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1 **Glomerular filtration rate by differing measures, albuminuria and prediction of cardiovascular**
2 **disease, mortality and end-stage kidney disease**

3

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18

19 **Abstract**

20

21 Chronic kidney disease is common in the general population and associated with excess
22 cardiovascular disease (CVD), but kidney function does not feature in current CVD risk prediction
23 models. We tested three formulae for estimated glomerular filtration rate (eGFR) to determine the
24 most clinically informative for predicting CVD and mortality. Using data from 440,526 participants
25 from UK Biobank, eGFR was calculated using serum creatinine, cystatin C (eGFR_{cys}) and creatinine-
26 cystatin C. Associations of each eGFR with CVD outcome and mortality were compared using Cox
27 models adjusting for atherosclerotic risk factors (per relevant risk scores), and predictive utility was
28 determined by the C-statistic and categorical Net Reclassification Index. We show that eGFR_{cys} is
29 most strongly associated with CVD and mortality, and along with albuminuria adds predictive
30 discrimination to current CVD risk scores, whilst traditional creatinine-based measures are weakly
31 associated with risk. Clinicians should consider measuring eGFR_{cys} as part of cardiovascular risk
32 assessment.

33 **Introduction**

34

35 Chronic kidney disease (CKD), characterized by gradual loss of kidney function over time, is
36 associated with progression to end-stage kidney disease (ESKD) and is an indicator for renal
37 replacement therapy, premature cardiovascular disease (CVD) and mortality. CKD is common in the
38 general population and affects around 10% of the total population, although estimates of its
39 prevalence vary.¹ Although impaired kidney function indicates patients at risk of future requirement of
40 renal replacement therapy, the vast majority of patients do not require renal replacement therapy, and
41 for these the greatest risk of CKD is the excess in cardiovascular risk associated with CKD.² The
42 excess CVD risk appears to begin in the relatively early stages of CKD and increases with CKD
43 stage, especially in CKD stages G3b-5.²⁻⁴ The most pronounced effects are seen in those with
44 advanced CKD or ESKD when the complications specific to CKD - including renal anaemia,
45 disordered acid-base balance and CKD mineral and bone disorder - are most apparent.⁵⁻⁷ Though the
46 atherosclerotic risk factors for CVD – such as hypertension, diabetes, dyslipidaemia and smoking –
47 are prevalent in those with CKD, there is an excess CVD risk seen in CKD beyond that captured by
48 atherosclerotic risk factors alone.⁸⁻¹²

49 Large individual patient-level meta-analyses of cohorts combining observational and clinical trial data,
50 from the CKD Prognosis Consortium, confirm CKD as an additional, independent risk factor for
51 CVD.^{13,14} Albuminuria, an important feature of CKD and glomerular damage, is thought to be
52 associated with increased CVD risk and all-cause mortality. In additional analyses from the CKD
53 Prognosis Consortium, albuminuria particularly improved CVD risk prediction above current methods
54 in patients with CKD.¹⁵ As far as we are aware, only one risk calculator includes presence of CKD 3-5
55 as a binary risk factor for CVD risk estimation (QRISK3¹⁶), whereas other modern calculators in
56 Europe and the United States do not include CKD as a risk (Systematic Coronary Risk Evaluation
57 (SCORE)¹⁷, American Heart Association/American College of Cardiology (AHA/ACC)¹⁸).

58 The CKD-EPI formula¹⁹ to estimate glomerular filtration rate (GFR) has traditionally used serum
59 creatinine values, adjusted for age, sex and ethnicity, however, in retrospective analyses, CKD-EPI
60 formulae using cystatin C, either alone or in combination with creatinine, perform better than those
61 including creatinine alone in estimating GFR²⁰ and in predicting risk of ESKD, CVD and mortality.²¹

62 Cystatin C testing has been available in UK National Health Service laboratories for over 10 years,
63 and has a number of potential advantages over serum creatinine: it is released by every cell in the
64 body, is freely filtered at the glomerulus, and is not influenced by body habitus, muscle mass or
65 gender. It is thus thought to be a more sensitive measure to estimate kidney function. However,
66 owing to the costs of the reagents²², cystatin C is around 10 times more expensive than serum
67 creatinine at £2.50 (USD 3.00) per test compared with £0.25 (USD 0.30) for serum creatinine²³.
68 While this seems a significant additional expense, cystatin C costs about the same or less other
69 standard tests conducted in patients with CKD, including parathyroid hormone, C-reactive protein and
70 vitamin D²². Despite being recommended by the National Institute for Health and Care Excellence
71 (NICE) in the United Kingdom for confirmatory testing for CKD²⁴, measurement of cystatin C has not
72 been widely adopted in clinical practice, presumably relating to uncertainty around the added value of
73 a more expensive biomarker.

74 UK Biobank is one of the largest prospective population-based cohorts anywhere with extensive
75 participant phenotyping and sampling of baseline biochemical measures including renal function
76 (creatinine and cystatin C), albuminuria (urine albumin:creatinine ratio; uACR) and lipids in around
77 500,000 participants. Using data from UK Biobank, we aimed to determine whether estimated GFR
78 (eGFR) and albuminuria improves risk prediction for all-cause mortality, and CVD, using serum
79 creatinine (eGFR_{cr}), cystatin C (eGFR_{cys}) and combined cystatin C-creatinine (eGFR_{cr-cys})
80 estimates of GFR.

81 **Results**

82 **Demographics of participants**

83 Of the 502,536 participants initially included, 31,283 had missing biochemical data, 260 had prevalent
84 ESKD, 38 had calculated eGFR (any measure) <15 ml/min/1.73m² and 30,112 had prior CVD and
85 were excluded from further analysis: therefore 440,526 participants were included in the models. Over
86 a median of 8.9 years (Q1-Q3 8.2-9.5 years) of follow-up, 15,469 participants died from any cause,
87 2552 of which were deaths from CVD (based on European SCORE¹⁷). There were 8662 incident fatal
88 or nonfatal CVD events (based on AHA/ACC definition¹⁸) and 336 cases of incident renal replacement
89 therapy.

90

91 Participants with lower eGFR measures tended to be older, male, smokers with lower diastolic blood
92 pressure (DBP), total and high-density lipoprotein (HDL) cholesterol, be on antihypertensive and
93 statin medications, to report diabetes, and to be in the highest category of uACR (Table 1 and
94 Supplementary Tables S1 and S2). eGFR_{cys} provided lower estimates than both eGFR_{cr} and
95 eGFR_{cr-cys}, resulting in a greater proportion of participants categorised as having CKD G3-5
96 (eGFR_{cr} 1.9%; eGFR_{cys} 4.0%; eGFR_{cr-cys} 1.3%).

97

98 All eGFR measures were strongly correlated, with the strongest being between eGFR_{cys} and
99 eGFR_{cr-cys} ($r=0.925$, $p<0.001$) and weakest between the eGFR_{cr} and eGFR_{cys} ($r=0.599$, $p<0.001$).

100 The correlation between eGFR_{cr} and eGFR_{cr-cys} was $r=0.828$ ($p<0.001$).

101 **Associations between eGFR measures and all-cause mortality, CVD and ESKD outcomes**

102 Unadjusted survival plots for the outcomes of interest across eGFR_{cr} categories are shown in
103 Extended Data Figures E1-4. The adjusted shape of the associations between each eGFR measure
104 and all-cause mortality plus both CVD outcomes and ESKD were largely linear and negative (Figure
105 1). This was most convincing for eGFR_{cys}, with eGFR_{cr} and eGFR_{cr-cys} demonstrating greater
106 inflections particularly at the lower extremes of eGFR. For ESKD, all three eGFR displayed an initially
107 steep negative linear association until around 70 mL/min, and thereafter continued to decrease but at
108 a lesser gradient (Extended Data Figure E5). Event rates per 100,000 person years for all outcomes
109 were greater with higher eGFR by all measures (Supplementary data, Table S3).

110 Adjusted hazard ratios for each eGFR-outcome combination were consistent with lower eGFR being
111 associated with higher risk of CVD, all-cause mortality, fatal CVD and ESKD (Table 2). For each
112 outcome, there was a trend that the HR was stronger for measures of eGFR incorporating cystatin C
113 for each 10ml/min/1.73m² increase (Table 2), which increased further alongside each 1 standard
114 deviation increase in eGFR. The strongest associations (lowest HR per 1 SD difference) were
115 consistently found for eGFR_{cys} (HRs 0.72, 0.80, 0.66, 0.14, for all-cause mortality, composite CVD,
116 fatal CVD and ESKD/ renal replacement therapy outcomes, respectively, Table 2).

117 **Prediction of all-cause mortality and CVD with albuminuria**

118 There was an association between decreasing eGFR and increased hazard ratio for all outcomes that
119 was sustained across uACR groups, though the magnitude of the association was similar in higher

120 compared with lower uACR groups (Figure 2 and Supplementary data Table S4). As expected, the
121 risk of ESKD increases with higher uACR category (Figure 2 and Supplementary data Table S4).
122 Addition of uACR to atherosclerotic risk factors and eGFR was associated with mortality and CVD
123 across the full cohort (Supplementary data Tables S5-S7). Figure 2 shows heat maps for prediction
124 of all-cause mortality and CVD events using AHA/ACC and SCORE criteria using eGFRcys and
125 uACR groups. Addition of albuminuria to eGFRcys and atherosclerotic risk factors did not improve
126 Net Reclassification Index across 7.5% 10 year risk threshold used in AHA/ACC guidelines¹⁸ (Table
127 3).

128 **Prediction of all-cause mortality and CVD outcomes by eGFR measures**

129 For all-cause mortality, atherosclerotic risk factors (age, sex, ethnicity, systolic blood pressure,
130 diastolic blood pressure, antihypertensive medications, statin use, smoking, diabetes, total and HDL
131 cholesterol) yielded a C-statistic of 0.7157 (95% CI 0.7115-0.7200), Figure 3). Addition of both
132 eGFRcys and eGFRcr-cys significantly improved discrimination, with the largest improvement seen
133 with eGFRcys (C-statistic +0.0103, 95% CI 0.0087-0.0121). Similarly, for the composite fatal/non-
134 fatal CVD outcome (based on AHA/ACC risk score), atherosclerotic risk factors yielded a C-statistic of
135 0.7387 (95% CI 0.7337-0.7439) which was improved by addition of eGFRcys (C-statistic + 0.0039,
136 95% CI 0.0025-0.0052). Similarly, and for fatal CVD (based on SCORE), atherosclerotic risk factors
137 (C-statistic 0.7828, 95% CI 0.7740-0.7917) were improved by addition of the eGFRcys (C-statistic
138 change +0.0085 (0.0049-0.0122).

139 For all-cause mortality, and both CVD outcomes, addition of eGFRcr did not improve discrimination.
140 For the CVD outcome, we tested improvement in risk classification across the 7.5% 10 year risk
141 threshold for statin therapy used in AHA/ACC guidelines.¹⁸ eGFRcr did not improve risk classification,
142 but measures incorporating eGFRcys did (Table 3).

143 **Subgroup analyses: prediction of all-cause mortality and CVD**

144 Across all subgroups, eGFRcr did not improve prediction of all-cause mortality or either CVD outcome
145 in addition to atherosclerotic risk factors. eGFRcys improved prediction of all-cause mortality in all
146 subgroups (Supplementary Table S5), and prediction of at least one CVD outcome in subgroups
147 except non-white ethnicity and CKD3b/4 (Supplementary Tables S6 and S7). In a model
148 incorporating atherosclerotic risk factors and uACR, eGFRcys improved prediction of mortality across

149 all subgroups (Supplementary Table S5); eGFRcys improved prediction of at least one CVD outcome
150 in subgroups except non-white ethnicity, CKD 3b/4 and BMI 30-35 kg/m² (Supplementary Tables S6
151 and S7). As a sensitivity analysis for data linearity, the performance of each eGFR was tested for all
152 outcomes in participants with eGFRcr <60ml/min/1.73m² with consistent conclusions: eGFRcr did not
153 improve outcome prediction for all-cause mortality or either CVD outcome; addition of eGFRcys most
154 strongly improved model prediction across all 4 outcomes (Supplementary Table S8). Across all
155 models in and in every subgroup, the likelihood ratio test comparing addition of uACR to models
156 containing atherosclerotic risk factors and eGFR produced p value <0.001.

157 **Discordance analysis**

158 There was absolute discordance $\geq 20\%$ between eGFRcr and eGFRcys measurements in 183,867
159 participants (41.47%). Baseline characteristics were broadly similar amongst those with discordant
160 compared with concordant eGFRcr and eGFRcys (Supplementary data Table S9). Amongst those
161 with discordant results, eGFRcys remained the marker with the greatest improvement in C-statistic,
162 adjusted for atherosclerotic risk factors and uACR, across mortality and both CVD outcome measures
163 (Table 4). None of the eGFR measures improved prediction over atherosclerotic risk factors and
164 uACR for ESKD (Table 4), though there were very few ESKD events (n=52) in this subgroup and the
165 impact of eGFR in this discordant group may be under-estimated.

166 **Discussion**

167 This is the largest prospective cohort study to demonstrate additional reclassification of
168 cardiovascular risk using eGFRcys, with no added predictive value of traditional eGFRcr. We also
169 demonstrate eGFRcys to be more strongly associated with future CVD events than eGFRcr for both
170 CVD outcomes, and this message is consistent across most subgroup analyses. It is not surprising
171 that baseline kidney function and albuminuria are associated with increased risk of ESKD. In this
172 cohort, broadly representative of the general UK population, eGFRcys was more closely associated
173 with this outcome than eGFRcr – the measure used in current clinical practice.

174 Estimation of renal function from creatinine has become more reliable since creatinine measurements
175 were standardised to IDMS-traceable techniques²⁵, however, creatinine remains an imperfect tool to
176 estimate GFR. Released from the breakdown of muscle tissue, creatinine will be more abundant in
177 those with extremes of muscle mass, body habitus or dietary habits. eGFRcr will therefore tend to

178 overestimate GFR (and underestimate presence or severity of CKD) in older individuals or those with
179 less than average muscle mass for their age and is an insensitive marker of kidney impairment in
180 early disease. Cystatin C is a small protein produced by all nucleated cells (so is less susceptible to
181 influence by individual patient characteristics) and is freely filtered at the glomerulus. Cystatin C is
182 thought to be a more sensitive blood marker of kidney function than is creatinine, and is not
183 influenced by muscle mass, age, gender or ethnicity.

184 Cystatin C, however, is correlated with oxidative stress and inflammation^{26,27}. Various
185 cardiometabolic conditions are associated with higher levels of cystatin C; we have adjusted for some
186 within the current risk prediction tools (diabetes, dyslipidaemia, hypertension). Patients with other
187 diagnosed or undiagnosed cardiometabolic conditions, such as thyroid disease, cancer and
188 glucocorticoid therapy, may have slightly higher levels of cystatin C, confounding the observed effect
189 of lower eGFR_{cys} on outcome. This may also explain the difference we observed in the strong
190 aetiological association between eGFR_{cys} and outcome with modest improvement in risk prediction²⁸.
191 Observed effects of lower eGFR on cardiovascular risk may be mediated by inadequate adjustment
192 for atherosclerotic risk factors in patients with CKD in current risk prediction models. The contribution
193 of non-traditional CVD risk factors in those with moderate CKD may be less pronounced than
194 previously thought and use of eGFR_{cys} may serve as a biomarker that better captures risk associated
195 with other cardiometabolic conditions.

196 Amongst patients with CKD, CVD risk-reduction relies on three important factors: 1) accurate
197 diagnosis of CKD; 2) recognition of the elevated risk of CVD; and 3) early identification and treatment
198 of modifiable risk factors.

199 In the UK, NICE guidelines recommend considering a cystatin C-based calculator to confirm or refute
200 a diagnosis of CKD²⁴. This recommendation is not being undertaken in routine clinical practice and
201 the diagnostic and prognostic utility of cystatin C in CKD stage G3 is currently being tested in a
202 prospective study.²⁹ We confirm the observation that reduced kidney function and albuminuria
203 indicate groups at extremely high risk of need for future renal replacement therapy, as outlined in the
204 Kidney Failure Risk Equation.³⁰ Our data show that eGFR_{cys} identified a greater proportion of
205 participants with CKD G3-5 than both eGFR_{cr} and eGFR_{cr-cys}, and is more strongly associated with
206 clinical outcome than either of the other measures. On this basis, we suggest that eGFR_{cys} should

207 be used for diagnosis of CKD. This could apply both for confirmation of CKD in patients with eGFRcr
208 <60ml/min/1.73m² or for screening for CKD in patients with CVD risk factors, such as diabetes,
209 hypertension or obesity.

210 A more accurate diagnosis of CKD should then prompt assessment and treatment of modifiable risk
211 factors including diabetes mellitus, hypertension, smoking and dyslipidaemia. The efficacy of statins
212 in reducing atherosclerotic CVD effects has been demonstrated in randomised clinical trials^{31,32}.
213 However, the effectiveness of statins in reducing risk of CVD in patients with CKD reduces as eGFR
214 declines, partly as the mode of CVD events in advanced CKD becomes less driven by atherosclerotic
215 effects and more by heart failure and sudden cardiovascular death. Therefore, cholesterol-lowering
216 treatments are recommended for primary prevention in all patients with non-dialysis dependent CKD
217 who are over 50 years or any patient with CKD and diabetes³³. Similar arguments pertain to the
218 treatment of blood pressure, which is a risk factor for both future CVD and CKD progression.

219 Accurate documentation of kidney function should prompt consideration of whether to aim for lower
220 blood pressure targets as informed by the SPRINT trial and its CKD subgroup analysis.^{34,35} Whilst the
221 added NRI of 1.54% for CVD in this study seems modest, 1.37% more cases were appropriately
222 identified which, when multiplied by how many people receive CVD risk scoring, would amount to a
223 large number of people worldwide. Indeed, in participants in the UK Biobank, addition of eGFRcys to
224 atherosclerotic risk factors improves prediction and reclassification of CVD more substantially than
225 does addition of total and HDL-C cholesterol to non-lipid atherosclerotic risk factors³⁶. Our results -
226 showing eGFRcys as the most appropriate measure of renal function and predicting cardiovascular
227 diseases - are therefore clinically important. eGFRcys should be incorporated into cardiovascular risk
228 prediction tools.

229 Supporting published data from other large cohort studies^{13,14,37}, we have shown that eGFR is
230 independently associated with CVD events and mortality, and CVD event rates were augmented as
231 eGFR declined³. Our conclusions assume a linear relationship between eGFR measures and
232 outcome, in keeping with other published analyses^{14,21,37}. Both eGFRcys and eGFRcr demonstrated
233 some increase in CVD, mortality and ESKD risk below 90 ml/min/1.73m², however, there was
234 pronounced flattening of the risk association across all outcomes for eGFRcr ~75-90 ml/min/1.73m².
235 eGFRcys demonstrated a much stronger, linear association with risk of mortality, CVD and ESKD

236 below eGFRcys 90 ml/min/1.73m². Within the group with eGFRcys 60-89 ml/min/1.73m², HRs for all-
237 cause mortality and both cardiovascular outcomes are mostly ~1.3 or higher, compared with HRs
238 close to the reference for eGFRcr at the same level. eGFRcys is therefore better suited for early
239 detection of increased risk of these outcomes, both through its strong linear association, and a
240 tendency to estimate lower GFR (and therefore higher risk). Within our population, 37.5% had
241 eGFRcr 60-89, increasing to 46.3% with eGFRcys. This represents a substantial group of patients
242 who could benefit from the added predictive value of eGFRcys - at relatively low cost of £2.50 (USD
243 3.00) per test - particularly if used judiciously at point of CKD diagnosis or for one-off use in
244 cardiovascular risk prediction tools. The UK Biobank population may not be representative of the UK
245 population with CKD G3-5 prevalence <2.0% by eGFRcr or ~4% by eGFRcys; across England, the
246 prevalence of CKD G3-5 in adults is around 6.1%³⁸. Applied to a real-life population with a greater
247 burden of kidney disease, the potential benefit of measuring eGFRcys for risk prediction may be
248 augmented.

249 In keeping with previous analyses from the Chronic Kidney Disease Prognosis Consortium
250 (combining administrative datasets, observational cohorts and clinical trials)^{13,14,37}, we have found
251 albuminuria to be associated with mortality and cardiovascular events. Our hazard ratios for all-cause
252 mortality and fatal cardiovascular disease were broadly similar across uACR groups and eGFR
253 categories to those found in a previous analysis from the Chronic Kidney Disease Prognosis
254 Consortium¹⁴, though we report lower hazard ratios for fatal CVD for those with eGFR >45 and uACR
255 >30mg/mmol, but higher HR for those with eGFR <45 and uACR >30mg/mmol than reported
256 previously¹⁴. In contrast to previous studies, we did not find albuminuria to improve risk prediction
257 over atherosclerotic risk factors and eGFRcys in reclassification models. In the UK Biobank, only
258 3.2% of 440,526 participants had uACR >30mg/mmol, compared to 11.5% of 105,715 participants
259 representing the general population in the Chronic Kidney Disease Prognosis Consortium¹⁴.
260 Furthermore, albuminuria was estimated from random spot urine samples, though early morning
261 samples have been shown to correlate more closely with 24-hour urinary albumin concentration³⁹.
262 The risk associated with heavy albuminuria in our study may have been underestimated.

263

264 The strengths of this study lie in the large scale, prospective population-based cohort containing
265 complete cases and a low proportion (0.06%) of excluded cases due to missing data. The outcomes
266 are obtained from linked health records and through self-reporting measures and therefore are likely
267 to have captured the majority of hard endpoints of interest (CVD and death⁴⁰). The data were
268 obtained from a single-protocol study conducted in accordance with published protocols and
269 measurement of biochemical data was centralised, affording consistent measurement of creatinine
270 and cystatin C values used in the GFR calculators across all participants⁴⁰.

271 We acknowledge some weaknesses in the data presented. Though the population studied was
272 broadly representative of a real-life cohort (i.e. many had comorbidities including diabetes mellitus,
273 hypertension and other comorbid disease), included participants were volunteers for the UK Biobank
274 resource and therefore may not be representative of the breadth of comorbidity seen in the general
275 population,⁴¹ particularly given the ~5% response rate. Reflecting the general population, there was a
276 relatively small proportion with advanced CKD (stages G4/5) and relatively few ESKD events. eGFR
277 measures may perform differently in later stage CKD, particularly with reference to prediction of
278 ESKD, and this subgroup warrants more specific study. Renal outcomes were identified from linked
279 health records, but are more reliably obtained from designated renal resources such as the UK Renal
280 Registry;⁴² renal outcomes were not linked to national registries and some ESKD outcomes may have
281 been missed. Similarly, cardiovascular endpoints obtained from linked health records will have
282 captured most events, but these events have not been validated with the same rigour as in dedicated
283 cardiovascular studies, and some cardiovascular endpoints may have been missed or incorrectly
284 coded. The ethnic groups were representative of a UK population with a bias towards white ethnic
285 groups. We did not find eGFR_{cys} to be predictive of CVD outcome in non-white ethnic groups,
286 though with limited non-white participants it is difficult to draw any firm conclusions in other ethnic
287 groups. We have not adjusted for all cardiometabolic conditions associated with higher levels of
288 cystatin C (including cancer, glucocorticoid therapy or thyroid disease), nor were participants with
289 these conditions excluded. Inclusion of participants with these conditions may have overestimated
290 the impact of eGFR_{cys} on outcome. Finally, GFR has been estimated from serum biomarkers and not
291 measured using radioisotope studies, iohexol clearance or formal inulin clearance studies.
292 Nevertheless, estimation of GFR is standard practice in the diagnosis of CKD. That noted, the size

293 and scope of UK Biobank means that our data particularly on CVD risk with eGFR data add strongly
294 to this literature on this important question.

295 In summary, we have shown that cystatin C-based calculations of GFR provide for more accurate
296 prediction of all-cause mortality and fatal/non-fatal CVD. eGFR_{cys} and albuminuria are independent
297 risk factors for fatal/non-fatal CVD and should be considered in cardiovascular risk prediction to
298 advise primary preventative treatment decisions. Consideration should be given to measuring serum
299 cystatin C and using eGFR_{cys} for diagnosis of CKD, prediction of CVD and thus making important
300 clinical decisions around implementation of CVD risk lowering therapies in addition to conventional
301 CVD risk factor calculators.

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307 **Author contributions**

308 J.S.L. and P.B.M. conceived of and designed the study. Data were analysed by C.E.W. and P.W.
309 under UK Biobank project 9310 led by N.S. and involving all authors. The first draft of the manuscript
310 was written by J.S.L. and C.E.W. All authors (J.S.L., C.E.W., P.W., P.B.M., N.S., C.A.C-M., D.M.,
311 S.R.G., J.G.C., J.M.G., P.S.J., J.L., D.M.L., J.P.) read, critically revised and approved the final
312 manuscript.

313 **Competing interests**

314 The authors declare no competing interests.

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Figure legends

Figure 1

Fully adjusted splines of estimated glomerular filtration rate (eGFR) against adjusted hazard ratio (with 95% confidence limits) for all-cause mortality (top row), composite fatal/non-fatal cardiovascular disease (CVD) (second row) and fatal CVD (third row) using eGFR based on creatinine (eGFRcr; left column), eGFR based on cystatin C (eGFRcys; middle column) and eGFR based on creatinine and cystatin C (eGFRcr-cys; right column).

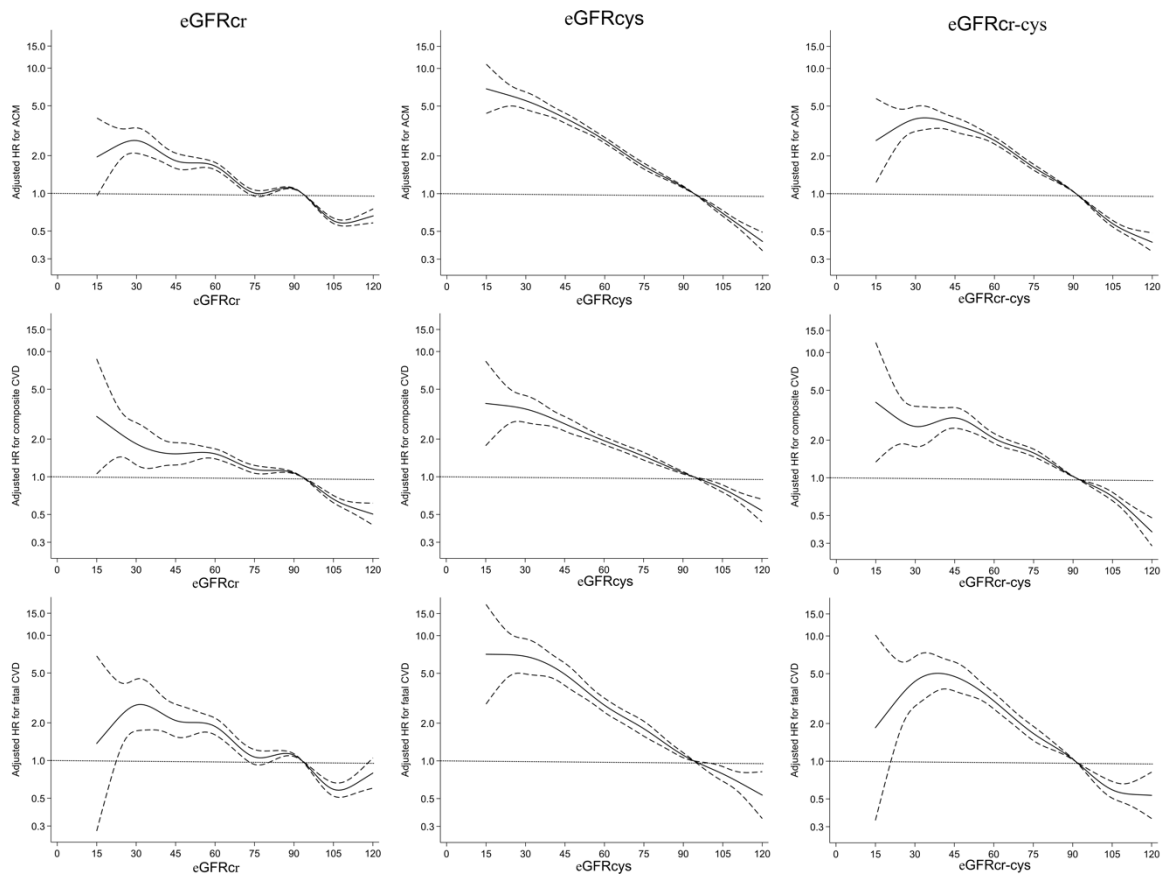


Figure 2

Heat maps for prediction of all-cause mortality, composite fatal/non-fatal cardiovascular disease (CVD), fatal CVD and end-stage kidney disease using eGFRcys and albuminuria (urine albumin:creatinine ratio; uACR) for estimated glomerular filtration rate (eGFR) based on creatinine (eGFRcr; top), eGFR based on cystatin C (eGFRcys; middle) and eGFR based on creatinine and cystatin C (eGFRcr-cys; bottom). No data were available for those with eGFR >90ml/min/1.73m² and uACR >3mg/mmol. Otherwise, hazard ratios adjusted for atherosclerotic risk factors (age, sex, ethnicity, systolic and diastolic blood pressure, antihypertensive medications, statin use, smoking, diabetes, total and high-density lipoprotein cholesterol) were ranked 1-13 (1 being the lowest risk), and heat maps were colour-coded for all outcomes: 1-4 (green), 5-7 (yellow), 8-10 (orange), 11-13 (red).



Figure 3

Change in C-statistic with 95% confidence intervals for composite fatal/non-fatal cardiovascular disease, fatal cardiovascular disease or all-cause mortality upon addition of each estimated glomerular filtration rate (eGFR) method: eGFR based on creatinine (eGFRcr), eGFR based on cystatin C (eGFRcys) and eGFR based on creatinine and cystatin C (eGFRcr-cys). The centre line (0.00) represents no change to C-index.

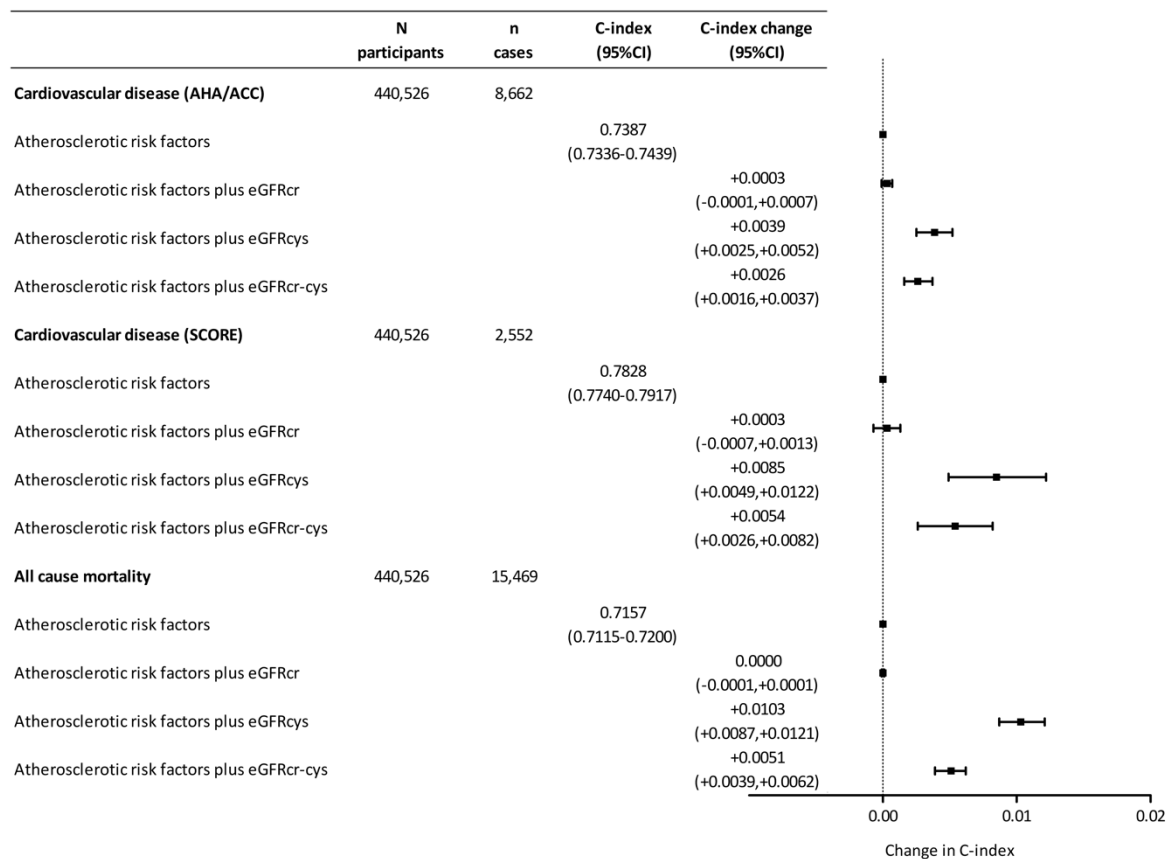


Table 1 – Distribution of atherosclerotic risk factors by category of eGFRcr

| Baseline variables | | eGFRcr range mL/min/1.73m ² | | | | | P value |
|--|-------------|--|-----------------|----------------|----------------|----------------|---------|
| | | >90 | 60-89 | 45-59 | 30-44 | 15-30 | |
| Number of participants | | 267122 (60.6%) | 165044 (37.5%) | 7148 (1.6%) | 991 (0.2%) | 221 (0.1%) | |
| Age (years) | | 54.15 (7.95) | 59.21 (7.25) | 62.59 (5.86) | 62.82 (6.08) | 60.76 (6.99) | <0.001 |
| Sex | Female | 150134 (56.2%) | 90057 (54.6%) | 4222 (59.1%) | 519 (52.4%) | 95 (43.0%) | <0.001 |
| | Male | 116988 (43.8%) | 74987 (45.4%) | 2926 (40.9%) | 472 (47.6%) | 126 (57.0%) | |
| Ethnicity | White | 250464 (93.8%) | 158834 (96.2%) | 6855 (95.9%) | 943 (95.2%) | 200 (90.5%) | <0.001 |
| | Black | 5009 (1.9%) | 1819 (1.1%) | 70 (1.0%) | 13 (1.3%) | 9 (4.1%) | |
| | South Asian | 4850 (1.8%) | 1752 (1.1%) | 85 (1.2%) | 16 (1.6%) | 8 (3.6%) | |
| | Other | 6799 (2.5%) | 2639 (1.6%) | 138 (1.9%) | 19 (1.9%) | 4 (1.8%) | |
| Smoking status | Non-smoker | 233701 (87.9%) | 151565 (92.3%) | 6556 (92.4%) | 891 (90.8%) | 203 (92.7%) | <0.001 |
| | Smoker | 32133 (12.1%) | 12679 (7.7%) | 542 (7.6%) | 90 (9.2%) | 16 (7.3%) | |
| Systolic blood pressure (mmHg) | | 138.38 (19.47) | 141.73 (19.76) | 143.06 (19.99) | 143.64 (20.35) | 142.64 (20.53) | <0.001 |
| Diastolic blood pressure (mmHg) | | 82.26 (10.73) | 82.71 (10.56) | 81.84 (10.82) | 80.27 (11.20) | 78.95 (11.11) | <0.001 |
| Antihypertensive medication | No | 227776 (85.3%) | 129458 (78.4%) | 3972 (55.6%) | 302 (30.5%) | 31 (14.0%) | <0.001 |
| | Yes | 39346 (14.7%) | 35586 (21.6%) | 3176 (44.4%) | 689 (69.5%) | 190 (86.0%) | |
| Statins | No | 240373 (90.0%) | 139578 (84.6%) | 5041 (70.5%) | 547 (55.2%) | 99 (44.8%) | <0.001 |
| | Yes | 26749 (10.0%) | 25466 (15.4%) | 2107 (29.5%) | 444 (44.8%) | 122 (55.2%) | |
| Self-reported diabetes mellitus at baseline | No | 255049 (95.5%) | 158240 (95.9%) | 6407 (89.6%) | 784 (79.1%) | 158 (71.5%) | <0.001 |
| | Yes | 12073 (4.5%) | 6804 (4.1%) | 741 (10.4%) | 207 (20.9%) | 63 (28.5%) | |
| Total cholesterol (mmol/l) | | 5.75 (1.10) | 5.79 (1.13) | 5.58 (1.23) | 5.29 (1.31) | 4.93 (1.27) | <0.001 |
| HDL cholesterol (mmol/l) | | 1.47 (0.38) | 1.46 (0.38) | 1.39 (0.38) | 1.30 (0.38) | 1.28 (0.46) | <0.001 |
| uACR (mg/mmol) | <3 | 247449 (92.6%) | 152917 (92.7%) | 6099 (85.3%) | 620 (62.6%) | 68 (30.8%) | <0.001 |
| | 3-30 | 11557 (4.3%) | 6906 (4.2%) | 648 (9.1%) | 228 (23.0%) | 65 (29.4%) | |
| | >30 | 8116 (3.0%) | 5221 (3.2%) | 401 (5.6%) | 143 (14.4%) | 88 (39.8%) | |
| HBA1c (mmol/mol) | | 35.67 (6.77) | 36.08 (5.61) | 38.11 (7.69) | 40.56 (10.83) | 42.25 (12.74) | <0.001 |
| Outcomes | | | | | | | |
| All-cause mortality | No | 259039 (97.0%) | 158433 (96.0%) | 6570 (91.9%) | 844 (85.2%) | 171 (77.4%) | <0.001 |
| | Yes | 8083 (3.0%) | 6611 (4.0%) | 578 (8.1%) | 147 (14.8%) | 50 (22.6%) | |
| Composite CVD* | No | 262682 (98.3%) | 161188 (97.7%) | 6862 (96.0%) | 928 (93.6%) | 204 (92.3%) | <0.001 |
| | Yes | 4440 (1.7%) | 3856 (2.3%) | 286 (4.0%) | 63 (6.4%) | 17 (7.7%) | |
| Fatal CVD** | No | 265868 (99.5%) | 163918 (99.3%) | 7022 (98.2%) | 956 (96.5%) | 210 (95.0%) | <0.001 |
| | Yes | 1254 (0.5%) | 1126 (0.7%) | 126 (1.8%) | 35 (3.5%) | 11 (5.0%) | |
| End-stage kidney disease | | 267064 (100.0%) | 164976 (100.0%) | 7110 (99.5%) | 914 (92.2%) | 126 (57.0%) | <0.001 |

| | | | | | |
|-----|----------|----------|-----------|-----------|------------|
| Yes | 58 (<1%) | 68 (<1%) | 38 (0.5%) | 77 (7.8%) | 95 (43.0%) |
|-----|----------|----------|-----------|-----------|------------|

*Composite fatal/nonfatal cardiovascular disease (CVD) outcome as per American Heart Association/American College of Cardiology (AHA/ACC) risk score.

**Fatal CVD outcome as per European Systematic Coronary Risk Evaluation (SCORE). CVD cardiovascular disease; HDL high-density lipoprotein; uACR urine albumin:creatinine ratio.

Table 2 - Adjusted* hazard ratios of each eGFR measure (per 10mL/min/1.73m²) for each of the four outcomes, among 440,526 UK Biobank participants.

| | N cases | HR (per 10ml/ min/1.73m ²) | 95% CI | 1 SD | HR (per 1 SD) | 95% CI |
|---------------------------------|---------|--|-----------|-------|------------------|-----------|
| All-cause mortality | 15649 | | | | | |
| eGFRcr | | 0.99 | 0.97-1.00 | 13.17 | 0.98 | 0.96-1.00 |
| eGFRcys | | 0.81 | 0.80-0.82 | 15.82 | 0.72 | 0.71-0.73 |
| eGFRcr-cys | | 0.84 | 0.82-0.85 | 12.66 | 0.80 | 0.78-0.81 |
| Composite CVD* | 8662 | | | | | |
| eGFRcr | | 0.95 | 0.93-0.97 | 13.17 | 0.93 | 0.91-0.96 |
| eGFRcys | | 0.87 | 0.86-0.88 | 15.82 | 0.80 | 0.78-0.82 |
| eGFRcr-cys | | 0.87 | 0.85-0.89 | 12.66 | 0.84 | 0.82-0.86 |
| Fatal CVD** | 2552 | | | | | |
| eGFRcr | | 0.92 | 0.89-0.95 | 13.17 | 0.90 | 0.86-0.94 |
| eGFRcys | | 0.77 | 0.75-0.79 | 15.82 | 0.66 | 0.63-0.69 |
| eGFRcr-cys | | 0.79 | 0.76-0.81 | 12.66 | 0.74 | 0.71-0.77 |
| End-stage kidney disease | 336 | | | | | |
| eGFRcr | | 0.33 | 0.31-0.35 | 13.17 | 0.23 | 0.21-0.25 |
| eGFRcys | | 0.29 | 0.28-0.31 | 15.82 | 0.14 | 0.13-0.16 |
| eGFRcr-cys | | 0.28 | 0.26-0.30 | 12.66 | 0.20 | 0.18-0.21 |

*Adjusted for age, sex, ethnicity, systolic blood pressure, diastolic blood pressure, antihypertensive medications, smoking, diabetes, statin use, total and high-density lipoprotein (HDL) cholesterol.

*Composite fatal/nonfatal cardiovascular disease (CVD) outcome as per American Heart Association/American College of Cardiology (AHA/ACC) risk score. **Fatal cardiovascular disease (CVD) outcome as per European Systematic Coronary Risk Evaluation (SCORE)

Table 3 – Net Reclassification Index for composite fatal/non-fatal CVD outcome for 3 eGFR measures: eGFRcr, eGFRcys and eGFRcr-cys

| Comparator | Addition | Case NRI (95%CI) | Non-case NRI (95% CI) |
|---|-----------------|-----------------------------|----------------------------------|
| Atherosclerotic risk factors | +eGFRcr | -0.08% (-0.41, +0.22%) | -0.01% (-0.03, +0.01%) |
| Atherosclerotic risk factors | +eGFRcys | +1.54% (+1.01, +2.10%) | -0.17% (-0.21, -0.14%) |
| Atherosclerotic risk factors | +eGFRcr-cys | +1.02% (+0.53, +1.48%) | -0.11% (-0.14, -0.08%) |
| Atherosclerotic risk factors +eGFRcr | +uACR | 0.08% (-0.13, +0.19%) | +0.01% (0.00, +0.02%) |
| Atherosclerotic risk factors +eGFRcys | +uACR | -0.03% (-0.15, +0.09%) | +0.02% (0.01, +0.03%) |
| Atherosclerotic risk factors +eGFRcr-cys | +uACR | 0.07% (-0.05, +0.18%) | +0.01% (0.00, +0.02%) |

Estimated glomerular filtration rate (eGFR) method: eGFR based on creatinine (eGFRcr), eGFR based on cystatin C (eGFRcys) and eGFR based on creatinine and cystatin C (eGFRcr-cys). uACR urine albumin:creatinine ratio

Table 4 – Change in Harrell’s C-statistic for prediction of all-cause mortality, composite and fatal CVD outcomes and end-stage kidney disease in those with $\geq 20\%$ absolute discordance between eGFRcr and eGFRcys

| | N cases | C-statistic | Change in C-statistic | P value |
|----------------------------|----------------|------------------------|------------------------------|----------------|
| All-cause mortality | 6557/183867 | 0.7207 (0.7140-0.7275) | na | |
| + eGFRcr | | 0.7270 (0.7203-0.7337) | 0.0063 (0.0042-0.0084) | <0.001 |
| + eGFRcys | | 0.7343 (0.7277-0.7410) | 0.0136 (0.0108-0.0165) | <0.001 |
| + eGFRcr-cys | | 0.7246 (0.7178-0.7312) | 0.0038 (0.0023-0.0054) | <0.001 |
| Composite CVD | 3565/183867 | 0.7422 (0.7338-0.7505) | na | |
| + eGFRcr | | 0.7425 (0.7341-0.7508) | 0.0003 (-0.0003-0.0009) | 0.347 |
| + eGFRcys | | 0.7469 (0.7386-0.7552) | 0.0047 (0.0027-0.0067) | <0.001 |
| + eGFRcr-cys | | 0.7448 (0.7365-0.7532) | 0.0027 (0.0012-0.0042) | <0.001 |
| Fatal CVD | 1059/183867 | 0.7947 (0.7806-0.8088) | na | |
| + eGFRcr | | 0.7965 (0.7823-0.8107) | 0.0018 (-0.0013-0.0049) | 0.264 |
| + eGFRcys | | 0.8036 (0.7897-0.8175) | 0.0089 (0.0036-0.0141) | 0.001 |
| + eGFRcr-cys | | 0.7979 (0.7839-0.8119) | 0.0032 (-0.0001-0.0065) | 0.057 |
| ESKD | 52/183867 | 0.7200 (0.6332-0.8068) | na | |
| + eGFRcr | | 0.7215 (0.6363-0.8067) | 0.0015 (-0.0027-0.0058) | 0.483 |
| + eGFRcys | | 0.7301 (0.6378-0.8223) | 0.0101 (-0.0473-0.0675) | 0.730 |
| + eGFRcr-cys | | 0.7446 (0.6649-0.8244) | 0.0246 (-0.0337-0.0830) | 0.408 |

C-statistics and change in C-statistic for Cox proportional hazard models adjusted for atherosclerotic risk factors (age, sex, ethnicity, smoking, systolic and diastolic blood pressure, antihypertensive medications, statins, total and high-density lipoprotein (HDL) cholesterol) and log urine albumin:creatinine ratio (uACR) with addition of the three estimated glomerular filtration rate (eGFR) methods: eGFR based on creatinine (eGFRcr), eGFR based on cystatin C (eGFRcys) and eGFR based on creatinine and cystatin C (eGFRcr-cys).

Methods

UK Biobank collected data from 502,536 consenting participants (age 37 to 73) from 2007-2010 across 22 assessment centres in the UK. Biological data and information from touch-screen questionnaires were collected at baseline as previously described.^{43,44} Ethical approval for UK Biobank was issued by the North West Multi-centre Research Ethics Committee (REC reference: 11/NW/03820). The study was conducted in accordance with the principles of the Declaration of Helsinki, and all participants gave written informed consent before enrolment.

Ethnicity was initially coded as white, black, south Asian or other, but for the purposes of eGFR calculators, ethnicity was coded as black or other. Smoking history was self-reported and categorised as current/previous or never smoker. We excluded those with prevalent ESKD or who were receiving renal replacement therapy in any form at baseline, defined from self-reported ESKD according to a pre-specified algorithm.⁴⁵ We further excluded any participant with a calculated eGFR from any measure of $<15 \text{ ml/min/1.73m}^2$. Participants with previous history of CVD (self-reported angina, myocardial infarction, stroke or transient ischaemic attack) were excluded.

Full details of the biochemistry sampling, handling and quality control protocol for UK Biobank has been described and validated previously.^{40,46-48} In brief, blood and spot urine samples were collected and analysed at a central laboratory, including creatinine, cystatin C, lipids (high-density, low-density and total cholesterol) and urine albumin content (urine albumin:creatinine ratio; uACR). The UK Biobank operated a high-turnover, clinic setup and thus samples were collected at various times of day. Serum and urine creatinine were measured using an enzymatic (creatinase), IDMS-traceable, method on Beckman Coulter AU5400 instrument.⁴⁹ Serum cystatin C was measured by latex enhanced immunoturbidimetric method on a Siemens ADVIA 1800 instrument.⁴⁹ Urine microalbumin was measured by immunoturbidimetric method using reagents and calibrators sourced from Randox Bioscience (UK)⁵⁰. Over 3 levels of control the coefficient of variation for creatinine was $<2.8\%$ ⁴⁸, for cystatin C, over 2 levels of control, was $<1.4\%$ ⁴⁸ and for urinary microalbumin and creatinine, over 2 levels of control, was $\leq 2.1\%$ ⁵⁰. Each assay was registered with an external quality assurance (EQA) scheme, and assay performance was externally verified via the results returned from participation in

these schemes. Data were adjusted by UK Biobank centrally before release to adjust for pre-analytical variables⁵¹. Estimated glomerular filtration rate was calculated by CKD-EPI using serum creatinine (eGFR_{cr})¹⁹, cystatin C (eGFR_{cys}) or cystatin C-creatinine (eGFR_{cr-cys}) equations as previously reported.²⁰

There were four outcomes of interest. First, all-cause mortality was defined as death from any cause, with date and cause of death obtained from death certificates held by the National Health Service (NHS) Information Centre (participants in England and Wales) or the NHS Central Register Scotland (for participants from Scotland). Second, composite fatal or non-fatal cardiovascular disease (CVD) events (nonfatal episodes of MI, stroke, or heart failure ICD10 codes I22, I24, I60, I61, I63 or I64, or fatal CVD ICD10 codes I20-I25, I60-I64) were identified by linkage with routine hospital data, and date and cause of death (where appropriate) were obtained from death certificates as for all-cause mortality. Third, fatal CVD events were identified from fatal CVD ICD10 codes (I20-I25, I60-I64) and from death certificates as for all-cause mortality. Last, ESKD was defined as reaching CKD stage G5 or requirement for renal replacement therapy, using hospital admission ICD10 (E85.3, N16.5, N18.0, N18.5, Q60.1, T82.4, T86.1, Y60.2, Y61.2, Y62.2, Y84.1, Z49.0, Z49.1, Z49.2, Z94.0, Z99.2) and OPCS4 (L74.1, L74.2, L74.3, L74.4, L74.5, L74.6, L74.8, L74.9, M01.2, M01.3, M01.4, M01.5, M01.8, M01.9, M02.3, M08.4, M17.2, M17.4, M17.8, M17.9, X40.1, X40.2, X40.3, X40.4, X40.5, X40.6, X40.7, X40.8, X40.9, X41.1, X41.2, X41.8, X41.9, X42.1, X42.8, X42.9, X43.1) codes, or ICD10 codes (N18.0, N18.5) listed in any position in a death record, according to a pre-specified algorithm.⁴⁵

The follow-up period started at the date of first assessment. The follow-up period ended with the date of death, first date of hospitalisation for non-fatal CVD or ESKD, or end of follow-up (whichever occurred first). For mortality endpoints, end of follow-up was recorded as the first of date of death, or the end of data collection for the attended assessment centre (30/11/2016 for centres in Scotland; 31/1/2018 for centres in England/Wales).

Statistical analysis

Each eGFR was categorised into the following five groups (ml/min/1.73m²) aligned with KDIGO staging of chronic kidney disease: ≥ 90 , 60-89, 45-59, 30-44, 15-30.¹¹ The distributions of

atherosclerotic risk factors (age, sex, ethnicity, smoking, systolic and diastolic blood pressure, use of antihypertensive medication and statins, total and HDL cholesterol) and uACR were investigated across eGFR categories and across each outcome. Continuous risk factors were displayed as mean (SD) if normally distributed and median (Q1-Q3) if skewed. Categorical risk factors were displayed as count (%). Tests for trends across categories were performed using chi-squared tests, ANOVA or Wilcoxon rank-sum tests where appropriate.

The event rate per 100,000 person years for each outcome in each category of eGFR was calculated using the Kaplan-Meier method, and the event-free survival rate per category was plotted using this method.

Pearson correlation coefficients between the three linear eGFR measures were calculated. To examine the relationship between each eGFR-outcome combination, restricted cubic splines (with knots at eGFR of 20, 30, 45, 60, 75, 90, 105 and 115 ml/min/1.73²) were constructed and each fully adjusted relationship (adjusted for age, sex, ethnicity, smoking, systolic and diastolic blood pressure, antihypertensive medications, statins, total and HDL cholesterol) was plotted.

To assess the effect of the addition of each eGFR measure and albuminuria to the discriminative ability of atherosclerotic risk factors, Cox-proportional hazard models (adjusted as above) were constructed for all-cause mortality, fatal/non-fatal CVD or fatal CVD, and the change in model discrimination was assessed using Harrell's C-statistics for each eGFR, in isolation and with addition of albuminuria. To understand better the relationship between eGFR and each outcome in the context of different uACR levels, uACR was split into clinically meaningful groups (uACR <3, 3-30 or >30 mg/mmol according to the KDIGO clinical practice guideline for definition and classification of CKD¹¹) and the hazard ratio for each category of eGFR within each uACR group was assessed.

eGFR and albuminuria were modelled linearly, with log transformation for uACR. Similar analyses were conducted in subgroups including gender, ethnicity (white vs. non-white), body mass index (<30, 30-35, >35 kg/m²) and CKD 3b/4 (eGFRcr 15-45 ml/min/1.73m²). As a sensitivity analysis, C-statistics for change in model discrimination for each outcome in the group of participants with

eGFR_{cr} <60ml/min/1.73m², to check the influence of data linearity above 60ml/min/1.73m², Categorical Net Reclassification Index was tested for reclassification of patients from “low” to “intermediate” risk, i.e. across the threshold 7.5% 10-year risk of CVD that would warrant initiation of statin therapy for CVD risk reduction.¹⁸

As a sensitivity analysis we assessed cases in which there was $\geq 20\%$ absolute discordance between eGFR_{cr} and eGFR_{cys}. Adjusted Cox-proportional hazard models (as above) were repeated for the group in which there was $\geq 20\%$ discordance for all 4 outcomes.

We excluded participants with missing data and who reported baseline CVD or ESKD, or who had eGFR (any measure) <15 ml/min/1.72m²; all analyses were performed on complete cases. Analyses were performed using STATA 14 (StataCorp LP, College Station, USA) and *nricens* for R statistical software package (version 3.5.3) for NRI. A p-value of <0.05 was considered statistically significant.

Data availability

The UK Biobank data that support the findings of this study are available from the UK Biobank (www.ukbiobank.ac.uk). This study was conducted under project code 9310.

Methods-only references

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