

Scenarios of diffusion and control of an influenza pandemic in Italy

C. RIZZO^{1,2*}, A. LUNELLI³, A. PUGLIESE³, A. BELLA¹, P. MANFREDI⁴,
G. SCALIA TOMBA⁵, M. IANNELLI³, M. L. CIOFI DEGLI ATTI¹ on behalf
of the EPICO Working Group

¹ National Centre for Epidemiology Surveillance and Health Promotion, Istituto Superiore di Sanità, Rome, Italy

² Department of Pharmaco-Biology, University of Bari, Italy

³ Department of Mathematics, University of Trento, Italy

⁴ Department of Statistic and Economical applied Mathematics, University of Pisa, Italy

⁵ Department of Mathematics, University of 'Tor Vergata', Roma, Italy

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SUMMARY

To predict the spread of a pandemic strain of influenza virus in Italy and the impact of control measures, we developed a susceptible–exposed, but not yet infectious–infectious–recovered, and no longer susceptible (SEIR) deterministic model with a stochastic simulation component. We modelled the impact of control measures such as vaccination, antiviral prophylaxis and social distancing measures. In the absence of control measures, the epidemic peak would be reached about 4 months after the importation of the first cases in Italy, and the epidemic would last about 7 months. When combined, the control measures would reduce the cumulative attack rate to about 4·2%, at best, although this would require an extremely high number of treated individuals. In accordance with international findings, our results highlight the need to respond to a pandemic with a combination of control measures.

INTRODUCTION

Following the emergence in 1997 of a new strain of avian influenza, A(H5N1), which is capable of infecting humans [1], and the spread of this strain to Europe in 2005 [2], concerns were raised over the occurrence of a pandemic caused by A(H5N1) or a closely related strain [3, 4]. Consequently, countries have been urged to strengthen their preparedness for an influenza pandemic [5], an important aspect of which is predicting the spread of infection.

According to the predictive models used to date [6–12], influenza would spread worldwide over a

period of 2–6 months, depending on the basic reproductive number (R_0), and reducing transmission would entail combining control measures, specifically, reducing contacts and performing both antiviral prophylaxis (AVP) and vaccination [7–9, 11, 13].

We developed a SEIR (susceptible–exposed, but not yet infectious–infectious–recovered, and no longer susceptible) deterministic model with a stochastic simulation component to predict the spread of pandemic influenza in Italy and to evaluate the impact of vaccination, AVP and social distancing measures.

METHODS

The SEIR model

We developed a SEIR model in which the population is structured according to age and region of

* Author for correspondence: Dr C. Rizzo, National Centre for Epidemiology, Surveillance and Health Promotion, Istituto Superiore di Sanità, Viale Regina Elena, 299, Rome, Italy.
(Email: caterina.rizzo@iss.it)

residence. We defined six age classes: infants (0–2 years), children (3–14 years), teenagers (15–18 years), young adults (19–39 years), adults (40–64 years) and the elderly (≥ 65 years). In the model, the national population (56 995 744 inhabitants) was also distributed in Italy's 20 regions, according to national demographic data obtained from the 2001 Census [14]. The contact matrix was defined by considering, separately, household, school/work-place and random contacts, and by using data on household composition, school attendance, and employment status. The transportation matrix was defined using data on national airline traffic [15]. The model consists of a system of differential equations, reported in the Appendix.

We also introduced a stochastic component that takes into account all of the random effects that are important during a pandemic's initial and final stages, when the number of infected individuals would be low. Precisely, whenever the deterministic prediction of the number of infected individuals in an age class/region was below the threshold value of 10, it was replaced by a Poisson variable with the same mean. In each simulation, the pandemic began with the introduction of five infected adults in the Lazio Region, where Rome's intercontinental airport is located.

Based on published studies [16] and using the method described by Diekmann & Heesterbeek [17], we computed an R_0 of 1.8, which, when applying the contact matrix, corresponds to a cumulative infected attack rate (AR) of 35%.

Based on the literature [6, 8, 18], in the model we assumed an incubation period of 1 day and an infectious period of 3–9 days. The results were obtained by averaging over 200 simulations for each scenario. For all of the results, the 5–95 percentile values of the AR estimates were within 11%.

Control measures

We considered both single and combined control measures, most of which are included in the Italian National Plan for preparedness and response to an influenza pandemic [19]. We assumed that two doses of vaccine would be administered, 1 month apart. The target population was divided into four categories: (I) personnel providing essential services (15% of the working population, aged 25–60 years) [14]; (II) elderly persons (aged ≥ 65 years); (III) children and adolescents (aged 2–18 years); and (IV) adults (aged 40–64 years). We assumed a vaccination coverage of

60% of the target population, based on the 2005–2006 national influenza coverage [20]. We assumed that a period of 2 weeks would be necessary for administration of the vaccine to each target category. For vaccine effectiveness (VE) we made two different assumptions: (1) VE of 70% for all categories; and (2) VE of 50% for all categories; for both assumptions, we assumed that the VE would be reached starting 15 days after the second dose.

We considered different scenarios of vaccine availability; in one scenario, adequate VE would be reached 4 months after the first national case; in the second scenario, it would be reached after 5 months. An adequate VE at 4 months would be feasible only if the first dose contained an avian virus precursor of the pandemic strain [3], followed by a dose of pandemic vaccine; the actual VE of this regimen was assumed to be equal to that of two doses of the pandemic vaccine.

The AVP for uninfected individuals was assumed to reduce susceptibility by 30% and infectiousness by 70% [8]. We considered the administration of one course of antiviral drugs. We assumed that AVP would be provided to all household contacts of 80% of the clinical cases (66% of all infected individuals). We considered administration of AVP for the entire epidemic period; however, since the feasibility of actually doing this would be limited, we also considered other scenarios, i.e. administering AVP only for 2, 4, 8, or 16 weeks after the occurrence of the first Italian case. AVP was assumed to reduce the transmission rate among household contacts, based on the consideration that those household contacts already infected at the time of starting AVP would have a reduced infectiousness, therefore it would be as if only a fraction of them were actually infected; those not yet infected when starting AVP would benefit from both lower susceptibility and lower infectiousness.

We considered the nationwide closure of all schools, public offices, and public meeting places (e.g. restaurants, cinemas, and churches). We simulated the closure of schools for 3 weeks, public offices for 4 weeks and public meeting places for 8 weeks. We assumed that these measures would be introduced simultaneously, at different times (i.e. 2, 4 or 8 weeks after the start of the pandemic). In the model, school closure would reduce the contacts among children and teenagers (the school component of the transmission rate) by 75%; workplace closure would reduce the job component of the transmission rate by 16%; closure of public meeting places would reduce

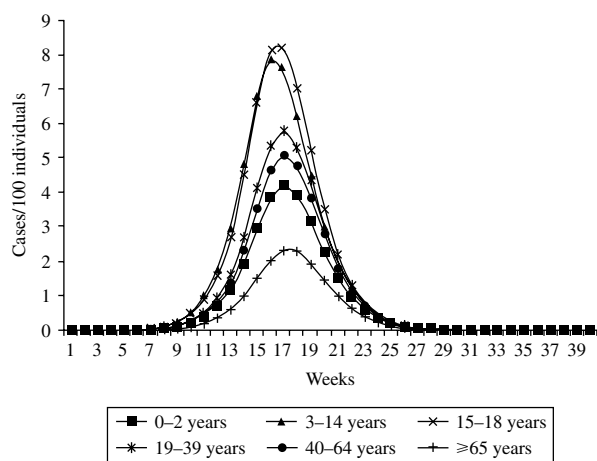


Fig. 1. Weekly attack rate, by age group, with no control measures.

the random component of the transmission rate by 50%.

Sensitivity analysis

We evaluated how the results would change depending on different levels of pathogen transmissibility, with a resulting R_0 value of 1.6, 1.8 or 2.0. We also considered the resulting AR for the three R_0 values for the baseline scenario and for scenarios that differed in terms of the specific control measures adopted.

RESULTS

Baseline dynamics

In the absence of control measures, the epidemic peak would be reached about 4 months after the identification of the first case, with a total of 3 million cases during the peak week. The epidemic would be over in 7 months, with a cumulative infected attack rate (AR) of 35% (about 20 million cases). The dynamics of the epidemic were similar in all age groups, whereas the AR varied markedly by age group. The incidence would be particularly high among those aged 15–18 years, with a AR of 54% (Fig. 1).

Because of the model's stochastic component, the introduction of few infectious individuals in the population did not always result in an outbreak; in fact, in around 40% of the simulations, the number of infected individuals in the early stages of the pandemic was insufficient for sustaining transmission, and the epidemic expired spontaneously.

Single control measures

The impact of single control measures is shown in Table 1. The introduction of control measures frequently increased the probability of stochastic extinction of the pandemic. Vaccination seems to be the most effective measure, especially when VE is reached at 4 months. Vaccinating three of the four target categories (i.e. personnel providing essential services; elderly persons; and those aged 2–18 years) would reduce the AR from 35% to 25%, with almost 5 million cases avoided by treating about 17 million individuals. Vaccinating the fourth target category (i.e. the 40–64 years age group) would not result in an important additional reduction in the AR. If protective VE were reached at 5 months (two doses of pandemic vaccine), vaccinating all four categories, the AR would be 32.5%. Assuming a VE of 50% for all categories would not greatly affect the AR; in fact, the AR would be only 2 or 3 percentage points higher than the AR when assuming a 70% VE for all categories (Table 1).

Social distancing measures and AVP were not effective in reducing the AR. However, providing AVP for 16 weeks after the identification of the first Italian cases and implementing social distancing measures starting at week 4 or week 8 would delay the epidemic peak by 1 or 3 weeks, respectively.

Combined control measures

The combination of control measures would be more effective than single measures (Tables 2 and 3). The highest reduction (from 35% to 4.2%) would be obtained by starting social distancing measures at week 4, providing AVP throughout the entire epidemic, and performing vaccination with a VE of 70% at 4 months (when combining measures, we assumed that vaccination would be provided for all categories). This would allow for 17 million cases to be avoided by vaccinating around 26 million individuals and by providing AVP to about 3 million individuals. The AR would be higher (11%) if VE were reached at 5 months, avoiding 13 million cases by treating 25 million individuals and 7 million individuals with vaccine and AVP respectively (Table 2). Providing AVP for 16 weeks, instead of for the entire epidemic period, would increase the AR to 8.4% or 16.6% if VE were reached at 4 or 5 months, respectively. However, this would determine an important reduction in the number of treated individuals (about 150 000). Combining control measures would also

Table 1. Effectiveness of single control measures on the dynamics of an influenza pandemic with an R_0 of 1.8 and an attack rate of 35%, for different values of vaccine effectiveness (VE)

Control measures	Attack rate*	Avoided cases	Treated individuals
Adequate VE at 5 months (VE = 70%)			
Categories I, II	33.0% (29.1–34.1)	974 151	12 076 619
Categories I–III	32.6% (26.4–34.0)	1 203 363	17 006 817
Categories I–IV	32.5% (25.9–34.0)	1 260 666	25 542 092
Adequate VE at 4 months (VE = 70%)			
Categories I, II	28.9% (27.1–30.5)	3 323 574	12 076 619
Categories I–III	25.3% (17.8–29.2)	5 386 482	17 279 633
Categories I–IV	24.4% (13.1–29.0)	5 902 209	25 814 908
Adequate VE at 5 months (VE = 50%)			
Categories I, II	33.4% (30.5–34.2)	744 939	12 076 619
Categories I–III	33.0% (28.3–34.2)	974 151	17 008 452
Categories I–IV	33.0% (27.8–34.1)	974 151	25 543 727
Adequate VE at 4 months (VE = 50%)			
Categories I, II	30.4% (29.1–31.6)	2 464 030	12 076 619
Categories I–III	27.5% (22.5–30.4)	4 125 818	17 278 523
Categories I–IV	26.6% (18.5–30.2)	4 641 545	25 814 799
Antiviral			
2 weeks	34.7% (34.7–34.7)	0	50
4 weeks	34.7% (34.7–34.7)	0	355
8 weeks	34.7% (34.7–34.7)	0	12 389
16 weeks	33.9% (33.3–34.6)	458 424	2 993 052
Entire epidemic	29.6% (29.6–29.6)	2 922 454	18 758 578
Social distancing measures			
From week 2	34.7% (34.7–34.7)	0	n.a.
From week 4	34.7% (34.6–34.7)	0	n.a.
From week 8	34.1% (33.3–34.7)	343 818	n.a.

n.a., Not applicable.

For age categories see text (Control measures section).

* Value in parentheses represent the 5–95 percentile values of the attack rate estimates.

increase the probability of stochastic extinction during the initial phases of the epidemic, due to a low number of infectious individuals. A VE of 50% for all categories considered would affect the AR estimates, but only when considering adequate VE at 4 months. In fact, the AR would be 6–8 percentage points higher than the AR assuming a VE of 70%, with a remarkable difference in terms of the number of avoided cases (Table 3). The impact of combined control measures (pre-pandemic vaccine in all categories and AVP and/or social distancing measures), compared to the baseline dynamics of the influenza pandemic, is shown in Figure 2.

Sensitivity analysis

The results of the sensitivity analysis are shown in Figure 3. For $R_0=1.6$, the epidemic could be

mitigated with moderate efforts; all strategies would be successful independently of the timing of vaccination, of the duration of providing AVP, and of the timing of social distancing measures. For $R_0=1.8$, vaccinating the target categories with a pre-pandemic vaccine, providing AVP for 16 weeks, and implementing social distancing measures for 4 weeks would reduce the AR from 35% to 10%. For $R_0=2$, this combination of control measures would result in a less marked decrease in the AR, from 42% to about 20%.

CONCLUSIONS

Our results, considering an R_0 value of 1.8, confirmed the need to combine different control measures [7–9]. In fact, none of the single measures was shown to be effective in containing the pandemic, with the AR

Table 2. Effectiveness of combined control measures on the dynamics of an influenza pandemic with an R_0 of 1.8 and an attack rate of 35%, with 70% vaccine effectiveness (VE)

Interventions	Attack rate*	Avoided cases	Treated individuals	
			With vaccine	With antiviral
Adequate VE at 5 months (VE = 70%)				
Social distancing measures from week 2				
Antiviral for 2 weeks	24.6% (17.5–29.2)	5 787 603	25 821 426	55
Antiviral for 4 weeks	23.6% (15.4–26.9)	6 360 633	25 825 375	182
Antiviral for 8 weeks	22.1% (15.5–26.0)	7 220 178	25 831 314	717
Antiviral for 16 weeks	18.3% (11.3–22.0)	9 397 697	25 837 928	2 58 992
Antiviral for the entire epidemic	13.0% (6.2–16.7)	12 434 757	25 837 928	8 224 930
Social distancing measures from week 4				
Antiviral for 2 weeks	23.7% (15.1–28.5)	6 303 330	25 824 246	51
Antiviral for 4 weeks	22.7% (12.7–27.6)	6 876 360	25 828 371	373
Antiviral for 8 weeks	20.5% (12.3–25.1)	8 137 026	25 835 232	1690
Antiviral for 16 weeks	16.6% (10.5–21.2)	10 371 848	25 837 926	1 59 521
Antiviral for the entire epidemic	11.3% (5.5–15.8)	13 408 909	25 837 928	7 177 152
Adequate VE at 4 months (VE = 70%)				
Social distancing measures from week 2				
Antiviral for 2 weeks	12.6% (8.7–16.9)	12 663 963	25 837 928	55
Antiviral for 4 weeks	11.9% (8.0–14.3)	13 065 084	25 837 928	182
Antiviral for 8 weeks	10.9% (7.9–13.3)	13 638 114	25 837 928	717
Antiviral for 16 weeks	9.0% (5.8–10.7)	14 726 878	25 837 928	247 028
Antiviral for the entire epidemic	5.0% (1.8–7.0)	17 018 999	25 837 928	3 193 698
Social distancing measures from week 4				
Antiviral for 2 weeks	12.0% (7.9–15.9)	13 007 781	25 837 928	51
Antiviral for 4 weeks	11.5% (6.8–14.9)	13 294 296	25 837 928	373
Antiviral for 8 weeks	10.1% (6.6–12.6)	14 096 538	25 837 928	1690
Antiviral for 16 weeks	8.4% (5.3–10.2)	15 070 696	25 837 928	1 52 056
Antiviral for the entire epidemic	4.2% (1.4–6.4)	17 477 424	25 837 928	2 673 736

* Value in parentheses represent the 5–95 percentile values of the attack rate estimates.

decreasing at most from 35% to 24%. Combining measures would be more effective, especially if using the pre-pandemic vaccine (reaching VE at 4 months). In this case, the AR would be 4.2%, but this would require an extremely high number of AVP doses. Providing AVP for 16 weeks only would increase AR to 8.4%, which is similar to that observed during severe seasonal epidemics [21], with a considerable reduction in the number of doses provided. Moreover, if the time to reach adequate VE were 5 months, assuming a different VE (i.e. 70% or 50% in all categories) would not substantially affect the AR. However, if the time to reach adequate VE were 4 months, a VE of 70% would result in an AR of 4.2%, compared to 11.0% if assuming a VE of 50% (i.e. a 50% difference in the AR). In any case, using a less effective vaccine (i.e. with a VE of 50%) would nonetheless allow the pandemic to be contained, with an AR below 18% (range 11.0–18.1%).

Combining different measures markedly increased the probability of stochastic extinction during the early phases of the pandemic. To the best of our knowledge, most of the SEIR models used to simulate a pandemic do not consider the stochastic factors, which can strongly influence the dynamics of the pandemic in its early phases. However, we assumed that no other infectious individuals would enter the country after the few initial cases. If we were to assume that infectious individuals continued to enter the country, then stochastic extinction would be less important.

Another important finding is that the decrease in the AR would depend on which target groups were vaccinated. If a pandemic were to occur, vaccine supplies would be limited and the target groups would have to be prioritized (i.e. personnel of essential services, elderly persons and persons with chronic disease, children and young adults, and healthy adults)

Table 3. Effectiveness of combined control measures on the dynamics of an influenza pandemic with an R_0 of 1.8 and an attack rate of 35%, with 50% vaccine effectiveness (VE)

Interventions	Attack rate*	Avoided cases	Treated individuals	
			With vaccine	With antiviral
Adequate VE at 5 months (VE = 50%)				
Social distancing measures from week 2				
Antiviral for 2 weeks	26.4% (18.8–30.0)	4 756 151	25 821 812	55
Antiviral for 4 weeks	26.0% (19.9–28.8)	4 985 363	25 826 670	183
Antiviral for 8 weeks	25.0% (17.8–27.8)	5 558 394	25 830 892	745
Antiviral for 16 weeks	22.1% (17.5–24.8)	7 220 182	25 837 928	258 992
Antiviral for the entire epidemic	16.5% (11.9–19.4)	10 429 151	25 837 928	10 494 921
Social distancing measures from week 4				
Antiviral for 2 weeks	26.0% (20.7–29.8)	4 985 363	25 824 256	51
Antiviral for 4 weeks	25.4% (19.5–29.0)	5 329 182	25 828 178	367
Antiviral for 8 weeks	23.8% (19.2–27.2)	6 246 030	25 835 286	1665
Antiviral for 16 weeks	20.9% (16.9–24.1)	7 907 818	25 837 926	159 520
Antiviral for the entire epidemic	15.4% (11.5–18.7)	11 059 485	25 837 928	9 763 649
Adequate VE at 4 months (VE = 50%)				
Social distancing measures from week 2				
Antiviral for 2 weeks	18.1% (15.7–20.7)	9 512 302	25 837 928	55
Antiviral for 4 weeks	17.7% (15.8–19.5)	9 741 515	25 837 928	183
Antiviral for 8 weeks	17.3% (15.4–18.7)	9 970 727	25 837 928	745
Antiviral for 16 weeks	16.2% (15.1–17.0)	10 601 060	25 837 928	250 341
Antiviral for the entire epidemic	11.4% (8.7–12.6)	13 351 606	25 837 928	7 232 024
Social distancing measures from week 4				
Antiviral for 2 weeks	17.9% (15.9–20.3)	9 626 909	25 837 928	51
Antiviral for 4 weeks	17.6% (15.8–19.8)	9 798 818	25 837 928	367
Antiviral for 8 weeks	16.9% (15.7–18.3)	10 199 939	25 837 928	1664
Antiviral for 16 weeks	16.0% (15.1–16.7)	10 715 666	25 837 928	154 130
Antiviral for the entire epidemic	11.0% (8.6–12.2)	13 580 818	25 837 928	6 983 830

* Value in parentheses represent the 5–95 percentile values of the attack rate estimates.

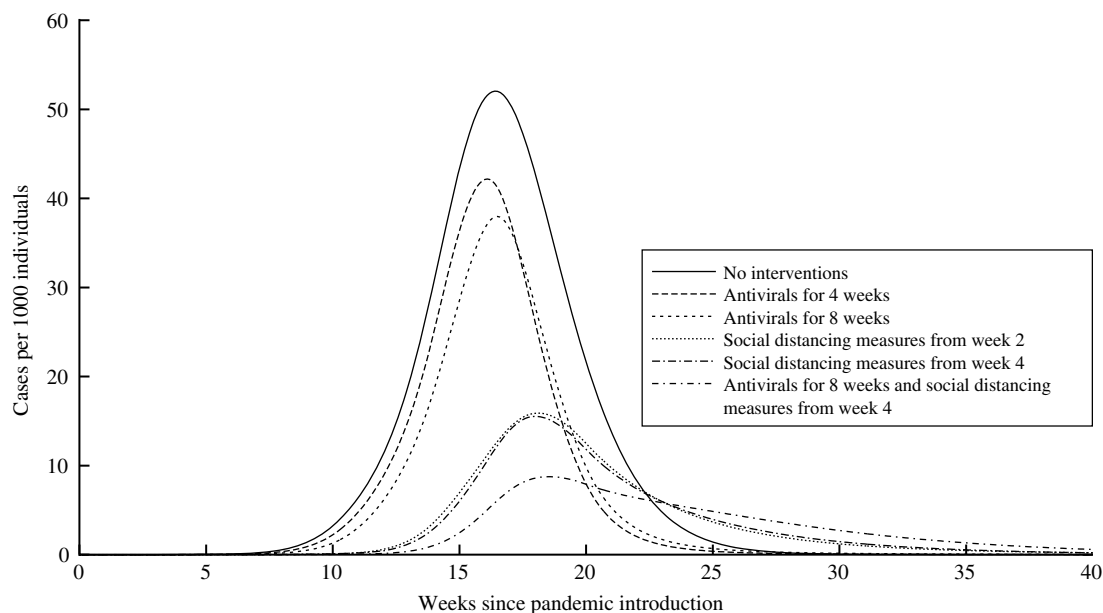


Fig. 2. Impact of different combinations of control measures considering the use of a pre-pandemic vaccine provided to all categories (I–IV).

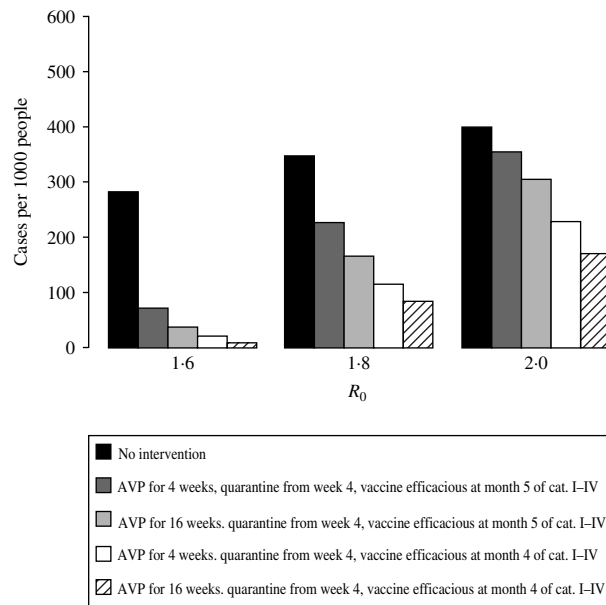


Fig. 3. Total attack rates for different values of R_0 , with no control measures or selected control measures.

[19], requiring the vaccination of 26 million persons with two doses, that would be very difficult to put in practice if a pandemic did occur. However, as reported in other studies [9], our results showed that, independently of the VE, the vaccination of children and young adults would considerably reduce the incidence in other age groups (i.e. resulting in ‘herd immunity’), probably because of the important role of children and adolescents in the spread of influenza, as also observed in inter-pandemic periods [22].

In interpreting our results, some limitations need to be considered. First, we assumed that AVP provided to household contacts would decrease transmission within households but not in other contexts, which could have resulted in an underestimate of the effect of this measure. Second, the parameters used in our model obviously influenced the time estimated for the pandemic to evolve, although our estimate is similar to those obtained in other studies based on deterministic SEIR models on a global [13] or local [23] scale or individual-based models [6–9]. We examined this issue by performing a sensitivity analysis; clearly, the success of control strategies would be strongly influenced by the R_0 : for $R_0 = 1.6$, all strategies would be quite successful, whereas for $R_0 = 2$ only the combined strategy with a pre-pandemic vaccine would satisfactorily mitigate the pandemic. Although the absolute effect of control

strategies is strongly influenced by the different values of R_0 , the relative worth of strategies are independent from the different R_0 values.

Another important limitation is that mathematical models cannot take into account the fact that the past influenza pandemics in Europe and Italy occurred over two consecutive winters, with the highest AR in the second winter [24–26]. This two-wave pattern is probably an effect of the closing of schools during the summer. Thus our model probably depicts a ‘worst case scenario’, which could be useful in evaluating control measures [9].

Our simulations show that appropriate and prompt measures, when combined, could be effective in containing an influenza pandemic. Timing is also essential, and measures that at first glance appear to be less important, such as increasing social distancing, could be extremely useful in delaying the epidemic peak and thus providing more time for vaccines to be produced. Implementing such measures, however, would entail organizing a variety of both medical and non-medical resources, and some measures, such as the closure of schools, would also have a social impact.

APPENDIX

The equations of the model are

$$\left. \begin{aligned} \dot{S}_i^p &= -\dot{S}_i^p \sum_{j,q} \beta_{i,j}^{p,q} \frac{I_j^q}{N_j^q} \\ \dot{E}_i^p &= \dot{S}_i^p \sum_{j,q} \beta_{i,j}^{p,q} \frac{I_j^q}{N_j^q} - \eta E_i^p \\ \dot{I}_i^p &= \eta E_i^p - \gamma I_i^p \\ \dot{R}_i^p &= \gamma I_i^p \end{aligned} \right\}$$

where $1/\eta$ and $1/\gamma$ represent, respectively, the mean length of the latent and the infectious periods and $\beta_{i,j}^{p,q}$ is the transmission rate between an individual of class i in region p and an individual of class j in region q .

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DECLARATION OF INTEREST

None.

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