

Validation of the corticomedullary difference in magnetic resonance imaging-derived apparent diffusion coefficient for kidney fibrosis detection: a cross-sectional study

BERCHTOLD, Lena, *et al.*

Abstract

Background: Kidney cortical interstitial fibrosis (IF) is highly predictive of renal prognosis and is currently assessed by the evaluation of a biopsy. Diffusion magnetic resonance imaging (MRI) is a promising tool to evaluate kidney fibrosis via the apparent diffusion coefficient (ADC), but suffers from inter-individual variability. We recently applied a novel MRI protocol to allow calculation of the corticomedullary ADC difference (Δ ADC). We here present the validation of Δ ADC for fibrosis assessment in a cohort of 164 patients undergoing biopsy and compare it with estimated glomerular filtration rate (eGFR) and other plasmatic parameters for the detection of fibrosis. Methods: This monocentric cross-sectional study included 164 patients undergoing renal biopsy at the Nephrology Department of the University Hospital of Geneva between October 2014 and May 2018. Patients underwent diffusion-weighted imaging, and T1 and T2 mappings, within 1 week after biopsy. MRI results were compared with gold standard histology for fibrosis assessment. Results: Absolute cortical ADC or cortical T1 values correlated poorly to IF [...]

Reference

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1 **Original NDT**

2 **Validation of the cortico-medullary difference in MRI-derived apparent diffusion**
3 **coefficient for kidney fibrosis detection: a cross-sectional study**

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19 Running title: Diffusion MRI for fibrosis evaluation

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25 **Abstract:**

26 **Background:** Kidney cortical interstitial fibrosis (IF) is highly predictive of renal prognosis,
27 and is currently assessed by the evaluation of a biopsy. Diffusion MRI is a promising tool to
28 evaluate kidney fibrosis via the apparent diffusion coefficient (ADC), but suffers from inter-
29 individual variability. We recently applied a novel MRI protocol to allow calculation of the
30 cortico-medullary ADC difference (Δ ADC). We here present the validation of Δ ADC for
31 fibrosis assessment in a cohort of 164 patients undergoing biopsy and compare it to eGFR and
32 other plasmatic parameters for the detection of fibrosis.

33 **Methods:** This monocentric cross-sectional study included 164 patients undergoing renal
34 biopsy at the Nephrology Department of the University Hospital of Geneva between October
35 2014 and May 2018. Patients underwent diffusion-weighted imaging, and T1- and T2-
36 mappings, within one week after biopsy. MRI results were compared to gold standard histology
37 for fibrosis assessment.

38 **Results:** Absolute cortical ADC or cortical T1 values correlated poorly to IF assessed by the
39 biopsy, whereas Δ ADC was highly correlated to IF ($r=-0.52$, $p<0.001$) and eGFR ($r=0.37$,
40 $p<0.01$), in both native and allograft patients. Δ T1 displayed a lower, but significant, correlation
41 to IF and eGFR, whereas T2 did not correlate to IF nor to eGFR. Δ ADC, Δ T1 and eGFR were
42 independently associated with kidney fibrosis, and their combination allowed detecting
43 extensive fibrosis with good specificity.

44 **Conclusion:** Δ ADC is better correlated to IF than absolute cortical or medullary ADC values.
45 Δ ADC, Δ T1 and eGFR are independently associated to IF and allow the identification of
46 patients with extensive IF.

47 **Keywords:** MRI, fibrosis, diffusion, cortex, chronic kidney disease

48 **Introduction:**

49 Chronic kidney disease (CKD) is defined as abnormal kidney structure and/or function lasting
50 for more than 3 months^{1,2}. Whereas kidney function may be evaluated using creatinine and
51 cystatin based equations, kidney structure is more difficult to appreciate non-invasively. The
52 histological hallmark of CKD is the presence of cortical interstitial fibrosis (IF). IF is better
53 correlated to renal function and to long term renal outcome than glomerulosclerosis or any other
54 histological lesions^{3,4}. Evaluation of IF is therefore used to tailor treatment and judge renal
55 prognosis⁵⁻⁷. This evaluation is currently performed by the visual inspection of a kidney biopsy
56 using specific stains such as Masson trichrome and/or Sirius Red⁸. Recent evidence has shown
57 that the extent of interstitial fibrosis is one of the main factor predicting renal function evolution,
58 even independently of eGFR⁹.

59 In several organs, noninvasive ways to evaluate fibrosis are available. The kidney possesses
60 specific features rendering it more difficult to image. It is a heterogeneous organ, and its global
61 evaluation may be difficult¹⁰. In addition, native kidneys are located quite deep, move with
62 respiration, and are close to air/tissue interfaces (intestines) limiting image quality and
63 subsequent analysis. Non-invasive evaluation of fibrosis would be useful to avoid kidney
64 biopsies in cases of extensive fibrosis, to follow the evolution of kidney disease non-invasively,
65 and to identify patients at risk of CKD with still preserved renal function. Imaging would be
66 complementary to eGFR estimation for the detection of early kidney lesions. Finally, imaging
67 the whole kidney may also point to the presence of scars that may be missed or, conversely,
68 overrepresented by a biopsy.

69 Diffusion Weighted Magnetic resonance imaging (DW-MRI) has been described as promising
70 for evaluation of renal fibrosis, since it may easily be performed on clinical scanners¹¹⁻¹³. In
71 both human disease and experimental kidney disease models, DW-MRI could identify diseased

72 versus healthy kidneys^{11,14-21}. In experimental models, the apparent Diffusion Coefficient
73 (ADC) derived from DW-MRI showed a good negative correlation to fibrosis^{22,23}. In human
74 kidneys, Inoue et al. showed that diffusion MRI was correlated to renal function and to IF in 37
75 diabetic patients having undergone biopsy¹¹. In another study, ADC correlated to cortical IF
76 and eGFR in 25 patients¹². Although promising, diffusion MRI of abdominal organs is still
77 difficult to use clinically because of the artifacts associated with image acquisition, as well as
78 the inter-individual variations of the absolute ADC values²⁴. Finally, although correlation to IF
79 is observed, the additional role of perfusion in these associations is debated²⁵.

80 Given the limitations described above, we recently adapted renal diffusion with the application
81 of a readout-segmented echo planar (EPI) sequence (RESOLVE)²⁶. In healthy volunteers, we
82 could demonstrate that this diffusion sequence led to better discrimination between the cortical
83 and medullary parts of the kidney²⁶. The use of the cortico-medullary ADC difference (Δ ADC)
84 reduced inter-individual variation, allowing for better comparison between subjects²⁶. In a pilot
85 study, Δ ADC was very well correlated to fibrosis assessed by standard histology in 29 kidney
86 allograft patients having undergone kidney biopsy²⁷.

87 We aimed here to perform an external validation of Δ ADC for IF detection in a larger and mixed
88 population of patients having undergone biopsy, using a different scanner to the pilot study. We
89 performed a multivariable analysis to improve IF detection. We investigated the identification
90 of patients with extensive fibrosis in this cohort.

91 **Methods**

93 **Patients**

94 We designed a cross-sectional study, including adult kidney allograft recipients and CKD
95 patients who were planned for a kidney biopsy for clinical purposes. MRI was scheduled on the

96 same day as the biopsy whenever possible, or within one week. Patients, 18 years of age or
97 older, who were followed at the University Hospital of Geneva, were eligible for enrollment.
98 Exclusion criteria were the presence of a pacemaker or other MR incompatible device,
99 pregnancy, claustrophobia, and patient refusal. In all patients, additional fasting serum and urine
100 were collected and stored at -80 °C. The study was approved by the local ethical committee for
101 human studies of Geneva, Switzerland (CER 11-160, Commission Cantonale d’Ethique de la
102 Recherche) and performed according to the Declaration of Helsinki principles. All the patients
103 were contacted to provide written informed consent to participate in this prospective study.
104 None of the patients were from a vulnerable population and all patients or next of kin provided
105 written informed consent which was freely given.

106 **Laboratory measurement**

107 Baseline characteristics, including medical history, co-morbidities and treatment, were
108 collected through patient records. Patients’ blood pressure, weight and size were measured
109 routinely during follow-up visits. Serum creatinine and other standard laboratory values were
110 measured during routine follow-up visits or hospitalizations. Standard biochemical analyses
111 were performed in a Geneva University Hospital Laboratory using routine automated analyzers.
112 The eGFR was calculated using the Chronic Kidney Disease Epidemiology Collaboration
113 equation (CKD-EPI). Creatinine was measured by Jaffé-kinetics using IDMS-traceable
114 methods.

115 **Histological fibrosis quantification**

116 Renal fibrosis was assessed quantitatively on the kidney biopsy specimen by the Pathology
117 Department of the University Hospital of Geneva, using Masson trichrome stained kidney
118 sections. The expert pathologist (S.M.) was blinded to the other results, including eGFR and
119 MRI. Expert evaluation of fibrosis is recommended to evaluate IF and is reproducible. It is the

120 current gold standard in most pathology services^{9,28}. The severity of renal fibrosis was scored
121 from 0 to 100% for each patient and reported on the clinical biopsy report independently of our
122 study. To verify the reproducibility of this evaluation, 60 random sections were evaluated
123 blindly by two experienced nephrologists. This repeated fibrosis evaluation displayed a good
124 correlation to pathological evaluation (ICC 0.92; 95%CI 0.87 to 0.95). Furthermore, renal
125 fibrosis was quantified using the BANFF criteria in renal allograft patients: ci (interstitial
126 fibrosis) and ct (tubular atrophy) with a minimal score of 0 and maximal score of 6. Due to a
127 good correlation between the two methods ($r = 0.86$; $p < 0.001$), we used subjective histological
128 renal fibrosis as a continuous variable (0 to 100%) for all analyses. In our predictive models,
129 we also use the fibrosis in categories (<10; 10-25; 25-50; >50 %) in both native and allograft
130 patients, as recently proposed for renal prognosis⁹.

131 **MR imaging**

132 Patients were scanned on a PRISMA 3T MR (Siemens AG, Erlangen Germany) with the
133 standard 32-element spine coil and the 18-element phased-array abdominal coil. MRI protocol
134 parameters are summarized in Table 1. ROI were determined as previously described^{26,27} for
135 diffusion-weighted ADC, T1 and T2 mapping, and the cortico-medullary differences were
136 calculated. ADC was measured directly on the ADC map produced by the Siemens MR system,
137 which uses a monoexponential fitting model. The analysis of the MRI images was also blinded
138 to all other markers. The MRI was performed in 55% of the cases before the biopsy. In the
139 remaining patients the biopsy was performed one week before MRI. All focal pathological areas
140 (cyst, scar, hematomas ...) were avoided in the ROI placement aiming to cover a large and
141 representative part of the cortex and medulla.

142 **Statistical analysis**

143 Continuous variables are expressed as mean \pm standard deviation or median and interquartile
144 range according to the distribution. Categorical variables are expressed as numbers and
145 percentages. The statistical significance was determined as a p value of less than 0.05 and all
146 tests were two-sided. For simple correlation analyses, we performed Pearson's tests, after
147 controlling the linearity of associations with scatterplots. We conducted univariable and
148 multivariable linear regression analyses to assess the associations with IF²⁹. Univariable and
149 multivariable logistic regression models were used to investigate the capacity of parameter to
150 predict different levels of fibrosis and vascular lesions. The discriminative performance of
151 markers and logistic regression models to predict different levels of fibrosis and vascular lesions
152 were assessed by using receiver operating characteristic (ROC) curves. We reported AUC
153 values with 95%CI. Statistical analyses were performed using STATA 13.1 (StataCorp, College
154 Station, TX, USA).

155

156 **Results:**

157 **Characteristics of the study population**

158 From October 2014 to May 2018, we included 164 CKD patients, mainly Caucasian (91%) and
159 male (67%), undergoing kidney biopsy for clinical reasons. Of the 164 patients, 118 (72%)
160 were kidney allograft patients and 46 (28%) were native kidney patients (Figure 1). Baseline
161 characteristics are presented in Table 2. Biopsy indications were made by the nephrologist in
162 charge of the patients, as clinically justified, and independently of the present study. For native
163 kidney disease, most of the indications were an abnormal urinary microscopy and proteinuria
164 and/or acute or chronic renal dysfunction. For allograft patients, biopsy indications were routine
165 biopsies (at one year, after steroid withdrawal), elevation of creatinine levels, and apparition of
166 proteinuria or de novo donor specific antibodies.

167 **Univariable analysis of predictors of fibrosis**

168 **MRI indexes for IF evaluation: Δ ADC, Δ T1 and Δ T2**

169 Images for 97% of the patients were of sufficient quality to allow measurement of the difference
170 between ADC of the cortex and medulla (Δ ADC values [$\times 10^{-6}$ mm²/s]) (Figure 2). In order to
171 validate Δ ADC for IF evaluation in this population, we correlated Δ ADC with IF assessed by
172 the gold standard clinical IF evaluation method. We confirmed a statistically significant and
173 high correlation between these parameters ($r = -0.52$, $p < 0.001$) (Figure 3A). Absolute cortical
174 ADC values correlated moderately to IF ($r = -0.22$, $p = 0.01$), whereas medullary ADC did not
175 correlate with IF (Supplementary 1A-B). The correlation of Δ ADC to IF was stronger in native
176 kidney patients ($r = -0.64$, $p < 0.001$) than in kidney allograft patients ($r = -0.42$, $p < 0.001$)
177 (Supplementary Figure 2). Δ ADC correlated to eGFR ($r = 0.37$, $p < 0.001$) (Figure 1C).

178

179 In patients with relatively preserved normal renal function ($eGFR \geq 60 \text{ ml/min}$), ΔADC still
180 correlated to IF ($r = -0.27$, $p = 0.03$), whereas the correlation was even stronger in patients with
181 $eGFR < 60 \text{ ml/min}/1.73 \text{ m}^2$ ($r = -0.53$, $p < 0.01$). Cortical and medullary ADC values did not
182 correlate to IF in patients with $eGFR \geq 60 \text{ ml/min}/1.73 \text{ m}^2$, and the correlations were not
183 statistically significant, with a limit p-value, in patients with an eGFR lower than 60
184 $\text{ml/min}/1.73 \text{ m}^2$. Cortical and medullary ADC did not correlate significantly to eGFR ($r = 0.15$,
185 $p = 0.07$) and ($r = -0.04$, $p = 0.58$) respectively).

186 A moderate correlation was found between absolute T1 values and IF with $r = 0.26$, $p = 0.005$
187 for the cortex (Supplementary Figure 1C). Medullary T1 was inversely correlated to IF $r = -$
188 0.20 , $p = 0.03$ (Supplementary Figure 1D). We further calculated the cortico-medullary
189 difference for T1 values ($\Delta T1$). $\Delta T1$ displayed a better correlation to IF ($r = 0.49$, $p < 0.001$)
190 than absolute values (Figure 3B). The correlation between IF and $\Delta T1$ was stronger in native
191 kidney patients than in kidney allograft patients (supplementary Figure 2). Cortical and
192 medullary T1 did not correlate with eGFR ($r = -0.13$, $p = 0.09$ and $r = 0.15$, $p = 0.06$ respectively)
193 whereas $\Delta T1$ did ($r = -0.30$, $p < 0.001$) (Figure 3D).

194 Neither T2 nor $\Delta T2$ correlated with renal function nor with IF, in both native kidney and kidney
195 allograft patients (Supplementary Figure 3).

196 **Biological parameters**

197 In order to test whether combining plasmatic and MRI variables could improve the detection of
198 fibrosis, we tested the association between fibrosis and different biological parameters in
199 univariable analysis (Supplementary Figure 4). Parameters eGFR, PTH, 25-OH vitamin D,
200 proteinuria, phosphate and hemoglobin displayed good correlation to IF as shown in Table 3.

201 **Multivariable model.**

202 In the complete multivariable analysis presented in Table 3, only $\Delta T1$, ΔADC and eGFR were
203 independently associated with fibrosis.

204 The coefficient R^2 of the complete multivariable model was 0.54 ($R=0.74$) (Table 3), indicating
205 that the combination of parameters improved the detection of IF. No significant interaction was
206 observed between $\Delta T1$, ΔADC and eGFR. Using the multivariable model, the higher the
207 fibrosis category, the higher our predictive score (Figure 4).

208 When considering only the three independently associated factors (ΔADC , $\Delta T1$, eGFR), the R^2
209 was also 0.54.

210 **Identifications of patients by fibrosis categories**

211 With a logistic model aiming to identify patients with low fibrosis (10% or less), the obtained
212 combination of ΔADC , $\Delta T1$ and eGFR showed an AUC of 0.840 (Figure 5A). 89 patients had
213 a high level of risk to have a fibrosis predicted by the model greater than 10%, among which
214 85 had actual biopsy-measured fibrosis $>10\%$ (positive predictive value, PPV=95.5%).
215 However, thresholds clinically relevant to rule-out patients with low fibrosis (i.e. thresholds
216 with a high sensitivity) identified only a small subgroup of patients.

217 With a logistic model aiming to identify patients with a significant fibrosis (more than 25%),
218 the AUC was 0.840 (Figure 5B). 18 patients were identified by the model with a low level of
219 risk to have a fibrosis greater than 25%, among which 16 had actual biopsy-measured fibrosis
220 $\leq 25\%$ (negative predictive value, NPV=88.9%). 41 patients were identified by the model with
221 a high level of risk to have a fibrosis greater than 25%, among which 37 had actual biopsy-
222 measured fibrosis $> 25\%$ (PPV=86.3%).

223 With a logistic model aiming to identify patients with a significant fibrosis (50% or more), the
224 AUC was 0.905 (Figure 5C). 127 patients were identified by the model with a low level of risk

225 to have a fibrosis of 50% or more, among which 127 had actual biopsy-measured fibrosis <
226 50% (NPV=96.2%). 9 patients were identified by the model with a high level of risk to have a
227 fibrosis of 50% or more, among which 8 had actual biopsy-measured fibrosis \geq 50%
228 (PPV=88.8%). The ROC curves using the same threshold, but with only Δ ADC as predictor,
229 are represented in Supplementary Figure 5.

230 **Discussion:**

231 In this study, we externally validated an improved diffusion MRI sequence allowing the
232 calculation of the cortico-medullary ADC difference for fibrosis detection in a mixed
233 population of 164 patients who had undergone kidney biopsy for clinical purposes. We showed
234 also Δ ADC's superiority to absolute cortical ADC values. We used a different scanner than in
235 our previous studies. We demonstrated that MRI parameters add to eGFR for IF detection.
236 Finally, we showed that MRI parameters combined to eGFR identify patients with extensive
237 fibrosis with a good specificity.

238 Although several studies have used diffusion MRI as a tool to evaluate fibrosis, differences
239 between sequences and ADC values precluded clear comparison³⁰. Our study represents, to the
240 best of our knowledge, the largest study validating diffusion MRI to predict fibrosis in patients
241 undergoing biopsies. The difference in cortical and medullary ADC correlated well to fibrosis
242 in our mixed population of native and allograft kidneys, with various types of primary diseases,
243 therefore validating our previous observation in a small homogeneous population. In addition,
244 the difference index was stable between different brands and types of scanner (Friedli, ISMRM,
245 2017, abstract#3298). Diffusion MRI also did not require the use of contrast medium, an
246 advantage in the CKD population. Interestingly, Δ ADC correlated to fibrosis even with patients
247 with preserved renal function, which may indicate that early detection of lesions is possible.
248 Absolute ADC values were less correlated to fibrosis than Δ ADC. Fibrosis usually affects the

249 cortex. Normalization to the medulla was technically easier and more efficient than to
250 surrounding tissues outside the kidney, since the close proximity of the medulla decreased
251 errors related to B1 and B0 heterogeneity as well as to the coil sensitivity profile ²⁷. Since
252 medullary ADC was not correlated to fibrosis, subtracting it from the cortical ADC improved
253 reproducibility and likely corrected for the baseline physiological inter-individual variability of
254 the ADC²⁷. The lower correlation between absolute ADC values and fibrosis compared to the
255 existing literature is probably related to the mixed population we included, and this therefore
256 calls for normalization of absolute cortical ADC values as an important tool in this research.
257 We used here monoexponential fit for ADC calculation with all the b-values and not
258 biexponential fit since we previously demonstrated that parameters derived from the
259 biexponential fit did not improve detection of IF³¹. As perfusion may also be reduced in case
260 of IF, we still believe that the whole range of b-values is useful for IF detection. As emphasized
261 by a recent review³², the monoexponential model is still preferred by the majority of studies on
262 renal diffusion as the superiority of biexponential model in renal diffusion remains to be better
263 demonstrated.

264 Fibrosis evaluation was more accurate in native kidney patients, which may be related to the
265 lower number of patients in this group. Alternatively, the vasoconstriction usually observed in
266 allograft patients, related to the use of calcineurin inhibitors, may modulate perfusion and affect
267 diffusion MRI independently of fibrosis, lowering the association to fibrosis.

268 T1 mapping measures the longitudinal (spin-lattice) relaxation time and has been used to
269 evaluate cardiac fibrosis³³. We showed here that T1, in particular $\Delta T1$, were also associated to
270 renal IF, although not as strongly as ΔADC . Interestingly, the combination of $\Delta T1$ and ΔADC
271 in multivariable analysis improved fibrosis detection by imaging variables alone, showing that
272 the two values measure slightly different phenomena. These two parameters may thus be
273 complementary to predict fibrosis in the kidney.

274 We further demonstrated that adding Δ ADC values to eGFR improves the correlation in a
275 multivariable model suggesting that Δ ADC and eGFR measure different parameters associated
276 to IF, and are thus complementary. Whether Δ ADC and ADC measure structural parameters or
277 modifications of water movement of filtrate is much debated and difficult to demonstrate, but
278 we showed here that diffusion correlated to IF, at least independently of glomerular filtration
279 rate. Modifications of ADC may still be influenced by perfusion and other parameters that were
280 not measured here. The important question of the origin of ADC change induced by IF remains
281 to be addressed by further studies. In this respect, diffusion tensor imaging (DTI) that can assess
282 the renal anisotropy may bring new insights^{15,17,34}. Nevertheless, our aim was to evaluate Δ ADC
283 as an independent marker of IF, whatever the primary cause of the modification in signal.

284 Categories of IF have recently been demonstrated to predict renal function evolution⁹. We
285 studied the value of MRI parameters in combination to eGFR to identify patients in four fibrosis
286 categories. Our model was able to identify patients with more than 10% fibrosis with a great
287 sensitivity, corresponding to early detection of structural lesions in relatively healthy kidneys.
288 Our model could identify patients with extensive (>50%) IF with a good specificity. Although
289 not perfect, addition of MRI to clinical evaluation may thus avoid biopsies or unnecessary
290 treatment in selected cases, or could help tailor follow-up.

291 One limitation of our study is its monocentric design, despite the large number of patients
292 included. Another source of error could be related to manual, therefore subjective, placement
293 of ROIs. This procedure is still standard in the field of diffusion MRI and we have shown, in a
294 previous study that our methodology had a good inter and intra-observer reproducibility²⁷. We
295 used the evaluation of a biopsy by a pathologist blinded for eGFR as gold standard for IF
296 evaluation. To secure our evaluation, we performed a blinded second reading of the IF in 60
297 sections chosen randomly by two nephrologists. The agreement between the second reading
298 and the pathologist reading was good (ICC: 0.92; 95%CI 0.87 to 0.95). Although automatic

299 kidney biopsy evaluation has been suggested to be useful in fibrosis estimation, it is still rarely
300 performed routinely and correlated less well to eGFR than pathological evaluation⁸ in this study
301 population. This is likely because of the non-exclusion of glomeruli and vessels in these
302 automatized quantifications (data not shown). We however observed a relatively good
303 correlation between the pathological and automated evaluation of IF ($r=0.4$, $p<0.01$). Finally,
304 subjective assessment of tubulo-interstitial fibrosis has been shown to have very high inter-
305 reader agreement and is the current gold standard for IF assessment in pathology services^{9,28}.
306 Given these limitations, novel, more objective tools to quantify fibrosis are being developed,
307 but are not routinely available^{35,36}. Sampling error may also occur in random biopsies. This last
308 limitation is however inherent to kidney biopsies. Finally, given the design of our study and the
309 need to have MRI performed on a research timetable, we could not include many emergency
310 biopsies and our population principally represents semi-elective biopsies (planned within one
311 week) in native kidney and kidney allograft patients.

312 Overall, we externally validated the Δ ADC as an excellent index to evaluate cortical fibrosis
313 non-invasively, with much better accuracy than absolute cortical or medullary ADC values. We
314 show that Δ ADC is strongly associated to IF in both native and allograft patients. We further
315 show that Δ ADC may be used in combination with Δ T1 and eGFR to evaluate fibrosis, and that
316 MRI parameters significantly improve the detection to IF. Finally, we show that our model is
317 able to identify patients with extensive fibrosis with good specificity. Further studies on the
318 prognostic value and the longitudinal follow-up of patients would be of interest.

319 **Disclosures:**

320 The authors have nothing to disclose

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329 **Authors's contributions:**

330 SdS, LB, JPV, IF: study design, data acquisition, statistical analysis, manuscript writing, LC,
331 TdP: data acquisition, manuscript writing, PYM: study design, manuscript revision, CM: data
332 acquisition, SM: data acquisition, manuscript revision, KH: manuscript revision, CC: statistical
333 analysis, manuscript revision

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415 interstitial fibrosis in renal transplantation. *Transplantation* **92**, 890-899 (2011).

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418

419 **Table 1:** MRI parameters used in this study

420 DWI: diffusion weighted imaging; T1: Longitudinal (spin-lattice) relaxation time; T2:

421 Transverse (spin-spin) relaxation time, RESOLVE: Readout Segmentation of Long Variable

422 Echo Trains; MOLLI: Modified Look-Locker Inversion-recovery; ADC: apparent diffusion

423 coefficient

424

	RESOLVE DWI (for ADC)	MOLLI T1 mapping	T2
Resolution [mm³]	2×2×5	2×2×5	2×2×5
Echo time/ repetition time [ms]	68/2000	1.2/1500	1.21/392
Acceleration factor (GRAPPA)	3	2	2
Bandwidth [Hz/pixel]	1040	1085	1202
Readout segments	5	-	-
Echo Spacing [ms]	0.69	2.7	2.9
Inversion Scheme	-	3(3)3(3)5	-
Starting TI [ms]	-	117	-
TI Increment [ms]	-	80	-
Flip Angle [°]	180	35	12
b-values [s/mm²]	0, 50, 100, 150, 200, 250, 300, 500, 700, 900	-	-
Diffusion gradient scheme	Bipolar	-	-
Respiratory gating	Belt	Belt	Breath hold

425

426

427 **Table 2: Baseline characteristics of the study population (n = 164): clinical parameters,**
 428 **medication, laboratory measurements, biopsy diagnosis and chronic histological lesions.**

Characteristics	Total (n=164)	Native (n=46)	Allograft(n=118)
Clinical parameters			
Age, years	54 ± 14	51 ± 16	55 ± 13
Male, n (%)	110 (67.1)	33 (71.7)	77 (65.3)
Body mass index, kg/m ² (n=124)	25.7 ± 4.0	25.9 ± 4.1	25.6 ± 4.0
Caucasian, n (%)	149 (90.9)	39 (84.8)	110 (93.2)
Histological lesions			
Fibrosis in %	27.2 ± 17.7	33.3±24.1	24.9 ± 14.0
BANFF score			
IF/TA (ci+ct), min 0 – max 6 (n=116)	-	-	2 (2.0-4.0)
Medication, n (%)			
ACEi/ARB	71 (44.6)	28 (60.9)	43 (36.4)
Calcium channel blockers	66 (40.2)	14 (30.4)	52 (44.1)
Diuretics	22 (13.4)	14 (30.4)	8 (6.8)
Beta-blockers	66 (40.2)	12 (26.1)	54 (45.8)
Statins	76 (46.3)	21 (45.7)	55 (46.6)
Calcium supplementation	77 (47.0)	8 (17.4)	69 (68.5)
1.25OH-vitamin D supplementation	12 (7.3)	0 (0)	12 (10.2)
25OH-vitamin D supplementation	109 (66.5)	16 (34.8)	93 (78.8)
Anticalcineurin	-	-	111 (94.1)
Mycophenolate mofetil	-	-	94 (79.7)
Corticosteroids	-	-	85 (72.0)
Others (Azathioprine, m-Tor inhibitor, ...)	-	-	11 (9.3)
Laboratory measurements			
Creatinine, micromol/l	119 [96 – 152]	120 [81-187]	119 [101-147]
eGFR ml/min per 1.73m ² *	57.2 ± 24.2	59.2 ± 33.7	56.4± 19.4
Hemoglobin, g/l	128.6 ± 18.3	124.7 ± 23.1	130.1 ± 15.9
Calcium, mmol/l (n=142)	2.4 ± 0.1	2.3 ± 0.1	2.4 ± 0.1
Phosphate, mmol/l (n=153)	1.01 ± 0.25	1.12 ± 0.34	0.98 ± 0.21
Magnesium, mmol/l (n=118)	0.68 ± 0.11	0.80 ± 0.16	0.67 ± 0.09
25-hydroxyvitamin D, nmol/l (n=135)	71.1 ± 25.7	47.3 ± 25.7	77.0 ± 22.0
Parathyroid hormone, pmol/l (n=131)	8.8 [5.6-13.0]	6.4 [4.1-8]	10.0 [6.0 ± 13.]
Albumin, g/l (n=152)	40.3 ± 4.4	38.3 ± 5.6	41.1 ± 3.7
Proteinuria/créatinine, g/g (n=144)	0.15 [0.06-0.55]	1.00 [0.21-2.58]	0.08 [0.05-0.21]
Biopsy diagnosis**, n (%)			
Rejection	13 (7.9)	-	13 (11.0)
Positive C4D			11(9.3)
Tubular lesions	29 (17.7)	6 (13.0)	23 (19.5)
- Interstitial nephritis	6 (4.6)	5 (14.3)	1 (1.1)
Glomerulonephritis incl FSGS	42 (25.6)	23 (50.0)	19 (16.1)
Diabetic nephropathy	10 (6.1)	10 (21.7)	0 (0)
Vascular nephropathy	21 (12.8)	17 (37.0)	4 (3.4)
Anticalcineurin toxicity	40 (24.4)	-	32 (27.1)
Chronic allograft nephropathy	3 (1.8)	-	3 (2.5)
Others (oxalate, amyloidosis, ...)	6 (3.7)	3 (6.5)	3 (2.5)

429

430 Values reported as numbers and %, mean±SD, or median with interquartile ranges, as
431 appropriate. *eGFR (estimated Glomerular Filtration Rate) was calculated according to the
432 Chronic Kidney Disease Epidemiology Collaboration equation. ACEi/ARB, angiotensin-
433 converting enzyme inhibitor/angiotensin II receptor blocker. ** One biopsy may have more
434 than one diagnosis.

435 **Table 3: univariable and multivariable analysis**

436 r^2 value for the multivariable analysis was 0.54. ADC: apparent diffusion coefficient
437 [$\times 10^6 \text{mm}^2/\text{s}$]; T1[ms]; eGFR: estimated glomerular filtration rate;

	Univariable models			Multivariable model	
	Coefficient (95%CI)	r^2	p value	Coefficient (95%CI)	p value
Δ ADC	-0.09 (-0.11 to -0.06)	0.27	<0.001	-0.05 (-0.07 to -0.03)	<0.001
Δ T1	0.06 (0.05 to 0.08)	0.23	<0.001	0.03 (0.01 to 0.05)	<0.001
eGFR	-0.41 (-0.50 to -0.31)	0.30	<0.001	-0.22 (-0.32 to -0.12)	<0.001
Phosphate	29.64 (19.61 to 39.66)	0.18	<0.001	9.35 (-0.91 to 19.6)	0.074
Hemoglobin	-0.48 (-0.61 to -0.35)	0.24	<0.001	-0.09 (-0.23 to 0.04)	0.173
Calcium	-23.97 (-46.08 to -1.85)	0.02	0.034	11.40 (-9.99 to 32.80)	0.293
Albumin	-0.98 (-1.61 to -0.35)	0.03	0.003	-0.38 (-1.06 to 0.29)	0.265
Proteinuria	3.58 (1.91 to 5.25)	0.11	<0.001	1.34 (-0.23 to 2.92)	0.093

438 **Figure legends**

439 **Figure 1:** Flowchart illustrating patient recruitment

440 **Figure 2:** Representative MRI images showing ADC, T1 and T2 maps in a kidney with low
441 (<20%, upper row), and diffuse (>60%, lower row), cortical fibrosis. Masson trichrome sections
442 are displayed for histological comparison.

443 **Figure 3:** Correlations between MRI indices and Fibrosis and eGFR. Scatter plots of Δ ADC
444 (A), Δ T1 (B) versus IF. **Scatter** plot of Δ ADC (C), and Δ T1 (D) versus eGFR. Each symbol
445 represents one patient. The continuous line indicates least-square linear regression. ADC:
446 apparent diffusion coefficient. Correlation coefficient (r) and significance (p) are displayed in
447 each scatter plot.

448 **Figure 4:** Boxplot comparison of predicted fibrosis using a multivariable model containing
449 eGFR, Δ ADC and Δ T1 and histological fibrosis in four categories (<10; \geq 10-<25; \geq 25-<50;
450 >50%). The horizontal bar inside each box is the median, the top and bottom of the box indicate
451 the interquartile range, the T bars indicate the 95th percentiles.

452 **Figure 5:** ROC curves of multivariable model (Δ ADC, Δ T1, eGFR) in predicting fibrosis for
453 cutoffs of 10 % (A), 25% (B), and 50% (C) AUC: Area under the Curve; ROC: Receiver
454 Operating Characteristic.

455 **Supplementary figure 1:** Correlations between MRI indices and Fibrosis. Scatter plots of
456 absolute cortical ADC (A), Medullary ADC (B), cortical T1 (C) and Medullary T1 (D) versus
457 IF. The continuous line indicates least-square linear regression. ADC: apparent diffusion
458 coefficient. Correlation coefficient (r) and significance (p) are displayed in each scatter plot.

459 **Supplementary Figure 2:** Correlations between MRI indices and fibrosis in native and
460 allograft patients. Scatter plots of Δ ADC (A), and Δ T1 (B) versus IF, in native kidney (solid
461 circles) and kidney allograft (open circles) patients. Each symbol represents one patient.

462 **Supplementary Figure 3:** Correlations between T2 and fibrosis in native and allograft patients.
463 Scatter plots of absolute cortical T2 (A), medullary T2 (B), Δ T2 (C), cortical T1 (E) and cortical
464 fibrosis. Scatter plot of Δ T2 versus eGFR (D). Scatter plots of Δ T2 versus fibrosis in native
465 versus kidney allograft kidneys (E and F). Each dot represents one patient. The continuous line
466 indicates least-square linear regression. eGFR: estimated glomerular filtration rate Correlation
467 coefficient (r) and significance (p) are displayed in each scatter plot.

468 **Supplementary Figure 4:** Correlations between laboratory values and fibrosis. Scatter plots of
469 ln Creatinine (A), ln PTH (B), 25-hydroxyvitamin D (C), ln proteinuria (D), albumin (E),
470 eGFR(F), calcium (G), phosphate (H) and hemoglobin (I) versus interstitial fibrosis. Each
471 symbol represents one patient. The continuous line indicates least-square linear regression.
472 Correlation coefficient (r) and significance (p) are displayed in each scatter plot.

473 **Supplementary Figure 5:** ROC curves of Δ ADC in predicting fibrosis for cutoffs of 10 % (A),
474 25% (B), and 50% (C) AUC: Area under the Curve; ROC: Receiver Operating Characteristic.

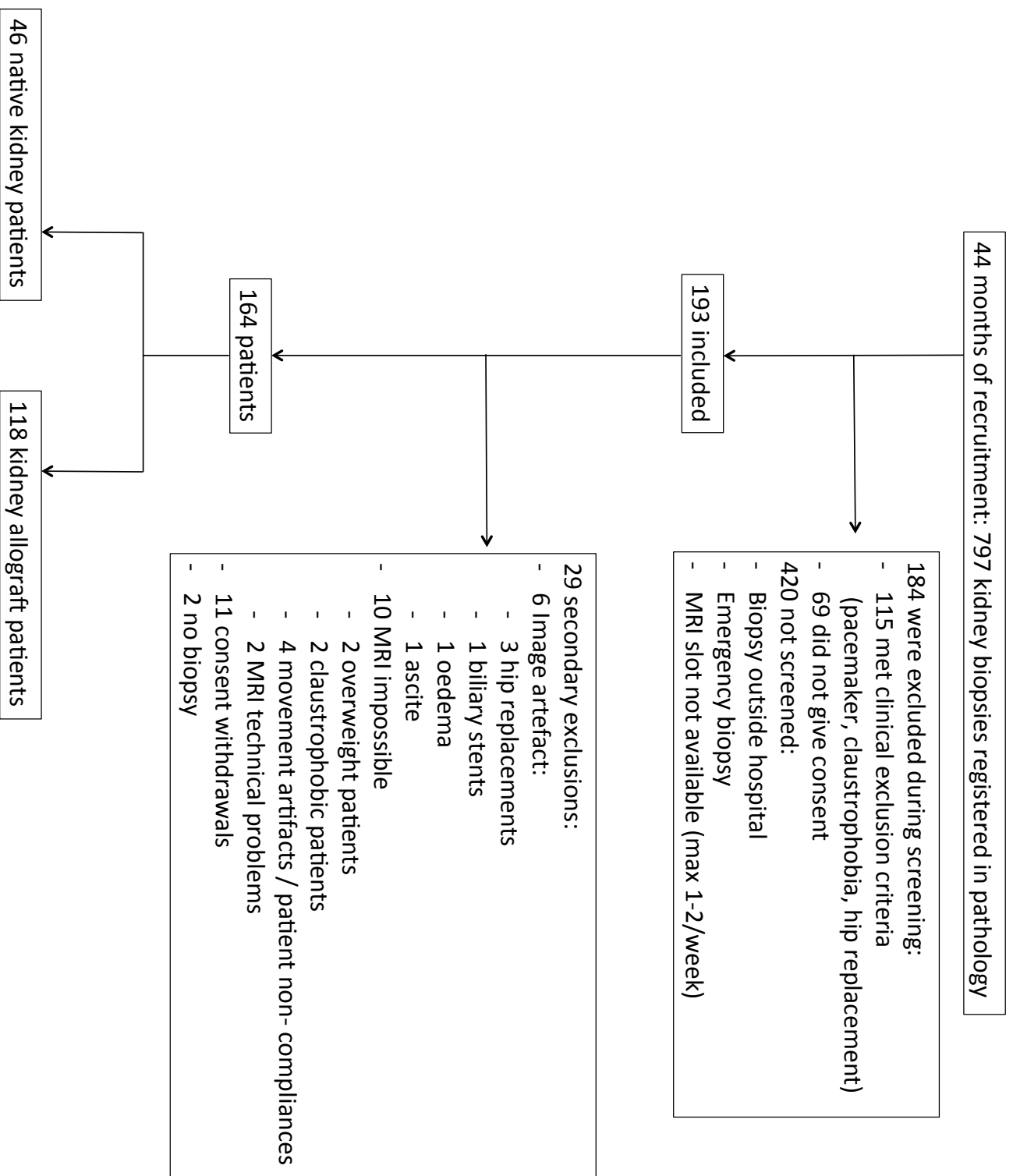


Figure 1

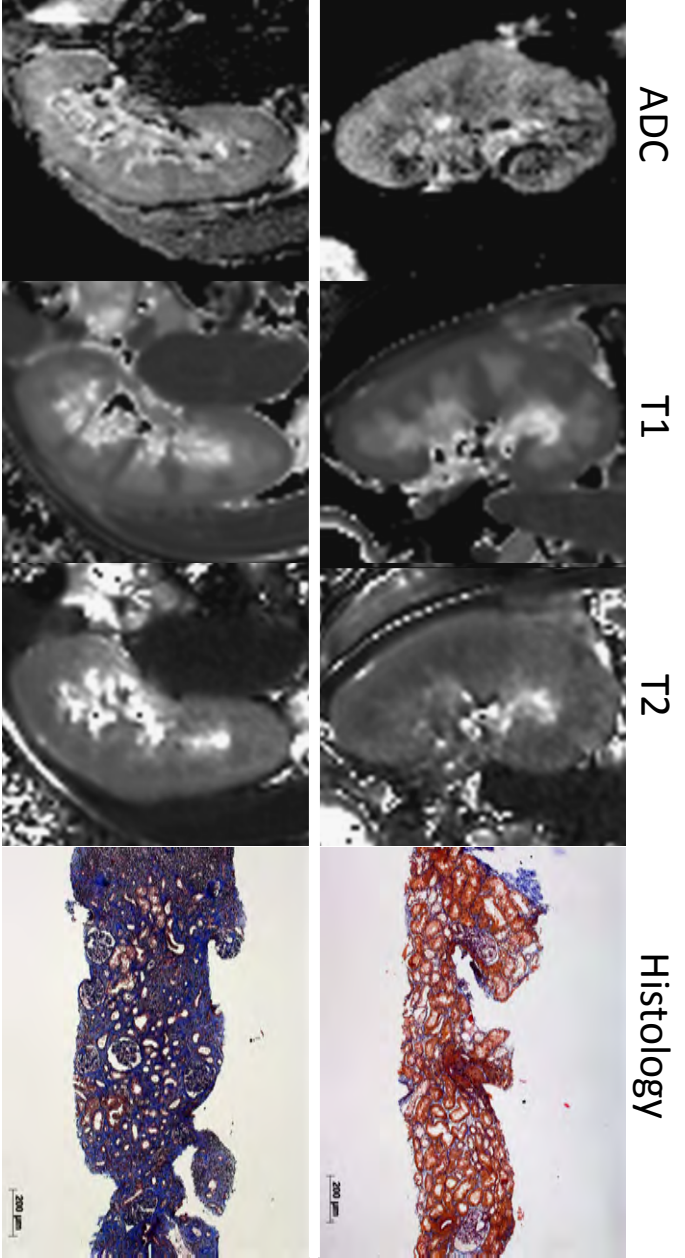


Figure 2

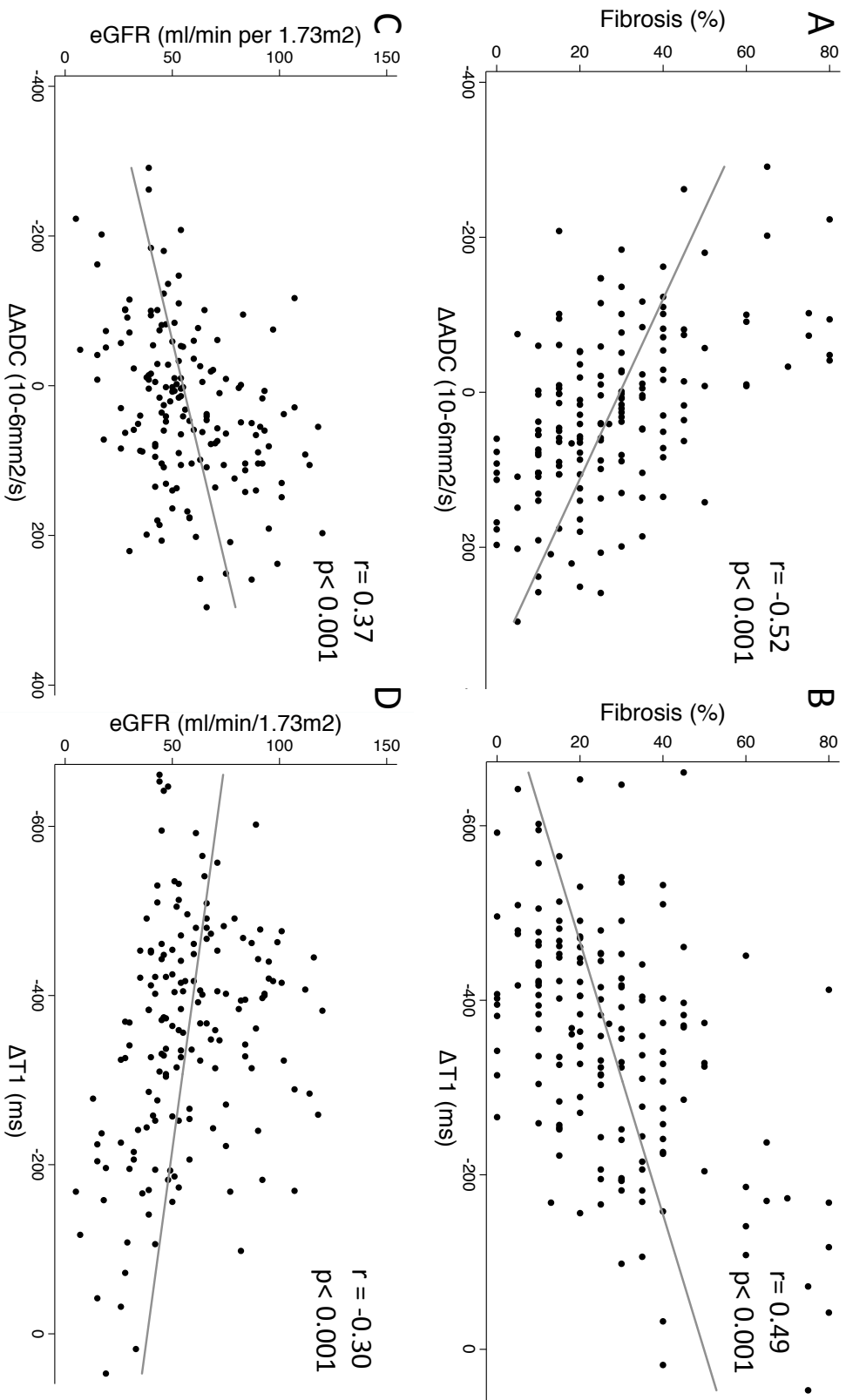


Figure 3

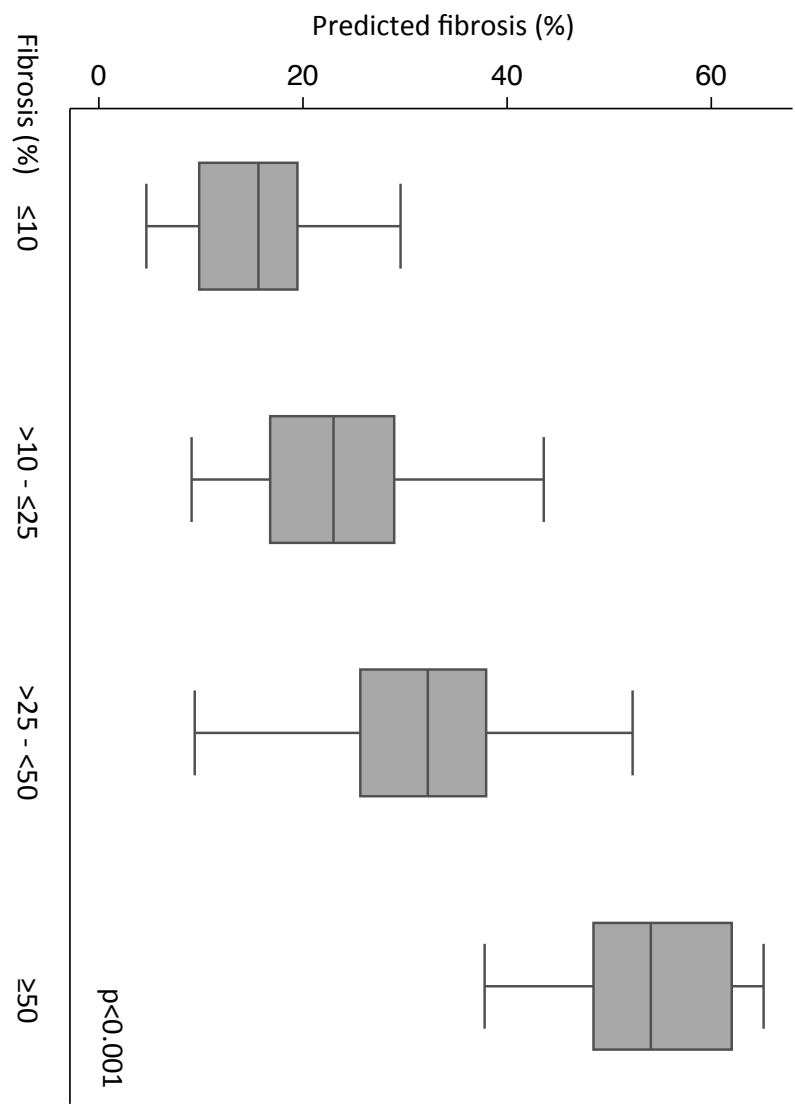


Figure 4

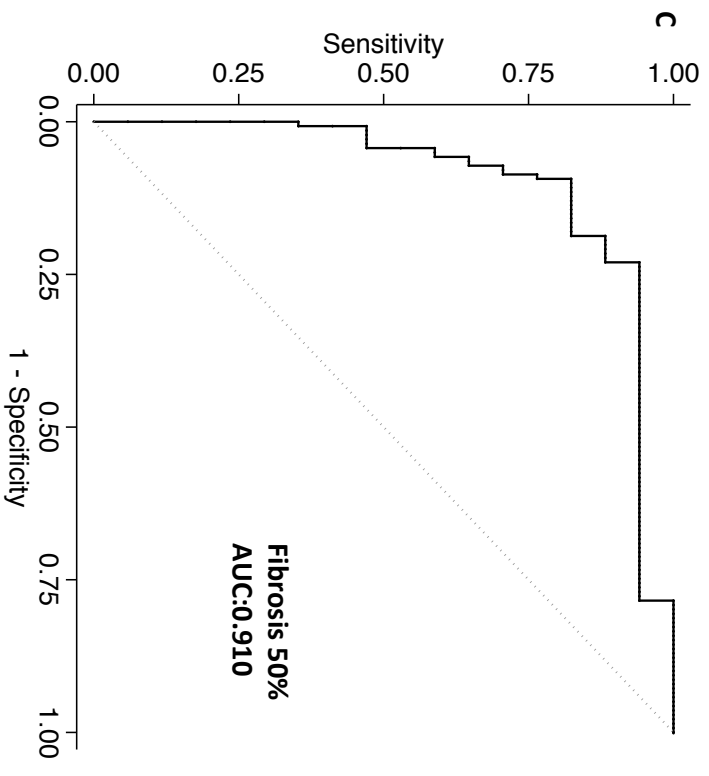
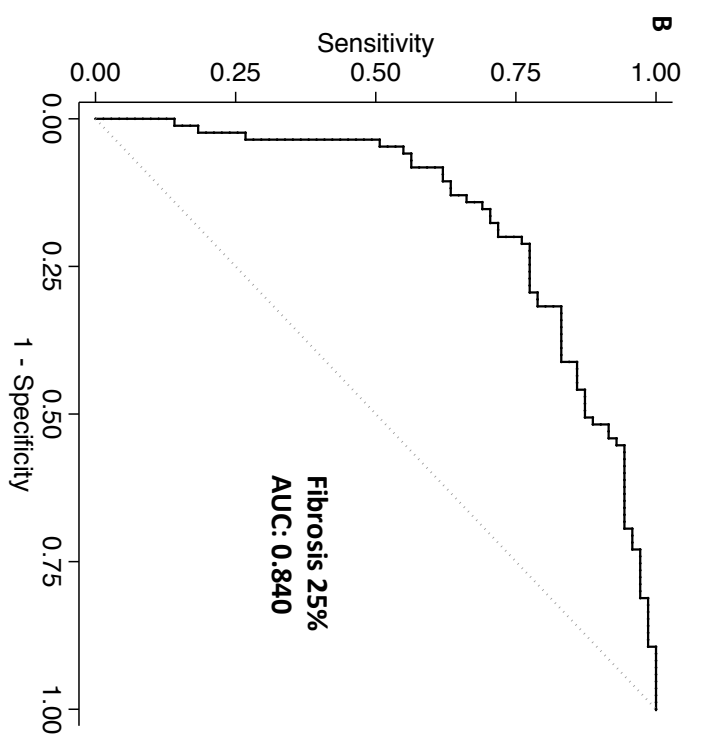
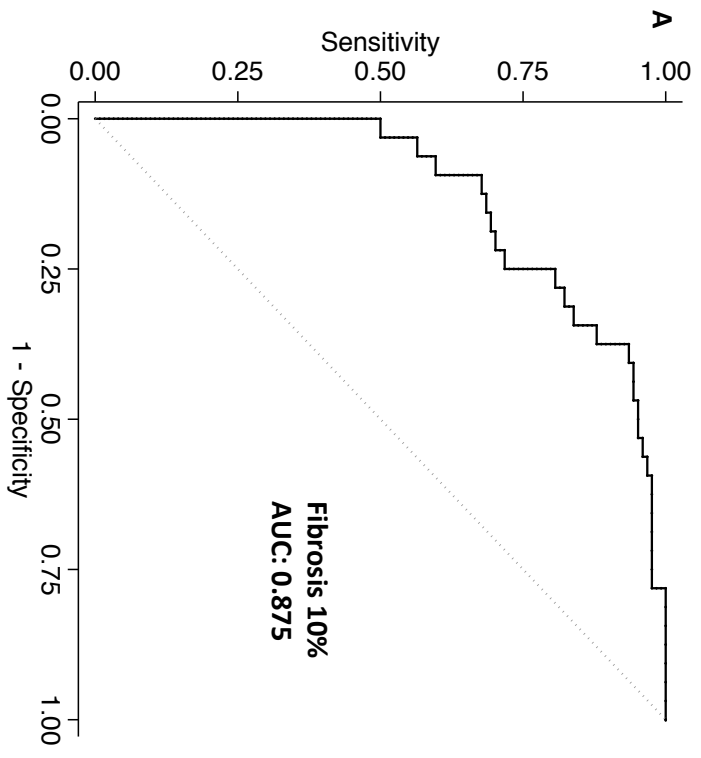
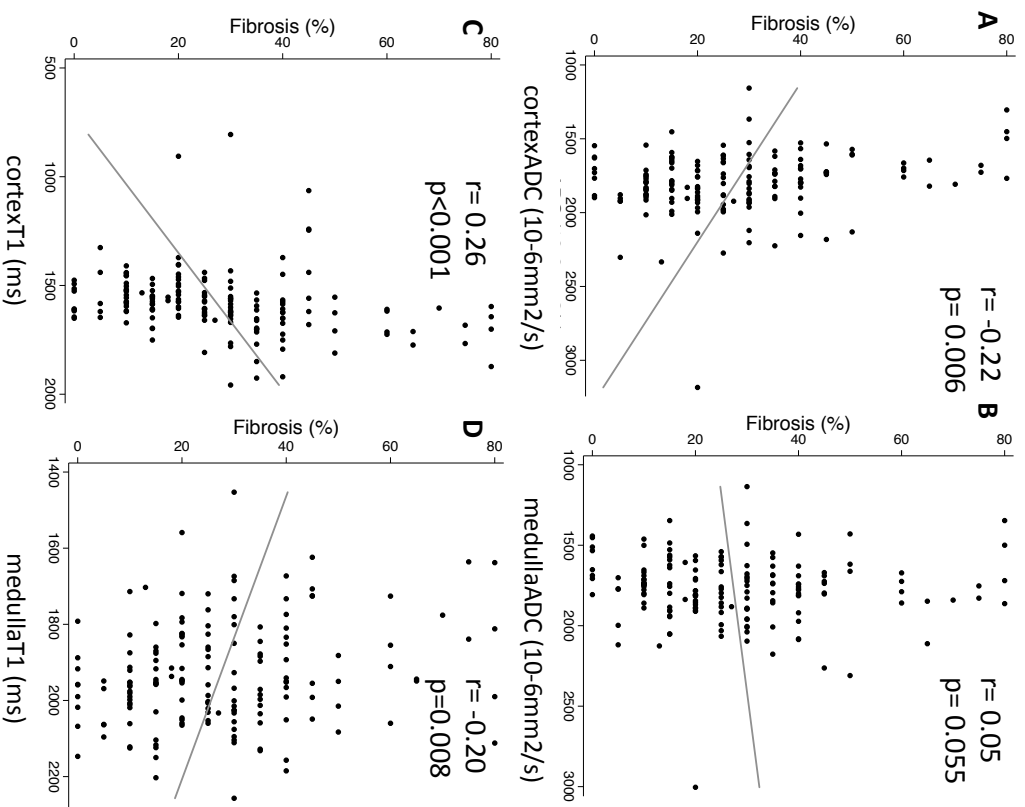
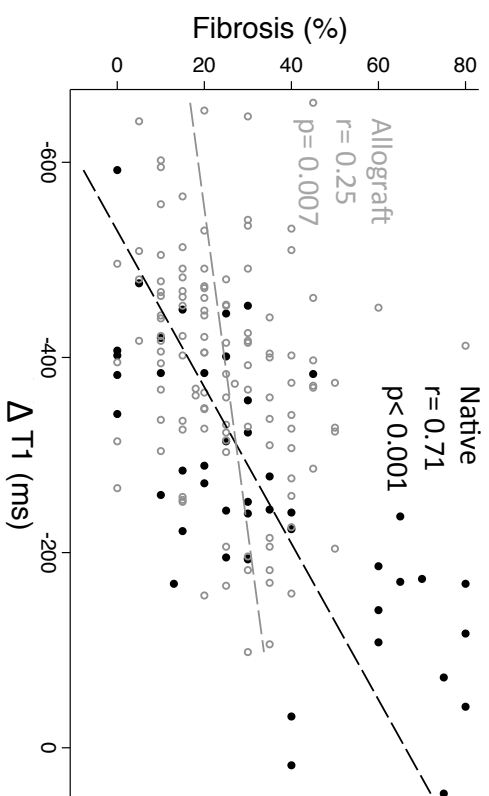
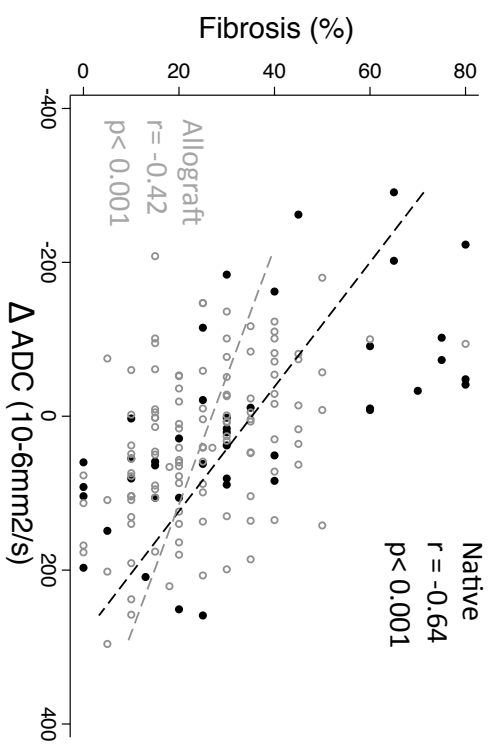


Figure 5

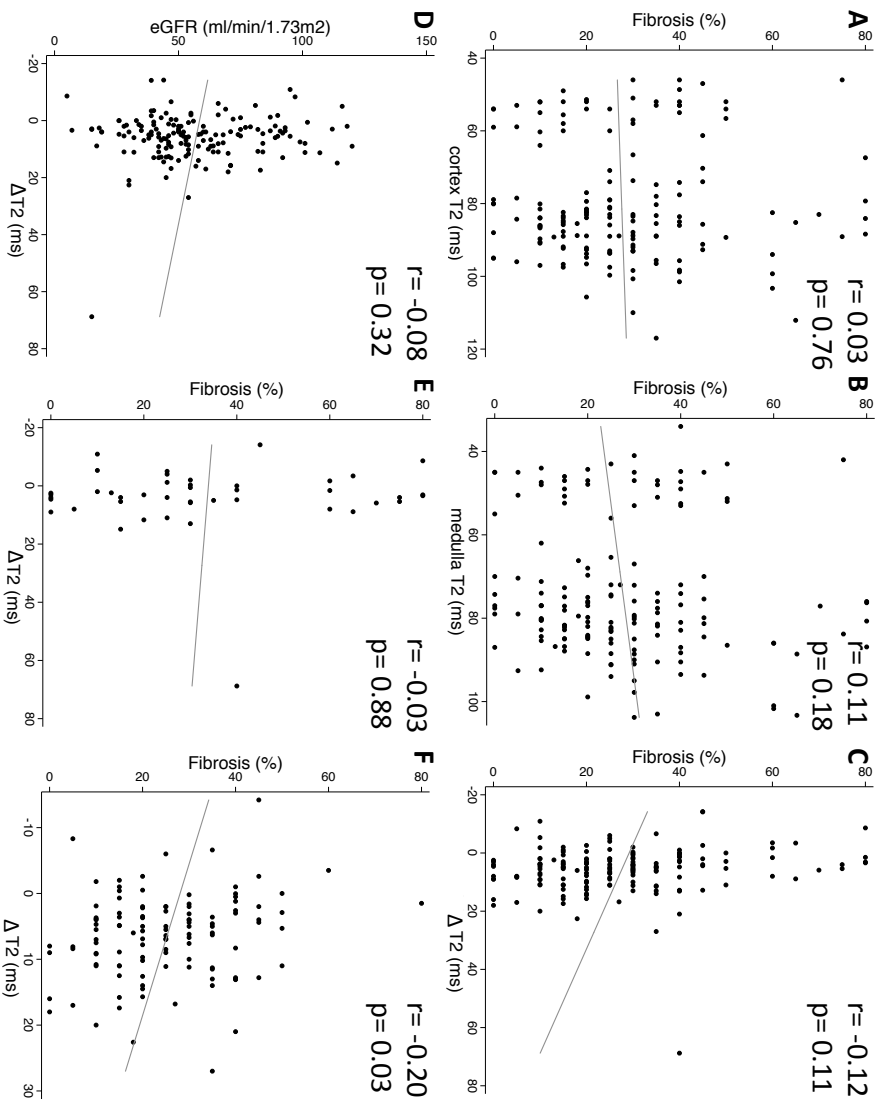
Supplementary



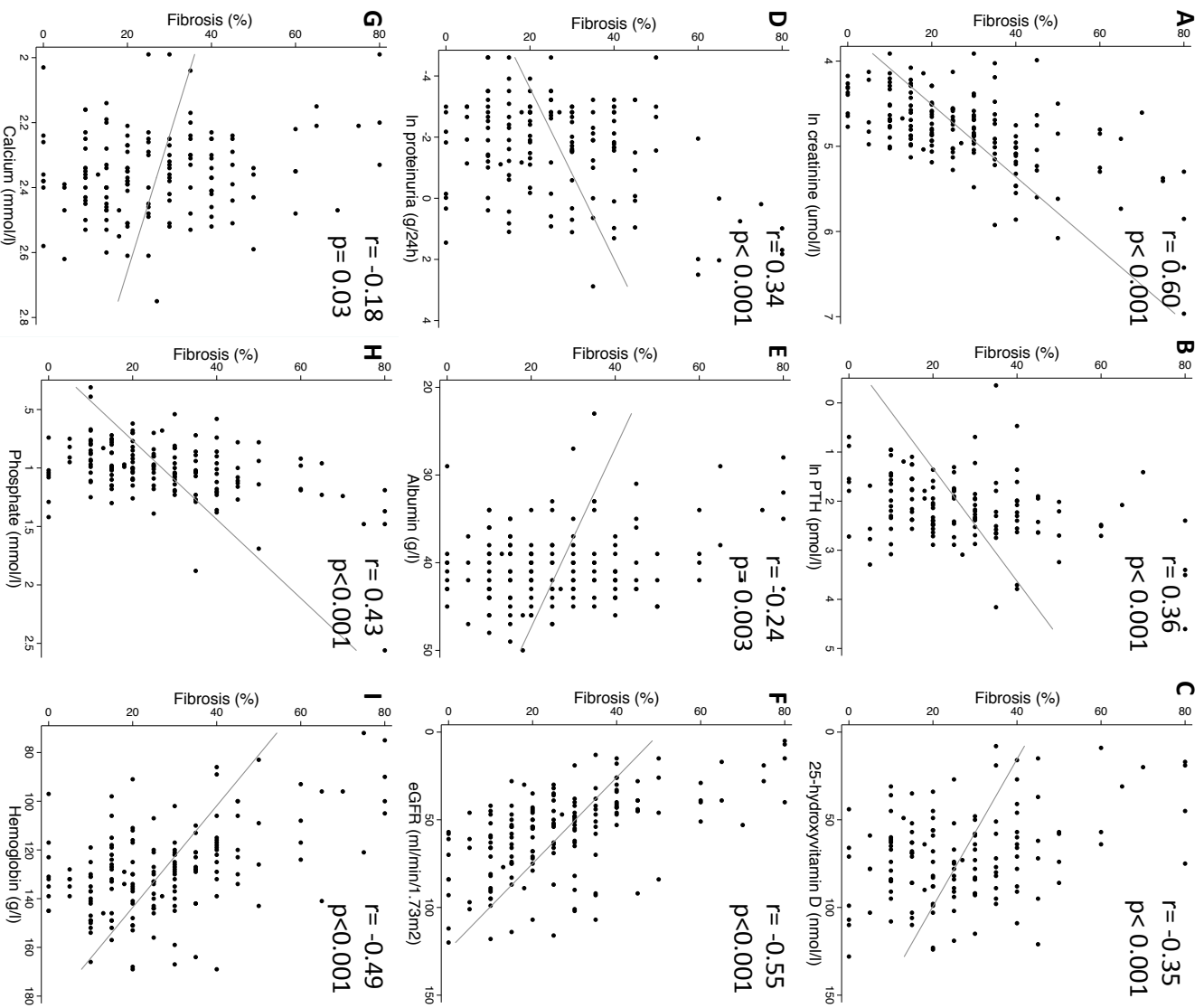
Supplementary Figure 1



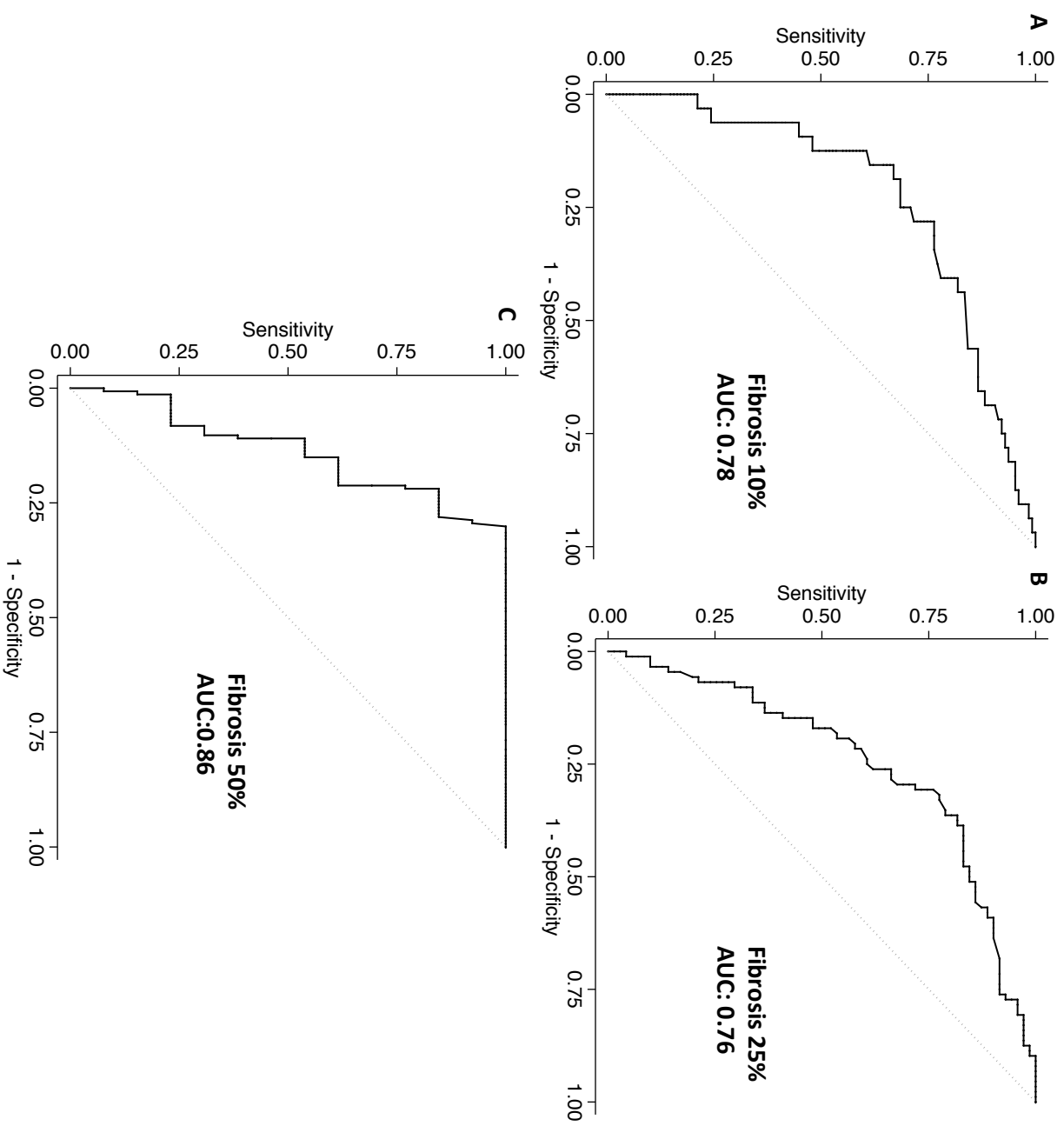
Supplementary Figure 2



Supplementary Figure 3



Supplementary Figure 4



Supplementary Figure 5