## The forgotten form of stratification: Sexual Orientation in large social survey research

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#### Social Stratification

 Typically dominated by discussions of social class, gender, and ethnicity (Big Three)

 Sexual Orientation is often left out of models of social stratification

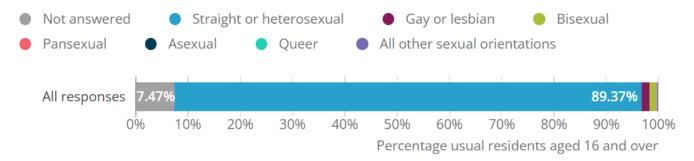
#### Forgotten?

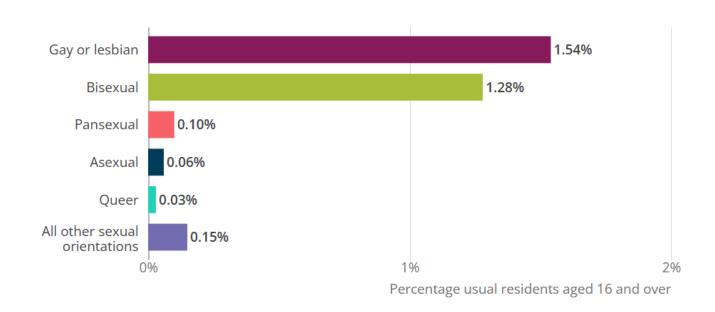
- Left out or forgotten entirely?
- There are a limited number of large social surveys in the UK that collect sexual orientation data
  - Even fewer that collect it well

#### Issues

- Data Collection strategies
  - Pretty hard to ask
  - Pretty obvious candidate for high levels of missingness
  - 'Moral panic' over asking young people
- Real world issues
  - The 'Queer' population is small

Figure 1: Sexual orientation, 2021, England and Wales





Source: Office for National Statistics - Census 2021

#### Issues

- Real world issues
  - The 'Queer' population is small
- Very small sample size
- To my knowledge NO large social surveys in UK that collect sexual orientation construct appropriate weights for this population

#### Issues

- Bad Collection and Recording
  - Couple data
  - Adults only
  - Only asking binary questions
  - Special License Locked

## Making the most out of what is left

- Enter the UKHLS
  - Contemporary large social survey in the UK
  - Has a robust question on sexual orientation that is collected from wave 3 and asked every other wave since then
  - Small, but nationally representative sub-populations
  - Unfortunately, no weights include sexual orientation this was indicated on UKHLS forums at some point but seems to be forgotten
    - (Wouldn't that be a lovely postdoc...)

#### Sexuality Pay Gap

Working Paper

#### Sexuality Pay Gap

- Using UKHLS data to study the Sexuality Pay Gap
  - Very few papers on this topic worldwide
  - Almost exclusively US led
  - A few papers from the UK
  - Consensus is far from settled

- UKHLS also allows us to go beyond Pay gap analysis and look at growth curves
  - This has never been done for the study of sexual orientation pay gaps before

## UK Sexuality Pay Gap

- Aksoy (2018)
- Arabsheibani (2005)
- Arabsheibani (2007)
- Booth and Frank (2008)
- Bridges and Mann (2019)
- Bryson (2017)
- Frank (2008)

## UK Sexuality Pay Gap

- 7 studies
- 2 not nationally representative (Teacher survey + Academic survey)
- 5 nationally representative studies between 2005-2018 (13 years)
- Seminal research on this topic began in 1995 (Badgett 1995)
- If we add (Klawitter 2015; Drydakis 2022) 22% of the papers written on UK sexuality pay gaps are meta-analyses on the topic...

#### Meta-analysis (Klawitter, 2015)

- From the earliest paper on sexuality income gaps (Badgett 1995) to 2015
- Consistently shows gay men earn less than straight men
- Lesbian women are sometimes more likely to outearn straight women
- Non-US studies report smaller earnings gaps
- Sexual orientation measured through self-identity rather than couple status or sexual behaviour reports smaller gaps
- Annual earnings rather than hourly reports larger earnings gap %
- Limiting to full-time workers increases earnings gap

### Metaanalysis (Klawitter, 2015)

- Earnings penalty of 11% for gay men
- Earnings premium of 9% for lesbian women
- No information for bisexuals...

## Working Data

- Sample consists of Waves 3-14 of UKHLS
- Sample includes those aged 16-66 AND those in some form of employment AND not in full time training or education
- Leaves us with a N=294,377 over 11 waves of data

#### Non-starter Analysis?

Sexuality	Frequency
	(%)
Heterosexual	42,258
	(14.36%)
Homosexual	738
	(0.25%)
Bisexual	676
	(0.23%)
Don't Know	29
	(0.01%)
Inapplicable	75,041
	(25.49%)
Missing	174,579
	(59.30%)
Refused	1,056
	(0.36%)

#### Across Waves

	UKHLS wave														
	1	2	3	4	5	6	7	8	9	10	11	12	13	14	Total
N	27,340 (9.3%)	27,244 (9.3%)	24,783 (8.4%)	23,108 (7.8%)	22,284 (7.6%)	22,728 (7.7%)	21,670 (7.4%)	20,732 (7.0%)	19,197 (6.5%)	18,262 (6.2%)	17,100 (5.8%)	15,594 (5.3%)	15,196 (5.2%)	19,139 (6.5%)	294,377 (100.0%)
Sexuality															
Heterosexual	0 (.%)	0 (.%)	21,546 (97.5%)	0 (.%)	1,034 (94.8%)	0 (.%)	998 (94.3%)	0 (.%)	17,414 (96.7%)	0 (.%)	719 (91.2%)	0 (.%)	547 (88.1%)	0 (.%)	42,258 (96.8%)
Homosexual	0 (.%)	0 (.%)	341 (1.5%)	0 (.%)	31 (2.8%)	0 (.%)	25 (2.4%)	0 (.%)	318 (1.8%)	0 (.%)	10 (1.3%)	0 (.%)	13 (2.1%)	0 (.%)	738 (1.7%)
Bisexual	0 (.%)	0 (.%)	215 (1.0%)	0 (.%)	26 (2.4%)	0 (.%)	35 (3.3%)	0 (.%)	280 (1.6%)	0 (.%)	59 (7.5%)	0 (.%)	61 (9.8%)	0 (.%)	676 (1.5%)

# Nonstarter Analysis?

- From base data sexuality is collected every other wave from wave 3
- Not a whole lot to work with
- Using Last Observation Carried Forward can fill in a lot of blanks here
- Last Observation Carried Backwards to get to wave 1-2 data is dangerous when dealing with sexual orientation data
  - We have no way of knowing when someone has 'come out'

	Male	Female	Total			
Sexuality	Frequency	Frequency	Frequency			
	(%)	(%)	(%)			
Heterosexual	82,312	96,547	178,859			
	(59.51%)	(61.87%)	(60.76%)			
Homosexual	1,831	1,395	3,226			
	(1.32%)	(0.89%)	(1.10%)			
Bisexual	865	1,527	2,392			
	(0.63%)	(0.98%)	(0.81%)			
Missing	53,310	56,580	109,890			
	(38.54%)	(36.62%)	(37.33%)			
Total	138,318	156,049	294,367			
	(100%)	(100%)	(100%)			

	Male	Female	Total
Sexuality	Frequency	Frequency	Frequency
	(%)	(%)	(%)
Heterosexual	82,312	96,547	178,859
	(96.83%)	(97.06%)	(96.95%)
Homosexual	1,831	1,395	3,226
	(2.15%)	(1.40%)	(1.75%)
Bisexual	865	1,527	2,392
	(1.02%)	(1.54%)	(1.30%)
Total	85,008	99,469	184,477
	(100%)	(100%)	(100%)

• Getting close to 2021 census data

• Still large amounts of missingness

- Could derive orientation from couple data
  - Bad idea
  - Erasure of bisexuality often occurs

	UKHLS wave														
	1	2	3	4	5	6	7	8	9	10	11	12	13	14	Total
N	27,340 (9.3%)	27,244 (9.3%)	24,783 (8.4%)	23,108 (7.8%)	22,284 (7.6%)	22,728 (7.7%)	21,670 (7.4%)	20,732 (7.0%)	19,197 (6.5%)	18,262 (6.2%)	17,100 (5.8%)	15,594 (5.3%)	15,196 (5.2%)	19,139 (6.5%)	294,377 (100.0%)
Sexuality															
Heterosexual	0 (.%)	0 (.%)	21,546 (97.5%)	17,035 (97.7%)	16,325 (97.4%)	13,974 (97.5%)		12,444 (97.3%)		15,345 (96.6%)	14,345 (96.4%)	12,633 (96.4%)	12,239 (96.0%)		178,860 (97.0%)
Homosexual	0 (.%)	0 (.%)	341 (1.5%)	265 (1.5%)	270 (1.6%)	231 (1.6%)	231 (1.6%)	211 (1.7%)	332 (1.8%)	298 (1.9%)	277 (1.9%)	259 (2.0%)	261 (2.0%)	250 (2.1%)	3,226 (1.7%)
Bisexual	0 (.%)	0 (.%)	215 (1.0%)	144 (0.8%)	160 (1.0%)	126 (0.9%)	152 (1.1%)	128 (1.0%)	294 (1.6%)	234 (1.5%)	256 (1.7%)	206 (1.6%)	250 (2.0%)	227 (1.9%)	2,392 (1.3%)

- Pretty healthy numbers from wave 3-14
- Remember the longitudinal context
  - Is sexual orientation fixed?
    - Common question that props its head every so often with supposedly 'fixed' variables such as ethnicity as well
- IFF orientation is not fixed it would be inappropriate to use LOCF to fill in item missingness across waves for individual pidp units

	MALE	FEMALE	TOTAL
Total Sexuality Switchers	Frequency	Frequency	Frequency
	(%)	(%)	(%)
Bisexual -> Homosexual	-	-	-
Bisexual -> Heterosexual	26	31	57
	(38.24%)	(27.68%)	(31.67%)
Homosexual -> Bisexual	-	-	-
Homosexual -> Heterosexual	-	-	-
Heterosexual -> Bisexual	24	58	82
	(35.29%)	(51.79%)	(45.56%)
Heterosexual -> Homosexual	10	10	20
	(14.71%)	(8.93%)	(11.11%)
Total	68	112	180
	(100%)	(100%)	(100%)

- Out of 184,477 cases of sexual orientation only 180 people across 11 waves of UKHLS data have 'switched' their sexual orientation
- Out of these 180 people 178 of them have only switched once
  - This is capturing people 'coming out'
    - The lower mean age of switchers at 35.5 compared to sample average of 42 seems to corroborate this
  - There are a handful of sexual minority -> heterosexuality switchers however
    - These people appear to have 'dipped their toes'
    - This is dominated by the bisexual -> heterosexual and vice versa categories, which are themselves dominated by women

#### Switchers

 Given the very low number of people that appear to switch sexual orientations over the life course it is fair in my view to consider sexuality as 'fixed'

## Occupational Sorting

 The occupations that individuals sort into has a knock on impact when assessing wages

• Beyond wages, occupational sorting is a clear indicator of societal and cultural expectations, norms, and influences on individual behaviours that are constituted via larger unit groups

Using SOC 2000 codes

## Occupational Sorting

- Women from all sexual orientations and bi men appear to be more heavily concentrated in a few occupational titles compared to straight and gay men
- Bi men are concentrated in <4000 occupations. Possible driver of a wage gap?
  - 7% of bi men are software professionals...
- Women across sexuality appear to focus on soft-skilled labour
  - Though gay and bi women appear to concentrate in authority-related occupations. Possible driver of a pay premium?
- Gay men concentrate in similarly soft-skilled labour positions (teachers)
  - Possible driver of a pay penalty?

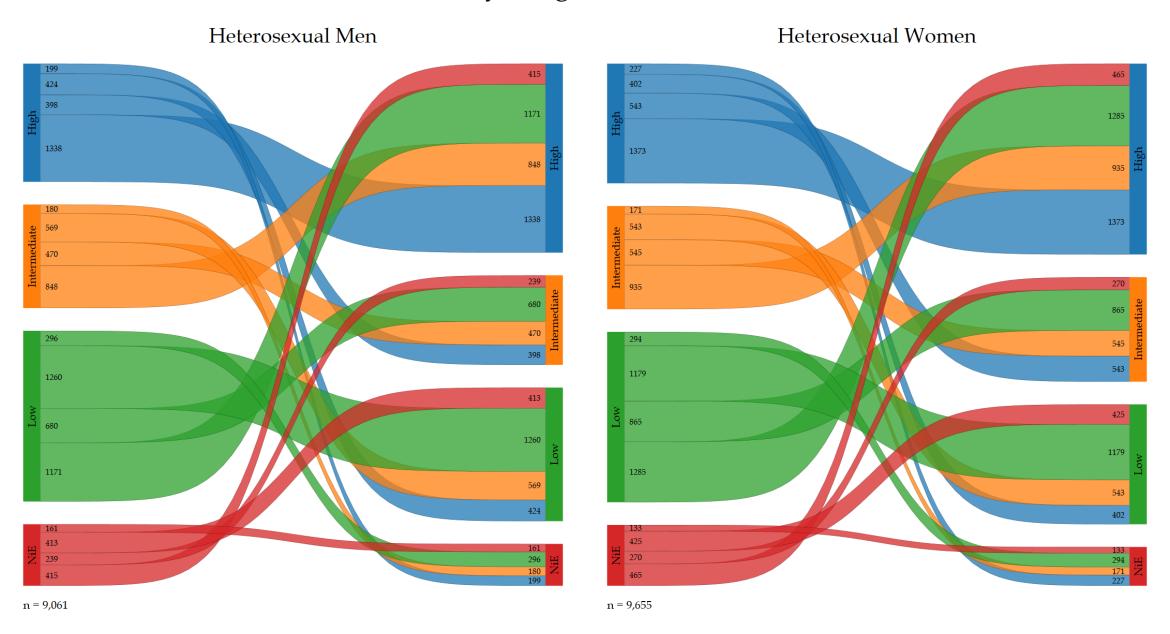
#### Social Mobility

 Prior to modelling some introspection on the lives of sexual minorities is called for

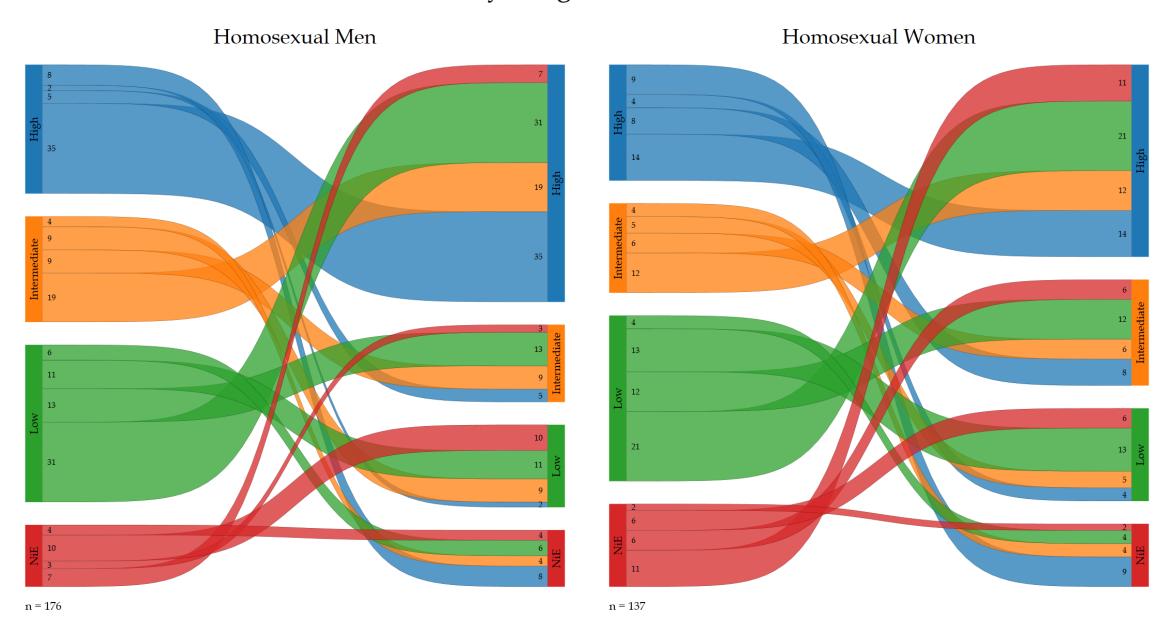
Social Mobility is a good starting point here

How much do origins actually matter for current destinations

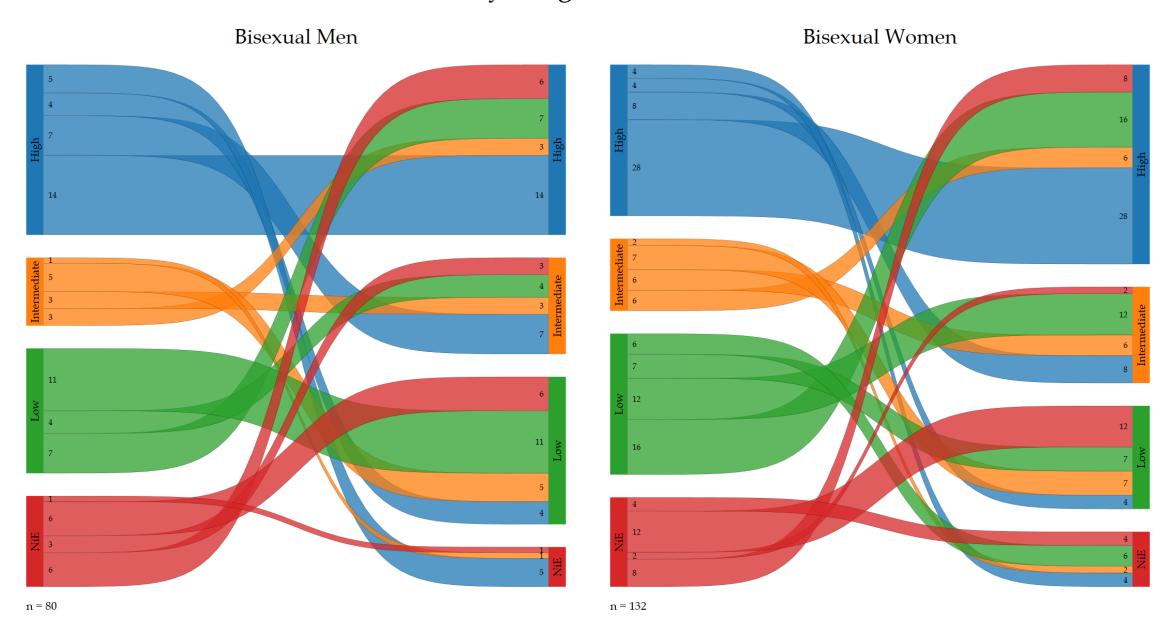
#### Social Mobility: Origin to Latest Destination



#### Social Mobility: Origin to Latest Destination



#### Social Mobility: Origin to Latest Destination



• Slightly larger proportions of gay male individuals 'leapfrogging' in terms of upward social mobility

• Is this an origins effect or an 'in-spite of origins' effect?

 Supportive households are presumably more likely to have a net positive origin effect on individuals

## Modelling the sexuality pay gap

- Four components to this modelling strategy
- 1: Demographics
- 2: Human Capital
- 3: Geography
- 4: Time/Growth

#### Demographics

- Social Origins
- Sexuality
- Age
- Ethnicity
- Sex
- Housing Tenure
- Marital Status
- Parental Status
- Long Term Illness

# Human Capital

- Education
- Work Hours
- Current NS-SEC
- Sector
- Industry
- Size of Firm

# Geography

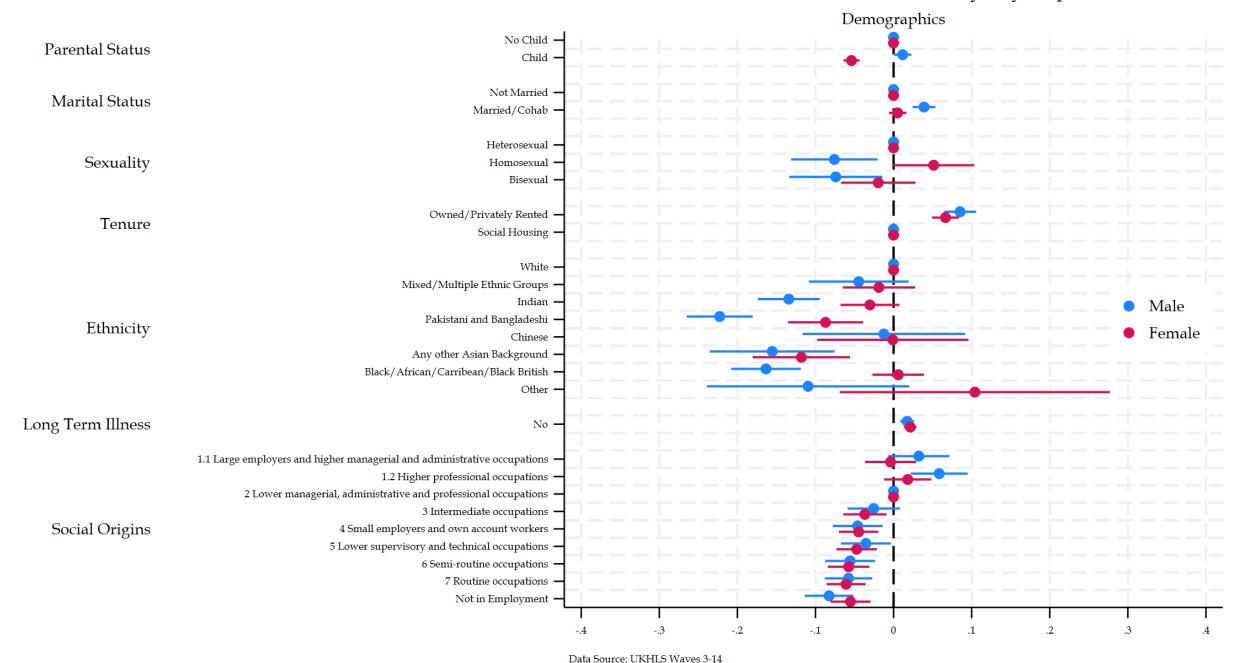
- Urban
- Region

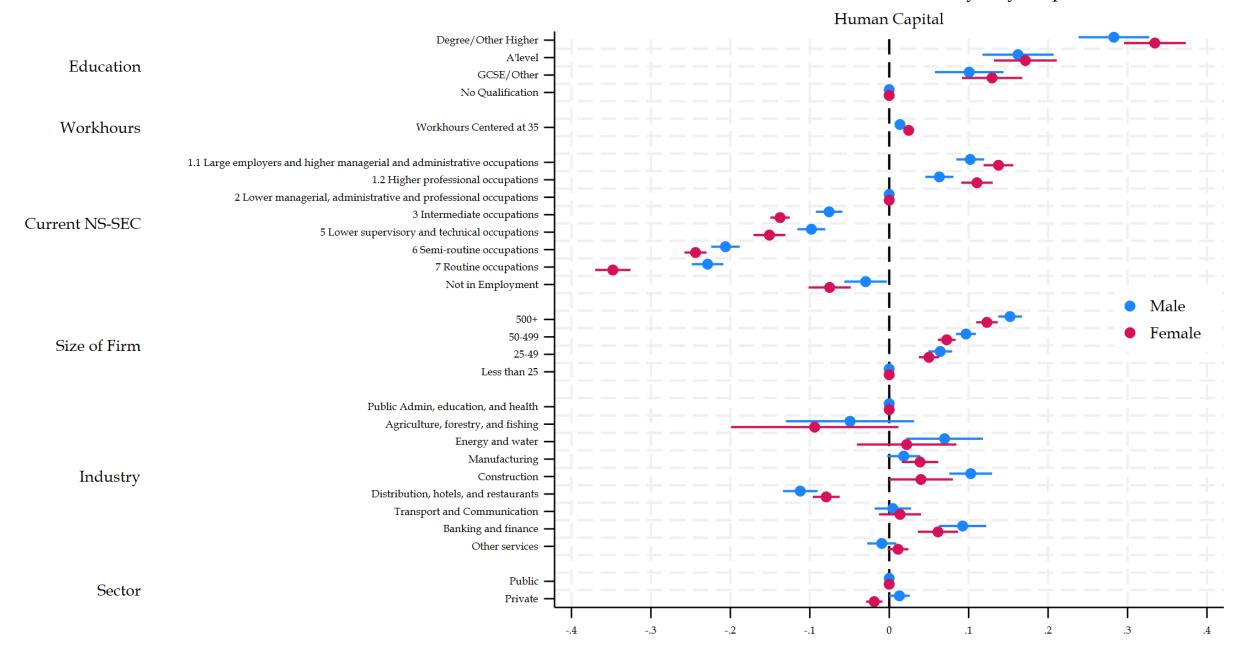
# Time/Growth

- Wave as a function of time
- Modelling through an appropriate panel set up via a unique individual pidp
- Measure change over time

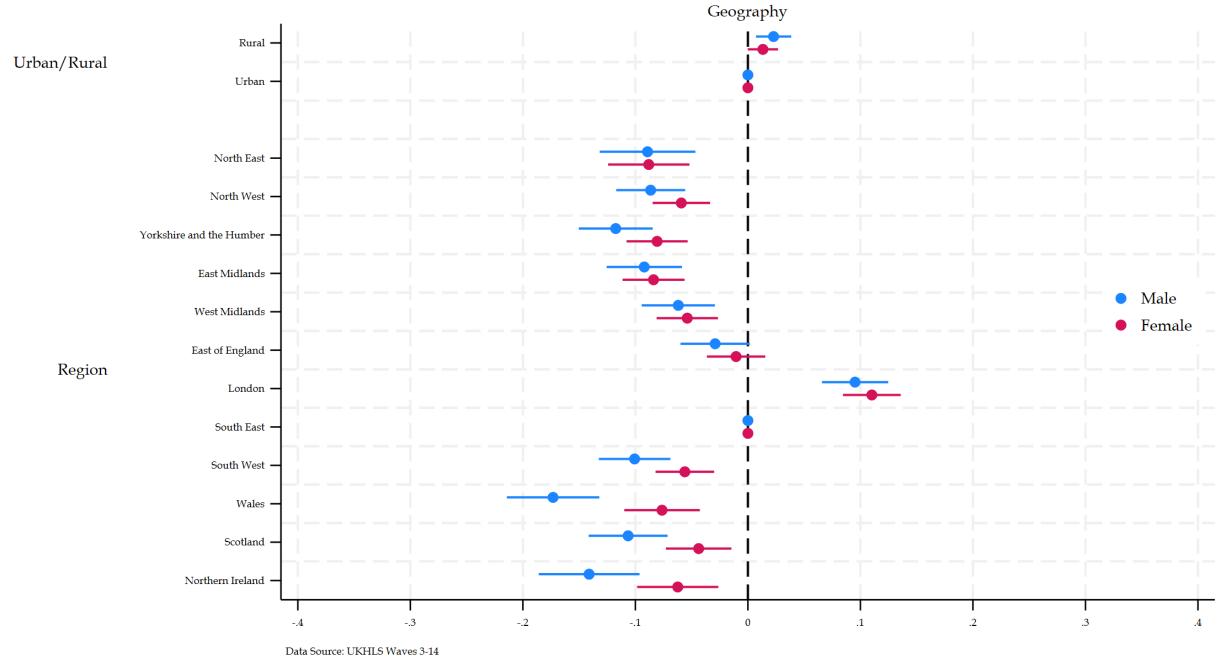
# Regression Models

- Big model
- Shown in the form of coefficient plots and predictive marginal effects
- Shown in 'clusters'
- Not interpreting all effects (Shout at me if you want me to go back or look at a specific effect)

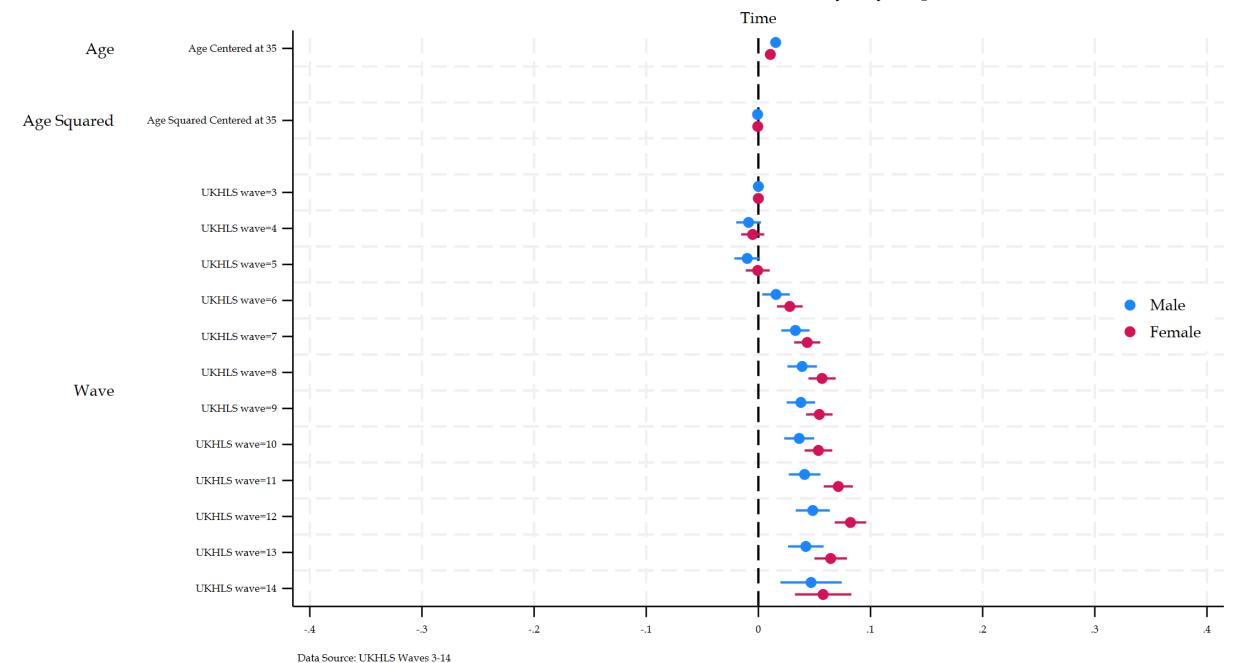




Data Source: UKHLS Waves 3-14



Male Model n=40,135 & Female Model n=48,917



• Gay men earn 8% less than straight men annually and bi men 7% less controlling for all other variables

• Gay women and bi women not statistically significant

- Remember the meta-analysis?
  - Gay men 11% penalty + Gay women 9% premium
- My model:
  - Gay men 8% penalty
  - Bi men 7% penalty
- Reduction in penalty

- Converting this into real £s would also be beneficial
- Set a new constant:
  - Straight white men with no children that are married age 35 and work 35 hours a week and own their own home with a degree in an urban region of London with NS-SEC 2 parents in an NS-SEC 2 job for a size 500+ company in the private banking industry at wave 3.
  - New constant = 10.28 OR £30,000
- Straight men = £30,000 gross OR £2,500 a month
- Gay men = £26,900 gross OR £2241 a month
- Bi men = £27,173 gross OR £2264 a month

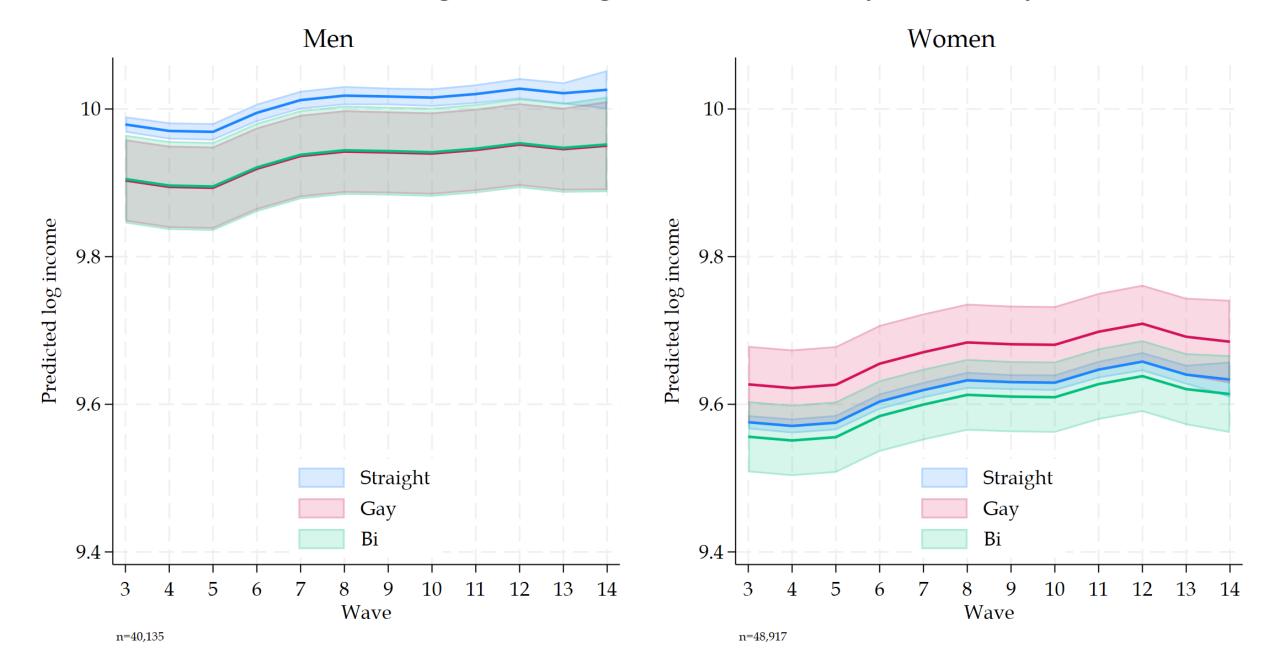
- Straight men = £30,000 gross OR £23,303 net (£1,942 a month)
- Gay men = £26,900 gross OR £21,501 net (£1,792 a month)
- Bi men = £27,173 gross OR £21,660 net (£1,805 a month)

- Compared to women?
  - Women earn less than men of all orientations across the board

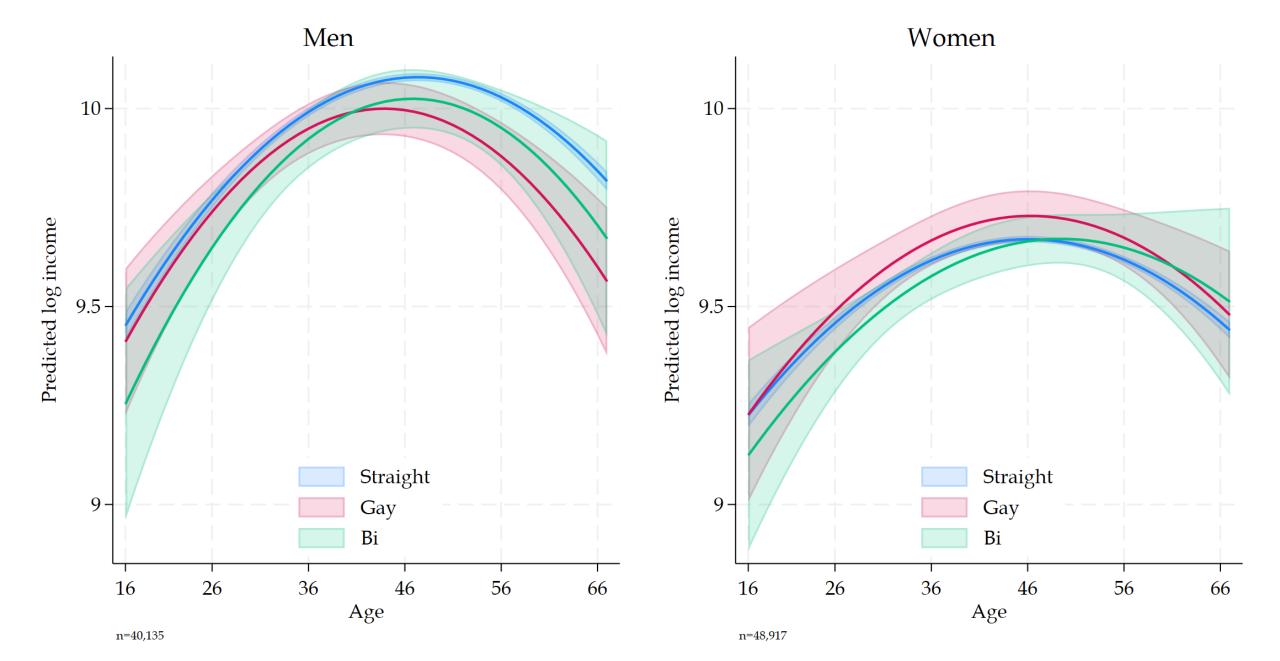
### **Growth Curves**

- Using wave as a function of time
- Possibility to look at age and age2 for curvilinear affects
- Careful of age=period-cohort effects
- Model may need tweaking
- Synthetic cohorts + age instead of waves?

### Predicted log income growth curves by sexuality



### Predicted log income growth curves by sexuality



### Future Plans

- Growth curve needs tweaking a little
  - Stay with xtreg or move over to mixed
    - Possibility to random slopes
- Complex survey design adjustments
- Handling missing data
- Experimental longitudinal decomposition techniques
  - Used to explain possible 'discrimination'

# Questions