# Modelling social care receipt using linked administrative data

A presentation at interview for Post-Doctoral Research Fellow

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### Project funding and supervision

- Scottish Government/ESRC
- Nick Bailey, Colin McCowan, Stewart Mercer



# Background

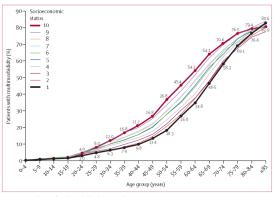
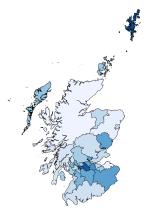


Figure 2: Prevalence of multimorbidity by age and socioeconomic status
On socioeconomic status scale, 1=most affluent and 10=most deprived.

### Barnett et al (2012)

# Background

Rate of social care for over 65s 2011/12 by Local Authority



Rate per thousand people over 65 receiving any form of social care



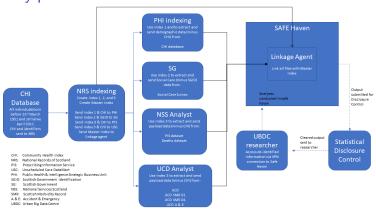
### Research Questions

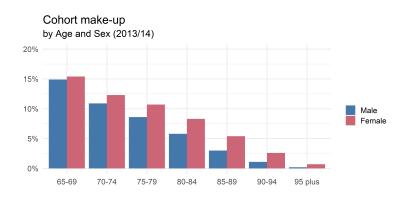
#### In people over the age of 65 in Scotland:

- 1 (a) What are the socioeconomic, demographic, and geographic patterns in the use of social care?
  - (b) Is there an association between multimorbidity status and the amount and type of social care use over time? Does this vary by the patterns described in 1(a)?
- 2 Is there an association in the use of social care services, multimorbidity status and unscheduled health care use?

# Background

#### Study period 2011-2016

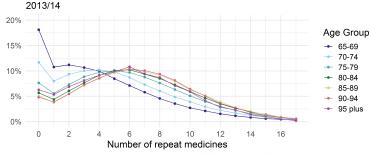




### Percentage of individuals in SIMD deciles by Local Authority (2013/14)

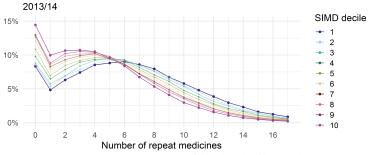


# Percentage of individuals with specific count of repeat medicines



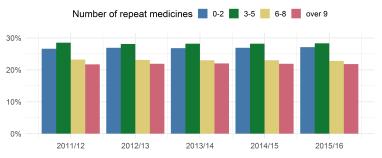
Outlying values > 17 medicines removed

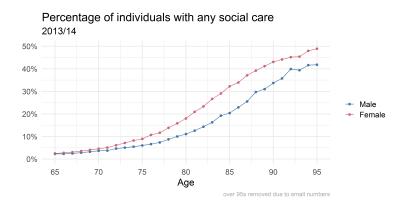
# Percentage of individuals with specific count of repeat medicines



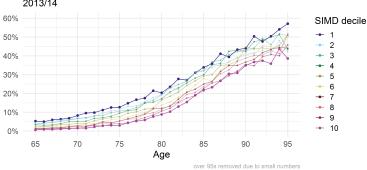
Outlying values > 17 medicines removed

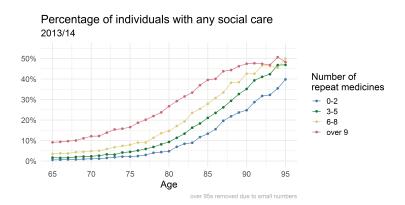
# Percentage of population receiving repeat medicines by medicine group





# Percentage of individuals with any social care 2013/14





# Modelling

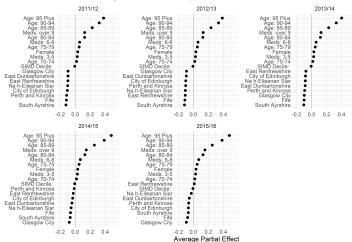
- Logistic regression model applied to each separate year of data
- Model applied to subsets of individuals with similar linkage rates to population spine
- Effects reported as Average Partial Effects (APEs)
- Indicate the marginal effect of each variable on outcome

#### Regression code

```
glm(social_care ~ sex*age_grp + simd*meds_grp +
age_grp*meds_grp + council*simd, family = binomial())
```

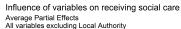
### Results - Modelling

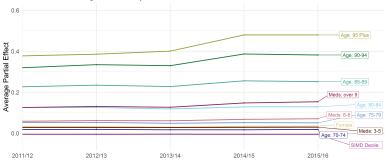
Influence of variables on receiving social care Average Partial Effects ordered high to low by year Councils with linkage rate of 92%-95.9% only



Reference group Age: 65-69 Reference group Sex: Male Reference group SIMD Decile: Decile 1 Reference group Meds: 0-2 medicines Reference group Council:Shetland Islands

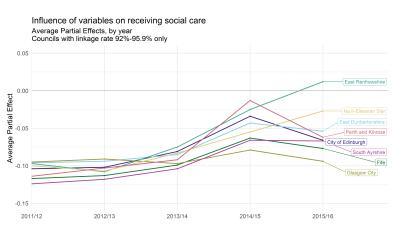
### Results - Modelling





Reference group Age: 65-69 Reference group Sex: Male Reference group SIMD Decile: Decile 1 Reference group Meds: 0-2 medicines

# Results - Modelling



Reference group: Shetland Islands

#### Limitations

- No measure to capture those living alone
- Variation in linkage rate of SCS to population spine at the local authority level
- No longitudinal analysis here
- Untested multimorbidity measure

### **Implications**

- Age has the biggest effect on receipt of social care -Multimorbidity also important
- Females also more likely to receive care, but a lot of the difference seen in overall numbers is accounted for when adjusting for age and multimorbidity
- After adjusting for age and multimorbidity status, there remains variation in levels of care at the local authority level.
   Further investigation is warranted.
- Again, deprivation is shown to to be a driver of service use.
   The difference in effect between decile 1 and decile 10 is of similar magnitude to the difference between age groups 65-69 and 75-79