

# The confounded effects of age and exposure history in response to influenza vaccination



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## ARTICLE INFO

### Article history:

Received 18 May 2015

Received in revised form 31 October 2015

Accepted 19 November 2015

Available online 5 December 2015

### Keywords:

Influenza vaccine

Elderly

Vaccine efficacy

Immunosenescence

Repeated vaccination

## ABSTRACT

Numerous studies have explored whether the antibody response to influenza vaccination in elderly adults is as strong as it is in young adults. Results vary, but tend to indicate lower post-vaccination titers (antibody levels) in the elderly, supporting the concept of immunosenescence—the weakening of the immunological response related to age. Because the elderly in such studies typically have been vaccinated against influenza before enrollment, a confounding of effects occurs between age, and previous exposures, as a potential extrinsic reason for immunosenescence.

We conducted a four-year study of serial annual immunizations with inactivated trivalent influenza vaccines in 136 young adults (16 to 39 years) and 122 elderly adults (62 to 92 years). Compared to data sets of previously published studies, which were designed to investigate the effect of age, this detailed longitudinal study with multiple vaccinations allowed us to also study the effect of prior vaccination history on the response to a vaccine.

In response to the first vaccination, young adults produced higher post-vaccination titers, accounting for pre-vaccination titers, than elderly adults. However, upon subsequent vaccinations the difference in response to vaccination between the young and elderly age groups declined rapidly. Although age is an important factor when modeling the outcome of the first vaccination, this term lost its relevance with successive vaccinations. In fact, when we examined the data with the assumption that the elderly group had received (on average) as few as two vaccinations prior to our study, the difference due to age disappeared.

Our analyses therefore show that the initial difference between the two age groups in their response to vaccination may not be uniquely explained by immunosenescence due to ageing of the immune system, but could equally be the result of the different pre-study vaccination and infection histories in the elderly.

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## 1. Introduction

The World Health Organization and many national health authorities recommend yearly influenza vaccination for people at risk of developing serious complications, including elderly persons

over a defined age limit (typically 60 or 65 years). Various studies have described lower serological responses to vaccination in elderly than in young human adults [1–4]. For example, Beyer et al. [5] described how ten studies revealed a better immune response in young subjects than in elderly, 16 could not detect a significant difference, and four found an increased response in the elderly. Another quantitative meta-analysis of 31 studies consistently found lower seroprotection and seroconversion rates in the elderly compared to younger adults [6], findings that are in agreement with results from a database of 48 serological trials performed for regulatory purposes [6,7]. Thus most but not all published studies of serological comparisons report a lower antibody response to influenza vaccination in the elderly than in the young adults. A weakened

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immunological response related to age is known as immunosenescence, and this explanation is commonly used to explain the lower antibody response in elderly cohorts to vaccination.

Here we consider two different mechanistic drivers for immunosenescence. One mechanism concerns intrinsic drivers towards immunosenescence based on the ageing of the immune system, a complex process that is not yet fully understood, and may involve the age-dependent functioning of T-cells and a decreased output of naïve T-cells as a result of involution of the thymus [8–11]. Such an intrinsic immunosenescent process has been observed in studies of influenza-naïve rhesus macaques, where ageing results in declined antibody response to influenza vaccination [12,13].

The effects of such intrinsic immunological drivers may be compounded by extrinsic, or environmental, drivers of immunosenescence. An example of such an environmental contribution towards immunosenescence in reactions to influenza vaccine is previous infection with cytomegalovirus (CMV). CMV antibodies have been reported to increase pro-inflammatory potential, which contributes to unresponsiveness of the immune system. Because the presence of CMV antibody strongly correlates with age, this would also explain lower serological responses to vaccination against influenza in the elderly [14–16]. Similarly, studies on the effect of repeated vaccination in the elderly have proposed the explanation that prior vaccination may attenuate subsequent immune responses upon re-exposure to influenza [17,18].

Because humans partaking in vaccination studies are not naïve to influenza infection and their history of vaccination prior to enrollment is typically unknown, it is difficult to establish the relative contribution and possible interdependence of age and exposure history on immunosenescence. We designed a four-year cohort vaccination study to delineate the intertwined effect of age and repeated exposures on the response to influenza vaccination.

## 2. Materials and methods

### 2.1. Subjects and study design

The study was performed from 1996 to 1999 in healthy community-dwelling young and elderly adults living in Hampton Roads, Virginia, United States. The young adults had never received influenza vaccine, and older adults may have been vaccinated previously, but not for at least two years prior to their enrollment in the study. Subjects consented upon enrollment to participation for the duration of the study. The Institutional Review Board of Eastern Virginia Medical School, Norfolk, VA, approved the study protocol and informed consent form. All study participants received an intramuscular injection of the standard dose of trivalent seasonal influenza vaccine (Fluzone®, Sanofi) in each of the study years in which they were enrolled. The health status of all participants of both age groups was very good. All subjects were contacted in the fall of each year to schedule a vaccination visit. Post-vaccination follow-up visits were scheduled in October of each year. Blood samples included 5 cm<sup>3</sup> of serum collected just prior to vaccination, and four weeks post-vaccination.

142 healthy young adults (20–40 years) and 122 healthy older adults (≥65 years) completed the study, i.e., their sequence of vaccinations was uninterrupted during the years, and their pre- and post-vaccination antibody titers were available for all vaccination events and influenza strains involved. The two age groups consisted of four cohorts each, as each year a new cohort of young and elderly adults entered the study. Table 1A shows the numbers of vaccinees per year and cohort. The vaccine strains changed once for

**Table 1A**

Numbers of volunteers, according to age, year of entering the study, and vaccinations within the study. The table also shows the compliance of participants during the study. For example, the cohort of young adults started with 55 young individuals in year 1 (1996), of whom 30 also participated in the second year, 22 in the third year, and 18 in the final year (diagonal).

Age group	Number of vaccinations (NV)	Number of vaccination events				
		1996	1997	1998	1999	All
Young adults	1	55	25	32	24	136
	2		30	14	15	59
	3			22	8	30
	4				18	18
	All	<b>55</b>	<b>55</b>	<b>68</b>	<b>65</b>	<b>243</b>
Elderly adults	1	33	42	32	15	122
	2		27	38	25	90
	3			24	32	56
	4				21	21
	All	<b>33</b>	<b>69</b>	<b>94</b>	<b>93</b>	<b>289</b>

**Table 1B**

Vaccine strain for each year and subtype for both age groups. Vaccine and titration strains were taxonomically identical, except for the A-H1N1 subtype in 1997 (vaccine: A/Beijing/262/95, titration: A/Johannesburg/33/94).

Year	A-H3N2	A-H1N1	B
1996	Nanchang/933/95	Texas/36/91	Harbin/7/94
1997	Nanchang/933/95	Beijing/262/95	Harbin/7/94
1998	Sydney/5/97	Beijing/262/95	Harbin/7/94
1999	Sydney/5/97	Beijing/262/95	Yamanashi/166/98

each of the three (sub)types in the course of the study, as shown in Table 1B.

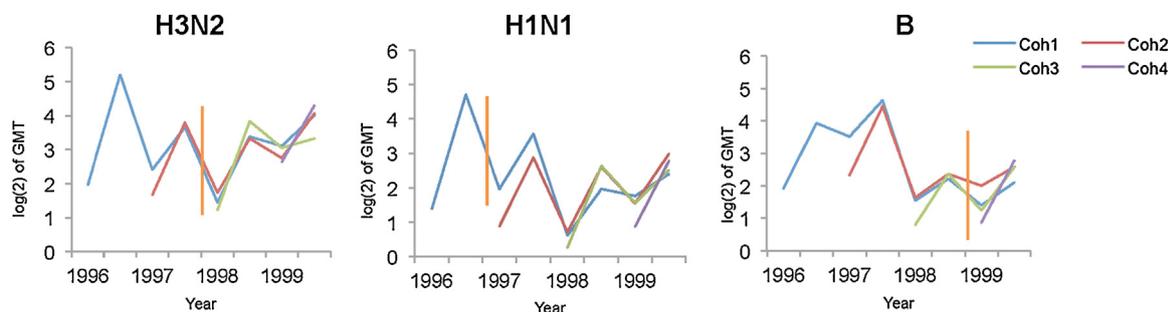
### 2.2. Serum antibody titers

Hemagglutination inhibition assays (HIA) were performed using a single stock source for each of the hemagglutinin antigens (supplied by Centers for Disease Control) and representing the strains of virus contained in the vaccine. HIA was performed as previously described [19] using two-fold dilutions of serum from 1/10 to 1/1024. Titers of <1/10 were calculated as 1/5. Geometric mean titers were calculated using log conversion for each dilution.

### 2.3. Linear regression models

Heteroscedasticity robust ordinary least squares, a type of linear regression model, was used to determine the effects of age and vaccination history on individual post-vaccination titers,  $T_{post}$ , using the heteroscedasticity robust regression (option r) in Stata 12 software. In all calculations pre- and post-vaccination HI titers ( $T_{pre}$  and  $T_{post}$ -values) were log<sub>2</sub>-transformed logarithms of measured titer levels. For an undetectable HI titer (<10, indicating a 'seronegative' person), a value of 5 was imputed. Group log titer means were re-exponentiated and presented as geometric mean titers GMTs throughout the text.

The initial regression model was  $T_{post} = A + B_{pre} * T_{pre}$ , where A is the y-axis intercept,  $T_{pre}$  the pre-vaccination titer, and  $B_{pre}$  the regression coefficient (additional increase in  $T_{post}$  per unit increase of  $T_{pre}$ ). Subsequently, age group (G: young adults = 0, elderly adults = 1) and number of vaccinations within the study (NV: values from 1 to 4) were then added to the regression models as independent variables:  $T_{post} = A + B_{pre} * T_{pre} + B_{agegroup} * G$  and  $T_{post} = A + B_{pre} * T_{pre} + B_{agegroup} * G + B_{NV} * NV$ . The respective regression coefficients were designated  $B_{agegroup}$  and  $B_{NV}$ . All analyses were run for the three virus (sub)types separately, and for all (sub)types combined.



**Fig. 1.** Titers as  $\log(2)$  of GMT by cohort are shown for the pre-vaccination and post-vaccination data per year. There are two data points for each year in the plots, the first for pre-vaccination titer, and the second for the post-vaccination titer. Vertical lines in orange represent changes in vaccine strain. (For interpretation of the references to color in this figure legend, the reader is referred to the web version of this article.)

### 3. Results

#### 3.1. The effects of previous vaccinations and age on response to vaccination

We studied the antibody levels from individuals vaccinated as part of the four year cohort, based on their hemagglutination inhibition (HI) assay titers. Fig. 1 shows the geometric mean titer (GMT) values by vaccine (sub)type, year and cohort. When all titers were grouped without distinction by virus (sub)type, cohort, or trial year, the pre-vaccination GMTs were virtually identical in both age groups (16.4 in young adults versus 15.5 in elderly adults), but the post-vaccination GMTs were a little over twice as high in the young adults (80.3) than in the older adults (38.8), indicating an age effect. The A/H1N1 subtype was associated with a much larger age group ratio (3.1 ratio) than the other two vaccine components (1.6 ratio for A-H3N2 and 1.8 ratio for B).

A first look at the effect of age as an explanatory variable in regression models of the response to vaccination suggests that age is an important determinant of the titer increase induced by vaccination. In the regression model including only pre-vaccination

titers as an independent variable to predict the post-vaccination titers, the predictive power of the model ( $R^2$ ) is 23.7%, but increases by about 7% when age is included (Table 2).

As also reported in other studies [1,4,20], we find that the number of previous vaccinations received also has an effect on the titer increase in response to vaccination. Table 3 shows in its top section that number of vaccinations alone can explain 4% of the variance in post-vaccination titers. In a regression model with pre-vaccination titer as a predictor of post-vaccination titer, the number of previous vaccinations adds 7% to the percentage of variance explained by the model, and in a regression model with pre-vaccination titer and age group as predictors of post-vaccination titer, number of previous vaccinations adds a further 6% to the percentage of variance explained.

#### 3.2. The age difference is limited to a difference between age groups

Age appears to be an important factor determining the post-vaccination titer, based on the pre-vaccination titer and number of vaccinations. We subsequently wanted to test if this age effect

**Table 2**

Regression analysis. Predicting post vaccination titer, by pre vaccination titer (left), including age as an independent variable (right)<sup>\*</sup>.

Independent variables	Estimates of regression model		Independent variables	Estimates of regression model	
Pre-titer	A (y-intercept)	2.433 <sup>***</sup>	Pre-titer	A (y-intercept)	3.535 <sup>***</sup>
	$B_{pre}$	0.600 <sup>***</sup>		$B_{pre}$	0.594 <sup>***</sup>
	$R^2$	0.257		Age	$B_{Age}$
			$R^2$	0.328	

<sup>\*</sup> Notice that the inclusion of age as an independent variable contributes a highly significant effect and increases the  $R^2$ .

<sup>\*\*\*</sup>  $P < 0.001$ .

**Table 3**

Results of regression models with post vaccination titer  $T_{post}$  as the dependent variable in every case<sup>^</sup>.

Variables	Estimates of regression model		Variables	Estimates of regression model		
Pre-titer	Nr vaccinations	3.738 <sup>*</sup>	Pre-titer	A (y-intercept)	2.777 <sup>*</sup>	
	$B_{NV}$	-0.369 <sup>*</sup>		$B_{pre}$	0.642 <sup>*</sup>	
	$R^2$	0.039		Nr vaccinations	$B_{NV}$	-0.502 <sup>*</sup>
Pre-titer	A (y-intercept)	2.433 <sup>†</sup>	Pre-titer	A (y-intercept)	3.248 <sup>*</sup>	
	$B_{pre}$	0.600 <sup>†</sup>		$B_{pre}$	0.692 <sup>*</sup>	
	$R^2$	0.257		Age group	$B_{agegroup}$	-0.911 <sup>*</sup>
Pre-titer	A (y-intercept)	2.993	Pre-titer	A (y-intercept)	3.248 <sup>*</sup>	
	$B_{pre}$	0.592 <sup>†</sup>		Age group	$B_{pre}$	0.692 <sup>*</sup>
	Age group	$B_{agegroup}$		-1.006 <sup>*</sup>	Nr vaccinations	$B_{NV}$
	$R^2$	0.335		$R^2$	0.390	

<sup>^</sup>The regression models on the right include number of previous vaccinations  $B_{NV}$ . Notice that number of previous vaccinations has a statistically significant negative effect and contributes to the  $R^2$  in every case.

<sup>\*</sup>  $P < 0.001$ .

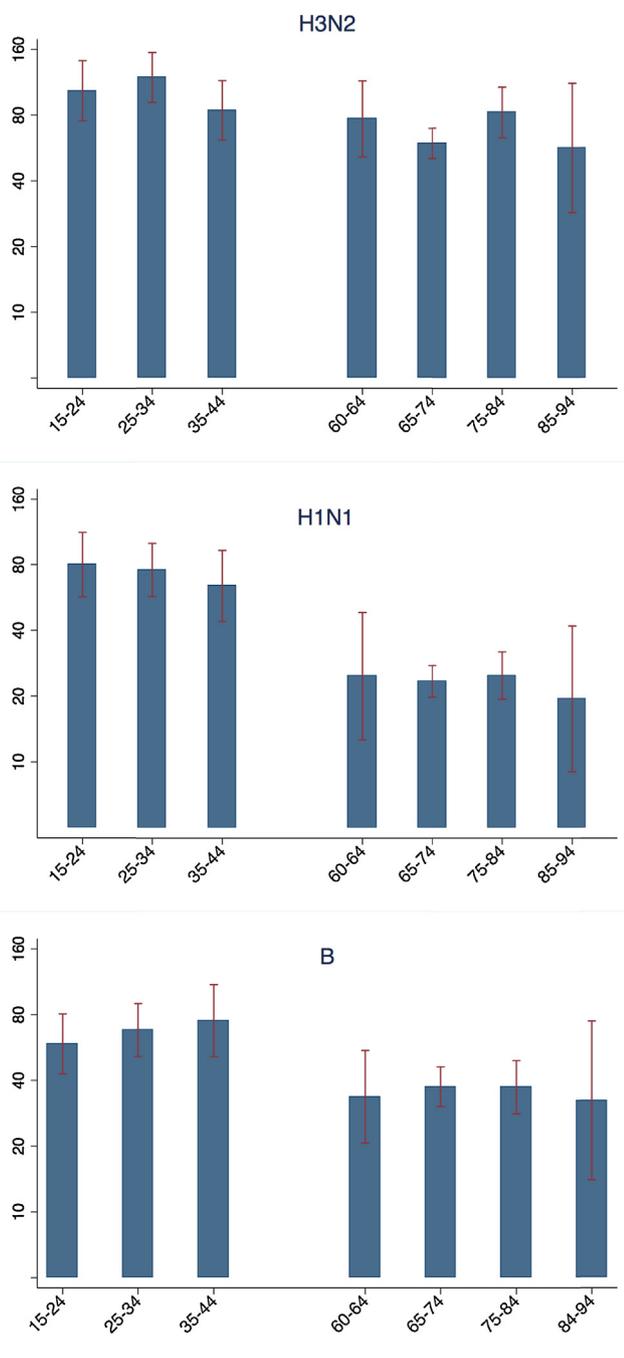


Fig. 2. Post-vaccination antibody levels (GMT) for different age classes are shown.

was also observed within an age group. Fig. 2 shows the post-vaccination GMTs, per (sub)type and decade of age. While there is a clear difference between the two age groups, there is little variation in the response to vaccination within each of the two age groups.

In particular, the very old persons (84–94 years old) reacted similarly to vaccination as the other elderly persons (60–83 years old). It is worth noting however that there are only 13 individuals in the 84–94 grouping, and that these elderly may be fitter than average as this study deliberately only includes ambulatory elderly.

The difference in the effect of age within age group versus between age groups is also seen in the linear regression analyses shown in Table 4. Age is an insignificant factor when regressing post-vaccination titer on pre-vaccination titer and age within either of the two age groups: it explains none of the variation in post-vaccination titer data. However, in the dataset comprising all individuals from both age groups, age can explain 7.6% of the observed post-vaccination titer variation. In summary, we find no effect of age on post-vaccination titer within the age groups, but a marked effect between the two age groups.

### 3.3. The effect of the number of previous vaccinations on the response to vaccination

The previous analyses suggest that age is an important determinant of the response to vaccination, but the difference in response to vaccination is only present between age groups. Since young and elderly groups also differ in their vaccination history and because the results in Table 3 indicated that the number of previous vaccinations influences the response to vaccination, we further investigated the individual responses in relation to the number of administered vaccinations for each person.

Table 5A lists, for each number of previous vaccinations, the results of regressions where the post-vaccination titer was predicted from pre-vaccination titer and an “age group” term (young or old). In this analysis the first regression uses the first vaccination event recorded for all individuals, the second regression uses the second vaccination event for all those individuals that have a second vaccination, and so on until the fourth regression which only includes the first cohort, because this is the only cohort that is vaccinated four times. It can be seen in Table 5A that as subjects in the elderly adult group enter the study in any of the four years and receive their first vaccination, their post-vaccination titer will be significantly lower than the average post-vaccination titer for young adults and elderly combined ( $B_{\text{agegroup}}$  equals  $-1.46$  2-fold HI units). As the vaccination history of the elderly and young groups converges over 2nd, 3rd and 4th vaccinations, the difference caused by age status declines monotonically from  $-1.46$  to  $-0.16$  to  $0.075$ , and age group as a predictor of post-vaccination titer gradually becomes insignificant.

Comparing the regression results of Table 5B with those of Table 5A similarly shows the vanishing age effect: this comparison reveals that age group adds to the explanatory power of the model in the first vaccination event, because the  $R^2$  increases from 0.16 to 0.29 when we include age group in the model predicting the post-vaccination titer from the pre-vaccination titer for the first vaccination. In the second vaccination event the difference only increases from 0.40 to 0.42, and for the third and fourth events the effect of age on the regression model has completely disappeared.

Fig. 3 shows the age group ratios of young adult post-vaccination titers to elderly adult post-vaccination titers by number of

Table 4

Summarized results of the regression post-vaccination titer =  $A + B_{\text{Age}} * \text{Age}$  for each of the two age groups, and for the combined dataset.\*

Independent variables	Estimates of regression model			
		Old adults	Young adults	All
Age	$A$ ( $y$ -intercept)	3.330*	3.909*	4.565*
	$B_{\text{Age}}$	0.006 ( $P=0.490$ )	0.003 ( $P=0.739$ )	$-0.022$ ( $P=0.000$ )
	$R^2$	0.001	0.000	0.077

\* Notice that age only has a predictive effect in the combined dataset and not within either age group.

**Table 5A**  
Regression analysis predicting post-titer, by pre-titer and including age group, by number of vaccinations<sup>†</sup>.

Independent variables	Estimates of regression model	Total	Number of vaccinations (NV)			
			1	2	3	4
Pre-titer	A (y intercept)	2.43 <sup>***</sup>	3.54 <sup>***</sup>	2.36 <sup>***</sup>	1.79 <sup>***</sup>	1.04 <sup>***</sup>
Age group	$B_{pre}$	0.60 <sup>**</sup>	0.65 <sup>***</sup>	0.60 <sup>***</sup>	0.68 <sup>***</sup>	0.84 <sup>***</sup>
	$B_{agegroup}$		−1.46 <sup>***</sup>	−0.43 <sup>**</sup>	−0.16	0.075
	$R^2$		0.29	0.42	0.46	0.73

<sup>†</sup> Notice that the significance for age-group, and the difference in  $R^2$  between Table 4A and Table 4B decline with number of vaccinations.

\*\*  $P < 0.01$ .

\*\*\*  $P < 0.001$ .

**Table 5B**  
Regression analysis predicting post-titer, by pre-titer and excluding age group, by number of vaccinations.

Independent variables	Estimates of regression model	Total	Number of vaccinations (NV)			
			1	2	3	4
Pre-titer	A (y intercept)	2.43 <sup>***</sup>	3.00 <sup>***</sup>	2.03 <sup>***</sup>	1.69 <sup>***</sup>	1.09 <sup>***</sup>
	$B_{pre}$	0.60 <sup>**</sup>	0.56 <sup>***</sup>	0.64 <sup>**</sup>	0.69 <sup>**</sup>	0.84 <sup>**</sup>
	$R^2$	0.257	0.16	0.40	0.46	0.73

\*\*  $P < 0.01$ .

\*\*\*  $P < 0.001$ .

vaccinations, for all three virus (sub)types. Because these age-group titer ratios may be biased by pre-vaccination level, as mean pre-vaccination titers generally rise with increasing number of vaccinations, we adjusted post-vaccination titers to correct for pre-vaccination titer levels using the  $B_{pre}$  estimates of the regression models [21]. The resulting adjusted age group titer ratios shown in Fig. 3 (red) become more precise, as seen by narrower confidence intervals, than the unadjusted age group post-vaccination titer ratios. The age effect is observed for the first vaccination, yet declines after further vaccinations, though not monotonically, for all (sub)types. For A/H3N2, it takes only two vaccinations for the two age groups' serological responses to converge, and for A/H1N1, four vaccinations. The same pattern of declining difference between age groups with successive vaccinations is evident in a graph of the difference between the intercepts of the regression model post-vaccination titer =  $A + B * \text{pre-vaccination titer}$ , per age group and per cohort for the two age groups (Supplementary Fig. S1).

#### 3.4. The two variables age and number of previous vaccinations are confounded

The younger adults in our study had never been vaccinated against influenza; the elderly adults may have received multiple influenza vaccinations up to two years before the study. The two age groups start with a clear difference in post-vaccination titer level after the first vaccination within the study, which may be attributable to this difference in vaccination history, rather than age: the age term will proxy some of the negative effect attributable to previous vaccination history. Since there is no parameter that independently captures history, the exposure history difference will be attributed to the only parameter that distinguishes the two groups, namely age. Thus, we next investigated whether the observed age difference is likely to be caused by age *per se* (intrinsic immunosenescence), or by the extrinsic effect of these previous vaccinations.

As we have seen in Table 5A and Fig. 3, once individuals received two or more vaccinations within the study, the age effect disappeared. We therefore tested if the regression model could be used to estimate the unknown number of previous vaccinations in the elderly group. To this end, we varied the regression model as shown in Table 2 (right side) by replacing the number of vaccinations (NV) with a new variable, the augmented number of vaccinations ( $NV_{aug}$ ). For young adults,  $NV_{aug}$  was the same as NV because these

people had not been vaccinated prior to the study; for elderly adults however,  $NV_{aug}$  was set to  $NV + 2$  to account for vaccinations received prior to the study.

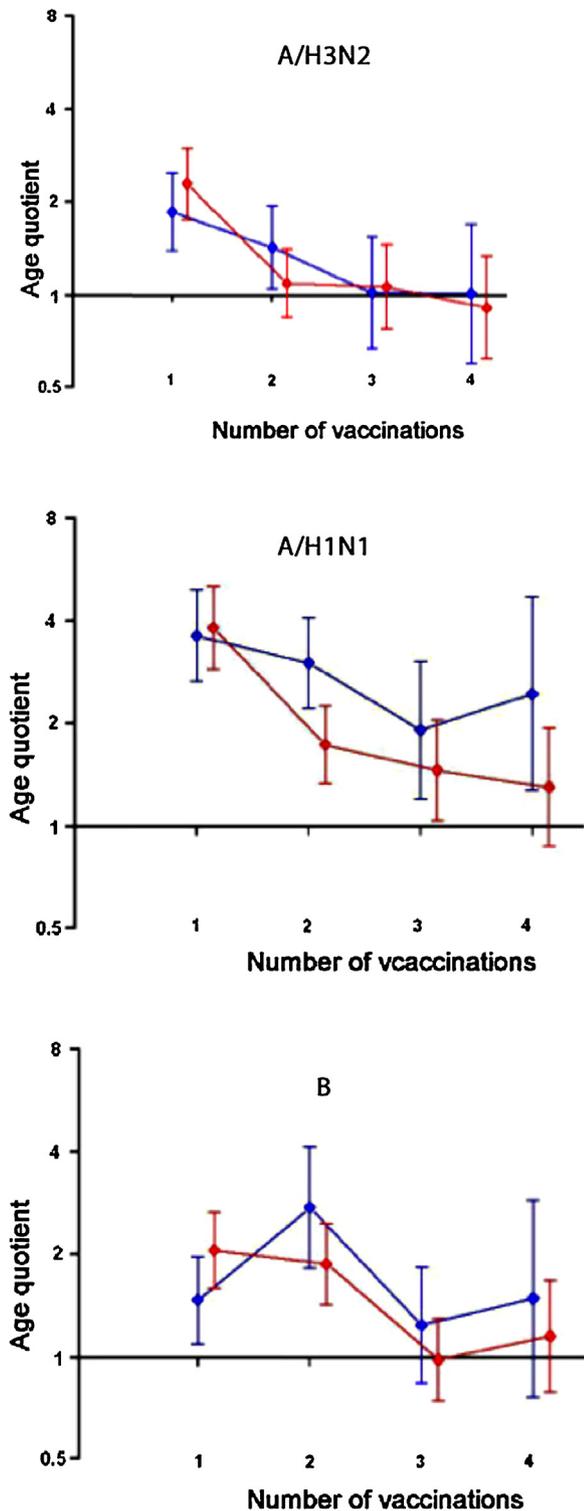
Table 6 shows the regression results using the augmented number of previous vaccinations, and can be compared with Table 2 (right side), the same regression model using the documented, non-augmented number of previous vaccinations. The estimates for the y-intercept, the slopes for pre-vaccination titers and (augmented) number of vaccinations, and  $R^2$  were exactly the same in Table 6 and Table 2. However, the age group coefficient changed dramatically: it was highly negative in the non-augmented model ( $-0.911$ ,  $P < 0.001$ ), but close to zero and insignificant in the augmented model ( $-0.014$ ,  $P = 0.893$ ).

Comparison of these tables therefore demonstrates that age loses significance as a predictor of the response to vaccination, whilst the  $R^2$  increases when using the augmented number of previous vaccinations. A number of previous vaccinations augmented by anywhere between 1.3 and 2.3 has the same explanatory effect as a combination of the documented number of previous vaccinations and age group (Supplementary Table S1 and Supplementary Fig. S2). These results show that the influence of age *per se* disappears when supposing two previous vaccination events in the elderly before entering the study, and that age as a predictor of the antibody titer in response to vaccination is thus equivalent to vaccination history.

## 4. Discussion

When antibody response after a single vaccination is studied in groups of young and elderly adults, usually a clear difference with a larger response to vaccination in young adults is observed. Importantly, in our study as well as many others, elderly participants had already been vaccinated against influenza prior to the study, to various degrees, and young participants usually had not. At least since the 1980s, vaccination of the entire elderly population is a common target in many developed and developing countries [22], where such vaccination studies are performed. Thus, it is difficult to enroll representative, previously unvaccinated, groups of elderly persons. As a result, any effect of age on the immune response is intrinsically correlated and necessarily closely linked to vaccination history.

The present cohort study where individuals were repeatedly vaccinated allowed us to analyze the effect of repeated vaccination separately from the effect of age in the same cohort study.



**Fig. 3.** Young adults' post-vaccination titers as multiples of elderly adults' post-vaccination titers according to virus (sub)type and number of vaccinations, raw (blue) and adjusted for pre-vaccination log titer (red). (For interpretation of the references to color in this figure legend, the reader is referred to the web version of this article.)

Using these data, we inferred the effect of repeated exposure on the response to vaccination, and showed that only two vaccinations prior to the study can account for the entire observed difference between the young and elderly age groups.

A limitation of this study is the lack of reliable infection history preceding this experiment, which is expected to differ between

**Table 6**

Regression analysis on post-vaccination log titer, by pre-vaccination log titer, augmented number of vaccinations, and age group<sup>a</sup>.

Variables included	Estimates of regression model	
Pre-titer	A (y-intercept)	3.696 <sup>*</sup>
	$B_{pre}$	0.630 <sup>*</sup>
Augmented number of vaccinations	$B_{Nvaug}$	-0.448 <sup>*</sup>
Age group	$B_{age}$	-0.014 ( $P=0.893$ )
	$R^2$	0.390

Notice that the coefficient on age group becomes insignificant when the augmented number of previous vaccinations is included.

<sup>\*</sup>  $P < 0.001$ .

the age groups, and the lack of data on post-vaccination infection. Serum antibody titers are a standard accepted [23] though indirect and necessarily imperfect measure of vaccine efficacy.

When an analysis leaves out an important explanatory factor, an omitted variable bias is created. Regression models compensate for the omitted variable by over- or underestimating included explanatory variables that are correlated with the omitted variable. In the present case, the lack of a variable for number of previous vaccinations is compensated for by a distortion in the correlated age variable. As a result, being elderly appears to be a determinant of the antibody response to vaccination, whereas at least part of actual determinant could be vaccination history. It should also be noted that any source of difference in exposure history, including prior infections, may lead to a similar situation, in which age and exposure history are confounded.

Our results are in line with previous studies, describing a weakened serological response to vaccination in elderly adults, i.e. an immunosenescent effect. However, we put forward the explanation that, in addition or instead of resulting from intrinsic ageing of the immune system, this immunosenescence effect could be enhanced, or more parsimoniously explained by an extrinsic driver: the previous vaccination and infection history of the elderly group, related to the yearly influenza vaccination campaigns. Of course, we do not deny the general existence of immunosenescence. It is well established that ageing modulates many immune functions. However, given the well-documented influence of repeated vaccination [20,24,25], it seems that comparative serological data sets containing an elderly group with an undocumented number of previous vaccinations prior to enrollment, and a vaccination-free young group, is an inaccurate way to explore immunosenescence. Indeed, a simple age effect would not explain how the titer difference between age groups changes for subsequent vaccinations. In our view, carefully controlled studies are needed to establish the effect of immunosenescence in this context, studies where either vaccination and infection history among the two age groups is comparable, or sufficient information on the history is known such that differences can be controlled for.

**Acknowledgments**

Supported by the NIH First Award R29AG11876 (PI: D. Powers), NIH Director's Pioneer Award, program grant P0050/2008 from the Human Frontier Science Program, European Union FP7 program EMPERIE (22349), National Institute of Allergy and Infectious Diseases–NIH Centers of Excellence for Influenza Research and Surveillance contracts HHSN266200700010C and HHSN272201400008C. This work was supported by the award of a Fellowship in Biomedical Informatics from the Medical Research Council (UK) and a Junior Research Fellowship from Homerton College Cambridge to JMF.

## Appendix A. Supplementary data

Supplementary data associated with this article can be found, in the online version, at <http://dx.doi.org/10.1016/j.vaccine.2015.11.058>.

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