

# Biomarker trajectories and health outcomes

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#### **Overview**

- Trajectories: methodology
- Individual trajectories of body size
- Group-based trajectories of an inflammation marker
- Association with ageing outcomes



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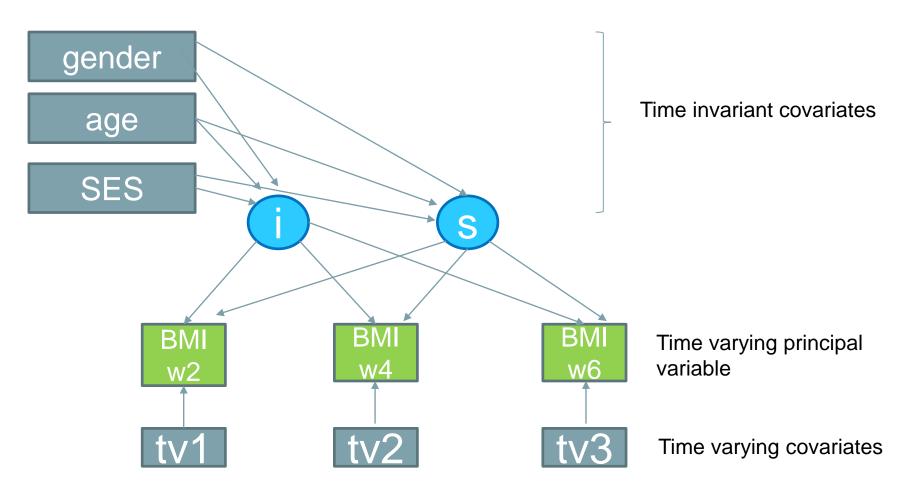


### **Growth curve modelling**

- Individual trajectories over time: Latent growth curve models
  - One trajectory for each person: Intercept and slope(s)
  - An average trajectory to characterize the entire sample
  - Variance trajectory = An indication of the extent to which individual trajectories deviate from the average trajectory
- Group-based trajectories: Latent Class Growth Analysis
  - Multiple sub-groups in a population → for each group you get an intercept and slope(s)
  - Sub-groups are NOT known a priori, can emerge from the data → exploratory and descriptive analysis
  - Sub-groups are called classes

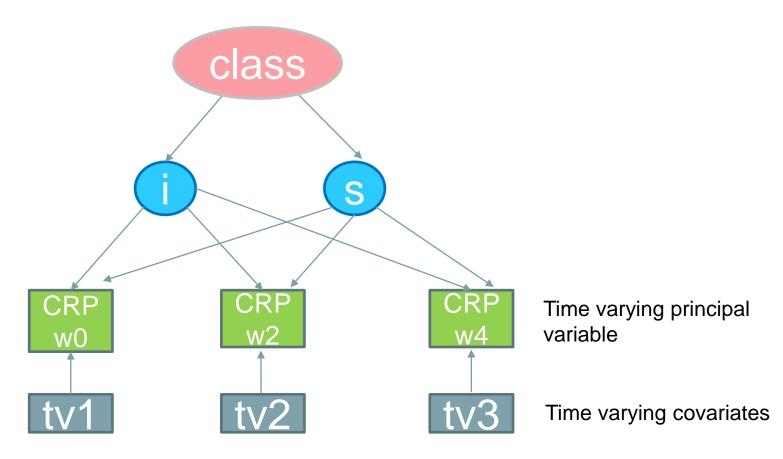


#### Individual: Latent growth curve model





### Group-based: Latent class growth analysis





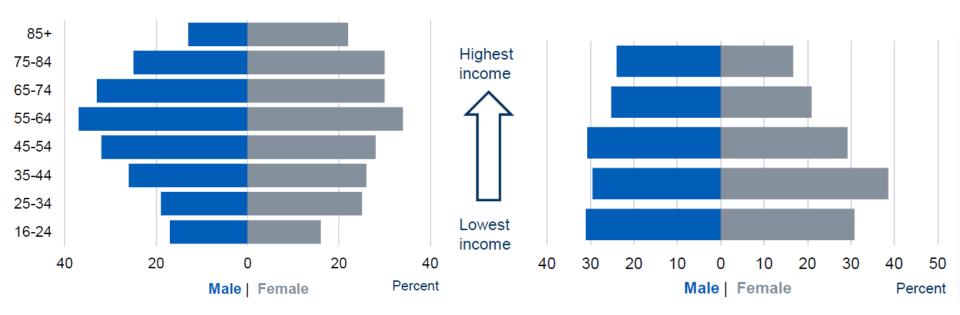
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### **Body size and health consequences**

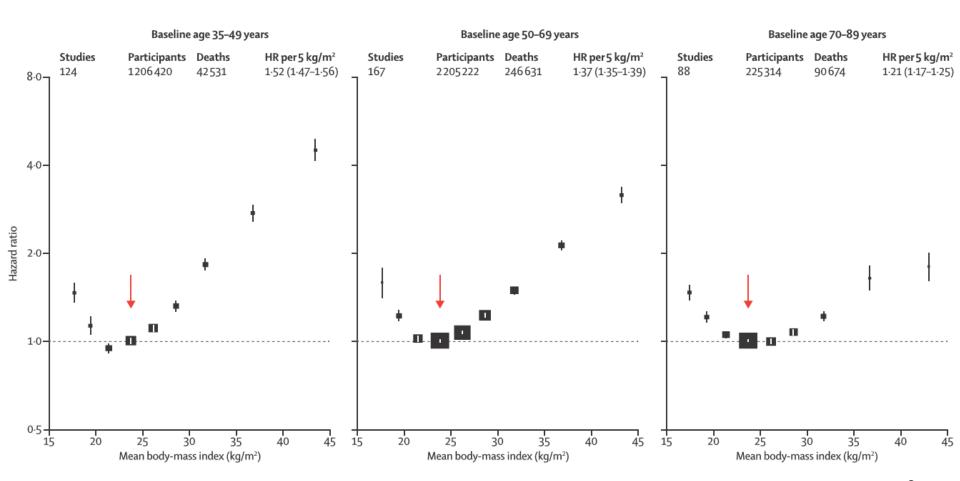
- ▶ Global obesity epidemic → diabetes, CVD, some cancers, other chronic diseases
- Prevalence in the UK in 2015: 58% women and 68% men overweight or obese
- Obesity particularly prevalent 45-74y then decline
- Socioeconomic gradient





### **Body size and health consequences**

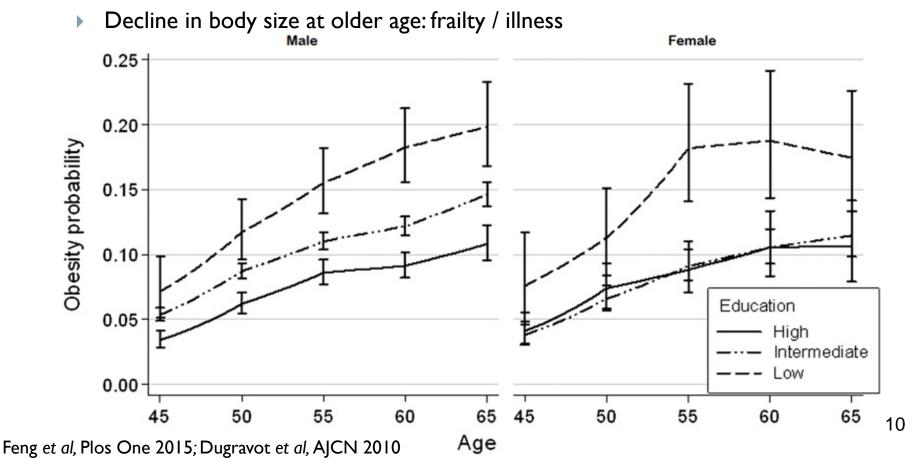
Both <u>obesity</u> and <u>underweight</u> are associated with higher mortality risk





### **Body size in relation to SES**

- Higher obesity rates in lower SES groups
  - Compelling evidence in adults throughout midlife
  - Less studies and more conflicting evidence in older age





### **Objective and Methods**

**Objective**: Describe and compare BMI and waist circumference trajectories in ELSA and assess the effect of socioeconomic status

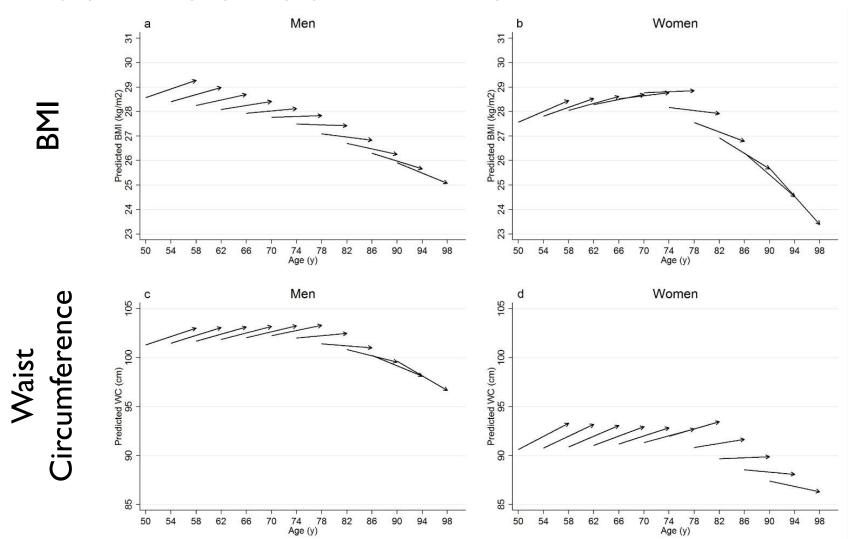
#### Population:

- Baseline year: 2004 (wave 2)
- ▶ Three time points : wave 2, wave 4, wave 6 (clinical examination)
  - N=3259 men, mean age at baseline 65.6 ± 9.2 y
  - N=3966 women, mean age 66. I ± 9.6 y
- BMI and Waist Circumference measured by nurse
- SES measure: tertiles of wealth in 2004
- Covariates: smoking, physical activity, limiting longstanding illness, marital status
- 8-year period
- Linear and quadratic term of age on I and S to describe age-specific trajectories = aging vectors



### Results (I)

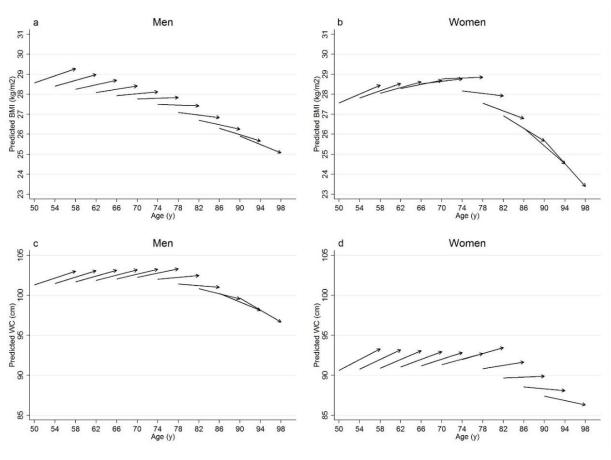
Vector graph showing 8-year aging vectors of anthropometric markers, ELSA 2004-5 to 2012-13





## Results (I)

Vector graph showing 8-year aging vectors of anthropometric markers, ELSA 2004-5 to 2012-13



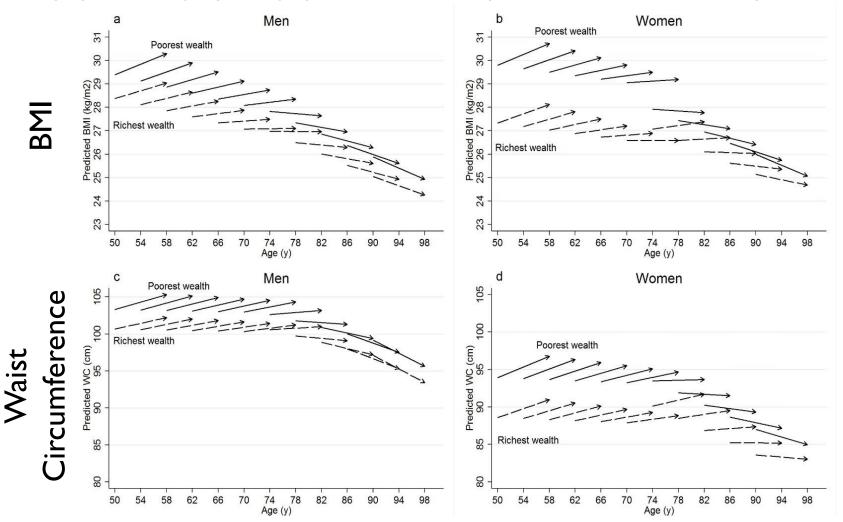
- Less variations in men than women
- Increase in BMI and WC (gain) from 50 to ~70y
- Decrease after 70 y stronger for BMI than WC
- Cohort effect: younger cohort bigger

E.g. a man who was 62 in 2004 has a lower BMI by 1kg/m<sup>2</sup> compared to a man who was 62 in 2012



### Results (II)

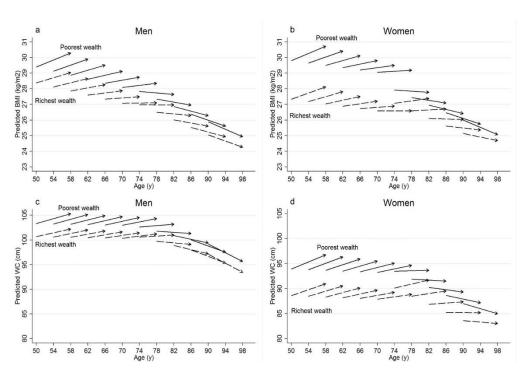
Vector graph showing 8-year aging vectors of anthropometric markers according to wealth





### Results (II)

Vector graph showing 8-year aging vectors of anthropometric markers according to wealth



- Strong effect of wealth on baseline BMI and WC
  - For men and women
  - For <70y and >70y
  - Poorer = greater body size
- No sig effect of wealth on slope
  - Men: parallel trajectories by SES group → the gap doesn't close
  - Women: in poorest wealth group decline faster

### Conclusions (I)

- Identification of body size trajectories in ELSA:
  - Participants aged 50 to 70 tend to gain weight
  - Participants aged >70 tend to lose weight, likely to be lean mass
  - Cohort effect: younger cohort bigger

#### Effect of wealth

- At any given age, lower wealth associated with higher BMI and higher waist circumference
- No significant effect of wealth on the change of body size: the gap doesn't close or the decline in BMI tend to be stronger in older women
- Our results support the cumulative disadvantage theory: social disadvantage over the life course is associated with unfavourable body size
- Paper under review Zaninotto P & Lassale C "Socioeconomic trajectories of body mass index and waist circumference: results from the English Longitudinal Study of Ageing"



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#### Inflammation and aging outcomes

- Compelling evidence of associations between elevated inflammation biomarkers (C-reactive protein, Inteleukin 6 and Fibrinogen mainly) with a range of age-related outcomes:
  - Cardioascular, Diabetes, Cancer
  - Cognitive function and dementia
  - Sarcopenia and osteoporosis
  - Frailty
- Most studies used only one measurement of biomarker
- Few studies have used two measurements to define "chronic inflammation"
  - ▶ Akbaraly et al, CMAJ 2013: Whitehall II, 5 years apart
  - ▶ Jenny et al, J Gerontol A Biol Sci Med Sci 2012: Cardiovascular Healthy Study All Stars, 9 years apart
- Promising but partial evidence
- Little is known about long-term changes in inflammation as predictor
- No study on trajectories on more time points



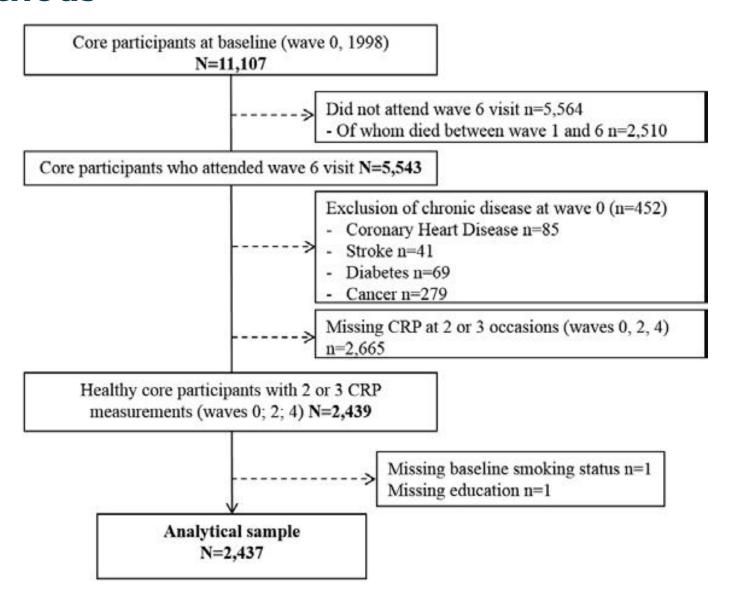
#### **Methods**

Objective: Identify group-based trajectories of inflammation and assess the associations with ageing outcomes

- ▶ N=2,439
- Exposure: hsCRP (mgL) measured at wave 0, 2 and 4 in ELSA
- Exclusion of existing cardiometabolic disease: CHD, stroke, diabetes
- Adjustment for: age, sex, education, smoking, BMI and NSAID use at baseline
- Distal outcomes measured at wave 6:
  - Physical functioning
  - Cardiometabolic health
  - Lung function
  - Cognitive function
  - Depression CES-D

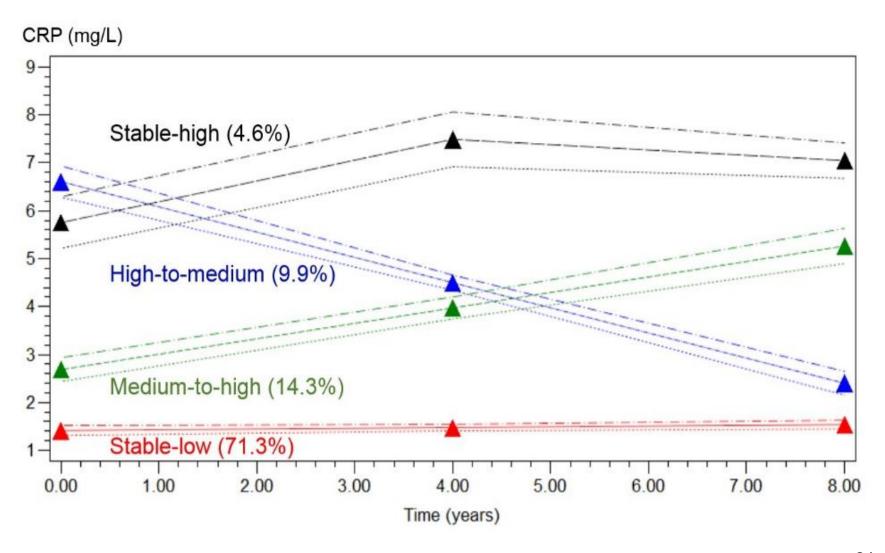


#### **Methods**



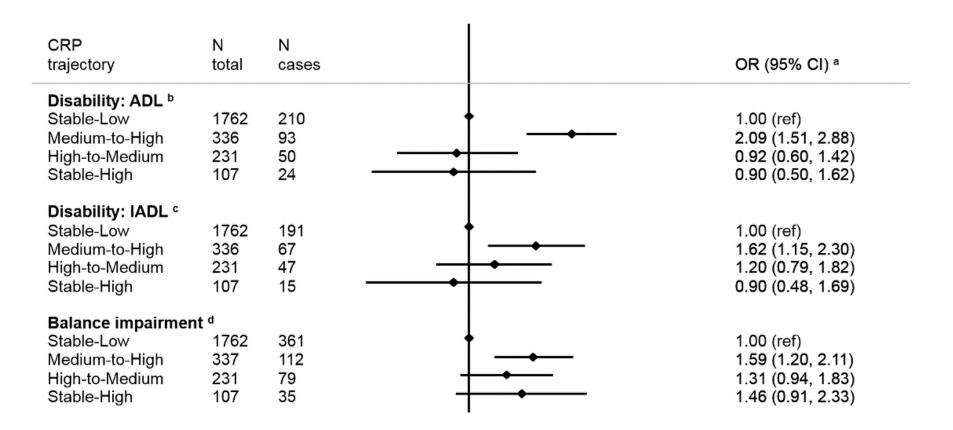


## **Results: trajectories of CRP**



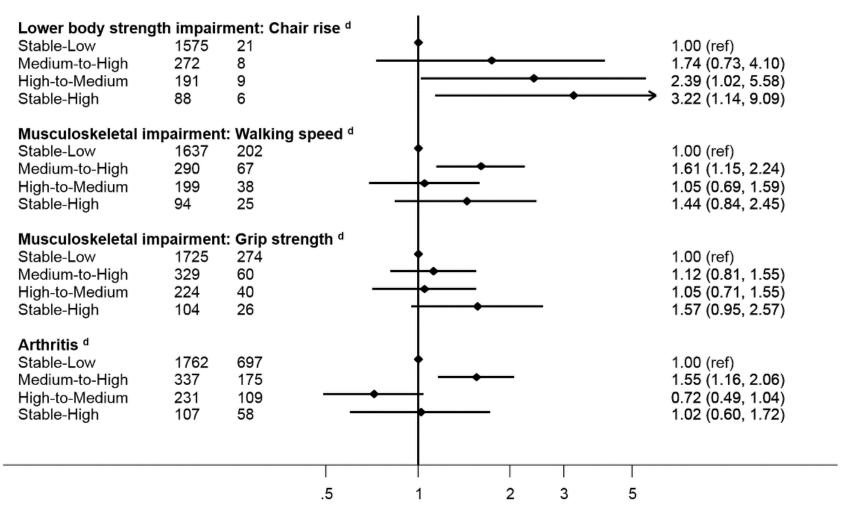


### Results: Association with physical functioning





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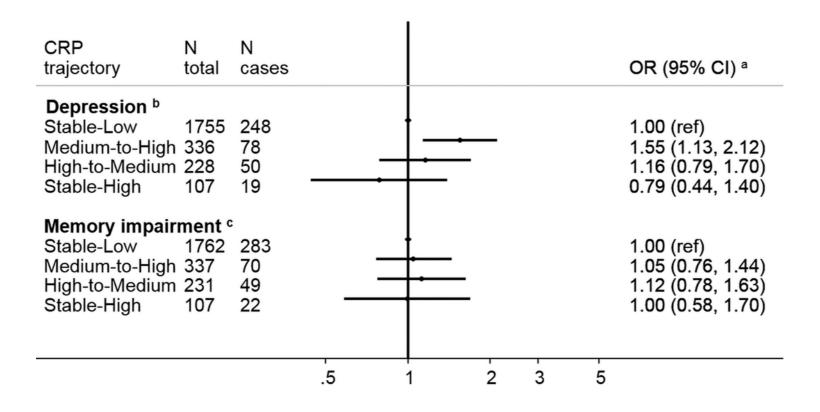


## Results: Association with CVD risk factors

CRP trajectory	N total	N cases		OR (95% CI) <sup>a</sup>
Hypertension b Stable-Low Medium-to-High High-to-Medium Stable-High	1691 321 223 104	808 205 127 58		1.00 (ref) 1.57 (1.21, 2.04) 1.06 (0.78, 1.44) 1.02 (0.67, 1.57)
Low HDL Stable-Low Medium-to-High High-to-Medium Stable-High	1493 282 183 92	143 45 27 12	<u>-</u>	1.00 (ref) 1.41 (0.97, 2.06) 1.24 (0.77, 1.99) 0.88 (0.45, 1.72)
Diabetes Stable-Low Medium-to-High High-to-Medium Stable-High	1761 337 231 107	145 51 31 21		1.00 (ref) 1.29 (0.89, 1.86) 1.03 (0.66, 1.63) 1.42 (0.81, 2.47)
Obesity Stable-Low Medium-to-High High-to-Medium Stable-High	1708 318 220 102	349 148 90 60		1.00 (ref) 1.95 (1.36, 2.80) 0.97 (0.63, 1.49) 1.82 (0.95, 3.49)
Low FEV1 Stable-Low Medium-to-High High-to-Medium Stable-High	1512 282 191 92	283 86 58 29	<u></u>	1.00 (ref) 1.84 (1.36, 2.50) 1.75 (1.22, 2.52) 2.16 (1.32, 3.55)



#### Results: Association with mental health





### **Conclusions (II)**

- Identified 4 long-term trajectories of CRP over a 10 year period
- Increasing CRP from medium to high levels associated with most adverse ageing outcomes:
  - poor cardiometabolic health
  - lower physical functioning and increased arthritis
  - Lower respiratory functioning
  - increased depressive symptoms
- Maintaining high levels of CRP associated with some outcomes
- Independent of health behaviours, SES, BMI, anti-inflammatory drugs
- Monitor inflammation levels over time can help prevent adverse ageing outcomes

Lassale C, et al. J Gerontol A Biol Sci Med Sci. 2018



#### **General conclusions**

- ▶ ELSA rich study with repeated measurements
- Latent growth curve modelling useful to identify both individual and group-based trajectories
- Useful to describe evolution of health markers over time and relate it to contextual factors
- Interesting to group individuals that follow similar pattern of trajectories over time and relate this to disease risk or any aspect of ageing



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