

1 **Associations between symptoms, donor characteristics and IgG antibody**  
2 **response in 2082 COVID-19 convalescent plasma donors**

3

4 Running title: donor characteristics and antibody levels

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29 **Complete text word count: 3437**

30 **Figures: 4**

31 **Tables: 3**

32 **Abstract (150 words)**

33 Many studies already reported on the association between patient characteristics on the severity of  
34 COVID-19 disease outcome, but the relation with SARS-CoV-2 antibody levels is less clear. To  
35 investigate this in more detail, we performed a retrospective observational study in which we used  
36 the IgG antibody response from 11,118 longitudinal antibody measurements of 2,082 unique COVID  
37 convalescent plasma donors. COVID-19 symptoms and donor characteristics were obtained by a  
38 questionnaire. Antibody responses were modelled using a linear mixed-effects model. Our study  
39 confirms that the SARS-CoV-2 antibody response is associated with patient characteristics like body  
40 mass index and age. Antibody decay was faster in male than in female donors (average half-life of 62  
41 versus 72 days). Most interestingly, we also found that three symptoms (headache, anosmia, nasal  
42 cold) were associated with lower peak IgG, while six other symptoms (dry cough, fatigue, diarrhoea,  
43 fever, dyspnoea, muscle weakness) were associated with higher IgG concentrations.

44

45 **Keywords**

46 Longitudinal; symptoms; antibodies; COVID-19; characteristics; CCP

## 47 Introduction

48 Severe acute respiratory syndrome corona virus 2 (SARS-CoV-2) emerged late 2019 in China, and by  
49 March 2020 was declared a pandemic by the World Health Organization (WHO). As of September  
50 2021, over 200 million individuals have been infected with COVID-19, which has inflicted an immense  
51 impact on the healthcare system worldwide. The virus mainly targets the respiratory tract, which can  
52 lead from mild symptoms to severe respiratory distress syndrome. Studies have shown that antibody  
53 responses against the SARS-CoV-2 spike protein can be first detected 1-3 weeks post symptom onset  
54 (1, 2) in most COVID-19 patients and remain in circulation for up to 1 year (3-6). There is however a  
55 substantial variation in antibody levels between individuals (5).

56 Many studies have reported on the association between disease severity and donor  
57 characteristics, such as sex, body mass index (BMI), age, and blood group. Males tend to be more  
58 susceptible to develop a severe course of the SARS-CoV-2 virus infection (7, 8). In addition, age above  
59 50 and obesity are also associated with increased risk of severe outcome (9-12). ABO blood type may  
60 also play a role in COVID-19 infection, but the exact influence remains unclear (13, 14).

61 Antibody responses also seem to be associated with symptoms and clinical information. In  
62 general, SARS-CoV-2 antibody levels are higher in patients with a severe disease outcome (15). A  
63 recent study in which COVID-19 convalescent plasma (CCP) donors were followed for three months  
64 after symptom resolution showed that greater disease severity, older age, male sex, and high BMI  
65 correlate with high SARS-CoV-2 antibody levels (7, 16). The same study also reported that particularly  
66 the symptoms fever, body aches, and low appetite correlate with high SARS-CoV-2 antibody levels.  
67 Limitations of this study include a small number of subjects and the low number of longitudinal data  
68 points available for each subject, which restricts the possibilities to analyse trends in antibody levels  
69 over time and the association with donor characteristics and symptoms.

70 Here, we aimed to gain a more detailed insight into individual symptoms and donor  
71 characteristics and their association with the IgG antibody response. Therefore, we analysed a  
72 longitudinal data set of 11,118 anti-RBD antibody measurements of 2,082 unique CCP donors.  
73 Interestingly, we found that three symptoms (headache, anosmia, nasal cold) were associated with  
74 lower peak IgG, while six other symptoms (dry cough, fatigue, diarrhoea, fever, dyspnoea, muscle  
75 weakness) were associated with higher IgG concentrations.

## 76 Materials and Methods

77

### 78 *Study population samples*

79 Between April 2020 and March 2021, Sanquin Blood Bank (Amsterdam, the Netherlands) collected  
80 samples from over 24,000 COVID-19 recovered adults who enrolled in the CCP programme. Within  
81 this programme plasma is derived from patients that recovered from COVID-19, with the aim to help  
82 patients recover from COVID-19. Donation was voluntary and non-remunerated, and donors provided  
83 written informed consent before their first donation. Donors were included based on either a positive  
84 PCR or presence of anti-RBD IgG antibodies above 80 Arbitrary Units per ml (AU/ml) and after being  
85 free of symptoms for at least two weeks. Donors donated plasma on average every two weeks, until  
86 antibody levels were below 4 AU/ml in two consecutive donations. Only donors with at least three  
87 consecutive antibody measurements and a complete questionnaire were included in the analyses,  
88 resulting in a study population of 2,082 donors (Supplemental Figure 1).

### 89 *Questionnaire*

90 Starting August 2020, donors that enrolled in the convalescent plasma programme were invited by e-  
91 mail to fill out an online questionnaire, programmed in Qualtrics (SAP, Walldorf, Germany). The  
92 questionnaire included questions about the possible origin of the infection, the reason why donors  
93 were tested and a list of 18 symptoms considered to be COVID-19-related according guidelines

94 specified by the Dutch National Institute for Public Health and the Environment (17). Participants could  
95 indicate if they experienced symptoms and, if the symptoms were present, how severe these  
96 symptoms were on a 4-point scale, from 1 (very mild) to 4 (severe). Additionally, participants were  
97 asked about the duration of their symptoms, whether they consulted a physician or were admitted to  
98 hospital and/or intensive care units. The full questionnaire is included as an online supplement.  
99 Donors were excluded from analysis if sex, age and/or date of illness was absent.

## 100 ***Antibody measurements***

101 IgG to RBD was measured essentially as described before (5, 6). In brief, plates were coated with  
102 recombinant RBD, incubated with samples, and bound IgG antibodies were detected using an anti-  
103 human IgG antibody (MH16, Sanquin); quantification was done relative to a plasma pool consisting of  
104 CCP donors and expressed as AU/mL.

## 105 ***Statistical model***

106 Longitudinal trends in antibody levels were analysed with a linear mixed-effects model, using log-  
107 transformed anti-RBD IgG levels as outcome variable. Timepoint 0 corresponds to 20 days post onset  
108 of symptoms (1) (1). As such, the estimated intercept of the model corresponds to a donor's estimated  
109 peak IgG level (18). The estimated slope of the model is used to calculate a donor's IgG half-life, in  
110 days:

$$111 \quad t_{1/2} = \log\left(\frac{1}{2}\right) / \text{slope}.$$

112 Only measurements within six months post onset of symptoms (1) were included, as in later stages of  
113 recovery antibody decline is expected to slow down and no longer expected to follow a loglinear  
114 decline (5).

115 A three-step approach was used to analyse the effects of the covariates. In the first step, a  
116 null-model was fit to the data, using time as the only predictor variable and allowing a random  
117 intercept and slope to be estimated for each donor. In the second step, we tried to explain the variance  
118 in random intercepts and slopes by including fixed effects for donor characteristics, i.e. sex, age,  
119 height, weight, BMI, and blood group (ABO and RhD), in addition to the random intercept and slope  
120 per donor. In the third step, fixed effects that were statistically insignificant in the second step were  
121 removed and additional donor information variables obtained from the questionnaires were added as  
122 fixed effects (Table S1 and S2). This information concerned data on hospitalization, ICU admission, co-  
123 morbidities, and the presence of 18 symptoms as shown in Table 1. This approach allowed separate  
124 estimation of the proportion of variance explained by donor characteristics and clinical information.

125 Significance levels of individual variables were estimated using Satterthwaite's approximation  
126 (19), as degrees of freedom cannot be calculated exactly in models that include both random and fixed  
127 effects. Because this approximation is slightly anti-conservative, an alpha-level of 0.01 was chosen to  
128 determine statistical significance. Non-significant predictors were excluded after each step. Relative  
129 quality, taking into account both goodness of fit and model complexity, of the models was assessed  
130 by comparing the Akaike information criteria (AIC) after each step.

131 Data were processed and analysed with the *R* programming language and environment for  
132 statistical computing (version 4.0.3), using packages *lme4* and *lmerTest* for analyses and *ggplot2* for  
133 generating graphs.

134

## 135 Results

### 136 *Study population characteristics*

137 We used 11,118 antibody measurements of 2,082 unique donors to study the associations between  
138 symptoms, donor characteristics, and IgG antibody response. The number of available antibody  
139 measurements per donor ranged from 3 to 18 measurements. In addition, each donor completed a  
140 questionnaire, which gave insight into symptoms and donor characteristics. Table 1 shows the  
141 distributions of donor and COVID-19 related disease characteristics in the study population.

142 Compared to all active whole-blood and plasma donors in 2020, donors in our study  
143 population are slightly older (46 vs 42 years for women, 52 vs 48 years for men). Median weight and  
144 height, as well as proportion of female donors and rhesus D blood group are similar to those of the  
145 active donor population. Blood group A is overrepresented in our study population (47% vs 39% for  
146 women, 45% vs 39% for men), while blood group O is underrepresented (39% vs 47% for women, 42%  
147 vs 47% for men).

148

### 149 *Null-model fit (step 1)*

150 In the first step we estimated an intercept and slope for each individual donor using the null model,  
151 describing the linear relationship between log-transformed IgG levels and time post onset of  
152 symptoms (1). The residuals, i.e. the difference between measured IgG and predicted IgG as estimated  
153 by the null model, follow a normal distribution with mean 0 and standard deviation of 0.21 log  
154 (AU/ml). This distribution is independent of time post onset symptoms, supporting the assumption  
155 that the relationship is linear after log-transformation. Given this assumption, the estimated peak IgG  
156 level set at 20 days POS is most likely an accurate extrapolation and allows for comparisons between  
157 donors. Supplemental Figure 2A shows the fitted line and actual measurements for four randomly  
158 selected donors (donors A to D). Supplemental Figure 2B shows the distribution of the residuals over  
159 all observations for all donors.

160 After analysing all samples, we found a median peak IgG concentration of 38.8 AU/ml (IQR  
161 20.9-78.6) and a median half-life of 66 days (IQR 50-94) (Figure 1). For the majority of donors, the  
162 estimated slope corresponds to a plausible antibody half-life. However, for 80 donors (3.8%), the fitted  
163 slope was positive, which results in a negative estimated half-life estimate. For an additional 59 donors  
164 (2.8%), the estimated half-life is extremely long (defined here as more than 365 days, but estimates  
165 ranged up to 16,000 days). This occurs when the estimated slope is very close to zero (but still  
166 negative), which may happen when IgG levels barely decrease between measurements and no decay  
167 in antibody levels are measured. Examples of donors with a negative half-life and very long half-life  
168 are given in Supplemental Figure 3. These donors were not excluded from the study in order not to  
169 overstate accuracy, and because there was no reason to assume the IgG measurements were  
170 incorrect.

171

### 172 *Associations with predictor variables (step 2)*

173 The results of step 2, where individual donor characteristics were added to the model as predictor  
174 variables, are shown in Figure 2A-C and Table 2. Sex was associated with the slope (Figure 2A), as the  
175 rate of antibody decay is faster in men: the median slope for men corresponds to a half-life of 62 days,  
176 while this is 72 days for women. Men displayed higher peak IgG levels than women, but this difference  
177 was not statistically significant ( $p = 0.68$ ). Age (Figure 2B) and BMI (Figure 2C) were both positively  
178 correlated with peak IgG concentration. A one-year increase in age corresponds to a 0.013 increase in  
179 the log-transformed IgG level, an increase of one BMI point corresponds to a 0.024 increase in log-  
180 transformed IgG level. No significant associations with antibody titres were found for variables blood

181 group, height, and weight. Random effects for peak IgG level and antibody half-life are positively  
182 correlated with a correlation coefficient of 0.29, indicating that higher peak IgG is moderately  
183 associated with higher (less negative) slope, and therefore with a longer half-life.

184

### 185 ***Associations with clinical information (step 3)***

186 After adding clinical information significant associations with peak IgG concentration were found for  
187 hospital admission and various clinical symptoms (Figure 2D, Figure 3, Figure 4 and Table 2). Hospital  
188 admission was significantly associated with both higher peak IgG level and shorter half-life (Figure 2D).  
189 Nasal cold, headache, and anosmia were associated with lower peak IgG levels, while dry cough,  
190 fatigue, fever, dyspnoea, diarrhoea, and muscle weakness were associated with higher peak IgG levels.  
191 Figure 3 shows the estimated peak IgG level when these symptoms are present. Note that values on  
192 the y-axis are the predicted peak IgG levels when all continuous variables are equal to their average  
193 value, and all binary variables (hospital admission and all other symptoms) equal zero.

194 The largest difference was found for the 'hospital admission' variable. Donors admitted to the  
195 hospital had considerably higher antibody levels, with an estimated difference of 77.8 AU/ml on the  
196 peak IgG concentration. These donors also have a faster rate of antibody decay, corresponding to an  
197 estimated half-life of 48 days (95% CI: 40-58 days) for men, and 60 days (95% CI: 49-80 days) for  
198 women.

199

### 200 ***Variance explained by model***

201 In the null-model that was fitted in step 1 (without any fixed effects), all variation in peak IgG  
202 and half-life was attributed to the individual variation per donor. As fixed effects were added in step  
203 2, part of this variation was now explained by these fixed effects, and the variation explained by the  
204 random effects decreased. Table 3 shows the variance of the random effects per donor in the null-  
205 model, as well the variance of the random effects as after adding donor characteristics as covariates  
206 (step 2), and after adding the clinical information (step 3). The variance reduction relative to the null-  
207 model (step 1) by the addition of extra explanatory variables in each step is also provided. Model fit  
208 was compared using the Akaike Information Criterion (AIC) and tested for statistical significance using  
209 a nested ANOVA, results of which are shown in Table 3.

## 210 **Discussion**

211 In this retrospective observational study, we investigated potential associations between SARS-CoV-2  
212 specific antibody kinetics and various donor characteristics and COVID-19 symptoms. To our  
213 knowledge, this is currently the largest study that describes such associations. Individual antibody  
214 responses were modelled using a linear mixed-effects model, from which peak IgG concentration and  
215 antibody half-life were determined. Symptoms and donor characteristics were obtained from a  
216 questionnaire. Our study shows that the SARS-CoV-2 antibody response is associated with patient  
217 characteristics like sex, age, and BMI. Of note, we also found that specific COVID-19 symptoms are  
218 associated with antibody levels.

219 As reported earlier, we found a large variation in anti-RBD antibody peak levels. A strength of  
220 our study are the longitudinal measurements, which enabled us to reliably estimate the peak level of  
221 each individual donor independent on the timing of the first antibody measurement. Only a quarter  
222 of the variation in peak IgG concentration between patients can be explained by associations with  
223 donor characteristics and disease symptoms. To a lesser degree, donor characteristics were also  
224 associated with differences in antibody half-life, which was also variable between donors, albeit less  
225 than the peak level. The antibody half-life was not correlated to peak levels. Whether these

226 differences in antibody half-life reflect differences in protection for reinfection will be investigated,  
227 and this thoroughly characterized donor cohort can serve as bench mark for those studies.

228 Six symptoms (dry cough, fatigue, diarrhoea, fever, dyspnoea, muscle weakness) were  
229 associated with higher IgG concentrations and three symptoms (headache, anosmia, nasal cold) were  
230 associated with lower peak IgG concentrations against RBD. This association between symptoms and  
231 antibody levels may possibly reflect the fact that the SARS-CoV-2 virus frequently initiates infection in  
232 the upper airways (mild symptoms and low IgG levels) before spreading through the body (severe  
233 symptoms and high IgG levels). Headache, anosmia and nasal cold were common symptoms, each  
234 present in at least 50% of patients in our population. Fatigue was present in more than 70% of patients  
235 and was associated with higher peak IgG concentration, suggesting more severe illness. A previous  
236 study in a hospital cohort found that fatigue and dyspnoea are prognostic for severe infection, and a  
237 stuffed nose (comparable to nasal cold) for mild infection, which is in line with our findings (20).

238 Furthermore, we found higher age and BMI to be associated with higher peak IgG  
239 concentrations. Sex was not associated with peak IgG concentration, but men had significantly shorter  
240 antibody half-lives than women (62 vs 72 days respectively). The small group of patients that had been  
241 admitted to hospital displayed both higher peak IgG concentrations and shorter half-lives. Probably  
242 this effect is the result of the presence of short-lived plasmablasts that produce high levels of  
243 antibodies. Previous studies found sex differences in COVID-19 immune responses, with higher IgG  
244 concentrations associated with male sex, older age, and hospitalization (21-23). Although our results  
245 are consistent with these findings for age and hospitalization, we found that the association between  
246 male sex and higher peak IgG concentration was not significant after correction for age and BMI. This  
247 suggests that the previously found association with male gender was possibly due to the increased  
248 risk of severe disease in men. Most studies on differences in antibody response are performed in  
249 hospital cohorts, our study population consisted mainly of recovered patients that were not admitted  
250 to hospital (96.7%), and therefore disease severity is expected to be lower. Consistently, BMI in the  
251 non-hospitalized group was 25.9 compared to 28.8 in hospitalized patients.

252 A strength of our study is the large number of recovered patients included in our study  
253 population. The status of Sanquin as the only blood bank in the Netherlands, combined with well-  
254 established connections with municipal health services, allowed us to invite people with a positive  
255 PCR test to become CCP donors after recovery. This allowed us to both include non-hospitalized and  
256 hospitalized patients in the cohort. However, we could only include donors who were healthy enough  
257 to regularly donate plasma, which means that our results are mainly applicable towards patients with  
258 a mild outcome. As a result our study is more representative of the total COVID-19 patient population  
259 than studies on hospitalized patient cohorts. It should also be noted that some bias may be present in  
260 our data, as symptoms are self-reported by patients after recovery. Relatively mild symptoms, such as  
261 nasal cold, may therefore be underreported by patients who at the same time experienced more  
262 severe symptoms, such as fever or dyspnoea. However, this explanation is unlikely to negate the  
263 association we found, as all symptoms associated with lower peak IgG were present in more than 50%  
264 of patients.

265 In conclusion, our study indicates that several COVID-19 symptoms are associated with SARS-  
266 CoV-2 antibody levels in addition to the previously described association with sex, age, and BMI.  
267 Discovery of these associations aids us in understanding why antibody responses differ between  
268 patients. The predictive value of IgG concentrations could also be used by blood banks to pre-select  
269 individuals with high and/or stable antibody levels as potential CCP donors.

## 270 Acknowledgements

271 We would like to thank all the donors who have kindly donated plasma.

## 272 Authorship contributions and conflict of interest

273 EvS, MPJ, TR conceived the study

274 AtB, CvH, MPJ, MvL, MV, FS, HV, LvdW, FQ, KvdH, TR, BH and EvS were involved in study design and  
275 organization  
276 MV conducted experiments  
277 MV, MPJ analyzed data  
278 MV, MS wrote the paper  
279 MPJ, TR, EvS supervised the study  
280 All authors provided critical revision of the paper.  
281 Authors declare no conflict of interest.

## 282 Funding

283

284 This study was supported by the European Commission (SUPPORT-E, grant number 101015756), by  
285 ZonMW (Protective Immunity, grant number 10430 01 201 0012), and Sanquin Blood Supply  
286 Foundation, PPOC grant 18-14/L2337.

## 287 Figure legends

288

289 **Figure 1. Anti-RBD IgG peak and half-life.** (A) Distribution of estimated peak IgG concentration (at 20  
290 days POS) and (B) estimated half-life of 2,082 COVID convalescent plasma donors, as estimated by the  
291 null model. Please note that since both distributions have an extremely long right tail, the horizontal  
292 axes are truncated at (A) 300 AU/ml and (B) 365 days, excluding 70 and 139 donors from left and right  
293 histograms, respectively.

294 **Figure 2. Associations between donor/clinical characteristics and antibody levels.** The effects of  
295 variables (A) sex, (B) age, (C) BMI, and (D) hospital admission on predicted antibody decline. Note that  
296 age and BMI are included in the model as continuous predictors; for clarity, the associations are only  
297 plotted for three values. Grey bands represent 95% confidence intervals.

298 **Figure 3. Predicted impact of various symptoms on anti-RBD IgG peak concentration.** Estimated peak  
299 IgG concentrations when different symptoms are displayed. For each of the symptoms here, the  
300 difference in peak IgG as compared to the group without this symptom is statistically significant with  
301  $p < 0.001$ .

302 **Figure 4. Effect size and 95% confidence intervals of fixed effects on anti-RBD IgG peak concentration**  
303 **(log-transformed) and the slope.**

304

## 305 Supplemental Figure legends

306

307 **Supplemental Figure 1. Flowchart showing the criteria for inclusion and exclusion of donors.**

308 **Supplemental Figure 2. Null model fit (step 1).** (A) Measured anti-RBD IgG levels (points) and fitted  
309 line as estimated by the linear model for four randomly selected donors, with (B) distribution of  
310 residuals over all observations, for all donors.

311 **Supplemental Figure 3. No decay in antibody levels.** Example of a donor with increasing IgG levels  
312 (left panel) and one with near-constant IgG levels (right panel). Estimated slopes for these donors are  
313 0.0024 and -0.0151, corresponding to estimated half-lives of -292 and 1843 days, respectively.

314

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**Table 1. Study population characteristics.** Continuous variables are represented by their median and interquartile range (IQR), categorical variables by absolute count and percentage.

Predictor variable	Female		Male	
	Median value or count	IQR or percentage	Median value or count	IQR or percentage
Number of donors (proportion of total)	1236	59.4%	846	40.6%
Number of donations per donor	6	4 – 8	6	4 – 10
Days POS at first donation	48	33 – 77	47	32 – 77
Days POS at last donation*	122	97 - 151	126	103 - 157
Age (years)	45.9	28.0 – 55.3	51.8	39.6 – 59.3
Height (cm)	171	167 – 176	184	180 – 189
Weight (kg)	73	65 – 83	88	80 – 97
BMI (kg/m <sup>2</sup> )	24.8	22.6 – 28.4	26.4	24.0 – 28.2
Blood group ABO				
- A	581	47%	381	45%
- B	120	9.7%	84	9.9%
- O	484	39%	352	42%
- AB	51	4.1%	29	3.4%
Blood group RhD				
- Positive	1024	83%	691	82%
- Negative	212	17%	155	18%
Hospital admission	19	1.5%	50	5.9%
Intensive care	4	0.3%	8	0.9%
Symptoms				
<i>Asymptomatic</i>	8	0.6%	7	0.8%
Fatigue	979	79%	597	71%
Anosmia/ageusia	853	69%	471	56%
Headache	820	66%	467	55%
Myalgia	705	57%	445	53%
Nasal cold	692	56%	424	50%
Fever	621	50%	507	60%
Dry cough	560	45%	396	47%
Sore throat	519	42%	307	36%
Chills	499	40%	356	42%
Sneezing	461	37%	381	45%
Dyspnoea	461	37%	297	35%
Muscle weakness	426	34%	260	31%
Diarrhoea	221	18%	102	12%
Nausea	184	15%	72	8.5%
Sputum production	178	14%	152	18%
Altered mental status	127	10%	80	9.5%
Skin rash	69	5.6%	27	3.2%
Vomiting	49	4.0%	28	3.3%

380 \* The maximum value is 182, as only donations within six months POS are included.

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383 **Table 2: Point estimates and 95% confidence intervals of fixed effects on log-transformed IgG levels.**

<b>Estimates of average intercept (log peak IgG) and slope</b>		
<b>Term</b>	<b>Estimate</b>	<b>95% CI</b>
Intercept (log peak IgG)	2.382	2.274 – 2.490
Slope ( $\Delta\log(\text{IgG})$ per day)	-0.010	-0.011 – -0.010
<b>Fixed effects on the intercept</b>		
	<b>Estimate</b>	<b>95% CI</b>
<i>*Sex: female</i>	-0.017	-0.063 – 0.096
Age (per 10 years increase)	0.128	0.100 – 0.157
BMI (per 5 points increase)	0.119	0.075 – 0.164
Hospital admission: yes	1.156	0.934 – 1.379
Headache: yes	-0.113	-0.193 – -0.032
Anosmia: yes	-0.111	-0.189 – -0.033
Nasal cold: yes	-0.101	-0.177 – -0.025
Dry cough: yes	0.095	0.019 – 0.171
Fatigue: yes	0.140	0.044 – 0.236
Diarrhoea: yes	0.148	0.043 – 0.252
Muscle weakness: yes	0.172	0.083 – 0.261
Shortness of breath: yes	0.196	0.111 – 0.280
Fever: yes	0.228	0.149 – 0.308
<b>Fixed effects on the slope</b>		
	<b>Estimate</b>	<b>95% CI</b>
Sex: female	0.003	0.002 – 0.004
Hospital admission: yes	-0.004	-0.007 – -0.001

384 \*The effect of sex on the intercept (peak IgG) was not statistically significant, but the variable is not excluded from the model  
 385 due to its effect on the slope.

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387 **Table 3. Variance of random effects in models of all three steps. Percentual variance decrease**  
 388 **relative to the null-model is given in brackets.**

Model	Variance of random effect on peak log(IgG) (reduction relative to the null-model)	Variance of random effect on slope $\Delta\log(\text{IgG})/\text{day}$ (reduction relative to the null-model)	AIC (ANOVA p-value relative to previous step)
Step 1: null-model	0.8814	0.0497	11886
Step 2: donor characteristics	0.7758 (-12%)	0.0485 (-2.4%)	11615 (p < 0.001)
Step 3: donor characteristics + clinical information	0.6610 (-25%)	0.0481 (-3.2%)	11290 (p < 0.001)

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