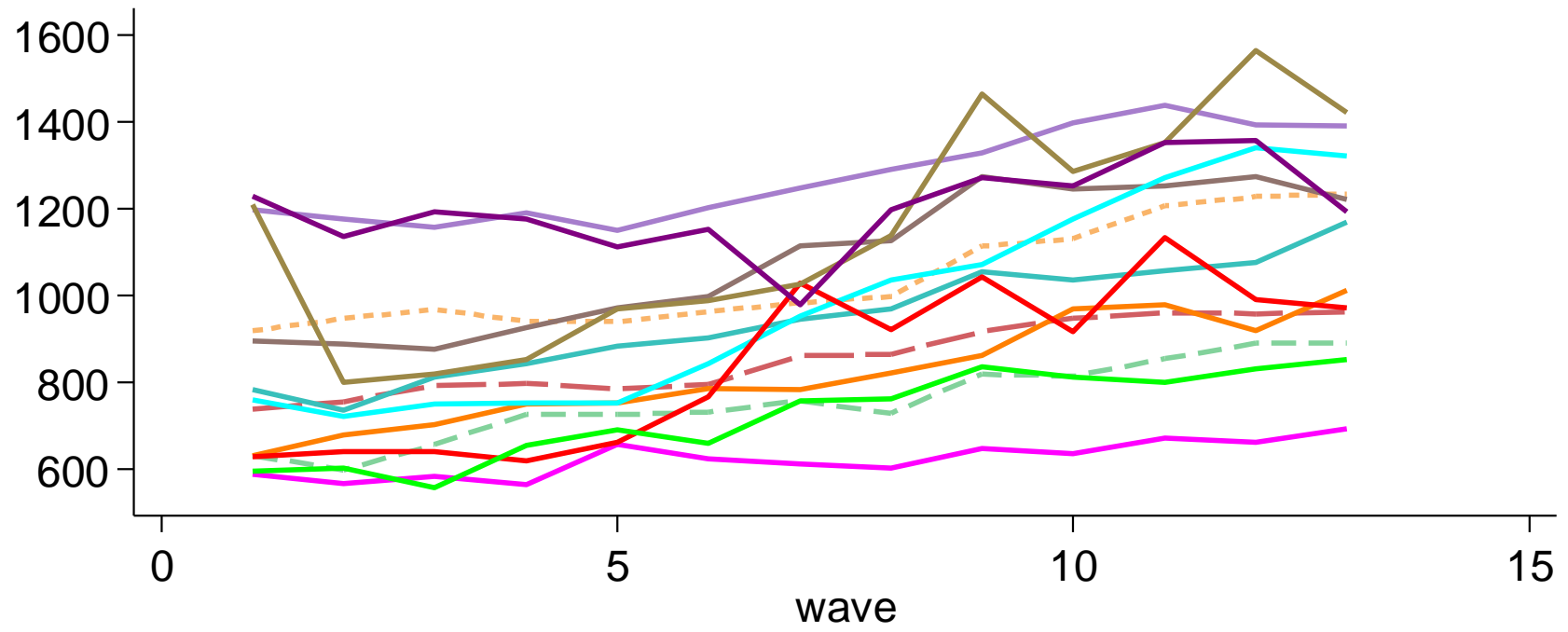
# APPENDIX: INCOME



- |                |                          |
|----------------|--------------------------|
| [11] Sydney    | [19] Balance of NSW      |
| [21] Melbourne | [29] Balance of Victoria |
| [31] Brisbane  | [39] Balance of QLD      |
| [41] Adelaide  | [49] Balance of SA       |
| [51] Perth     | [59] Balance of WA       |
| [61] Tasmania  | [71] Northern Territory  |
| [81] ACT       |                          |

Plotted is the 2% trimmed mean of labour income, calculated on those with positive labour income

# APPENDIX: RENTS



- [11] Sydney
- [21] Melbourne
- [31] Brisbane
- [41] Adelaide
- [51] Perth
- [61] Tasmania
- [81] ACT
- [19] Balance of NSW
- [29] Balance of Victoria
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Graph plots the 2% trimmed mean household rent

