

When to vaccinate for seasonal influenza? Check the peak forecast

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1 When to vaccinate for seasonal influenza?
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20 **Abstract**

21 **Background:** Seasonal influenza infects 5-20% of people every year in the United
22 States, resulting in hospitalizations, deaths, and adverse economic impacts. To
23 mitigate these impacts, influenza vaccines are developed and distributed annu-
24 ally; however, growing evidence suggests that vaccine effectiveness (VE) wanes
25 over the course of a flu season. Delaying influenza vaccination for older adults
26 has attracted attention as a potential public health strategy. However, given the
27 uncertainties in seasonal peak, vaccine effectiveness, and waning rates, postpon-
28 ing vaccination could also lead to increased morbidity, motivating an evaluation
29 of a range of potential scenarios.

30 **Methods:** We systematically investigated a broad range of vaccination start
31 times for five age groups under six combinations of initial effectiveness and waning

32 rates, based on influenza cases and vaccine uptake data from 10 influenza seasons.
33 We defined the most favorable vaccination schedule as the one that resulted in
34 the greatest reduction in disease burden.

35 **Results:** In scenarios with fast waning, all age groups benefit from delaying vac-
36 cination regardless of initial VE and peak timing. In scenarios with slower waning,
37 results are mixed. For the ≥ 65 group, high initial VE and slow waning suggests
38 that in early-peaking seasons, early vaccination most effectively reduces disease
39 burden, while in late-peaking seasons delaying vaccination is most effective. For
40 the ≥ 65 group in medium and low initial VE, and slow waning scenarios, delay-
41 ing vaccination appears to prevent the greatest number of cases, regardless of
42 whether the season peaks early or late.

43 **Conclusion:** The most favorable vaccination schedule is sensitive to changes
44 in initial VE, waning rate, and peak timing. Given estimates of these quanti-
45 ties from statistical and immunological models and observations, our methods
46 can inform vaccination recommendations in order to most effectively reduce the
47 annual disease burden caused by seasonal influenza. Specifically, accurate peak
48 timing forecasts for the upcoming season have the potential to guide decisions
49 on when to vaccinate.

50 **Keywords:** influenza, vaccine, vaccine effectiveness (VE), VE waning, influenza
51 forecasting

52 1 Background

53 Seasonal influenza causes substantial health and economic burden in the United States
54 each year. The Centers for Disease Control and Prevention (CDC) estimates that
55 from 2010-2020, annual influenza epidemics have resulted in 9-41 million illnesses,
56 140,000–710,000 hospitalizations, and 12,000–52,000 deaths [1]. Seasonal vaccination
57 has been the most effective strategy to reduce influenza transmission and mitigate
58 its potential impacts [2]. The U.S. Advisory Committee on Immunization Practices
59 (ACIP) publishes recommendations regarding the use of influenza vaccine every flu
60 season. The ACIP has recommended influenza vaccination to be offered to everyone
61 over 6 months of age by the end of October [3]. For certain populations, flu vaccines
62 begin to be administered as early as July, when the vaccines first become available
63 [4, 5]. As a result, approximately 30% of all adults in the US are vaccinated by the
64 end of October [6].

65 Several recent studies have presented evidence suggesting that intra-seasonal wan-
66 ing of influenza vaccine effectiveness (VE) exists and is an epidemiologically important
67 phenomenon [7–9]. VE waning is defined as the reduction of vaccine-induced immu-
68 nity during an influenza season while the virus is still actively circulating. Rambhia
69 and Rambhia [10] and Roy and MacDougall [11] summarized a series of recent stud-
70 ies regarding intra-seasonal waning of influenza VE [12–25]. These studies found some
71 degree of VE waning, although the estimated degree varied substantially. These results
72 suggest that early influenza vaccinations (e.g., during summer) may be suboptimal
73 since protection may be diminished during peak months of influenza activity [10].

74 Postponing influenza vaccination has attracted attention in the flu research com-
75 munity as a potential public health strategy to counteract VE waning. A few recent
76 studies have discussed the risks and benefits of delaying influenza vaccination [26–
77 28]. Using linear VE waning functions for each season from 2007 to 2016 with 2009
78 excluded, Costantino et al. [26] studied the impact of the influenza vaccination timing
79 change and reduced vaccine coverage on health outcomes for two age groups, < 65 and
80 ≥ 65 , in Australia. They found that delaying vaccination could have a net negative
81 impact, if it results in missed vaccination. However, it is unclear how sensitive their
82 results are to different VE waning functions. In contrast, Newall et al. [27] investigated
83 the impact of delaying vaccination using two VE waning scenarios (both VE waning
84 functions start at 50%, one wanes over 26 weeks, and the other wanes over 52 weeks)
85 in older adults in the U.S. They found net benefits of delaying vaccination based
86 on the 2010/11 to 2015/16 seasons, even if the vaccine coverage is lowered in some
87 cases. Ferdinands et al. [28] selected a single influenza season, the 2012/2013 season,
88 and evaluated the impact of potential vaccination timing change for older adults in
89 the U.S. They showed that delaying vaccination until October could lead to negative
90 outcomes (i.e., more influenza hospitalizations), if that strategy resulted in a $> 14\%$
91 reduction in the total number of vaccinated older adults compared with what would

92 have otherwise been expected during that period (i.e., prior to October). Limitations
93 of these studies indicated a need for further investigations: more complete age strati-
94 fication, analyses spanning multiple seasons, and a broader range of VE assumptions
95 under more realistic scenarios.

96 To address this need, we investigated a broad range of influenza vaccination sched-
97 ules for five age groups, assessed how the recommendations change under different
98 seasonal VE and waning scenarios, and quantified how influenza activity timing (early
99 or late peaking season) impacts the schedule. Using empirical data and five age cohorts,
100 we modeled the first week of influenza vaccination to begin such that the largest per-
101 centage of disease burden (i.e., influenza-like illness cases for people < 65 and influenza
102 hospitalizations for people ≥ 65) can be prevented in the US for each age group. Then,
103 we used these results to assess the possibility of an age-tiered vaccination schedule.
104 Finally, we explored the impact of an early or late peaking season on various schedules
105 under different VE scenarios.

106 **2 Methods**

107 We estimated the proportion of disease burden prevented at the state level for each
108 age cohort and each influenza season from 2010/2011 to 2019/2020 under different
109 vaccination schedules and VE scenarios. We assumed that the historic patterns of vac-
110 cination uptake (the proportion of people getting vaccinated) remained unchanged,
111 and we evaluated schedules shifted to begin from 1 – 20 weeks earlier (i.e., advancing
112 vaccination) or 1 – 20 weeks later (i.e., postponing vaccination) relative to historic
113 uptake patterns for each state in each season. We defined the most favorable vacci-
114 nation schedule as the one that resulted in the highest proportion of disease burden
115 prevented.

116 2.1 Data Sources

117 To estimate the national influenza burden for each age cohort, we used six data
118 sources: Influenza-like Illness (ILI) [29], laboratory-confirmed influenza hospitaliza-
119 tions (FluSurv-NET) [30], National Ambulatory Medical Care Survey (NAMCS) [31],
120 Flu Near You (FNY) Survey results (see ILI Case Data Calibration, Additional File
121 1), vaccine coverage data from FluVaxView [32], and age-specific population estimates
122 [33].

123 We assumed that vaccination timing, surveillance reports of ILI cases, and influenza
124 hospitalizations represent the distribution of disease burden in each season. Due to low
125 flu incidence during the summer, fewer providers report ILI data during this time. The
126 Weekly U.S. Influenza Summary Update used by CDC to monitor influenza activity is
127 updated each week from October through May of each year [34]. Thus, we limited our
128 analysis to ILI data from surveillance week 40 (first week of October) to surveillance
129 week 22 (end of May or beginning of June) to represent the disease burden distribution
130 for the flu season as framed by the CDC.

131 We further assumed (1) that the Flu Near You (FNY) survey respondents (see
132 ILI Case Data Calibration, Additional File 1) constitute a representative sample of
133 the population of each U.S. Census Region (note that this assumption—which is likely
134 incorrect—will be addressed through a sensitivity analysis in Section 2.4); (2) that the
135 data summarized from four seasons from 2015 to 2019 can be considered reasonable
136 time-invariant approximations that may be applied to the ten seasons from 2010 to
137 2020, and (3) that the ratio of FNY symptom reports consistent with ILI to total
138 FNY symptom reports equals the ratio of survey participants with ILI to total survey
139 participants.

140 Based on ILINet’s reporting standards [29], we partitioned the population into five
141 age cohorts: 0-4 years, 5-24 years, 25-49 years, 50-64 years, and ≥ 65 years (Table 1).

142 For disease burden data, we used weekly ILI data from the U.S. Outpatient Influenza-
 143 like Illness Surveillance Network (ILINet) for people < 65 years [29] and laboratory-
 144 confirmed influenza hospitalizations (FluSurv-NET) for people ≥ 65 years [30]. We
 145 used two different types of data because ILI outpatient visits better reflect disease
 146 burden in people < 65 , whereas hospitalizations better represent disease burden for
 147 people ≥ 65 . Hospitalizations for a particular virus more accurately represent disease
 148 burden than do syndromic ILI case counts, which underestimate burden specifically for
 149 older adults [35, 36]. Note that this assumption is consistent with previously published
 150 studies such as Ferdinands et al. [28].

Table 1 Age group correspondences in years from different data sources for people < 65 years old

Historic Case Bur- den and Patient Data (ILINet) ¹	Vaccine Coverage Data (FluVaxView)	National Ambulatory Medical Care Survey (NAMCS)	Flu Near You (FNY) Survey
0 - 4	0.5 - 4	0 - 4	0 - 17
5 - 24	5 - 17	5 - 24	0 - 17
25 - 49	18 - 49	25 - 44	18 - 49
50 - 64	50 - 64	45 - 64	50 - 64

¹Used as the standard for this study.

151 In the ILINet system, ILI is defined as a fever (temperature of 100°F [37.8°C] or
 152 greater) and a cough and/or a sore throat [37]. ILINet includes participating outpatient
 153 healthcare providers in all U.S. states, Puerto Rico, the District of Columbia and the
 154 U.S. Virgin Islands that report the total number of patients seen for any reason and
 155 the number of those patients with ILI by age group each week. All states were included
 156 in this analysis except Florida, as ILINet data are not publicly available for this state.
 157 Unlike ILINet, FluSurv-NET only covers selected states from season to season. We
 158 define a data point as a state in a particular season. Taking into account 49 states and
 159 10 seasons, there are 116 data points for people from FluSurv-Net for the age group
 160 of ≥ 65 years and 490 data points from ILINet for each of the under 65 age groups.

161 For people aged ≥ 65 years, weekly rate of influenza hospitalizations is available
162 from FluSurv-NET [30]. To calculate total hospitalizations, we multiplied the hospi-
163 talization rate by the total population of age ≥ 65 years at the state level. However,
164 ILI case rates by age cohort at the state level are not available. ILINet data are only
165 reported as total ILI cases from all patients seen by ILINet participating providers for
166 any reason. Further, these data only represent people seen in healthcare settings and
167 not the whole population. We accounted for these data constraints by calculating ILI
168 case rate by age cohort at the state level as shown in Figure 1:

- 169 1. We calculated weekly cases by age cohort for each state (steps 1-3 in Figure 1).
170 We used ILINet data at the Health and Human Services (HHS) region level, which
171 is provided by age cohort. We applied the HHS ILI percentages for each cohort to
172 the state total ILI cases to get state level ILI cases seen in healthcare settings by
173 age cohort in each season. In doing so, we assumed that the HHS region level ILI
174 prevalence represents the state level ILI prevalence among different age cohorts.
- 175 2. To estimate total prevalence, we used the results of the FNY survey to calculate the
176 weekly state level ILI cases we would have seen if 100% of people experiencing ILI
177 had sought health care (step 4 in Figure 1), and adjusted the state level ILI preva-
178 lence accordingly (step 5 in Figure 1). See ILI Case Data Calibration, Additional
179 File 1, for details.
- 180 3. We multiplied the state case rate by state population for each age cohort to get
181 total state ILI case rate (steps 6 and 7 in Figure 1).

182 **2.2 Vaccine Coverage and Seasonal Influenza Peaks**

183 In addition to seasonal VE, influenza vaccine uptake and the timing of influenza activ-
184 ity are the other two major factors that impact the outcome of our model. We used
185 monthly data by age group and state from FluVaxView for vaccine coverage data
186 [32]. See Monthly Coverage Data Preparation, Additional File 1, for our treatment of

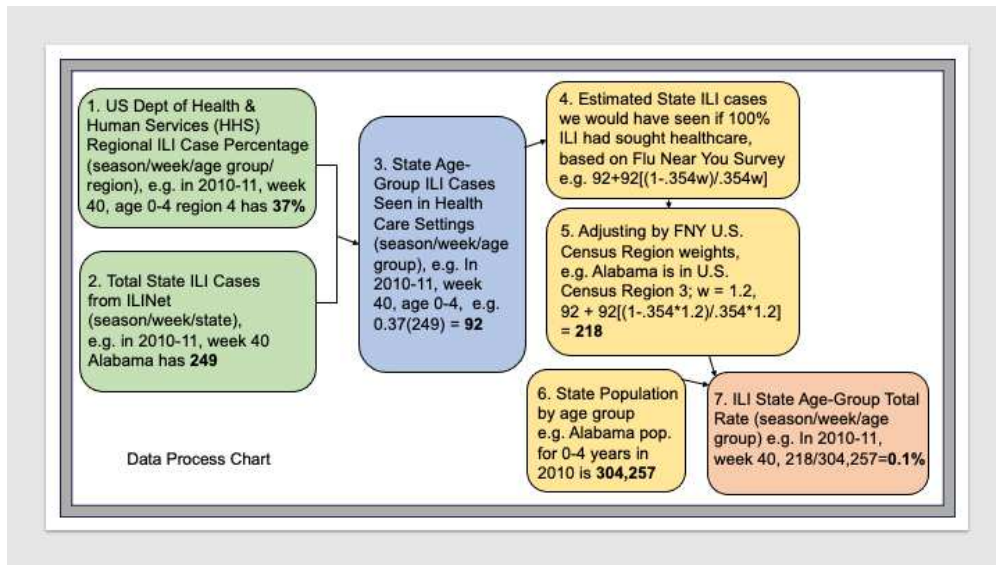


Fig. 1 Process chart depicting data used to calculate state level total ILI cases by age cohort, using 0-4 age group and the state of Alabama as an example

187 missing data problems. We then interpolated monthly coverage to estimate weekly coverage as follows. We assumed that vaccination begins at the end of July (surveillance week 30), and we assigned all vaccine coverage data for July (we assumed zero if not reported) to surveillance week 30. Cumulative monthly coverage data from surveillance week 30 to surveillance week 22 (end of May) of the following year were interpolated by fitting monotonic cubic splines to estimate weekly coverage. We assumed that a full immune response is achieved two weeks after vaccination, thus a vaccine received on surveillance week 30 becomes fully effective on surveillance week 32.

195 We adopted the age cohorts from ILINet as the standard for our age cohorts; Table 196 1 shows the age cohort structure for the coverage, NAMCS, and FNY data used in this study. Vaccine coverage data is reported for the age groups of 6 months to 4 years 198 and 6 months to 17 years. We derived coverage for the age group of 5 to 17 years from the difference of the above two age groups using coverage and population data. 199 The vaccine uptake pattern and total vaccine coverage (Figure 2) showed substantial 200

201 variation across age groups and seasons (see Note on Data Consistency, Additional File
 202 1). Compared to the other three age groups, the youngest and the oldest age groups
 203 have much higher total coverage and faster vaccination uptake from July to November,
 204 at which point for all groups, coverage typically starts to flatten out. This uptake is
 205 represented by the slope of the cumulative coverage between July and November.

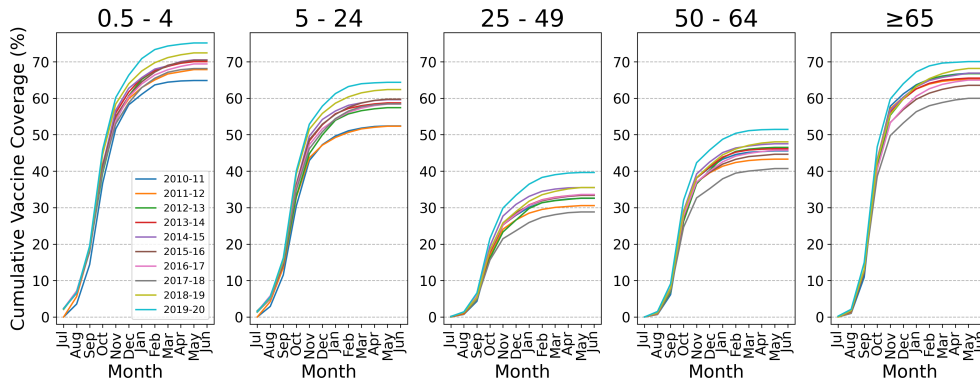


Fig. 2 Cumulative nationwide monthly coverage by age group for seasons 2010/2011 to 2019/2020

206 To assess the potential impact of influenza peak timing, we separated the results
 207 for states that experienced early peaking (peak week at or before the third week of
 208 January) from those for states that had late peaking (peak week after the third week
 209 of January) across all ten seasons. Based on historical data, influenza activity typically
 210 peaks between December and February [34]; however, a flu season can have multiple
 211 peaks. We defined the peak week for each season as the week at which the largest
 212 number of ILI cases are observed. When more than one equivalent week existed, we
 213 chose the first one. As shown in Figure 3, historical influenza peak weeks at the state
 214 level varied among seasons.

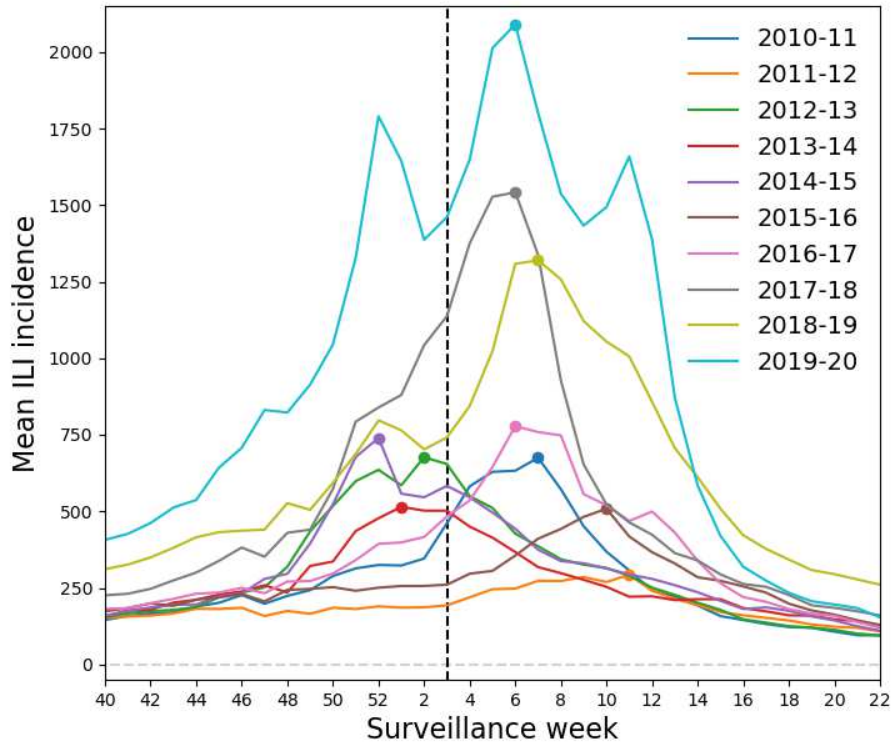


Fig. 3 Mean incidence of influenza-like illness for ten seasons, 2010-2020. Influenza surveillance weeks from approximately October to May are shown on the x-axis; mean incidence is shown on the y-axis. Color indicates season. The dotted vertical line at Week 3 marks the partition between early and late peaking seasons, and the points on the curves indicate peak weeks.

215 2.3 Model

216 We simulated six scenarios for each age group. Each scenario uses a different VE
 217 waning function and has either 490 (< 65 years) or 116 (≥ 65 years) data points
 218 corresponding to a total of 10 seasons. Each data point is defined as one state in a
 219 specific season for that age group (for example, Alabama in the 2010/2011 season for
 220 age group 0 – 4 years old). For each data point, we calculated the percentage of disease
 221 burden averted under the actual vaccination schedule and 40 shifted schedules (i.e.,

222 beginning from one to 20 weeks earlier or later than the actual schedule). Although
 223 the vaccination schedule is unlikely to be shifted by five months earlier due to various
 224 factors such as vaccine production and logistics, we included this wide range to ensure
 225 that the best shift would fall within the explored range. We calculated the percentage
 226 of disease burden prevented under the current and the shifted schedules for the season
 227 (i.e., surveillance week 40 to week 22 of the following year in our study). We defined
 228 the most favorable vaccination start week, that is, the optimal vaccination strategy,
 229 as the one resulting in the the greatest estimated reduction in disease burden.

230 Following Newall et al. [27] with revised notations, the proportion of prevented
 231 cases from vaccination (p_t) for each week is estimated as

$$p_t = \sum_{s=0}^{t-2} c_{(t-2-s)} e_s \quad (1)$$

232 where t is time in weeks and s represents the number of weeks that have elapsed
 233 since the development of a full immune response (i.e., if it has been t weeks since the
 234 vaccine was administered, $s = t - 2$). Note that we restrict $t \geq 2$ to account for the
 235 two-week lag of immune response development (we assume people are fully vaccinated
 236 two weeks after receiving the shot).

237 $\sum_{s=0}^{t-2} c_{(t-2-s)}$ is the cumulative vaccine coverage at week t while accounting for the
 238 two-week delay in the development of an immune response. Figure 2 shows cumulative
 239 vaccine coverage c_t , without the two-week delay. e_s is the value of the vaccine efficacy
 240 function (i.e., VE waning function) on week s , and it ranges $[0,1]$. e_0 is the initial VE
 241 value, where $e_s = 0$ indicates that the vaccine is not protective and $e_s = 1$ indicates
 242 that the vaccine is 100% effective.

243 For example, if the cumulative vaccine coverage is 40% on a given week of the
 244 influenza season for a given age group and $e_s = 1$, 40% of cases will have been averted,

245 whereas (in a more realistic situation), if cumulative vaccine coverage remains the
 246 same and $e_s = 0.5$, 20% of cases will have been averted.

247 To calculate the baseline disease burden without vaccine at week t , D'_t , we removed
 248 the vaccine effect from the reported disease burden, D_t , as follows,

$$D'_t = \frac{D_t}{1 - p_t} \quad (2)$$

249 Then seasonal disease burden without vaccine, from surveillance week 40 to week
 250 22 of the following year in our analysis, is represented by $\sum_t D'_t$. Thus the seasonal
 251 percentage of disease burden prevented is represented by

$$\frac{\sum_t D'_t - \sum_t D_t}{\sum_t D'_t} \quad (3)$$

252 We considered six hypothetical vaccine effectiveness scenarios as shown in Figure 4
 253 to account for three levels of starting VE (high, medium, and low) and two effectiveness
 254 waning functions (fast and slow). Note that VE and waning are assumed to be the same
 255 across all age groups and that we constrain negative values of VE to zero, although a
 256 value of 0 for VE is likely a pessimistic assumption.

257 **2.3.1 High VE and Fast Waning**

258 For the first scenario, we adapted the waning model fitted to empirical data by Fer-
 259 dinands et al. [28]. We modified their original equation by starting waning two weeks
 260 after vaccination, and by changing the time unit to one week instead of bi-week for t .
 261 This scenario assumes a high VE of 55% and fast waning over 27 weeks with a season
 262 average waning rate of approximately 7%.

$$VE_1 = \max(0, 55 - 1.37t + 0.18t^2 - 0.03t^3) \quad (4)$$

263 **2.3.2 High VE and Slow Waning (Best Scenario)**

264 In the second scenario, we assume the same high initial VE as the first scenario and
265 slower waning, dropping to zero after 37 weeks.

$$VE_2 = \max(0, 55 - 0.5t + 0.05t^2 - 0.01t^3) \quad (5)$$

266 **2.3.3 Medium VE and Fast Waning**

267 We also adapted the third scenario from Ferdinands et al., which was fitted to empirical
268 data [28]. It has an initial VE of 30.85% and wanes quickly, over 22 weeks.

$$VE_3 = \max(0, 30.85 - 1.37t + 0.18t^2 - 0.03t^3) \quad (6)$$

269 **2.3.4 Medium VE and Slow Waning**

270 In the fourth scenario, we assume the same initial medium VE as in the third scenario,
271 with slower waning of 31 weeks.

$$VE_4 = \max(0, 30.85 - 0.5t + 0.05t^2 - 0.01t^3) \quad (7)$$

272 **2.3.5 Low VE and Fast Waning (Worst Scenario)**

273 The fifth scenario has an initial VE of 20% and wanes over 19 weeks.

$$VE_5 = \max(0, 20 - 1.37t + 0.18t^2 - 0.03t^3) \quad (8)$$

274 **2.3.6 Low VE and Relatively Slow Waning**

275 The sixth scenario has an initial VE of 20% and wanes over 26 weeks.

$$VE_6 = \max(0, 20 - 0.5t + 0.05t^2 - 0.01t^3) \quad (9)$$

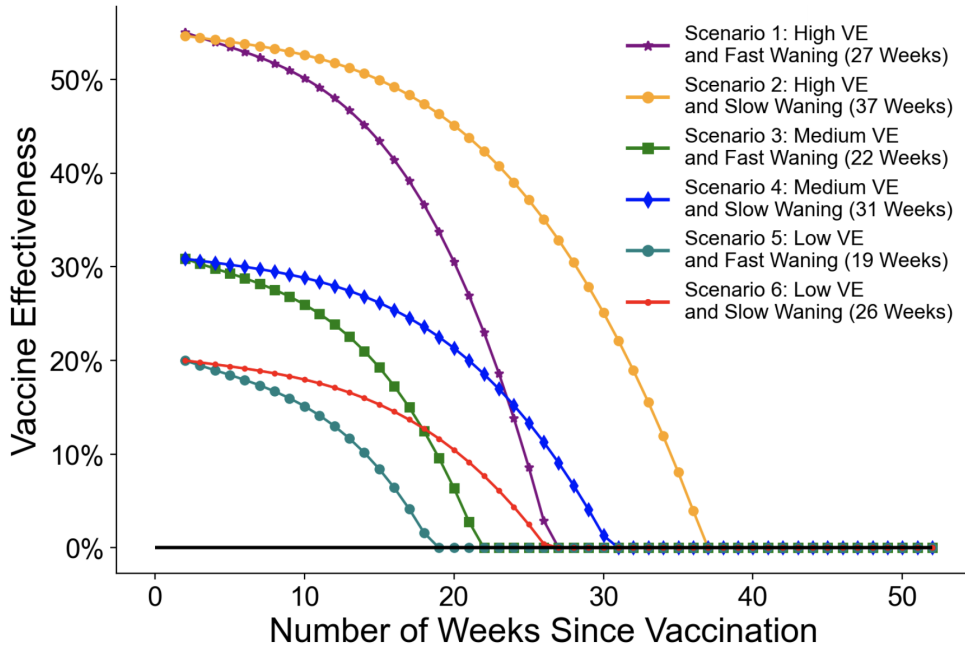


Fig. 4 Vaccine effectiveness waning scenarios. The initial vaccine effectiveness is set two weeks after vaccination, when the waning process begins.

276 **2.4 Sensitivity Analysis**

277 Given the existence of uncertainty in weekly influenza case count estimates for ten
 278 seasons from 2010/2011 to 2019/2020, we conducted a sensitivity analysis to evalu-
 279 ate the robustness of our results with respect to the precise case estimation method
 280 used. For people < 65, we defined a lower bound (extreme underestimation) as the
 281 number of cases seen in health care settings divided by the entire population. In the

282 underestimation, we are assuming that ILI cases seen in health care settings repre-
283 sent all cases that occurred. We defined an upper bound (extreme overestimation) by
284 assuming that the proportion of ILI to people seen in health care settings equals the
285 proportion of ILI in the entire population. These lower and upper bounds on the case
286 count estimates take into account that the FNY survey may not be a representative
287 sample of the underlying populations and may under or over estimate healthcare seek-
288 ing behavior. The lower bound takes into account that ILI is likely an overestimate of
289 influenza, since many viruses contribute to ILI. The sensitivity analysis is not appli-
290 cable to people ≥ 65 , as FluSurv-NET provides laboratory-confirmed influenza case
291 counts for that age group.

292 **3 Results**

293 In general, when VE waning is fast, delaying vaccination is beneficial in most states and
294 seasons for most age groups. When waning is slow, results are mixed. The distributions
295 of favorable schedule shifts for starting vaccination are shown in Figure 5 and the mean
296 favorable shifts are shown in Table 2. See Additional File 2 for individual histograms
297 for each age group.

298 In Scenario 1, for the case of late peaking seasons for ≥ 65 , the average favorable
299 shift indicates postponing vaccination by roughly seven weeks (Table 2). The absolute
300 shift size is less than 4 weeks for all other age groups. In early peaking seasons, the
301 best shift is postponing vaccination by about 2 weeks for ≥ 65 , and absolute shift size
302 is less than or equal to one week for all other age groups.

303 In Scenario 2, the schedule that prevents the most cases in the majority of the
304 circumstances has vaccination starting earlier than historic uptake for all age groups
305 with the exception of ≥ 65 for late peaking seasons (Figure 5). On average, the best
306 shift for all age groups is to begin vaccination earlier by 3 to 4 weeks in the case of
307 early peaking seasons, and by about one to two weeks in late peaking seasons, with

308 the exception of the ≥ 65 group (Table 2). For ≥ 65 , the average beneficial shift is to
 309 postpone vaccination by about three weeks for late peaking seasons.

310 Scenario 3 has the same waning rate (7% average monthly waning over 22 weeks)
 311 as does Scenario 1, but has a lower initial VE of 30.85%. In this scenario, the schedule
 312 shift that reduces the most burden for early and late peaking seasons and all age
 313 groups is to postpone vaccination.

Table 2 Mean shift in weeks that averts maximum cases, aggregated by age group and early or late peaking seasons under six VE and waning scenarios. Negative numbers indicate beginning vaccination sooner than historic uptake; positive numbers indicate beginning vaccination later than historic uptake.

Scenario	0-4	0-4	5-24	5-24	25-49	25-49	50-64	50-64	≥ 65	≥ 65
	Early	Late	Early	Late	Early	Late	Early	Late	Early	Late
1	0.40	2.69	0.50	3.94	-0.32	2.36	1.04	3.64	1.82	7.10
2	-3.12	-1.76	-3.19	-0.98	-4.23	-2.45	-2.97	-1.29	-2.07	3.22
3	2.24	4.84	2.37	6.04	1.85	4.99	3.10	5.95	3.63	8.99
4	-0.96	0.74	-0.90	1.75	-1.85	0.27	-0.51	1.52	0.59	5.69
5	3.56	6.20	3.62	7.04	3.28	6.53	4.35	7.19	4.70	10.31
6	0.57	2.64	0.66	3.89	0.03	2.50	1.28	3.71	2.13	7.27

314 In Scenario 3 for early peaking seasons, the average best shift for age groups 0–4 years,
 315 5–24 years, and 25–49 years is to postpone vaccination by about one to two weeks,
 316 and for ≥ 65 by about three weeks. In the case of late peaking seasons, the average
 317 best shift for age groups < 65 is to postpone vaccination by about five to six weeks,
 318 and for ≥ 65 by nine weeks.

319 In Scenario 4, with initial VE of 30.85% and waning of 31 weeks, the schedule
 320 that begins about one to two weeks earlier than historic uptake averts the most cases
 321 in early peaking seasons, with the exception of ≥ 65 , for which the best schedule is
 322 postponed by about half a week. The schedule that begins about one to six weeks
 323 later than historic uptake averts the most cases for late peaking seasons.

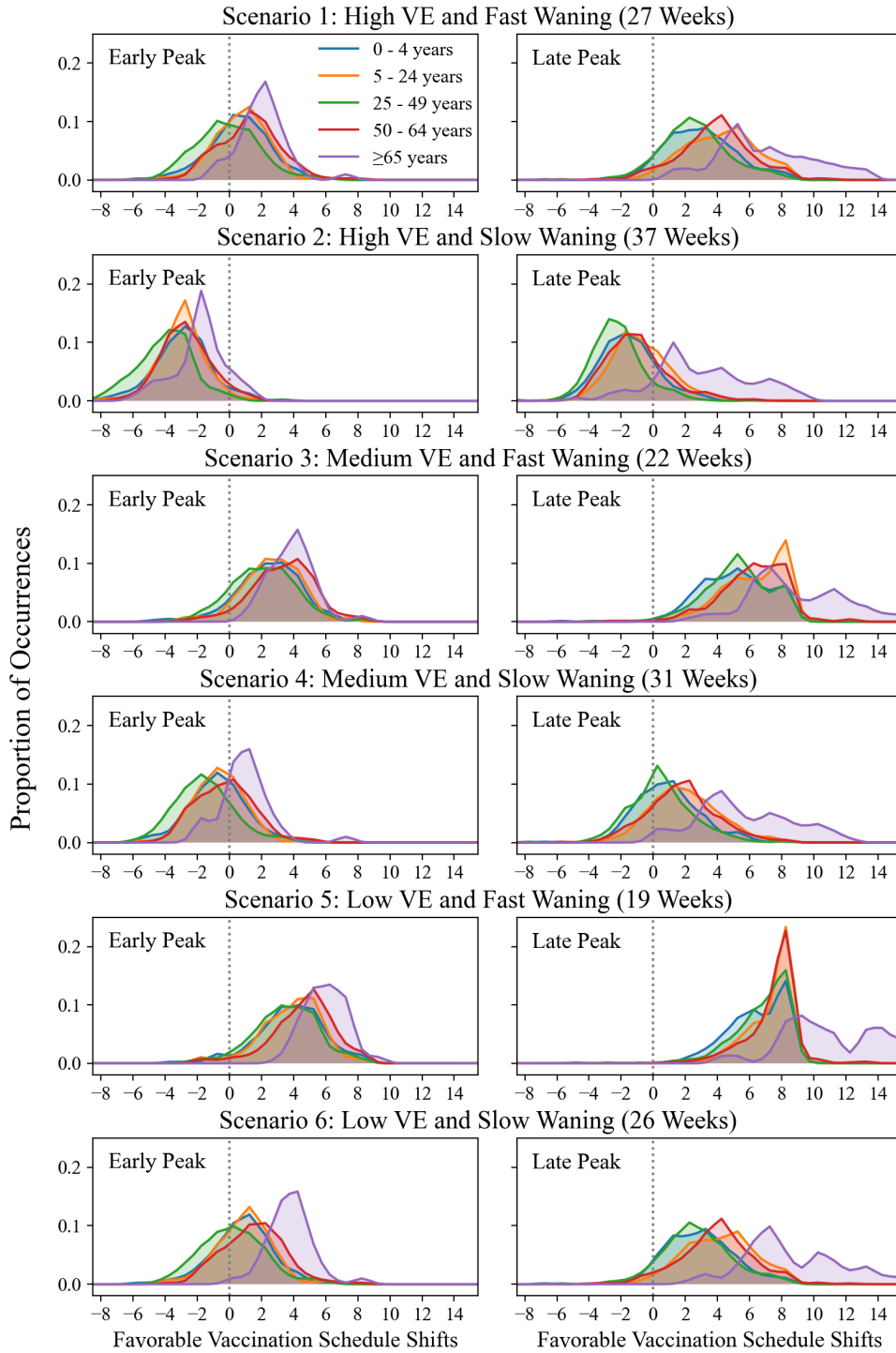


Fig. 5 Distributions of favorable vaccination schedule shifts compared to historic uptake, under six different initial VE and waning scenarios. Data points are states/seasons. On the x-axes, zero indicates the starting historic vaccination uptake week for each state and season. Negative numbers indicate early vaccination in weeks; positive numbers indicate delayed vaccination in weeks. The y-axes show the proportion of occurrences of a particular schedule shift for all states and seasons in our data set. For visualization purposes, shift distributions are smoothed using a Gaussian with 0.5 week standard deviation.

324 In both Scenarios 5 and 6, the initial VE is 20%. Scenario 5 has the shortest
325 duration of protection at 19 weeks (i.e., it is the worst case among the six scenarios
326 analyzed). In both scenarios, postponing vaccination reduces influenza burden for most
327 states and seasons in all age groups, with the exception of the 25-49 group in Scenario
328 6, which benefits from starting vaccination at approximately the historic uptake point
329 in the early peak case.

330 For each age group under all scenarios, there are differences in the average favor-
331 able shift when comparing late peaking seasons to early peaking seasons. In general,
332 late peaking seasons suggest delaying vaccination, especially under the fast-waning
333 scenarios. It should also be noted that the distribution of schedules is wider in the
334 case of late peaking seasons because the window of early peaking (as early as mid-
335 November to mid-January) is shorter than that of late peaking (mid-January to as
336 late as mid-April). Additionally, we observe that for the ≥ 65 age cohort, postpon-
337 ing vaccination appears to be the most favorable recommendation across all scenarios,
338 except for Scenario 2, early peak.

339 **3.1 Sensitivity Analysis Results**

340 Our sensitivity analysis shows that model outputs (distributions and means of favor-
341 able schedule shifts) for all scenarios are robust to variation in influenza case count
342 estimates (see Additional File 3). Slight variation in the results exists for the overes-
343 timation, which is based on the assumption that the proportion of ILI cases seen in
344 health care settings equals the proportion of ILI cases in the entire population. How-
345 ever, 100% of the 24 mean favorable schedule shifts resulting from this upper bound
346 of ILI case estimates are within one week of those from the main analysis. Two out
347 of these 24 mean schedule shifts differs in sign from those of the main analysis. As
348 expected, the mean schedule shifts of the lower bound of ILI case estimates, which are
349 a linear transformation of the main case estimates, are identical to those of the main

350 case. Overall, despite a wide range of variation in the absolute numbers of estimated
351 cases, the resulting optimal vaccination strategy remains the same. See Additional
352 File 3 for table and plots.

353 4 Discussion

354 An ideal way to reduce influenza burden would be by means of a highly effective
355 vaccination program that provides protection for the whole season. However, vaccine
356 effectiveness varies and wanes. In this study, we quantified the impact of different
357 vaccine effectiveness and waning scenarios based on 10 seasons of influenza case and
358 vaccine uptake data to determine vaccination roll-out schedules that avert the great-
359 est number of cases. Strengths of our study are stratification by five age categories,
360 analyses of a broad spectrum of waning scenarios, and consideration of 40 vaccination
361 start weeks, 1-20 before and 1-20 after the historic start week.

362 Our results show that it may be worthwhile to consider delaying vaccination if the
363 waning rate is fast; yet if the VE wanes slowly, it is more challenging to determine
364 an ideal influenza vaccination schedule. Additionally, our findings show that starting
365 vaccination at the same time for all age groups may not be optimal. Nevertheless, our
366 analysis supports current ACIP timing of vaccination recommendations that children
367 can get their vaccine earlier in situations when initial VE is high and waning is slow,
368 but that adults should avoid earlier vaccinations in most scenarios [5]. Our results
369 indicate that postponing vaccination is favorable in most circumstances for the ≥ 65
370 group, with the exception of Scenario 2 (high VE and slow waning in early-peaking
371 seasons). Thus, a general conclusion cannot be drawn in the absence of improved
372 ability to predict the peak of the flu season. A tiered vaccination strategy can be
373 implemented under the existing schedule to improve outcomes. Nonetheless, the deci-
374 sion of postponing or advancing vaccination for each age group cannot be determined

375 without knowing how the vaccination coverage may potentially change, how the VE
376 will wane, and when the influenza season will peak.

377 A challenge to consider is our limited understanding of VE waning and how to slow
378 the waning of VE. Recently, many studies have evaluated how immunity wanes after
379 vaccination [7–28]. However, results are largely inconsistent in terms of the estimated
380 VE and the degree of waning. Better estimates of VE and its waning rate, as well
381 as the data needed to support this, will be important in the future. To slow the
382 waning, Rambhia and Rambhia (2018) recommended a mid-flu-season booster vaccine,
383 and vaccine adjuvants or use of high-dose vaccines for susceptible populations. As
384 of June 30, 2022, the CDC adopted ACIP’s recommendation that people > 65 be
385 preferentially given high dose or adjuvanted flu vaccines [38]. However, limited studies
386 on the outcomes of this recommendation have shown mixed results [39]. It remains to
387 be seen how VE waning is impacted.

388 Finally, this work shows that accurate peak timing forecasts for flu seasons with
389 actionable lead times can play an important role in vaccination start time decisions.
390 Accurate and reliable long-lead peak time forecasts could guide public influenza vac-
391 cination campaigning efforts. In particular, seasonal forecasts can guide when people
392 should start getting vaccinated (e.g., before Halloween). Such accurate peak time fore-
393 casts at this lead time do not currently exist (e.g., accurate early/late peak forecasts
394 made in July), but infectious disease forecasting is an active area of research with
395 real-time forecasting [40–42]. Additionally, new initiatives by the CDC Center for
396 Forecasting and Outbreak Analytics [43], are promising in making these forecasts a
397 reality. This work provides a practical and concrete example in which reliable fore-
398 casting efforts could help reduce the burden of seasonal influenza through annual fine
399 tuning of vaccination start dates.

400 Note that our study did not specifically assess the impact of vaccination schedule
401 changes on missed vaccination when the schedule is postponed. Rather, we focused on

402 the fundamental issue of whether or not we have, or could acquire, enough information
403 to change the existing vaccination schedule. However, our analysis does implicitly
404 account for missed vaccinations due to shifts in vaccination start dates. Based on our
405 assumption that the vaccine uptake pattern persisted under the shifted schedule, some
406 portion of the population who were normally vaccinated late in the season would be
407 shifted out of the postponed schedule in certain cases, because they effectively missed
408 vaccination for that season. The impact of this issue, however, is minor relative to the
409 large and important uncertainty in the estimate of the VE waning function itself.

410 There are several limitations to our study. First, we applied the same VE wan-
411 ing functions across seasons, while VE dynamics could differ from season to season
412 depending on the circulating influenza virus strains and the match between the viruses
413 and the vaccine [44–46]. Second, we did not specifically quantify the impact of pre-
414 cise influenza season peak timing on vaccination schedule; we only considered whether
415 a season peaked early or late (see Methods, Section 2). Third, we did not explore
416 potential differences in VE waning when history of repeated vaccinations is taken into
417 account [47, 48]. Our VE waning functions are based on existing literature but addi-
418 tional studies are needed to more accurately quantify vaccine effectiveness and waning.
419 Fourth, the VE functions assumed similar dynamics for all age groups and it is possi-
420 ble that each age group may have different dynamics [49]. Fifth, our models assume
421 that shifting vaccination timing does not change people’s vaccination behavior (i.e.,
422 a shift in time does not change the shape of the vaccine coverage curves). While this
423 assumption is unlikely to hold to high precision, by including data from 49 states and
424 10 seasons our results are unlikely to be appreciably affected by such changes in behav-
425 ior (see Additional File 1 for further discussion). Finally, we used ILI data for people
426 < 65 years of age, which includes non-influenza viral causes and as such, it is likely to
427 overestimate influenza burden and increase uncertainty in its temporal dynamics [50].

428 Nonetheless, according to our sensitivity analysis, the results of our study are robust
429 to wide variations in ILI case estimates, and the overall insights are unchanged.

430 **5 Conclusions**

431 Our results are consistent with current CDC influenza recommendations for vaccine
432 timing, which are intended to guide influenza vaccination campaigns. However, given
433 the uncertainty in VE effectiveness and waning rates, it may be beneficial to offer
434 an annual mid-season booster vaccination to people ≥ 65 in order to reduce both
435 morbidity and mortality in this population. Additionally, our results show that one
436 size does not fit all, and that a tiered vaccination strategy may lead to more favorable
437 outcomes.

438 **List of abbreviations**

439 **ACIP** US Advisory Committee on Immunization Practices

440 **CDC** Centers for Disease Prevention and Control

441 **HHS** Health and Human Services

442 **ILINet** Influenza-like Illness Surveillance Network

443 **NAMCS** National Ambulatory Medical Care Survey

444 **VE** Vaccine Effectiveness

445 **WHO** World Health Organization

446 **Declarations**

447 **Ethics approval and consent to participate**

448 All data are publicly available.

449 **Consent for publication**

450 Not applicable.

451 **Availability of data and materials**

452 All data are publicly available and fully cited.

453 **Competing interests**

454 No competing interests are declared.

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466 **Authors' contributions**

467 MZS, DO, and SYD conceived and designed the study. JAS, MZS, and PCA acquired
468 and curated data, carried out quantitative analyses, visualized results, and wrote the
469 manuscript. SYD and DO supervised the project and critically revised the manuscript.
470 SYD acquired funding for the project. All authors discussed the results, contributed

471 to reviewing and editing the final manuscript, and approved of the version to be
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666 **Additional File 1** This file contains the following supplementary data notes: (1)
667 ILI case data calibration tables; (2) Monthly coverage data preparation; (3) Note on
668 data consistency.

669 **Additional File 2** This file contains individual age group histograms of favorable
670 vaccination schedule shifts (Figs 1-5).

671 **Additional File 3** This file contains sensitivity analysis table and plots (Table 1
672 and Figs 1-2).

Supplementary Files

This is a list of supplementary files associated with this preprint. Click to download.

- [AdditionalFile1.pdf](#)
- [AdditionalFile2.pdf](#)
- [AdditionalFile3.pdf](#)