

**Title: Antibody landscapes after influenza virus infection or vaccination**

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37 **Abstract:** We introduce the antibody landscape, a method for the quantitative analysis of  
38 antibody-mediated immunity to antigenically variable pathogens, achieved by accounting for  
39 antigenic variation among pathogen strains. We generated antibody landscapes to study immune  
40 profiles covering 43 years of influenza A/H3N2 virus evolution for 69 individuals monitored for  
41 infection over six years and for 225 individuals pre- and post-vaccination. On infection and  
42 vaccination titers increased broadly, including previously encountered viruses far beyond the  
43 extent of cross-reactivity observed after a primary infection. We explored implications for  
44 vaccination, and found that use of an antigenically advanced virus had the dual benefit of  
45 inducing antibodies against both advanced and previous antigenic clusters. These results indicate  
46 that pre-emptive vaccine updates may improve influenza vaccine efficacy in previously-exposed  
47 individuals.

48

49 **One Sentence Summary:** Influenza virus infection or vaccination produces an antigenically  
50 broad increase of titers that can be exploited to improve vaccine design.

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52

53 **Main Text:**

54 Much of the global burden of infectious disease today is caused by antigenically variable  
55 pathogens, which escape immunity induced by prior infection or vaccination by changing the  
56 molecular structure recognized by antibodies. Human influenza viruses are notorious for their  
57 capacity to evolve and evade the adaptive immune response. This evolution has been progressive  
58 and step-wise (fig. S1)(1), with antigenically similar viruses circulating for a few years before  
59 strains with related but novel antigenic characteristics replace them (2). As a result, vaccine  
60 strain updates, based on analyses of circulating viruses, are necessary to maintain vaccine  
61 effectiveness.

62

63 The current vaccine strain selection strategy is to choose a virus that is antigenically  
64 representative of circulating viruses, mostly determined by testing a global selection of virus  
65 isolates against a panel of ferret antisera using the hemagglutination inhibition (HI) assay (3).  
66 The ferrets used in such studies are influenza-naïve prior to inoculation, and each antiserum has  
67 been raised by infection with only a single virus. Such post-inoculation ferret antisera provide  
68 well-understood data for the characterization of antigenic differences between influenza viruses  
69 (2, 4). However, this strategy does not account for the influence of prior immunity on the  
70 response induced by the vaccine when administered to humans.

71

72 The direct analysis of human serological data presents an opportunity to assess and understand  
73 immune responses in the context of differing background immunity and to use this information  
74 as the basis for improved vaccine strain selection and evaluation. Indeed, such data are used in

75 the vaccine strain selection process. Unfortunately, immunological patterns in human serological  
76 data are difficult to interpret because of complex, and usually unknown, exposure histories and  
77 the confounding factor of cross-reactivity due to antigenic relationships among strains. As a  
78 result, in-depth analyses of serological data have been difficult and, despite excellent cross-  
79 sectional seroepidemiology (5), our understanding of the typical characteristics of the human  
80 serological response to infection and vaccination has remained limited.

81

82 Results from the original, and seminal, studies on the antibody-mediated immune response to  
83 influenza virus infection and vaccination in humans (6-9) have often been interpreted as  
84 “original antigenic sin” — a hypothesis that proposes an anamnestic reinforcement of the level of  
85 antibody to the strain that first infected the individual that dampens the serologic response to the  
86 current virus (9-11). This definition is, however, far from concrete and the historical literature on  
87 the effect of immune memory on the generation of responses to variant antigens has been  
88 particularly equivocal.

89

90 To increase our ability to quantitatively study human serological data of antigenically variable  
91 pathogens, we present a methodology that enables detailed analyses and visualization of complex  
92 serological data by plotting antibody-mediated immunity as a function of the antigenic  
93 relationships among viruses. To achieve this, we first used antigenic cartography (2) to  
94 determine the antigenic relationships among a selection of 81 viruses spanning 43 years of  
95 influenza A/H3N2 evolution, using HI titrations of first-infection ferret sera (Fig. 1A, fig. S2,  
96 Tables S1 and S8). Human serum samples were then titrated against the same viruses and their  
97 HI titers plotted in an extra dimension added to the antigenic map (Fig. 1B).

98

99 We found that HI titers of a given serum are related for antigenically similar viruses (fig. S3),  
100 and thus a representative smooth surface could be fitted through these HI titers. The resulting  
101 antibody landscape represents an immune profile for each serum with elevations corresponding  
102 to regions in the antigenic map with higher antibody levels (figs. S4-S5, S13). Since the  
103 landscape at any given point is a function of surrounding data points, antibody levels can be  
104 inferred for viruses not included in the titration set. For antibody landscapes of influenza  
105 A/H3N2 based on the HI assay, we found that the landscape predicted omitted HI titers with a  
106 root-mean-square error of 1.3  $\log_2$ -units, compared to an estimated error arising from HI assay  
107 repeatability alone of 0.9 (Table S10, figs. S6-S11, S14).

108

109 To aid the visual comparison of multiple landscapes, we used a path on the antigenic map that  
110 passes through each antigenic cluster in chronological order (Fig. 1C). The corresponding values  
111 of the landscape along this summary path were used to represent the three-dimensional landscape  
112 in two dimensions (Fig. 1D and fig. S12).

113

114 We used this methodology to study serological data we generated from samples taken annually  
115 between 2007 and 2012 from unvaccinated individuals in the Ha Nam household cohort study in  
116 Vietnam (12). More than 10,000 HI titrations were performed to construct a total of 324  
117 landscapes for 69 individuals born between 1917 and 2005, allowing us to assess the serological  
118 changes over time (Fig. 2, Tables S3, S4, fig. S15). Titers were highest for influenza viruses that  
119 circulated when an individual was approximately 6 years old (figs. S42-S43), corresponding with

120 the time-frame of first infection (13). Antibody levels against newly circulating viruses tended to  
121 be lower than against strains circulating earlier in an individual's lifetime, as reported previously  
122 (5,7-9,11). In addition, previous results found some cross-reactivity to strains that circulated  
123 before an individual's birth (5, 7-9,14) and based on the extent of detectable titers to viruses in  
124 circulation only before an individual's birth, we quantified this antibody cross-reactivity to be 0-  
125 2 antigenic clusters (Table S11). There was substantial heterogeneity among the antibody  
126 landscapes of different individuals; however, each individual's landscape shape was typically  
127 stable from one year to the next and had distinctive individual features (within-person  $r=0.86$   
128 (standard deviation $\pm 0.22$ ), between-person  $r=0.28\pm 0.21$ , figs. S16-S20).

129  
130 Infection with A/H3N2 resulted in a strikingly broad antibody response (Fig. 2 and figs. S21-  
131 S22) that was typically governed by the extent of the pre-exposure antibody landscape (fig. S45).  
132 This antibody response far exceeded the extent of cross-reactivity typically produced in the  
133 response following primary exposure with one of the circulating viruses (Fig S44, S47). For  
134 example, an individual born in 1970, infected in 2009 (Fig. 2, third row), had a substantial long-  
135 distance response back to the Hong Kong 1968 (HK68) antigenic cluster and all clusters in  
136 between, even though these older viruses had not circulated for decades. To illustrate the  
137 substantial breadth of this back-boost, there have been 13 antigenic cluster transitions from  
138 HK68 until Perth 2009 (PE09), each approximately 4.5 antigenic units (corresponding to a 24-  
139 fold dilution of antiserum in the HI assay). These antigenic changes have necessitated over 20  
140 vaccine strain updates, and are the result of changes in 69 of the 346 amino acid positions in the  
141 HA1 domain of the hemagglutinin gene between HK68 and the PE09 vaccine strain, including  
142 substitutions in all of the seven key antigenic positions identified by Koel *et al.* (15).

143

144 Because of the range of this response, and its dependence on the pre-exposure antibody  
145 landscape, we call it a “back-boost”. The magnitude of back-boost response declined with  
146 antigenic distance from the likely infecting virus (fig. S46). Although the response to older  
147 viruses was substantial, titer increases were largest for viruses from the contemporary antigenic  
148 cluster, in contrast to a common interpretation of the original antigenic sin hypothesis (fig. S47).  
149 Polymerase chain reaction confirmed infections with influenza B, A/H1N1 and A/H1N1(pdm09)  
150 often caused negligible changes in the A/H3N2 antibody landscape (fig. S23), indicating that the  
151 back-boost is type and subtype-specific.

152

153 Typically, the broad initial response was followed by a period of titer decay during which  
154 antibody titers stabilized to form an altered antibody landscape over the course of approximately  
155 one year (fig. S24). Comparison of the antibody landscapes of 2007 and 2012 (Fig. 2) shows that  
156 the antigenic region for which increased titers were maintained long-term was substantially  
157 narrower than that of the initial response to infection. This long-term persistence of increased  
158 antibody titers was more specific to the antigenic region of the likely infecting strain, but still  
159 spanned multiple antigenic clusters (fig. S46).

160

161 Next, we investigated whether the back-boost observed following infection could be used to  
162 improve vaccine effectiveness. In the vaccine strain selection process, it is sometimes unclear  
163 whether currently circulating strains or antigenically novel strains are most likely to predominate  
164 in the next influenza season. The resulting dilemma is whether it is more beneficial to leave the

165 vaccine strain unchanged, or to pre-emptively update the vaccine to match a novel strain, without  
166 certainty over which antigenic cluster of viruses will indeed circulate.

167

168 It would take a large, prospective, multi-year clinical trial comparing the two vaccination  
169 approaches to answer these questions definitively. However, we were able to retrospectively test  
170 the approach with the sera of 225 human vaccinees from two annual influenza vaccine re-  
171 registration studies, by identifying an antigenic cluster transition for which there was little  
172 circulation of the new cluster before a novel vaccine strain was first tested. Both groups had  
173 therefore received antigenically different vaccines, and yet there was no significant difference in  
174 the average pre-vaccination antibody landscapes of the two studies (figs. S30-S31). Individuals  
175 in the first study (n=102, Table S6), performed in 1997, received the A/Nanchang/933/95  
176 vaccine from the Wuhan 95 (WU95) antigenic cluster to which there had been some prior  
177 exposure, whereas individuals in the 1998 study (n=123, Table S7) received the A/Sydney/5/97  
178 vaccine from the antigenically advanced Sydney 97 (SY97) cluster to which there was  
179 substantially less pre-vaccination immunity – thus mimicking a pre-emptive update.

180

181 Individual antibody landscapes were constructed from serum samples taken pre-vaccination and  
182 four weeks post-vaccination (figs. S25-26, Table S5) and combined to give overall pre-  
183 vaccination and post-vaccination antibody landscapes (Fig. 3A,B). As expected following a  
184 vaccine update, average vaccination responses were significantly greater against later antigenic  
185 clusters following vaccination with the antigenically advanced SY97 strain (figs. S32-S35). The  
186 back-boost following infection was also observed for the vaccination studies, and interestingly,  
187 the magnitude and breadth of the response to infection and vaccination were comparable (figs.

188 S27-S29). Indeed, the back-boost in the SY97-vaccine study resulted in a slightly larger response  
189 to WU95 viruses than the response in the WU95-vaccine study (Fig. 3C). These findings also  
190 held when studying only elderly individuals (fig. S36), and individuals with a low pre-  
191 vaccination titer against WU95, typically considered the most susceptible (fig S37-S38) (16). We  
192 further tested a subset of vaccination sera with a neutralization assay, and these data support the  
193 results from the HI assay (figs. S40-S41). Despite differences in pre-vaccination landscapes, a  
194 second study of the WI05-PE09 cluster transition also demonstrated a similar back-boost upon  
195 vaccination (fig. S39).

196  
197 The mechanism behind the broad back-boost is currently unknown, but we considered several  
198 hypotheses (1). In summary, rather than resulting from the production of novel antibodies with  
199 extensive cross-reactivity, the back-boost appears most consistent with memory-cell stimulation  
200 and antibody recall. This pattern of recall is consistent with raw data from the mid-20<sup>th</sup> century  
201 studies on the response to infection or vaccination where studies on antigenically different  
202 A/H1N1 strains also show a broad sub-type specific back-boost (6, 8-9). However, this  
203 phenomenon was never quantified and put in relation to the antigenic difference among the  
204 viruses.

205  
206 Whether the original antigenic sin hypothesis refers to higher pre-exposure antibody titers, or  
207 also to a higher response to the first infecting virus is unclear, and both interpretations have been  
208 used over the past 60 years (17). We found no evidence for a predisposition in the antibody  
209 response towards the likely first infecting strain, and instead, we demonstrate that the increase in  
210 antibody titers is greatest to the most recently encountered strain. We do, however, corroborate

211 the finding that pre-exposure antibody reactivity tends to be highest against strains encountered  
212 earlier in life (fig. S37) (5, 7-9, 11). The presence of higher pre-exposure static titers, but not  
213 higher dynamic responses, to the first infecting strain may explain seemingly contradictory  
214 reports whereby cross-sectional studies have tended to describe a serological bias supportive of  
215 the original antigenic sin dogma (5, 11) while investigations into actual responses upon exposure  
216 frequently oppose it (17,18).

217

218 These findings also shed light on the growth of the serological immunity over time. Although  
219 responses were often present against the oldest strains, these long-distance back-boosts were  
220 typically not maintained beyond a year (Fig 2. right panel, fig S24). This is evidence against the  
221 hypothesis of long-term and progressive “reinforcement” of antibody titers against earlier viruses  
222 upon exposure to each subsequent antigenic variant over time. Instead, the pattern of higher  
223 static titers against antigenic clusters encountered early in life may also be explained if the  
224 immune response to primary exposure is larger than the responses to subsequent exposures (Fig  
225 S48).

226

227 As others have speculated, it is plausible that the decreased antibody responses to subsequent  
228 exposures may be a result of “antigen trapping”, a hypothesis according to which binding of  
229 antigen by pre-existing cross-reactive antibodies and memory-cells decreases the antigenic load  
230 available for priming naïve B-cells and leads to a diminished novel response (5, 7, 10, 19-20).

231 This would also explain why the closest antigenic match between the vaccine strain and the  
232 circulating strains does not necessarily generate the best antibody response against the  
233 corresponding cluster: the mismatch of an antigenically advanced strain is compensated for by a

234 greater novel response, as a result of reduced antigen trapping (21). The extent of interference by  
235 antigen trapping on the novel antibody response depends on the degree of antigenic relatedness  
236 and prior immunity (22). Note, when individuals have no prior immunity to a subtype, such as  
237 young children, or in a pandemic, the best vaccine is likely the closest antigenic match as there  
238 will be no prior immunity to avoid and no back-boost to exploit.

239

240 These findings highlight potentially important differences between the two types of vaccine  
241 mismatch in populations with prior immunity. Following a mismatch due to a delayed vaccine  
242 update (in which the vaccine strain, selected 10-14 months before the season in which it is used,  
243 lags behind influenza virus evolution), neither pre-existing nor newly induced antibodies provide  
244 immunity against the novel strains. Consequently, such vaccines have poor effectiveness in this  
245 mismatch situation (23-26). However, if there were a vaccine mismatch due to an incorrectly  
246 timed, pre-emptive antigenic update of the vaccine, then the data from our retrospective  
247 surrogate study indicate that the extensive back-boost would still induce equivalent titers against  
248 previous antigenic strains. Such vaccines would have the dual advantage of being effective  
249 against the antigenically novel viruses to which they were targeted while remaining effective, or  
250 being even more effective, for contemporary viruses if they continued to circulate.

251

252 Our results underscore the importance of accounting for antigenic variation to better understand  
253 multi-exposure sera, and provide a methodology for the direct visualization of otherwise  
254 complex serological patterns, allowing basic insights into the breadth of the adaptive humoral  
255 immune response to influenza and other antigenically variable pathogens. Antibody landscapes  
256 will be useful for the evaluation of evolutionary selection pressures (fig. S49) and the evaluation

257 of different vaccination techniques, including the effect of adjuvants, vaccine composition, dose  
258 sparing, and the durability, breadth and magnitude of responses to universal vaccines. Our results  
259 indicate that pre-emptive vaccine updates may substantially improve influenza vaccine  
260 effectiveness in previously-exposed individuals. Prospective clinical trials will further test the  
261 breadth and longevity of the immunological response and protection provided by antigenically  
262 advanced vaccine strains.

263

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366 Sequences of the influenza viruses used in this study will be made available on Genbank.  
367

368 **Figure legends:**

369

370 **Fig. 1** Creating an antibody landscape. **(A)** Antigenic map of A/H3N2 showing virus strains  
371 color-coded by antigenic cluster. Both axes represent antigenic distance, the spacing between  
372 grid lines is 1 antigenic unit, corresponding to a twofold dilution of antiserum in the HI assay.  
373 Two units correspond to fourfold dilution, three units to eightfold dilution, and so on (2). The  
374 gray line shows a path through the antigenic clusters in chronological order calculated by fitting  
375 a smoothing spline (*l*). **(B)** An additional dimension indicates the measured antibody titers as  
376 vertical impulses and a smooth surface is fitted using locally weighted multiple linear regression  
377 to create the antibody landscape within the convex hull bounded by the viruses titrated (RMSE  
378 of fit = 1.23 HI log<sub>2</sub>-units). **(C)** The height of the landscape along the path in (A) shows a slice  
379 through the landscape (*l*). **(D)** The height of the landscape along the antigenic summary path is  
380 plotted to create a rotation-independent 2D summary visualization of the landscape. Titrated  
381 virus strains are shown in their corresponding positions along the x-axis, symbol radius is  
382 inversely proportional to antigenic distance from the path, symbol color indicates antigenic  
383 cluster. The scale bar indicates 2 antigen units; each antigenic unit is a 2-fold dilution in the HI  
384 assay.

385

386 **Fig. 2.** Antibody landscapes from 2007-2012 for six individuals. The black line represents the  
387 landscape height for each position on the antigenic summary path through the antigenic clusters  
388 from Fig. 1A. The first sample taken after a confirmed A/H3N2 influenza virus infection is  
389 marked with a red box, and the red number gives the days from the start of influenza-like illness  
390 to serum collection. The red shading indicates increases, and beige decreases, compared to the

391 previous year. The blue-shaded area indicates antigenic clusters that circulated during an  
392 individual's lifespan until sample collection (Table S9). Dots along the x-axis indicate the subset  
393 of 30 viruses used to generate these landscapes - contemporary strains likely causing the  
394 infection are indicated with a red horizontal bar (Table S2). The rightmost column shows the  
395 difference between the landscape in 2012 compared to 2007. The scale bar indicates 2 antigenic  
396 units.

397

398 **Fig. 3.** Comparison of two different vaccines. (A) The mean pre-vaccination landscape (gray)  
399 and landscape after vaccination with A/Sydney/5/97 (blue) in the 1998 study (123 individuals),  
400 or (B) with A/Nanchang/933/95 (green) in the 1997 study (102 individuals) for each position on  
401 the antigenic summary path. Dots along the x-axis indicate the subset of 70 viruses used to  
402 generate these landscapes. The vertical dotted lines indicate the position of the SY97 (blue) and  
403 WU95 (green) wild type vaccine viruses. (C) Comparison of titer increase after vaccination with  
404 A/Nanchang/933/95 or A/Sydney/5/97 for each position along the antigenic summary path.  
405 Above the horizontal midpoint indicates higher response to the A/Sydney/5/97 vaccine, below to  
406 the A/Nanchang/933/95 vaccine. Data were calculated from the average titer increase between  
407 each individual's paired post-vaccination and pre-vaccination titers, with 95% (dark gray) and  
408 99% (light gray) t-test based confidence intervals. The scale bar indicates 2 antigenic units.

409

### 410 **Supplementary Materials:**

411 Materials and Methods

412 Figures S1-S49

413 Tables S1-S11

414 References (26-35)

415