

1 **Associations of meat and fish consumption with conventional and radiomics cardiovascular**
2 **phenotypes in the UK Biobank**

Zahra Raisi-Estabragh^{1,2}, Celeste McCracken¹, Polyxeni Gkontra³, Akshay Jaggi³,
Maddalena Ardissino⁴, Jackie Cooper¹, Luca Biasioli⁵, Nay Aung^{1,2}, Stefan K. Piechnik⁵,
Stefan Neubauer⁵, Patricia B. Munroe¹, Karim Lekadir³, Nicholas C. Harvey^{6,7}, Steffen E.
Petersen*^{1,2}

1. William Harvey Research Institute, NIHR Barts Biomedical Research Centre, Queen Mary University of London, Charterhouse Square, London, EC1M 6BQ, UK

2. Barts Heart Centre, St Bartholomew's Hospital, Barts Health NHS Trust, West Smithfield, EC1A 7BE, UK

3. Departament de Matemàtiques & Informàtica, Universitat de Barcelona, Spain

4. Imperial College London, Sir Alexander Fleming Building, Exhibition Road, SW7 2AZ, UK

5. National Institute for Health Research Oxford Biomedical Research Centre, Division of Cardiovascular Medicine, Radcliffe Department of Medicine, University of Oxford, Oxford, UK

6. MRC Lifecourse Epidemiology Unit, University of Southampton, Southampton, SO16 6YD, UK

7. NIHR Southampton Biomedical Research Centre, University of Southampton and University Hospital Southampton NHS Foundation Trust, Southampton, UK

***Corresponding author:** Professor Steffen E. Petersen. William Harvey Research Institute, NIHR Barts Biomedical Research Centre, Queen Mary University of London, Charterhouse Square, London, EC1M 6BQ, UK; Email: s.e.petersen@qmul.ac.uk; Telephone: +44-2078826902

Data availability statement

This research was conducted using the UKB resource under access application 2964. UK Biobank will make the data available to all bona fide researchers for all types of health-related research that is in the public interest, without preferential or exclusive access for any persons. All researchers will be subject to the same application process and approval criteria as specified by UK Biobank. For more details on the access procedure, see the UK Biobank website: <http://www.ukbiobank.ac.uk/register-apply/>.

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Short running head: Meat consumption and cardiovascular phenotypes

Abbreviations

ASI: arterial stiffness index

AD: aortic distensibility

BMI: body mass index

CI: confidence interval

CMR: cardiovascular magnetic resonance

LV: left ventricle

RV: right ventricle

NHS: national health service

ROI: region of interest

SI: signal intensity

TMAO: trimethylamine N-oxide

3 **Abstract**

4 **Background:** Greater red and processed meat consumption has been linked to adverse
5 cardiovascular outcomes. However, the impact of these exposures on cardiovascular
6 magnetic resonance (CMR) phenotypes has not been adequately studied.

7

8 **Objective:** We describe novel associations of meat intake with cardiovascular phenotypes
9 and investigate underlying mechanisms through consideration of a range of covariates.

10

11 **Design:** We studied 19,408 UK Biobank participants with CMR. Average daily red and
12 processed meat consumption was determined through food frequency questionnaires and ,
13 expressed as a continuous variable. Oily fish was studied as a comparator associated with
14 favourable cardiac outcomes. We considered associations with conventional CMR indices
15 (ventricular volumes, ejection fraction, stroke volume, left ventricular mass,), novel CMR
16 radiomics features (shape, first-order, texture), and arterial compliance measures (arterial
17 stiffness index, aortic distensibility). We used multivariable linear regression to investigate
18 relationships between meat intake and cardiovascular phenotypes, adjusting for confounders
19 (age, sex, social deprivation, educational level, smoking, alcohol intake, exercise) and
20 potential covariates on the causal pathway (hypertension, hypercholesterolaemia, diabetes,
21 body mass index).

22

23 **Results:** Greater red and processed meat consumption was associated with an unhealthy
24 pattern of biventricular remodelling, worse cardiac function, and poorer arterial compliance.
25 In contrast, greater oily fish consumption was associated with a healthier cardiovascular
26 phenotype and better arterial compliance. There was partial attenuation of associations
27 between red meat and conventional CMR indices with addition of covariates potentially on
28 the causal pathway, indicating a possible mechanistic role for these cardiometabolic
29 morbidities. However, other relationships were not altered with inclusion of these covariates
30 suggesting importance of alternative biological mechanisms underlying these relationships.
31 Radiomics analysis provided deeper phenotyping, demonstrating association of the different
32 dietary habits with distinct ventricular geometry and left ventricular myocardial texture
33 patterns.

34

35 **Conclusions:** Greater red and processed meat consumption is associated with impaired
36 cardiovascular health, both in terms of markers of arterial disease and of cardiac structure and
37 function. Cardiometabolic morbidities appeared to have a mechanistic role in the associations
38 of red meat with ventricular phenotypes, but less so for other associations suggesting
39 importance of alternative mechanism for these relationships.

40

41 **Key words:** cardiovascular magnetic resonance; radiomics; cardiovascular disease; diet; red
42 meat; processed meat; fish; population health

43 **Introduction**

44 Multiple epidemiological studies have demonstrated associations between greater meat
45 consumption and worse cardiovascular outcomes(1–4). In particular, higher red and
46 processed meat intake has been associated with greater burden of atherosclerosis(5),
47 increased risk of incident ischaemic cardiovascular events(6), and heart failure(7).
48 Furthermore, murine studies link greater red meat consumption with pathological ventricular
49 remodelling and heart failure phenotypes(8).

50

51 It has been proposed, that these relationships may be mediated through adverse
52 cardiometabolic alterations(9,10). More recently, evidence has emerged for novel causal
53 pathways relating to cross-system interactions with the gut microbiome(11).

54

55 We studied novel associations between red and processed meat consumption and measures of
56 cardiovascular structure and function in the UK Biobank, including conventional
57 cardiovascular magnetic resonance (CMR) metrics, novel CMR radiomics features, and
58 measures of arterial compliance. We considered associations between oily fish intake as a
59 comparator previously linked with favourable cardiovascular endpoints(6). We considered a
60 wide range of confounder, including covariates that may lie on the causal pathway.

61 **Methods**

62 **Setting and study population**

63 The UK Biobank is a cohort study of over 500,000 participants. Recruitment was between
64 2006-2010 through postal invite of UK residents aged 40-69 years identified through
65 National Health Service (NHS) registers. Individuals who were unable to consent or complete
66 baseline assessment due to illness or discomfort were not recruited. Baseline assessment of
67 participants comprised characterisation of socio-demographics, lifestyle, environmental
68 factors, medical history, and a range of physical measures(12). The protocol is publicly
69 available(13). The UK Biobank imaging study, which includes detailed CMR imaging, was
70 launched in 2015 and aims to scan a random subset of 100,000 participants (approximately
71 50,000 completed, March 2021)(14).

72

73 **Ethics**

74 This study complies with the Declaration of Helsinki; the work was covered by the ethical
75 approval for UK Biobank studies from the NHS National Research Ethics Service on 17th
76 June 2011 (Ref 11/NW/0382) and extended on 10th May 2016 (Ref 16/NW/0274) with
77 written informed consent obtained from all participants.

78

79 **Definition of meat consumption variables**

80 Dietary intake was assessed using a self-report food frequency questionnaire at the baseline
81 UK Biobank visit. Participants were asked to estimate their weekly intake of a range of food
82 items over the preceding year. We considered beef, pork, lamb/mutton, processed meat, and
83 oily fish consumption (**Supplementary Table 1**). We considered each type of red meat
84 separately (beef, lamb/mutton, pork) and also as a composite category of ‘unprocessed red
85 meat’. Processed meat included products such as bacon, ham, sausages, meat pies, kebabs,
86 and burgers. Reported portion frequencies were converted into probabilities of daily
87 consumption and multiplied by standard portion sizes(15) to derive average daily
88 consumption in grams. Thus, we were able to consider the meat exposures as continuous
89 variables, as has been published previously using this dataset(16).

90

91 **Measures of cardiac structure and function**

92 ***Conventional CMR indices***

93 CMR scans were performed in dedicated UK Biobank imaging centres using 1.5 Tesla
94 scanners (MAGNETOM Aera, Syngo Platform VD13A, Siemens Healthcare, Erlangen,
95 Germany) according to a pre-defined acquisition protocol, which is detailed in a separate
96 publication(17). Assessment of the left and right ventricles (LV, RV) included a complete
97 short axis stack acquired using balanced steady-state free precession sequences. The first
98 5,000 CMR scans were manually analysed according to a pre-defined segmentation
99 protocol(18) using CVI⁴²® post-processing software (Version 5.1.1, Circle Cardiovascular
100 Imaging Inc., Calgary, Canada). Briefly, LV endocardial and epicardial borders were
101 manually contoured in end-diastole and end-systole in the short axis view. The first phase of
102 acquisition was selected as end-diastole. End-systole was defined as the phase with the
103 smallest mid-ventricular LV intra-cavity blood pool as determined by visual inspection. The
104 most basal slice for the LV was selected when at least half of the LV blood pool was
105 surrounded by myocardium. LV papillary muscles were excluded from LV mass. RV

106 endocardial borders were traced in end-diastole and end-systole with volumes below the
107 pulmonary valve plane considered as part of the RV. This ground truth manual analysis
108 dataset was used to develop a fully automated image analysis pipeline with inbuilt quality
109 control, which has been applied to the first 20,000 UK Biobank CMR studies(19). Details of
110 reproducibility performance of the automated algorithm are available in a dedicated
111 publication(18,19). For the present analysis, data was available from 19,408 CMR studies,
112 including the following metrics: LV and RV volumes in end-diastole and end-systole, LV
113 and RV ejection fraction, LV and RV stroke volume, and LV mass.

114

115 *Novel CMR radiomics features*

116 CMR radiomics is a novel image analysis technique permitting computation of multiple
117 indices of shape and texture(20). Radiomics features provide information that is
118 complementary and potentially incremental to conventional CMR indices(20). We used
119 contours from conventional CMR analysis, with image segmentation as described above, on
120 short axis cine images(19) to define three regions of interest for radiomics analysis in end-
121 diastole and end-systole: RV cavity, LV cavity, and LV myocardium. We extracted shape
122 features from the RV and LV cavity ROIs. From the LV myocardium, we extracted first
123 order and texture radiomics features. Radiomics features were extracted using the
124 PyRadiomics open source platform(21). The full list of radiomics features extracted is
125 presented in **Supplementary Table 2**. To reduce variation of image signal intensities relating
126 to the acquisition process, we performed intensity normalisation of CMR images through
127 histogram matching, using as reference one of the studies from the dataset(22). For grey level
128 discretisation, we used a fixed bin width of 25 intensity values.

129

130 **Measures of arterial compliance**

131 *Aortic distensibility*

132 Aortic distensibility is a measure of local aortic compliance. Lower aortic distensibility
133 indicates poorer vascular health and is a marker of arterial disease(23). Aortic distensibility
134 may be estimated by considering the relative cross-sectional area change of the thoracic aorta
135 from diastole to systole on transverse cine CMR images(24). In previous work, we derived
136 aortic distensibility using data generated from a fully automated image analysis pipeline
137 applied to the first 20,000 UK Biobank CMR studies, details of the automated pipeline are
138 presented in a separate publication(25).

139

140 *Arterial stiffness index*

141 Arterial stiffness index (ASI) provides an estimate of large artery stiffness derived from a
142 pulse waveform contour(23). Higher ASI indicates lower arterial compliance and is
143 associated with poorer cardiovascular, in particular ischaemic, outcomes(23,26). ASI was
144 measured at both baseline and imaging visits using finger photoplethysmography with the
145 PulseTrace PCA2 (CareFusion, USA) device, according to a standardised protocol(27).

146

147 **Statistical analysis**

148 Statistical analysis was performed using R Version 3.6.2(28) and RStudio Version
149 1.2.5019(29). We estimated the association of the dietary intake exposures with

150 cardiovascular metrics in individual multivariable linear regression models. To allow
151 derivation of easily assimilated effect sizes, we report change in cardiovascular metric per
152 100g increase in daily meat consumption alongside corresponding 95% confidence intervals
153 (CIs) and p-values. For ASI, we assessed associations with measures taken at both baseline
154 and imaging visits. We identified significant interval change in ASI from baseline to imaging.
155 Therefore, we also considered “change in ASI” as a separate outcome, expressed using
156 standardised residuals derived from regression of ASI at imaging on ASI at baseline. The
157 average time interval between baseline and imaging assessment was 7.5 years in the CMR set
158 and 8.2 years in the ASI set.

159

160 In order to compare the magnitude of change across different radiomics features, prior to
161 regression analysis, we performed z-score normalisation of the features. This resulted in
162 calculation of standardised beta coefficients per 100g daily increase in meat/fish intake. As the
163 number of texture features extracted from the LV myocardium was large (n=144), we
164 performed cluster analysis (**Figure 1**) to group inter-related features(30). We hierarchically
165 clustered features using complete linkage on Pearson correlation distance between features.
166 We determined the optimal number of clusters by computing the average silhouette, a measure
167 of cluster consistency using the cluster package in R(30). The silhouette statistic reflects the
168 average distance between data points in the same cluster compared against average data points
169 in other clusters and allows judgement of the optimal number of clusters within a sample, such
170 that distance between datapoints within clusters are minimised whilst maximising distance with
171 datapoints from other clusters. We computed average silhouette statistic for 2 to 10 clusters.
172 Maximum silhouette statistic was observed with 7 and 8 clusters. Hence, we take 7 clusters as
173 representing the optimal number of clusters within our samples (**Figure 1**). We assigned
174 descriptive names to each cluster based on properties represented by its constituent features.
175 Thus, for the texture features, we present the mean beta-coefficient and 95% CIs for each
176 cluster for the different meat/fish exposures. We compare effects between exposure categories
177 through testing for the difference in mean beta coefficients using Kruskal-Wallis statistical testing
178 followed by Dunn’s correction for multiple comparisons. As radiomics is a relatively new approach,
179 we provide a brief guide to radiomics and specific guidance on interpretation of results from this
180 analysis in Appendix II.

181

182 We selected covariates based on association with both exposure and outcome in preliminary
183 analyses and existing literature (**Figure 2**). We adjust for potential confounders (age, sex,
184 material deprivation, education, smoking, alcohol intake, exercise) in our main models to
185 estimate the magnitude of the exposure-outcome associations. We identified hypertension,
186 hypercholesterolaemia, diabetes, and body mass index (BMI) as covariates potentially on the
187 causal pathway (**Figure 2**). To test the impact of these variables, we estimated associations
188 with additional inclusion of these factors in the models, with the expectation that covariates
189 on the causal pathway would attenuate exposure-outcome associations.

190

191 Educational level, smoking status, and alcohol intake frequency were based on self-report.
192 Material deprivation is reported as the Townsend index, a measure of deprivation relative to
193 national averages.(31). A continuous value for the amount of physical activity measured in
194 metabolic equivalent (MET) minutes/week was calculated by weighting different types of
195 activity (walking, moderate or vigorous) by its energy requirements using values derived
196 from the International Physical Activity Questionnaire (IPAQ) study(32). BMI was
197 calculated from height and weight. Diabetes was ascertained from self-report of the

198 diagnosis, self-reported use of “medication for diabetes”, or serum glycosylated haemoglobin
199 >48mmol/mol. Hypertension was coded based on self-report of the diagnosis or self-reported
200 use of “medication for high blood pressure”. Hypercholesterolaemia was coded based on self-
201 report of the diagnosis, self-reported use of “medication for high cholesterol”, or serum total
202 cholesterol >7mmol/L.

203

204

205 **Results**

206 **Baseline population characteristics**

207 CMR data was available for 19,408 participants. Mean age was 55.0 (± 7.5) years, 52.1%
208 were women (**Table 1**). The majority 97% (n= 18,810) were of White ethnic background;
209 Black, Asian, and Other ethnicities made up 0.5%, 1.0%, and 1.5% of the analysis sample
210 respectively. The cohort was predominantly healthy, with only 5.5% (n=1,062) having a
211 history of pre-existing cardiovascular disease (ischaemic and non-ischaemic heart diseases,
212 valvular heart disease, significant arrhythmias). The rates of cardiometabolic morbidities
213 were also lower than the general population, in line with previous analyses of the UK
214 Biobank(33,34). The prevalence of hypertension, hypercholesterolaemia, diabetes, and
215 smoking were 13.9%, 23.0%, 3.1%, and 6.4% respectively. Average red meat intake
216 (lamb/mutton, beef, and pork combined) was 22.3 (± 15.2) grams/day; beef was the most
217 frequently eaten of the red meat types. Average intake of processed meat and oily fish were
218 15.7 (± 15.0) grams/day and 11.7 (± 10.8) grams/day respectively.

219

220 **Association of meat and fish intake with conventional CMR indices**

221 Greater consumption of red and processed meat was associated with smaller LV volumes in
222 end-diastole and end-systole, higher LV mass, and lower LV stroke volume (**Table 2**).
223 Greater oily fish consumption was associated with larger LV volumes in end-diastole and
224 end-systole, greater LV mass, and higher LV stroke volume (**Table 2**). The same pattern of
225 remodelling was observed in the RV, with greater red and processed meat intake linked to
226 smaller ventricular volumes and lower stroke volume, whilst greater oily fish consumption
227 was associated with larger cavity volumes and higher stroke volume (**Supplementary Table**
228 **3**). These relationships were consistent across the different red meat types for both the LV
229 and RV indices (**Supplementary Table 3**). There was attenuation of associations between
230 unprocessed red meat with all CMR indices other than LV mass with addition of
231 cardiometabolic covariates, whilst associations with processed meat and oily fish remained
232 largely unchanged (**Supplementary Table 4**).

233

234 **Association of meat and fish intake with LV and RV radiomics shape features**

235 13 radiomics shape features were extracted from each ventricle (LV and RV) in end-diastole
236 and end-systole. Greater oily fish consumption was associated with significantly larger LV
237 volumes, larger cavity dimensions in both the short and long axis, and greater surface area of
238 the LV cavity (**Figure 3, Figure 6**). Interpreted in conjunction with the previously observed
239 association with higher LV stroke volume (indicating better myocardial function), these
240 findings are in keeping with healthy cardiac structure and function. Greater red and processed
241 meat intake were associated with lower “flatness” (values range between 1 (sphere-like) and
242 0 (a flat object)), lower “elongation” (values range between 1 (non-elongated) and 0 (a
243 maximally elongated object: i.e., a 1 dimensional line)), and lower “sphericity” (a
244 dimensionless measure of the roundness of the ROI relative to a sphere. The value range
245 is $0 < \text{sphericity} \leq 1$, where a value of 1 indicates a perfect sphere). Thus, greater red and
246 processed meat intake is associated with a more elongated LV shape (**Figure 3, Figure 6**). In
247 contrast, greater oily fish consumption showed trends toward greater elongation and flatness
248 (not statistically significant) indicating a more spherical chamber.

249

250 Considering these relationships as well as association with lower LV stroke volume, the
251 overall picture suggests that greater red and processed meat intake are associated with of an
252 unhealthy LV phenotype with impaired myocardial contractility. The pattern of associations
253 of cardiac structure and function metrics with greater oily fish intake is distinctly different to
254 that of the meat exposures and, overall, suggestive of a healthy phenotype.

255

256 The same pattern of associations was observed across the different red meat types in end-
257 diastole and end-systole (**Supplementary Figure 1**) and consistent associations were
258 observed with RV shape radiomics (**Supplementary figure 2**). Results from individual
259 associations between meat and fish exposures and LV and RV radiomics features in end-
260 diastole and end-systole are presented in **Supplementary Tables 5-8**.

261

262 **Association of meat and fish intake with LV myocardium radiomics first-order features**

263 First-order features are histogram-based statistics describing the global distribution of signal
264 intensity values in the defined region and may signify global tissue-level myocardial
265 changes(20). 18 radiomics first-order features were extracted from the LV myocardium in
266 end-diastole and end-systole. The red/processed meat and fish exposures showed markedly
267 different, often reverse, associations with radiomics first-order features (**Figure 4, Figure 5**).
268 Greater red and processed meat consumption was associated with lower average intensity
269 levels and less variation in signal intensity values (consistent across all relevant metrics, such
270 as, lower mean, median, range, variance, entropy). The reverse of these trends was observed
271 with greater oily fish consumption: higher average signal intensity level, greater range of
272 intensity levels, higher number of extreme intensities (kurtosis), and greater randomness of
273 intensity values (entropy). These associations appeared consistent across different meat types
274 and in end-diastole and end-systole (**Supplementary Figure 3**). Thus, associations with the
275 global pattern of signal intensities in the LV myocardium are very different between the meat
276 and fish exposures. These findings suggest that these exposures may be associated with
277 different (reverse) global pattern of alterations at the myocardial level. Results from
278 individual associations between meat and fish exposures and LV myocardium first-order
279 features in end-diastole and end-systole are presented in **Supplementary Tables 9 and 10**.

280

281 **Association of meat and fish intake with LV myocardium radiomics texture features**

282 Radiomics texture features allow quantification of the pattern of inter-voxel signal intensities.
283 Applied to the LV myocardium, radiomics texture features may provide biologically
284 informative quantifiers about underlying tissue properties. We extracted 72 texture features
285 from the LV myocardium in end-diastole and end-systole (total 144 features per CMR study).
286 Cluster analysis identified seven inter-correlated groups of features (**Figure 1**), to which we
287 assigned descriptive terms based on the features within the cluster (**Table 3**). Comparison of
288 mean effects in these clusters showed different effect sizes and directions of effect across the
289 various meat exposures (**Figure 5**). Greater red meat consumption was associated with lower
290 intensity levels, lower variation in intensity levels, less local heterogeneity, and less skewness
291 in the distribution of signal intensity values (**Figure 5, Figure 6**). Greater oily fish
292 consumption associated with greater local heterogeneity and greater skewness in the intensity
293 level distribution. The pattern of associations of the meat and fish exposures with inter-voxel
294 relationships were also different, suggesting potential different alterations at the myocardium.
295 Results from individual associations between meat and fish exposures and individual LV

296 myocardium texture features in end-diastole and end-systole are presented in **Supplementary**
297 **Tables 11 and 12.**

298

299 **Association of meat and fish intake with arterial compliance measures**

300 There was record of ASI at the baseline, imaging, and at both time points for 167,432
301 (baseline characteristics: **Supplementary Table 13**), 30,474, and 10,436 participants
302 respectively. For the latter group, we considered interval “change in ASI”. Higher intake of
303 red and processed meat was associated with higher ASI, indicating greater vascular
304 resistance, at both the baseline and imaging visits (**Table 4, Figure 7**). In addition, higher
305 unprocessed red meat intake was associated with significantly greater interval increase in ASI
306 from baseline to imaging (**Table 4**). In contrast, greater oily fish consumption was associated
307 with lower ASI at both time points and with a smaller baseline to imaging interval increase in
308 ASI (not statistically significant).

309

310 Greater red and processed meat consumption was associated with lower aortic distensibility
311 and greater oily fish consumption with higher aortic distensibility (not statistical significance,
312 **Table 4**). Relationship with all arterial compliance outcomes were consistent across the three
313 red meat groups (**Supplementary Table 14**) and broadly unchanged with adjustment for
314 potential mediators (**Supplementary Table 15**).

315 **Discussion**

316 **Summary of findings**

317 In this study of 9,303 men and 10,105 women, greater red and processed meat consumption
318 was associated with impaired cardiovascular health, both in terms of markers of arterial
319 disease and of cardiac structure and function, in contrast, greater oily fish intake was linked
320 with a healthy cardiovascular phenotype.

321

322 Specifically, greater red and processed meat intake was associated with smaller ventricular
323 volumes, poorer myocardial function (lower LV/RV stroke volume), and poorer arterial
324 compliance (higher ASI, greater interval increases in ASI, lower aortic distensibility). By
325 comparison, greater oily fish consumption was associated with larger LV and RV volumes,
326 better myocardial function (higher LV/RV stroke volume), and better arterial health (lower
327 ASI, smaller interval increases in ASI, higher aortic distensibility). There was evidence that
328 cardiometabolic morbidities may have a mechanistic role in the associations of unprocessed
329 red meat with ventricular phenotypes, but less so for other associations suggesting
330 importance of alternative mechanism for these relationships. Radiomics analysis provided
331 complementary and incremental information demonstrating association of the different
332 dietary habits with distinct overall shape of the ventricles and LV myocardial texture. Greater
333 oily fish consumption was associated with a less elongated LV (more spherical), whilst
334 greater red and processed meat intake was associated with a more elongated LV. The
335 different dietary habits were also associated with different patterns of associations with signal
336 intensity based radiomics features (first-order, texture). Overall, greater red and processed
337 meat intake was associated with lower average global signal intensity and a more
338 homogenous signal intensity pattern both globally and when considering inter-pixel
339 relationships. In contrast, greater oily fish consumption was associated with, on average, a
340 brighter myocardium (global higher signal intensity), with greater range and variation in
341 signal intensities, and greater randomness in the pattern of intensity levels. These findings
342 indicate that meat and fish consumption are associated with different signal intensity patterns
343 at LV myocardium, suggesting possible differences at the tissue level associated with the
344 different exposures.

345

346 **Comparison with existing literature**

347 To the best of our knowledge, the specific impact of red or processed meat intake on CMR
348 imaging phenotypes has not been previously studied. The association between a number of
349 dietary patterns and CMR indices of cardiac structure and function have previously been
350 addressed in two studies of the Multi-Ethnic Study of Atherosclerosis (MESA) cohort. These
351 evaluated long-term effect on CMR measures of LV structure and function of two specific
352 dietary patterns: Mediterranean(35) and the Dietary Approaches to Stop Hypertension
353 (DASH)(36) diets, both of which associated with healthy cardiovascular phenotypes (larger
354 cavity volumes, higher LV mass, higher stroke volume, higher ejection fraction). Limited
355 further studies have focused on diet and cardiovascular structure assessed by
356 echocardiography. Maugeri et al.(37) reported higher rates of concentric left ventricular
357 hypertrophy in individuals following a ‘western’ dietary pattern. Similarly, Wagner et al.(38)
358 documented associations between unhealthy dietary behaviours and higher LV mass. Haring
359 et al.(5) report association of higher red and processed meat intake with poorer imaging
360 indicators of arterial health (greater intima medial thickness and atherosclerotic burden on
361 carotid ultrasound). Our findings corroborate existing evidence and contribute incremental

362 knowledge by demonstrating detailed cardiac phenotypic indices associated with
363 red/processed meat and oily fish consumption in the largest population to date, using both
364 conventional and novel radiomics CMR measures and measures of vascular compliance.

365

366 Existing literature suggests a number of possible explanations for the association between
367 higher red meat consumption and cardiovascular disease. Firstly, these observed effects may
368 be mediated through alterations of the cardiometabolic profile. Greater red and processed
369 meat intake is linked to adverse lipid profiles(10), higher blood pressure(9), adverse body
370 composition(39), and reduced insulin sensitivity(40). Interestingly, in our study, the observed
371 effects on cardiovascular structure and function were not fully explained with adjustment for
372 potential cardiometabolic mediators, suggesting a role for mechanistic pathways independent
373 of these morbidities. Alternative disease pathways such as the gut microbiome dependent
374 trimethylamine N-oxide (TMAO) pathway may play a role in this association: red meat
375 intake, rich in carnitine, is known to increase both plasma and urine TMAO levels, by
376 increased provision of the precursor, L-carnitine and reduced fractional renal TMOA
377 excretion(41). TMAO has, in turn, been mechanistically associated with atherosclerotic
378 disease(11). In our study, associations with arterial health were largely unchanged with
379 additional adjustment for cardiometabolic mediators, suggesting that alternative pathways,
380 such as the TMAO pathway, may be more important in mediating associations with arterial
381 disease, whereas cardiometabolic factors are more important in driving relationships with
382 cardiac health. The TMAO pathway may thus present potential novel therapeutic targets for
383 targeting arterial disease.

384

385 **Strengths and limitations**

386 The large sample and detailed characterisation of participants including CMR scanning and
387 objective measures of arterial health permitted a uniquely comprehensive assessment of the
388 relationship between the various meat exposures and cardiovascular phenotypes with
389 consideration of a range of confounders and mediators. The uniform scanning and analysis
390 procedures presented a high-quality standardised dataset. Common to all nutritional
391 epidemiology research, the measurement and tracking of dietary behaviours is extremely
392 difficult. Our exposures are defined on the basis of a self-report food frequency questionnaire
393 from a single time point, and thus do not account for potential changes in dietary behaviour
394 over time. However, a formal evaluation of the performance of the UKB dietary
395 questionnaire demonstrated good repeatability for the main food groups(42). Furthermore, as
396 potential measurement error is likely to be non-differential across the spectrum of meat
397 consumption and meat types, the risk of bias implied by this is low. The UK Biobank dietary
398 questionnaire gathers information on dietary habits over the preceding 12 months. It is
399 possible that duration of exposure to various healthy or unhealthy dietary habits may modify
400 the observed relationships. However, the available data does not permit such evaluations in
401 the current analysis. We were unable to consider more granular details regarding covariates
402 (e.g., hypertension) which may have important disease modifying effects, for example, we
403 are unable to distinguish individuals with poorly controlled disease or those with evidence of
404 end-organ damage. It is also possible that certain medications have a modifying effect on
405 associations with cardiovascular phenotypes. Information on medication in the UK Biobank
406 is recorded at baseline based on self-report, the completeness and accuracy of this data cannot
407 be verified against clinical records, nor can links be definitively made to specific conditions.
408 As such, we have not taken into account potential effect of medications on the observed
409 relationships. These would be important considerations in future work. Furthermore, as the

410 UK Biobank participants in this analysis are predominantly of White ethnic background
411 (97%), we cannot be certain that observed associations are generalisable across different
412 ethnicities. With regards the radiomics analysis, the reproducibility of these features is highly
413 susceptible to variations in image segmentation. This is a major challenge with radiomics
414 analysis, particularly when the goal is to develop generalisable clinical models. In the present
415 study, we use radiomics for characterising associations with deeper cardiac phenotypes, as
416 the goal is not to produce a clinical model for application to external datasets, the
417 reproducibility issues are less relevant here. The potential effect of poor reproducibility in the
418 present study would be to introduce noise into radiomics features with possible attenuation of
419 some association. Another limitation is that the relatively novel approach of radiomics,
420 although providing unique information, is difficult to interpret and so any conclusions will be
421 descriptive and rather speculative at this stage. Finally, due to the observational nature of the
422 study, we are unable to exclude residual confounding or infer causality.

423

424 **Conclusion**

425 Greater consumption of red and processed meat is associated with poorer cardiovascular
426 health characterised in terms of CMR cardiac structure and function, novel radiomics
427 features, and measures of arterial compliance from CMR and plethysmography. Our findings
428 support previous clinical associations and provide greater insight into potential mechanisms
429 of dietary impact on cardiovascular health.

430

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432 Data access was granted through UK Biobank access application 2964. Figure 2 was created
433 with BioRender.com. ZRE, SEP, and NCH conceived the idea and designed the study. CM
434 performed statistical analysis. JC cross-checked and advised on statistical analysis. PG led
435 the radiomics analysis. AJ contributed to the radiomics analysis. KL supervised radiomics
436 analysis. LB provided aortic distensibility measures from automated analysis pipeline. NA
437 provided cardiac measures from automated analysis pipeline. ZRE wrote the manuscript. All
438 co-authors provided read and provided critical appraisal of the manuscript.

439

440 **Conflict of interest:** None declared

441

References

- 442 1. Bovalino S, Charleson G, Szoeka C. The impact of red and processed meat
443 consumption on cardiovascular disease risk in women. *Nutrition* (2016) **32**:349–354.
444 doi:10.1016/j.nut.2015.09.015
- 445 2. Sinha R, Cross AJ, Graubard BI, Leitzmann MF, Schatzkin A. Meat Intake and
446 Mortality A Prospective Study of Over Half a Million People. *Arch Intern Med* (2009)
447 **169**:562–571. doi:10.1001/archinternmed.2009.6
- 448 3. Kontogianni MD, Panagiotakos DB, Pitsavos C, Chrysohoou C, Stefanadis C.
449 Relationship between meat intake and the development of acute coronary syndromes:
450 The CARDIO2000 case - Control study. *Eur J Clin Nutr* (2008) **62**:171–177.
451 doi:10.1038/sj.ejcn.1602713
- 452 4. Bernstein AM, Sun Q, Hu FB, Stampfer MJ, Manson JE, Willett WC. Major dietary
453 protein sources and risk of coronary heart disease in women. *Circulation* (2010)
454 **122**:876–883. doi:10.1161/CIRCULATIONAHA.109.915165
- 455 5. Haring B, Wang W, Fretts A, Shimbo D, Lee ET, Howard B V., Roman MJ, Devereux
456 RB. Red meat consumption and cardiovascular target organ damage (from the Strong
457 Heart Study). *J Hypertens* (2017) **35**:1794–1800.
458 doi:10.1097/HJH.0000000000001385
- 459 6. Zhong VW, Van Horn L, Greenland P, Carnethon MR, Ning H, Wilkins JT, Lloyd-
460 Jones DM, Allen NB. Associations of Processed Meat, Unprocessed Red Meat,
461 Poultry, or Fish Intake with Incident Cardiovascular Disease and All-Cause Mortality.
462 *JAMA Intern Med* (2020) **180**:503–512. doi:10.1001/jamainternmed.2019.6969
- 463 7. Wolk A. Potential health hazards of eating red meat. *J Intern Med* (2017) **281**:106–
464 122. doi:10.1111/joim.12543
- 465 8. Organ CL, Otsuka H, Bhushan S, Wang Z, Bradley J, Trivedi R, Polhemus DJ, Tang
466 WHW, Wu Y, Hazen SL, et al. Choline Diet and Its Gut Microbe-Derived Metabolite,
467 Trimethylamine N-Oxide, Exacerbate Pressure Overload-Induced Heart Failure. *Circ
468 Hear Fail* (2016) **9**:e002314. doi:10.1161/CIRCHEARTFAILURE.115.002314
- 469 9. Steffen LM, Kroenke CH, Yu X, Pereira MA, Slattery ML, Van Horn L, Gross MD,
470 Jacobs DR. Associations of plant food, dairy product, and meat intakes with 15-y
471 incidence of elevated blood pressure in young black and white adults: The Coronary
472 Artery Risk Development in Young Adults (CARDIA) Study. *Am J Clin Nutr* (2005)
473 **82**:1169–1177. doi:10.1093/ajcn/82.6.1169
- 474 10. Wolmarans P, Benadé AJ, Kotze TJ, Daubitzer AK, Marais MP, Laubscher R. Plasma
475 lipoprotein response to substituting fish for red meat in the diet. *Am J Clin Nutr* (1991)
476 **53**:1171–6. doi:10.1093/ajcn/53.5.1171
- 477 11. Koeth RA, Wang Z, Levison BS, Buffa JA, Org E, Sheehy BT, Britt EB, Fu X, Wu Y,
478 Li L, et al. Intestinal microbiota metabolism of l-carnitine, a nutrient in red meat,
479 promotes atherosclerosis. *Nat Med* (2013) **19**:576–585. doi:10.1038/nm.3145
- 480 12. Raisi-Estabragh Z, Petersen SE. Cardiovascular research highlights from the UK
481 Biobank: opportunities and challenges. *Cardiovasc Res* (2020) **116**:e12–e15.
482 doi:10.1093/cvr/cvz294
- 483 13. UK Biobank: Protocol for a large-scale prospective epidemiological resource. (2007)
484 Available at: <https://www.ukbiobank.ac.uk/wp-content/uploads/2011/11/UK-Biobank->

- 485 Protocol.pdf [Accessed December 13, 2019]
- 486 14. Raisi-Estabragh Z, Harvey NC, Neubauer S, Petersen SE. Cardiovascular magnetic
487 resonance imaging in the UK Biobank: a major international health research resource.
488 *Eur Hear J - Cardiovasc Imaging* (2020)jea297. doi:10.1093/ehjci/jeaa297
- 489 15. Schenker S. Portion sizes Food Fact Sheet. (2016) Available at:
490 www.bda.uk.com/foodfacts [Accessed April 25, 2020]
- 491 16. Anderson JJ, Darwis NDM, Mackay DF, Celis-Morales CA, Lyall DM, Sattar N, Gill
492 JMR, Pell JP. Red and processed meat consumption and breast cancer: UK Biobank
493 cohort study and meta-analysis. *Eur J Cancer* (2018) **90**:73–82.
494 doi:10.1016/j.ejca.2017.11.022
- 495 17. Petersen SE, Matthews PM, Francis JM, Robson MD, Zemrak F, Boubertakh R,
496 Young AA, Hudson S, Weale P, Garratt S, et al. UK Biobank’s cardiovascular
497 magnetic resonance protocol. *J Cardiovasc Magn Reson* (2015) **18**:8.
498 doi:10.1186/s12968-016-0227-4
- 499 18. Petersen SE, Aung N, Sanghvi MM, Zemrak F, Fung K, Paiva JM, Francis JM, Khanji
500 MY, Lukaschuk E, Lee AM, et al. Reference ranges for cardiac structure and function
501 using cardiovascular magnetic resonance (CMR) in Caucasians from the UK Biobank
502 population cohort. *J Cardiovasc Magn Reson* (2017) **19**:18. doi:10.1186/s12968-017-
503 0327-9
- 504 19. Attar R, Pereañez M, Gooya A, Albà X, Zhang L, de Vila MH, Lee AM, Aung N,
505 Lukaschuk E, Sanghvi MM, et al. Quantitative CMR population imaging on 20,000
506 subjects of the UK Biobank imaging study: LV/RV quantification pipeline and its
507 evaluation. *Med Image Anal* (2019) **56**:26–42. Available at:
508 <http://www.ncbi.nlm.nih.gov/pubmed/31154149> [Accessed October 7, 2019]
- 509 20. Raisi-Estabragh Z, Izquierdo C, Campello VM, Martin-Isa C, Jaggi A, Harvey NC,
510 Lekadir K, Petersen SE. Cardiac magnetic resonance radiomics: basic principles and
511 clinical perspectives. *Eur Hear J - Cardiovasc Imaging* (2020) **21**:349–356.
512 doi:10.1093/ehjci/jeaa028
- 513 21. Van Griethuysen JJM, Fedorov A, Parmar C, Hosny A, Aucoin N, Narayan V, Beets-
514 Tan RGH, Fillion-Robin JC, Pieper S, Aerts HJWL. Computational radiomics system
515 to decode the radiographic phenotype. *Cancer Res* (2017) **77**:e104–e107.
516 doi:10.1158/0008-5472.CAN-17-0339
- 517 22. Gonzalez R, Fittes B. 2nd Conference on Remotely Manned Systems: Technology and
518 Applications. in *Gray-level transformations for interactive image enhancement* (Los
519 Angeles, California), 17–19. Available at:
520 <https://ntrs.nasa.gov/archive/nasa/casi.ntrs.nasa.gov/19770022806.pdf>
- 521 23. Laurent S, Cockcroft J, Van Bortel L, Boutouyrie P, Giannattasio C, Hayoz D, Pannier
522 B, Vlachopoulos C, Wilkinson I, Struijker-Boudier H. Expert consensus document on
523 arterial stiffness: Methodological issues and clinical applications. *Eur Heart J* (2006)
524 **27**:2588–2605. doi:10.1093/eurheartj/ehl254
- 525 24. Resnick LM, Militianu D, Cunnings AJ, Pipe JG, Evelhoch JL, Soulen RL. Direct
526 Magnetic Resonance Determination of Aortic Distensibility in Essential Hypertension.
527 *Hypertension* (1997) **30**:654–659. doi:10.1161/01.HYP.30.3.654
- 528 25. Biasioli L, Hann E, Lukaschuk E, Carapella V, Paiva JM, Aung N, Rayner JJ, Werys
529 K, Fung K, Puchta H, et al. Automated localization and quality control of the aorta in

- 530 cine CMR can significantly accelerate processing of the UK Biobank population data.
531 *PLoS One* (2019) **14**:e0212272. doi:10.1371/journal.pone.0212272
- 532 26. Abdullah Said M, Eppinga RN, Lipsic E, Verweij N, van der Harst P. Relationship of
533 arterial stiffness index and pulse pressure with cardiovascular disease and mortality. *J*
534 *Am Heart Assoc* (2018) **7**:e007621. doi:10.1161/JAHA.117.007621
- 535 27. UK Biobank Arterial Pulse-Wave Velocity. (2011). Available at:
536 <https://biobank.ndph.ox.ac.uk/showcase/showcase/docs/Pulsewave.pdf> [Accessed
537 December 4, 2019]
- 538 28. R Core Team (2019). R: A language and environment for statistical computing. R
539 Foundation for Statistical Computing, Vienna, Austria. Available at: [https://www.r-](https://www.r-project.org/)
540 [project.org/](https://www.r-project.org/) [Accessed October 18, 2020]
- 541 29. RStudio: Integrated Development for R. RStudio, Inc., Boston, MA. Available at:
542 <https://rstudio.com/> [Accessed October 18, 2020]
- 543 30. Maechler M. “Finding Groups in Data”: Cluster Analysis Extended Rousseeuw et al. R
544 Packag. version 2.0. (2019) Available at:
545 <https://www.rdocumentation.org/packages/cluster/versions/2.1.0> [Accessed May 3,
546 2020]
- 547 31. Townsend P, Phillimore P, Beattie A. Health and Deprivation: Inequality and the
548 North. *Nurs Stand* (1988) **2**:34–34. doi:10.7748/ns.2.17.34.s66
- 549 32. Craig CL, Marshall AL, Sjöström M, Bauman AE, Booth ML, Ainsworth BE, Pratt M,
550 Ekelund U, Yngve A, Sallis JF, et al. International physical activity questionnaire: 12-
551 country reliability and validity. *Med Sci Sports Exerc* (2003) **35**:1381–95.
552 doi:10.1249/01.MSS.0000078924.61453.FB
- 553 33. Batty GD, Gale CR, Kivimäki M, Deary IJ, Bell S. Comparison of risk factor
554 associations in UK Biobank against representative, general population based studies
555 with conventional response rates: prospective cohort study and individual participant
556 meta-analysis. *BMJ* (2020) **368**:1–8. doi:10.1136/bmj.m131
- 557 34. Fry A, Littlejohns TJ, Sudlow C, Doherty N, Adamska L, Sprosen T, Collins R, Allen
558 NE. Comparison of Sociodemographic and Health-Related Characteristics of UK
559 Biobank Participants With Those of the General Population. *Am J Epidemiol* (2017)
560 **186**:1026–1034. doi:10.1093/aje/kwx246
- 561 35. Levitan EB, Ahmed A, Arnett DK, Polak JF, Hundley WG, Bluemke DA, Heckbert
562 SR, Jacobs DR, Nettleton JA. Mediterranean diet score and left ventricular structure
563 and function: The multi-ethnic study of atherosclerosis. *Am J Clin Nutr* (2016)
564 **104**:595–602. doi:10.3945/ajcn.115.128579
- 565 36. Nguyen HT, Bertoni AG, Nettleton JA, Bluemke DA, Levitan EB, Burke GL. Dash
566 eating pattern is associated with favorable left ventricular function in the multi-ethnic
567 study of atherosclerosis. *J Am Coll Nutr* (2012) **31**:401–407.
568 doi:10.1080/07315724.2012.10720466
- 569 37. Maugeri A, Hruskova J, Jakubik J, Hlinomaz O, Medina-Inojosa JR, Barchitta M,
570 Agodi A, Vinciguerra M. How dietary patterns affect left ventricular structure,
571 function and remodelling: evidence from the Kardiovize Brno 2030 study. *Sci Rep*
572 (2019) **9**:1–10. doi:10.1038/s41598-019-55529-5
- 573 38. Wagner S, Lioret S, Girerd N, Duarte K, Lamiral Z, Bozec E, Van den Berghe L, Hoge

- 574 A, Donneau A-F, Boivin J-M, et al. Association of Dietary Patterns Derived Using
575 Reduced-Rank Regression With Subclinical Cardiovascular Damage According to
576 Generation and Sex in the STANISLAS Cohort. *J Am Heart Assoc* (2020) **9**:e013836.
577 doi:10.1161/JAHA.119.013836
- 578 39. van Baak MA, Larsen TM, Jebb SA, Martinez A, Saris WHM, Handjieva-Darlenska
579 T, Kafatos A, Pfeiffer AFH, Kunešová M, Astrup A. Dietary intake of protein from
580 different sources and weight regain, changes in body composition and cardiometabolic
581 risk factors after weight loss: The DIOgenes study. *Nutrients* (2017) **9**:1326.
582 doi:10.3390/nu9121326
- 583 40. Shang X, Scott D, Hodge AM, English DR, Giles GG, Ebeling PR, Sanders KM.
584 Dietary protein intake and risk of type 2 diabetes: Results from the Melbourne
585 Collaborative Cohort Study and a meta-analysis of prospective studies. *Am J Clin Nutr*
586 (2016) **104**:1352–1365. doi:10.3945/ajcn.116.140954
- 587 41. Wang Z, Bergeron N, Levison BS, Li XS, Chiu S, Jia X, Koeth RA, Li L, Wu Y, Tang
588 WHW, et al. Impact of chronic dietary red meat, white meat, or non-meat protein on
589 trimethylamine N-oxide metabolism and renal excretion in healthy men and women.
590 *Eur Heart J* (2019) **40**:583–594. doi:10.1093/eurheartj/ehy799
- 591 42. Bradbury KE, Young HJ, Guo W, Key TJ. Dietary assessment in UK Biobank: an
592 evaluation of the performance of the touchscreen dietary questionnaire. *J Nutr Sci*
593 (2018) **7**:1–11. doi:10.1017/jns.2017.66
- 594

Table 1. Baseline population characteristics

Male	9,303 (47.9%)
Female	10,105 (52.1%)
Age (years)	55.0 (\pm 7.5)
Townsend deprivation index	-2.0 (\pm 2.6)
Body mass index (kg/m ²)	26.6 (\pm 4.2)
Smoking (current smoker)	1,238 (6.4%)
Diabetes	606 (3.1%)
Hypertension	2,690 (13.9%)
Hypercholesterolaemia	4,464 (23.0%)
IPAQ score (METS/week)	1525.00 [2396.25]
Educational level:	
Left school age 14 or younger without qualifications	53 (0.3%)
Left school age 15 or older without qualifications	1,394 (7.2%)
High school diploma	2,679 (13.8%)
Sixth form qualification	1,114 (5.7%)
Professional qualification	5,506 (28.4%)
Higher education university degree	8,456 (43.6%)
Alcohol intake frequency:	
Never	954 (4.9%)
Special occasions only	1,587 (8.2%)
1-3 times a month	2,103 (10.8%)
1-2 times a week	4,997 (25.7%)
3-4 times a week	5,496 (28.3%)
Daily or almost daily	4,260 (21.9%)
Unprocessed red meat intake (grams/day)	22.3 (\pm 15.2)
Beef	9.5 (\pm 9.0)
Lamb	6.3 (\pm 5.4)
Pork	6.5 (\pm 5.9)
Processed meat intake (grams/day)	15.7 (\pm 15.0)
Oily fish intake (grams/day)	11.7 (\pm 10.8)

Table 1 footnote: Results are frequencies and percentages for categorical variables; mean (standard deviation) or median [interquartile range] for continuous variable. IPAQ: international physical activity questionnaire; METS: metabolic equivalents.

Table 2. Multivariable linear regression models showing change in LV conventional CMR indices per 100g increase in daily meat/fish consumption

	LVEDVi (ml/m ²)	LVESVi (ml/m ²)	LVEF (%)	LVSVi (ml/m ²)	LVMi (g/m ²)
Unprocessed red meat	-2.18* [-3.36, -1.00] 2.91×10 ⁻⁴	-0.96* [-1.70, -0.23] 0.01	0.041 [-0.50, 0.59] 0.88	-1.22* [-1.97, -0.47] 1.50×10 ⁻³	1.57* [0.91, 2.23] 3.19×10 ⁻⁶
Processed meat	-2.88* [-4.12, -1.65] 4.70×10 ⁻⁶	-1.12* [-1.89, -0.35] 4.2×10 ⁻³	-0.12 [-0.69, 0.45] 0.68	-1.77* [-2.55, -0.98] 1.06×10 ⁻⁵	0.57 [-0.12, 1.26] 0.11
Oily fish	4.13* [2.46, 5.80] 1.28×10 ⁻⁶	1.75* [0.71, 2.79] 9.68×10 ⁻⁴	0.10 [-0.67, 0.88] 0.79	2.38* [1.32, 3.45] 1.13×10 ⁻⁵	2.38* [1.44, 3.31] 6.40×10 ⁻⁷

Table 2 footnote: Each cell represents a separate model, adjusted for: age, sex, social deprivation, educational level, smoking, alcohol intake, and exercise level. Results are presented as beta coefficient [95% confidence interval] p-value. CMR: cardiovascular magnetic resonance; LV: left ventricle; LVEDV: LV end-diastolic volume; LVESV: LV end-systolic volume; LVEF: LV ejection fraction; LVM: LV mass; LVSV: LV stroke volume; i denotes indexation to body surface area. *indicates p-value <0.05.

Table 3. Description of clusters identified from the radiomics texture features

Assigned cluster name	Exemplar feature from the cluster	Properties represented by cluster
Low Grey Level Emphasis	Low Grey Level Emphasis	Local distribution and clustering of low SI values
Spatial Non-Uniformity	Size Zone Non-Uniformity	Non-uniformity in the size of pixel groupings
Grey Level Variance	Grey Level Variance	Distribution of SI values
Coarseness	Run Percentage	Tendency to small groupings of pixels with similar SI values
Local Heterogeneity	Dependence Entropy	Randomness of neighbouring pixel SI values
Large Scale Emphasis	Large Area Emphasis	Larger areas of similar pixel SI values
Grey Level Skewness	Cluster Prominence	Skewness of the SI distribution

Table 3 footnote: The table summarises the seven distinct groups of radiomics texture features identified through cluster analysis of these features (n=144, Supplementary Figure 4). Each cluster incorporates a number of inter-correlated features. For each cluster, we provide an assigned name, an exemplar feature, and a general description of the properties represented. SI: signal intensity

Table 4. Multivariate linear regression models showing change in arterial compliance measures per 100g increase in daily meat/fish consumption.

	Aortic distensibility ($\times 10^{-3}$ mmHg ⁻¹)	ASI (baseline, m/s)	ASI (imaging, m/s)	Interval change in ASI (baseline- imaging, m/s)
Unprocessed red meat	-0.06 [-0.13, 0.02] 0.12	0.49* [0.41, 0.57] 2.26×10^{-31}	0.349* [0.15, 0.55] 5.46×10^{-4}	0.150* [0.03, 0.27] 0.02
Processed meat	-0.00 [-0.08, 0.08] 1.00	0.45* [0.36, 0.53] 4.47×10^{-24}	0.22* [0.02, 0.43] 0.03	0.05 [-0.07, 0.17] 0.43
Oily fish	0.01 [-0.09, 0.12] 0.81	-0.22* [-0.34, -0.11] 1.70×10^{-4}	-0.43* [-0.71, -0.16] 2.20×10^{-3}	-0.17 [-0.34, 0.01] 0.06

Table 4 footnote: Each cell represents a separate model, adjusted for: age, sex, social deprivation, educational level, smoking, alcohol intake, and exercise level. For ‘interval change in ASI’, results are average standard deviation change from that expected from baseline. Results are presented as beta coefficient [95% confidence interval]; p-value. ASI: arterial stiffness index. *indicates p-value <0.05.

FIGURE LEGENDS

Figure 1 footnote: (A) Average silhouette statistic for complete-linkage hierarchical clustering of texture feature correlations. The silhouette statistic reflects the average distance between data points in the same cluster compared against average data points in other clusters and allows judgement of the optimal number of clusters within a sample, such that distance between datapoints within clusters are minimised whilst maximising distance with datapoints from other clusters. We computed average silhouette statistic for 2 to 10 clusters. Maximum silhouette statistic was observed with 7 and 8 clusters. Hence, we take 7 clusters as representing the optimal number of clusters within our samples. (B) Correlation heatmap, rows and columns correspond to all texture features creating grid with all possible pairs of texture features, grid colour corresponds to Pearson Correlation between pair of features at that point. Grid rows re-ordered by hierarchical clustering of correlations with tree coloured for optimal seven cluster cut of the tree.

Figure 2 footnote: None required.

Figure 3 footnote: Each bar represents standardised beta coefficients corresponding to the indicated radiomics shape feature. Each bar is from a separate model adjusted for age, sex, social deprivation, educational level, smoking, alcohol intake, exercise level. Black lines represent half-length of confidence interval for the corresponding bar. Asterix denotes significant association. Bonferroni adjusted significance threshold p-value =0.001 (corrected for 39 comparisons). CMR: cardiovascular magnetic resonance; LV: left ventricle

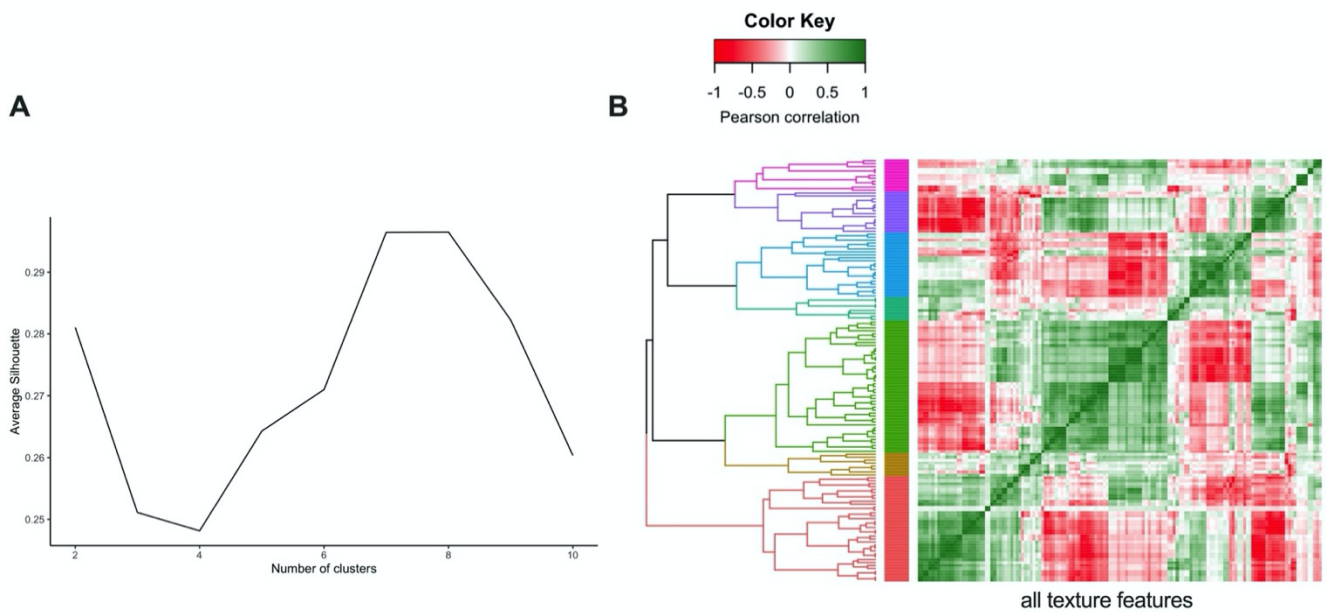
595 **Figure 4 footnote:** Each bar represents standardised beta coefficients corresponding to the indicated
596 radiomics shape feature. Each bar is from a separate model adjusted for age, sex, social deprivation,
597 educational level, smoking, alcohol intake, exercise level. Black lines represent half-length of
598 confidence interval for the corresponding bar. Bonferroni adjusted significance threshold p-value
599 =0.0009 (corrected for 54 comparisons).CMR: cardiovascular magnetic resonance; LV: left ventricle

600 **Figure 5 footnote:** Each bar represents mean standardised beta coefficients corresponding to the
601 indicated texture feature cluster. Models are adjusted for age, sex, social deprivation, educational
602 level, smoking, alcohol intake, exercise level (confounder adjusted model). Black lines represent half-
603 length of confidence interval for the corresponding bar. Bonferroni adjusted significance threshold p-
604 value =0.002 (corrected for 21 comparisons).CMR: cardiovascular magnetic resonance; LV: left
605 ventricle *denotes $p < 0.05$ in Mann-Whitney U Test between oily fish and unprocessed red meat and
606 between oily fish and processed red meat.

Figure 6 footnote: Greater red and processed meat intake was associated with smaller ventricular volumes, reduced short axis dimension, and a less elongated shape; lower global signal intensity levels, and less variation in SI levels within the LV myocardium. Greater oily fish consumption was associated with larger ventricles with overall more elongated shape, higher global myocardial intensity levels and more variation of myocardial intensities. CMR: cardiovascular magnetic resonance. *Histograms are from a selection of most illustrative cases and do not represent findings from the whole dataset.

607 **Figure 7 footnote:** Each bar is from a separate model adjusted for age, sex, social deprivation,
608 educational level, smoking, alcohol intake, exercise level (confounder adjusted model). AD: aortic
609 distensibility; ASI: arterial stiffness index.

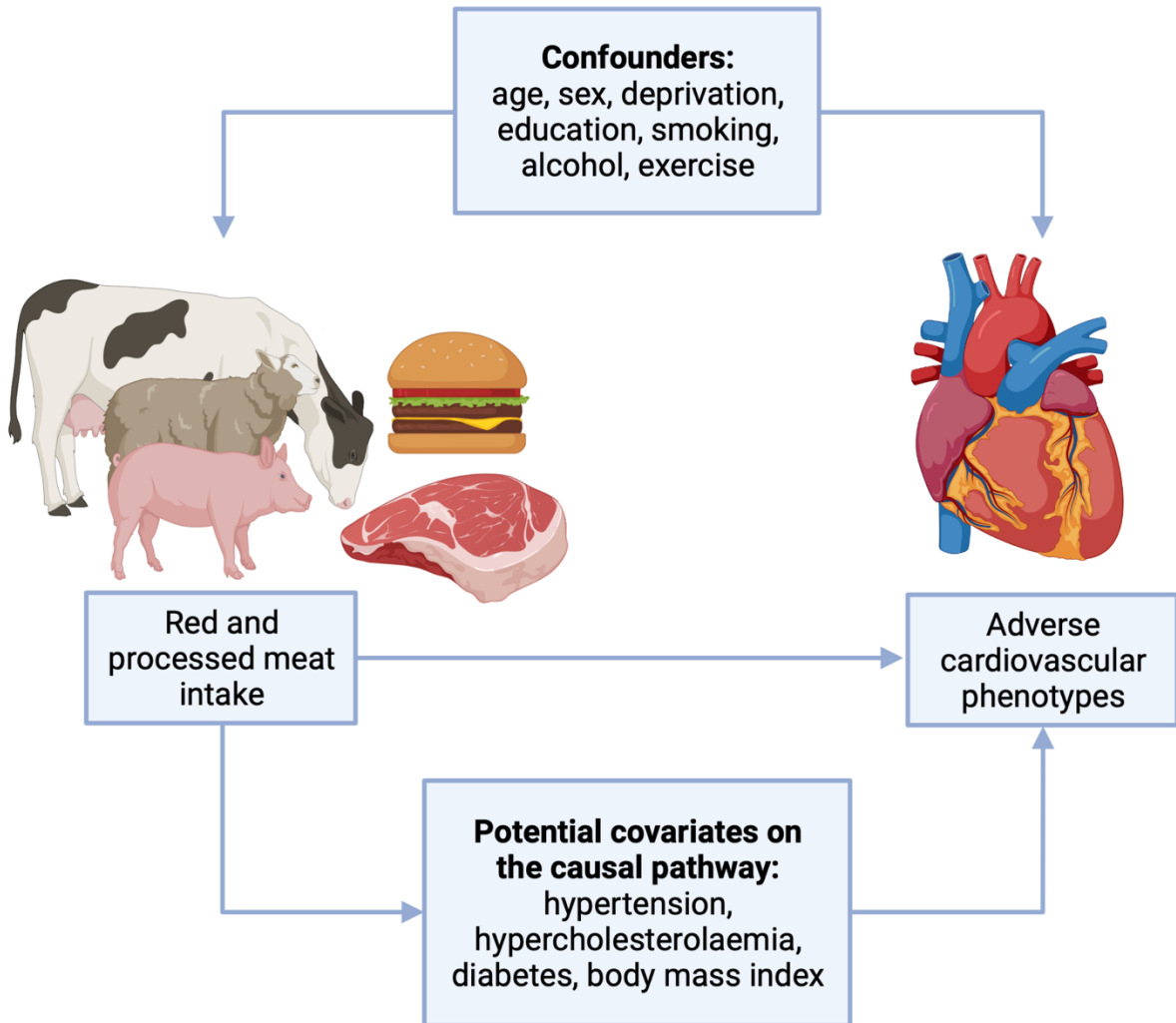
610 **Figure 1. Illustration of clustering method (hierarchical) and approach to defining the number**
611 **of clusters (average silhouette approach) for the LV myocardium radiomics texture features**



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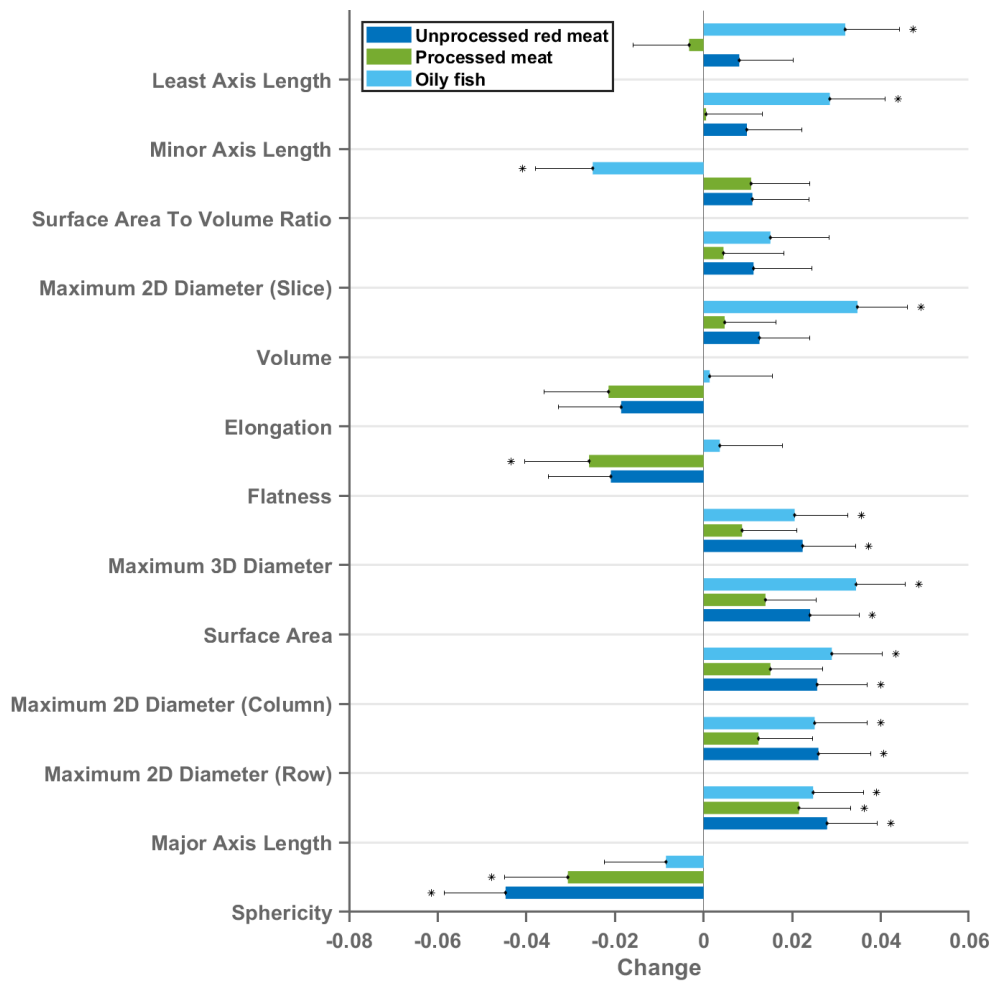
613 **Figure 1 footnote:** (A) Average silhouette statistic for complete-linkage hierarchical clustering of
614 texture feature correlations. The silhouette statistic reflects the average distance between data points
615 in the same cluster compared against average data points in other clusters and allows judgement of the
616 optimal number of clusters within a sample, such that distance between datapoints within clusters are
617 minimised whilst maximising distance with datapoints from other clusters. We computed average
618 silhouette statistic for 2 to 10 clusters. Maximum silhouette statistic was observed with 7 and 8
619 clusters. Hence, we take 7 clusters as representing the optimal number of clusters within our samples.
620 (B) Correlation heatmap, rows and columns correspond to all texture features creating grid with all
621 possible pairs of texture features, grid colour corresponds to Pearson Correlation between pair of
622 features at that point. Grid rows re-ordered by hierarchical clustering of correlations with tree
623 coloured for optimal seven cluster cut of the tree.

Figure 2. Covariates considered in the relationship between red and processed meat consumption and cardiovascular phenotypes



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Figure 3. Multivariable linear regression models showing change in LV cavity CMR shape radiomics (end-diastole) per 100g increase in daily meat consumption

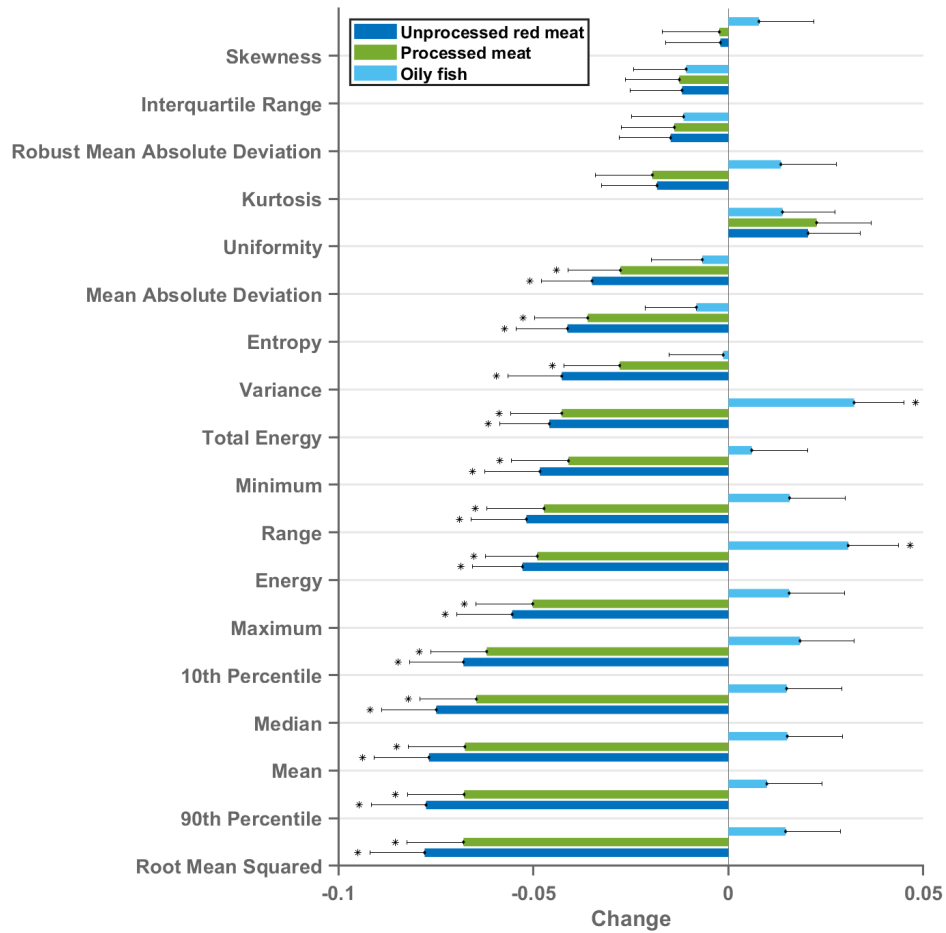


626

Figure 3 footnote: Each bar represents standardised beta coefficients corresponding to the indicated radiomics shape feature. Each bar is from a separate model adjusted for age, sex, social deprivation, educational level, smoking, alcohol intake, exercise level. Black lines represent half-length of confidence interval for the corresponding bar. Asterix denotes significant association. Bonferroni adjusted significance threshold p -value = 0.001 (corrected for 39 comparisons). CMR: cardiovascular magnetic resonance; LV: left ventricle

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Figure 4. Multivariable linear regression models showing change in LV myocardium CMR first-order radiomics (end-diastole) per 100g increase in daily meat consumption

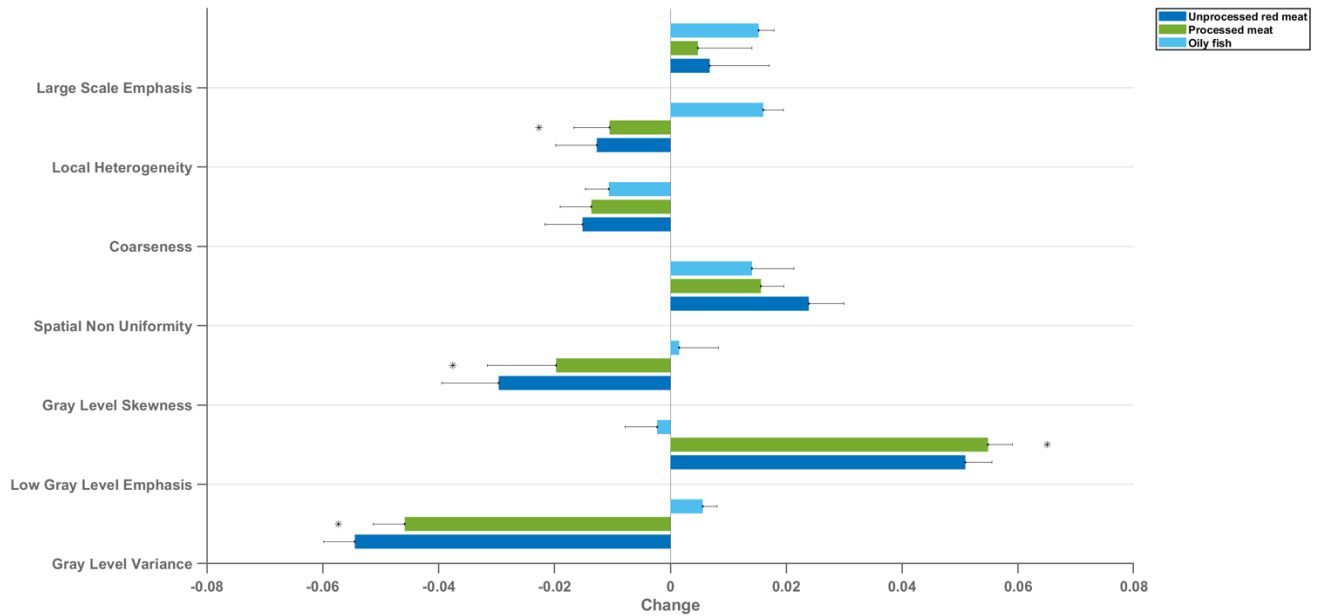


630

631 **Figure 4 footnote:** Each bar represents standardised beta coefficients corresponding to the indicated
 632 radiomics shape feature. Each bar is from a separate model adjusted for age, sex, social deprivation,
 633 educational level, smoking, alcohol intake, exercise level. Black lines represent half-length of
 634 confidence interval for the corresponding bar. Bonferroni adjusted significance threshold p-value
 635 =0.0009 (corrected for 54 comparisons).CMR: cardiovascular magnetic resonance; LV: left ventricle

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Figure 5. Mean change in LV myocardium CMR radiomics texture feature clusters per 100g increase in daily meat consumption



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641 **Figure 5 footnote:** Each bar represents mean standardised beta coefficients corresponding to the
642 indicated texture feature cluster. Models are adjusted for age, sex, social deprivation, educational level,
643 smoking, alcohol intake, exercise level (confounder adjusted model). Black lines represent half-length
644 of confidence interval for the corresponding bar. CMR: cardiovascular magnetic resonance; LV: left
645 ventricle *denotes $p < 0.05$ in using Kruskal-Wallis statistical testing followed by Dunn's correction
646 test for multiple comparisons. between oily fish and unprocessed red meat and between oily fish and
647 processed red meat.

Figure 6. Summary of the association of the oily fish, processed meat, and unprocessed red meat intake with the CMR radiomics shape and signal intensity-based features

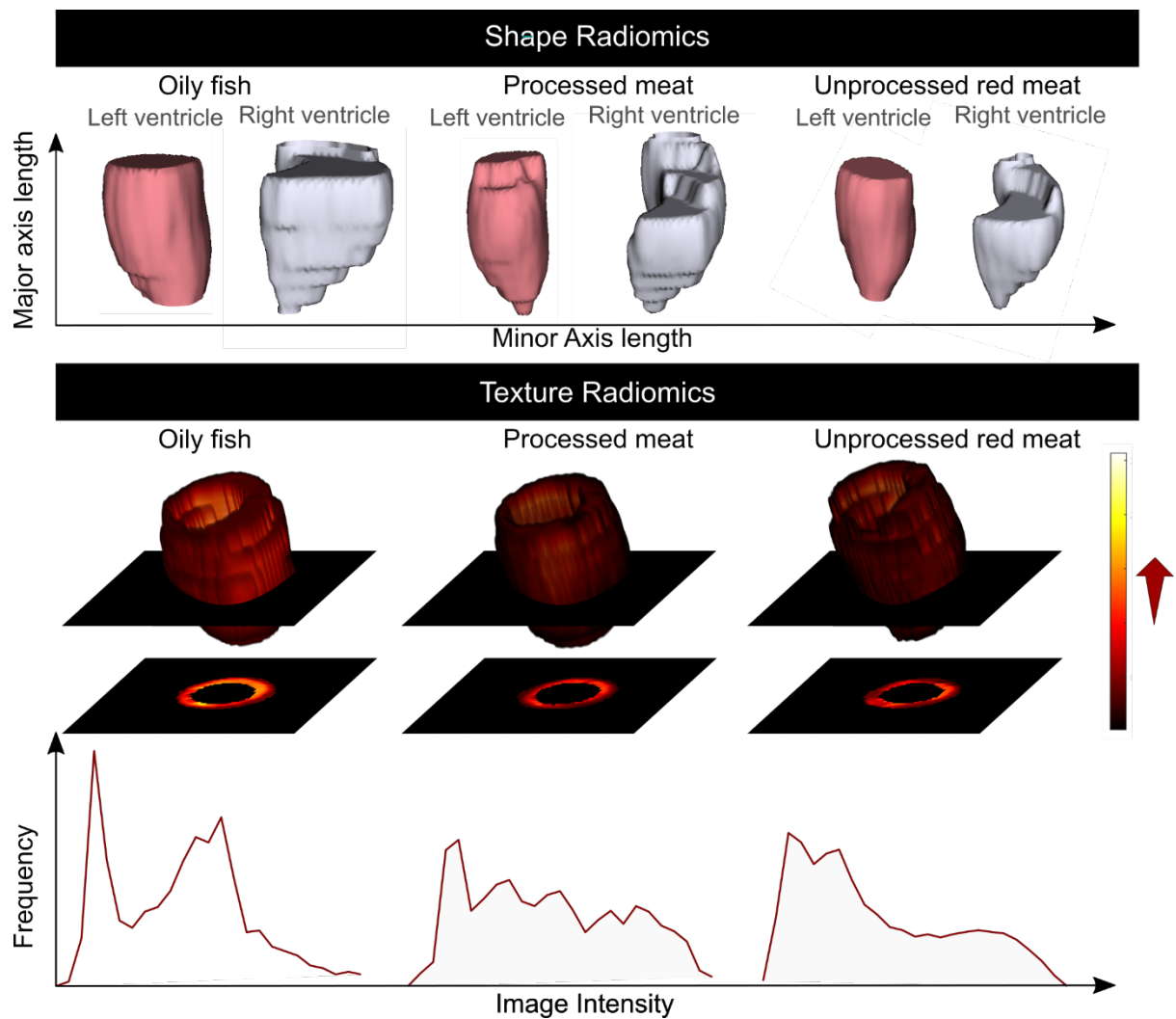


Figure 6 footnote: Greater red and processed meat intake was associated with smaller ventricular volumes, reduced short axis dimension, and a more elongated shape; lower global signal intensity levels, and less variation in SI levels within the LV myocardium. Greater oily fish consumption was associated with larger ventricles with overall less elongated (more spherical) shape, higher global myocardial intensity levels and more variation of myocardial intensities. CMR: cardiovascular magnetic resonance. *Histograms are from a selection of most illustrative cases and do not represent findings from the whole dataset.

Figure 7. Summary of multivariable linear regression results for arterial compliance measures displaying beta coefficients and 95% confidence intervals per 100g increase in daily intake of meat/fish

Linear Regression Coefficients (95% Confidence Intervals) - Main Model Covariates

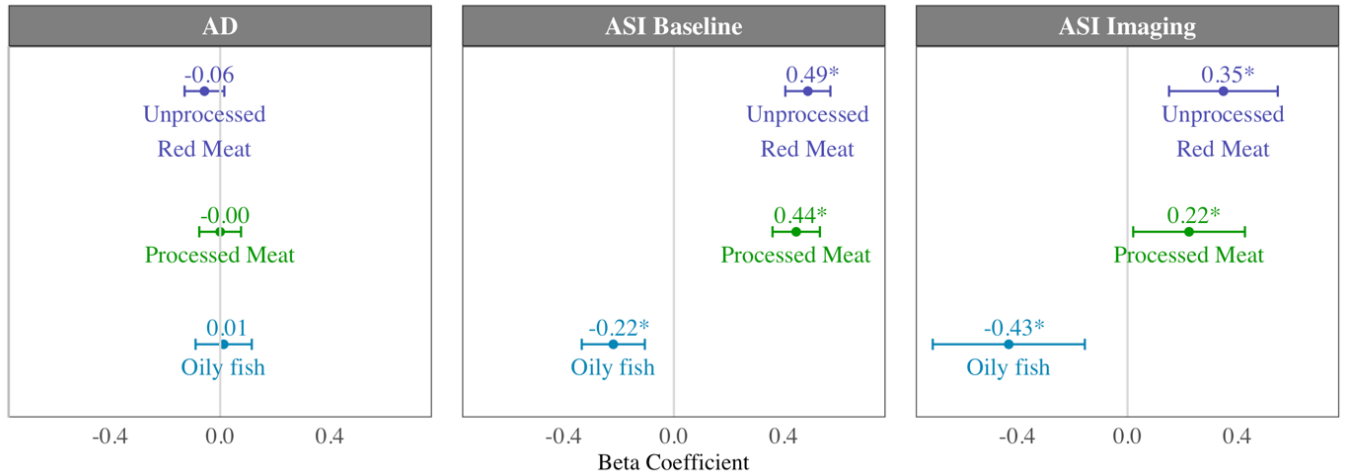


Figure 7 footnote: Each bar is from a separate model adjusted for age, sex, social deprivation, educational level, smoking, alcohol intake, exercise level (confounder adjusted model). AD: aortic distensibility; ASI: arterial stiffness index.